
Mapping benthic traits on the northwestern shelf of the Black Sea: a trait-based modelling approach

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The picture on the cover of this thesis, showing the Black Sea, is a personal photograph taken on October 24, 2024 from a beach in Constanța (Romania).

SUMMARY

The biological communities found at the bottom of the sea, the macrobenthos, play a pivotal role in marine ecosystems by influencing benthic-pelagic coupling and biogeochemical cycles in shallow coastal waters. Through ecological processes such as biomixing (transport of particles) and bioirrigation (transport of solutes), macrobenthos redistribute food resources and create micro-niches for other organisms, and thus are considered as key ecosystem engineers. Due to their relative long lifespans and limited mobility, macrobenthic animals integrate water and sediment quality over time and are widely used as bioindicators of marine ecosystem health. Therefore, macrobenthos are essential for maintaining biodiversity and ecosystem functioning, making them a primary conservation target for biodiversity management. We are currently witnessing a steady decline in marine biodiversity with serious damage to ecosystem functioning, stability and for human well-being. To address this issue, Ecosystem-Based Management (EBM) strategy has become as a central framework for marine conservation. This means that we need maps of biodiversity to design Marine Protected Areas (MPAs) based on ecological and environmental considerations.

Despite the availability of international databases compiling millions of benthic species records, benthic biodiversity remains poorly mapped at larger scale, constraining the establishment of MPAs, in contrast to the terrestrial realm where biodiversity mapping is far more developed. Indeed, the development of mapping and predictive modelling tools for the macrobenthos is particularly challenging due to the high heterogeneity of the benthic ecosystem and the fragmentary information available for benthic habitats. The lack of maps informing on the functions of the benthos (e.g. biomixing, bioirrigation) is also one of the major reasons why ocean models do not consider the impact of macrobenthos on benthic-pelagic coupling and biogeochemical cycles at the shelf scale.

This PhD thesis addresses this critical gap by developing predictive modelling tools capable of mapping the distribution of the characteristics of macrobenthos at the shelf scale. The following question is specifically addressed: **“How can we upscale local observations of benthic biodiversity to map benthic functions and vulnerabilities at the shelf scale?” (Figure 0.1)**. A functional approach to biodiversity has been adopted, meaning that living communities are defined by the main characteristics of their species (their traits) rather than their taxonomic composition. A trait-based approach is expected to be more closely related to environmental characteristics than species composition itself. The methodology is applied and developed for the northwestern shelf (NWS) of the Black Sea, a region strongly impacted by eutrophication and human pressures. The Black Sea is an ideal case for studying the relationships between the traits of the macrobenthos and the diversity of bottom abiotic conditions, particularly the oxygen gradient from well oxygenated coastal waters to oxygen-depleted deep waters.

Our methodology (**Chapter 2**) combines field sampling, functional biodiversity, statistical multivariate analyses as well as statistical (Trait Distribution Models, TDMs) and mechanistic models (ocean numerical model). In more detail, we compile macrozoobenthos occurrence as well as substratum type from 237 sampling stations covering various benthic habitats over the northwestern shelf of the Black Sea. Abiotic descriptors are provided by a biogeochemical - physics coupled model, at high temporal and spatial resolution, run in an operational mode by Copernicus Marine Service (CMEMS). We use a combination of a dozen biogeochemical (e.g. bottom oxygen and flux of organic carbon to the bottom) and physical drivers (e.g. bottom temperature and shear stress) as preliminary predictors of the distribution of traits. We compile traits with an effect on environmental conditions (e.g. burrowing depth, mobility) and traits that respond to environmental pressures (e.g. offspring type, age at sexual maturity). Statistical analyses allow us to identify the environmental variables that explain the spatial distribution of the chosen traits or combination of them into indicators. We combine RLQ analysis and the 4th corner method as proposed by Dray et al. (2014). Briefly, RLQ is a multivariate technique that ordinated scores to summarise the joint structure among abiotic conditions (matrix R), species (matrix L) and their respective traits (matrix Q), while the fourth-corner method tests for individual trait-environment relationships. From this information, we develop Trait Distribution Models to establish the link between the trait and the environmental conditions to which it is found to be suited and to predict its spatial distribution using maps of environmental conditions provided by an ocean numerical model. A trait-based approach is thus an adequate way of linking the wealth of information provided by ocean models to functional biodiversity considerations.

First, we are interested in the status of the Black Sea itself as it provides a unique opportunity to study the long-term relationships between the macrozoobenthos and environmental conditions (**Chapter 3**). Although the Black Sea is well-studied in terms of macrozoobenthic communities, there is still a lack of large-scale studies over the entire NWS investigating the relationships between species and abiotic conditions. This situation becomes even more complicated when it comes to comparing macrozoobenthos over decades due to the difference in sampling protocol and taxonomic issues. For example, the macrozoobenthos has experienced a well-established decline in biodiversity in relation to eutrophication and bottom hypoxia since the end of the sixties. However, nowadays, the nature of the current benthic communities remains questionable. In this PhD thesis, we investigate the long-term changes in the status of macrozoobenthic communities between the mid-nineties (the period of post-eutrophication) and the early 2010s. Through an appropriate multivariate analytical approach, we relate these benthic community changes solely due to organic pollution variations. Our key results show signs of recovery with an increase in biodiversity and the proportion of species vulnerable to organic enrichment (e.g. long-lived bivalves, some crustaceans). These changes are directly related to a decrease in riverine loads and subsequent eutrophication and bottom hypoxia. However, some long-lived bivalves typical of this area still do not exhibit noticeable recovery, and the flourishing of invasive species, both suggest that the recovery process has not yet been completely achieved.

In addition, we propose to fulfil the lack of trait data for the macrobenthos from the Black Sea by providing an open-access dataset of biological traits for the macrobenthos sampled during previous field campaigns. Specifically for this PhD work, we compile a comprehensive dataset of traits for over 120 macrozoobenthic species (**Chapter 4**). The response traits, which are the traits that respond to environmental pressures (e.g. age at maturity, offspring type), can be combined to derive life-history strategies. Life-history strategies are a set of traits that occur together and optimise an organism's fitness in a particular environment.

These strategies inform on species ability to withstand external pressures by sorting vulnerable species (i.e. slow-growing, long-lived) from those tolerant to disturbance (i.e. fast-growing, short-lived). Additionally, we compile effect traits, which are traits that impact ecosystem functions (e.g. mobility, burrowing depth), that can be combined to estimate key ecological processes such as biomixing and bioirrigation.

Based on this extensive compilation of macrobenthic species and their respective traits, we explore the relationships between environmental predictors and traits using multivariate approaches such as RLQ and fourth-corner analyses. These relationships were then upscale to the entire NWS using Trait Distribution Models (TDMs), based on artificial neural networks. Our TDMs enabling us to map benthic traits, life-history strategy groups and ecological processes (e.g. biomixing, bioirrigation) over the NWS shelf (**Chapter 5**). Our key findings show that bottom oxygen and the type of substratum are strong predictors for the spatial distribution of traits. Specifically, coastal areas with high fluxes of organic matter to the bottom, such as near the Danube Delta, show benthic communities with deeper burrowing depths and higher ventilation and pumping activities in muddy sand substratum. The potentials of biomixing and bioirrigation are higher near the coast and then strongly decrease along the depth gradient. At deeper depths (around 40 - 60 m), areas with mixed and coarse substratum host epifaunal sessile communities with higher biodeposition potential (i.e. filtering of suspended particles and excretion on the seabed). At the edge of the shelf, permanently hypoxic waters are characterised by very low sediment biomixing and bioirrigation potentials and very low benthic biodiversity. Only opportunistic species with specific traits (such as homogamy) can cope with those harsh conditions. Mapping vulnerabilities (here life-history strategies group) and functions (here proxy of ecological processes) can support marine management strategies and help for decisions aimed at preserving marine biodiversity.

Building on this PhD work, (**Chapter 6**), further research could predict changes in the spatial distribution of traits using our developed TDMs and climate change scenarios as new predictors (extension of hypoxia, increase in temperature). The potential modification in the macrozoobenthos trait distribution and bottom environmental conditions will impact ecosystem functioning, in a way that can be assessed through a new diagenetic model formulation that incorporates information on biota characteristics. Currently, most ocean models ignore or simplify the variability of macrozoobenthos functions and their potential impact on benthic-pelagic fluxes and biogeochemical budgets. Thanks to the maps of ecological processes (biomixing, bioirrigation) from this PhD work, the variability of macrozoobenthic activities could be included in the parametrisation of a diagenetic model, modulating spatially biogenic mixing and biodiffusion coefficients. A benthic model constrained by seafloor biodiversity will constitute a significant step for the development of ocean models that consider the impact of environmental changes on benthic life and its ability to deliver key ecosystem functions such as denitrification and carbon sequestration.

This thesis introduces a trait-based approach for mapping benthic functional biodiversity over the NWS of the Black Sea, bridging local species-level observations with large-scale applications in EBM. The coupling of functional biogeographic models and mechanistic biogeochemical models offers the way to connect environment, biodiversity, and functions at the ecosystem scale. Beyond its application to the Black Sea, this approach is transferable to other coastal systems with varying species compositions. By delivering maps of benthic functions and vulnerabilities, this PhD work directly supports the preservation of the Good Environmental Status (GES) of the Black Sea and provides a scientifically robust basis for the future design of MPAs design and the conservation of marine biodiversity.

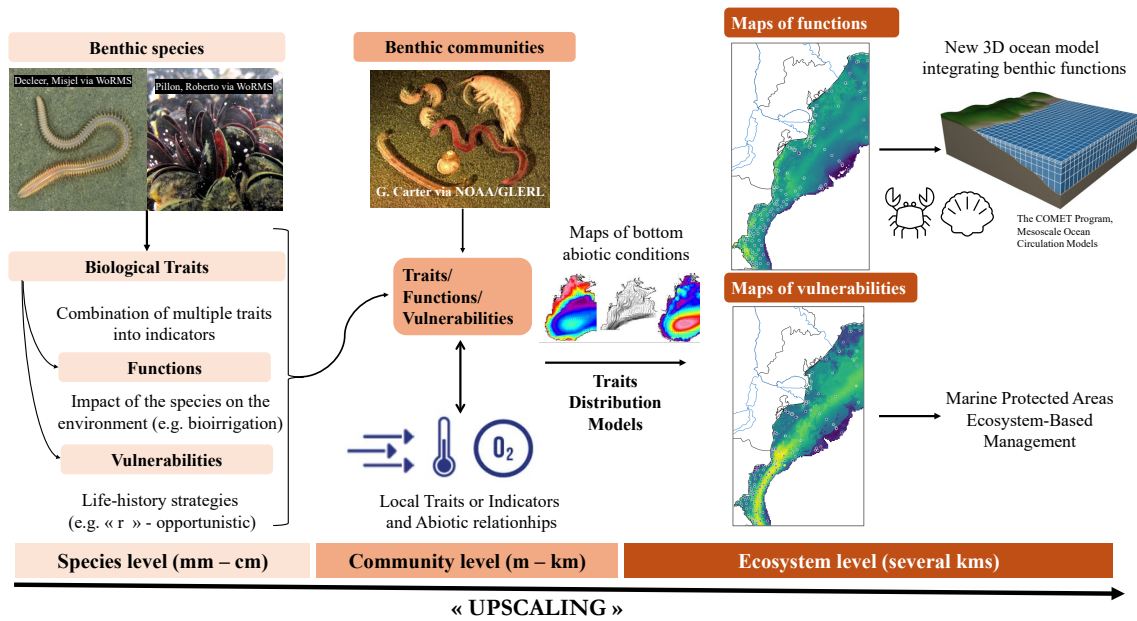


Figure 0.1: Graphical summary of the PhD thesis from benthic species to the maps of benthic vulnerabilities and functions through a trait-based modelling approach.

SAMENVATTING

De biologische gemeenschappen die op de zeebodem voorkomen, de macrobenthos, spelen een cruciale rol in mariene ecosystemen door hun invloed op de bentisch-pelagische koppeling en biogeochemische cycli in ondiepe kustwateren. Door ecologische processen zoals bioturbatie (transport van deeltjes) en bioirrigatie (transport van opgeloste stoffen) herverdelen macrobenthos voedselbronnen en creëren ze microniches voor andere organismen. Ze worden dan ook beschouwd als belangrijke ecosysteemingenieurs. Vanwege hun relatief lange levensduur en beperkte mobiliteit integreren macrobenthische dieren in de loop van de tijd de kwaliteit van water en sediment en worden ze veel gebruikt als bio-indicatoren voor de gezondheid van mariene ecosystemen. Daarom is macrobenthos essentieel voor het behoud van de biodiversiteit en het functioneren van ecosystemen, waardoor ze een primair doelwit zijn voor het behoud van de biodiversiteit. We zijn momenteel getuige van een gestage afname van de mariene biodiversiteit, met ernstige schade aan het functioneren en de stabiliteit van ecosystemen en aan het welzijn van de mens. Om dit probleem aan te pakken, is de strategie van ecosysteemgericht beheer een centraal kader voor mariene bescherming geworden. Dit betekent dat we kaarten van de biodiversiteit nodig hebben om mariene beschermde gebieden te ontwerpen op basis van ecologische en milieuoverwegingen.

Ondanks de beschikbaarheid van internationale databases met miljoenen gegevens over bentische soorten, blijft de bentische biodiversiteit op grotere schaal slecht in kaart gebracht, wat de instelling van mariene beschermde gebieden's belemmert, in tegenstelling tot het terrestrische domein, waar de in kaart brengen van biodiversiteit veel verder ontwikkeld is. De ontwikkeling van instrumenten voor het in kaart brengen en voorspellende modellen voor het macrobenthos is inderdaad bijzonder uitdagend vanwege de grote heterogeniteit van het bentische ecosysteem en de fragmentarische informatie die beschikbaar is over bentische habitats. Het gebrek aan kaarten met informatie over de functies van het benthos (bijvoorbeeld bioturbatie, bio-irrigatie) is ook een van de belangrijkste redenen waarom oceanmodellen geen rekening houden met de impact van macrobenthos op de koppeling tussen benthos en pelagische ecosystemen en op biogeochemische cycli op het continentaal plat.

Dit proefschrift pakt deze kritieke leemte aan door geavanceerde voorspellende modelleringsinstrumenten te ontwikkelen waarmee de verspreiding van de kenmerken van macrobenthos op het continentaal plat in kaart kan worden gebracht. De volgende vraag wordt specifiek behandeld: **“Hoe kunnen we lokale observaties van bentische biodiversiteit opschalen om bentische functies en kwetsbaarheden op het continentaal plat in kaart te brengen?”** (Figuur 0.2). Er is gekozen voor een functionele benadering van biodiversiteit, wat betekent dat levende gemeenschappen worden gedefinieerd aan de hand van de belangrijkste kenmerken van hun soorten (hun eigenschappen) in plaats van hun taxonomische samenstelling. Een op eigenschappen gebaseerde benadering zal naar verwachting nauwer verband houden met milieukenmerken dan de soortensamenstelling zelf.

De methodologie wordt toegepast en ontwikkeld voor het noordwestelijke plat van de Zwarte Zee, een regio die sterk wordt beïnvloed door eutrofiëring en menselijke druk. De Zwarte Zee is een ideaal geval om de relaties te bestuderen tussen de eigenschappen van het macrobenthos en de diversiteit van de abiotische omstandigheden op de bodem, met name de zuurstofgradiënt van goed zuurstofrijke kustwateren naar zuurstofarme diepe wateren.

Onze methodologie (**Hoofdstuk 2**) combineert veldbemonstering, functionele biodiversiteit, statistische analyses en statistische (Eigenschap Verspreiding Modellen) en mechanistische modellen (numeriek oceaanmodel). Meer in detail hebben we de aanwezigheid van macrozoöbenthos en het type substraat verzameld uit 237 bemonsteringsstations die verschillende bentische habitats op het noordwestelijke plat van de Zwarte Zee bestrijken. Abiotische descriptoren worden geleverd door een gekoppeld biogeochemisch-fysisch model met een hoge temporele en ruimtelijke resolutie, dat in operationele modus wordt uitgevoerd door Copernicus Marine Service (CMEMS). We gebruiken een combinatie van een tiental biogeochemische (bijv. zuurstofgehalte op de bodem en flux van organische koolstof naar de bodem) en fysische factoren (bijv. bodemtemperatuur en schuifspanning) als voorlopige voorspellers van de verspreiding van eigenschappen. We compileren eigenschappen die van invloed zijn op de omgevingsomstandigheden (bijv. graafdiepte, mobiliteit) en eigenschappen die reageren op omgevingsdruk (bijv. type nakomelingen, leeftijd bij geslachtsrijpheid). Statistische analyses stellen ons in staat om de omgevingsvariabelen te identificeren die de ruimtelijke verspreiding van de gekozen eigenschappen of combinaties daarvan in indicatoren verklaren. We combineren RLQ-analyse en de 4th corner-methode zoals voorgesteld door Dray et al. (2014). Kort gezegd is RLQ een multivariate techniek die scores coördineert om de gezamenlijke structuur tussen abiotische omstandigheden (matrix R), soorten (matrix L) en hun respectieve eigenschappen (matrix Q) samen te vatten, terwijl de vierhoekmethode individuele eigenschap-omgevingsrelaties test. Op basis van deze informatie ontwikkelen we Eigenschap Verspreiding Modellen om het verband tussen de eigenschap en zijn “niche” (d.w.z. de omgevingsomstandigheden waaraan hij geschikt blijkt te zijn) vast te stellen en om zijn ruimtelijke verspreiding te voorspellen met behulp van kaarten van omgevingsomstandigheden die door numerieke oceaanmodellen worden verstrekt. Een op eigenschappen gebaseerde benadering is dus een geschikte manier om de schat aan informatie die door oceaanmodellen wordt verstrekt, te koppelen aan functionele overwegingen op het gebied van biodiversiteit.

Ten eerste zijn we geïnteresseerd in de toestand van de Zwarte Zee zelf, omdat deze een unieke kans biedt om de langetermijnrelaties tussen het macrobenthos en de omgevingsomstandigheden te bestuderen (**Hoofdstuk 3**). Hoewel de Zwarte Zee goed bestudeerd is wat betreft macrozoöbenthische gemeenschappen, ontbreken er nog steeds grootschalige studies over de gehele noordwestelijke plat waarin de relaties tussen soorten en abiotische omstandigheden worden onderzocht. Deze situatie wordt nog ingewikkelder wanneer het gaat om het vergelijken van macrozoöbenthos over tientallen jaren, vanwege verschillen in bemonsteringsprotocollen en taxonomische kwesties. Zo heeft het macrozoöbenthos sinds het einde van de jaren zestig een duidelijk vastgestelde afname in biodiversiteit ondergaan in verband met eutrofiëring en bodemhypoxie. Tegenwoordig blijft de aard van de huidige bentische gemeenschappen echter onduidelijk. In dit proefschrift onderzoeken we de langetermijnveranderingen in de toestand van macrozoöbenthische gemeenschappen tussen het midden van de jaren negentig (de periode na eutrofiëring) en het begin van de jaren 2010. Door middel van een geschikte multivariate analytische benadering brengen we deze veranderingen in bentische gemeenschappen uitsluitend in verband met variaties in organische vervuiling.

Onze belangrijkste resultaten wijzen op tekenen van herstel, met een toename van de biodiversiteit en het aandeel van soorten die kwetsbaar zijn voor organische verrijking (bijvoorbeeld langlevende tweekleppigen en sommige schaaldieren). Deze veranderingen houden rechtstreeks verband met een afname van de rivierafvoer en de daaropvolgende eutrofiëring en hypoxie op de bodem. Sommige langlevende tweekleppigen die typisch zijn voor dit gebied vertonen echter nog steeds geen merkbaar herstel, en de bloei van invasieve soorten wijst erop dat het herstelproces nog niet volledig is voltooid.

Daarnaast stellen we voor om het gebrek aan gegevens over de eigenschappen van het macrobenthos uit de Zwarte Zee aan te vullen door een openbaar toegankelijke dataset aan te bieden met biologische eigenschappen van het macrobenthos dat tijdens eerdere veldcampagnes is bemonsterd. Specifiek voor dit doctoraatsonderzoek stellen we een uitgebreide dataset samen met eigenschappen van meer dan 120 macrozoöbentische soorten (**Hoofdstuk 4**). De responskenmerken, dat zijn de kenmerken die reageren op omgevingsdruk (bijv. geslachtsrijpe leeftijd, type nakomelingen), kunnen worden gecombineerd om levensgeschiedenisstrategieën af te leiden. Levensgeschiedenisstrategieën zijn een reeks kenmerken die samen voorkomen en de fitheid van een organisme in een bepaalde omgeving optimaliseren. Deze strategieën geven informatie over het vermogen van soorten om externe druk te weerstaan door kwetsbare soorten (d.w.z. langzaam groeiende, langlevende soorten) te onderscheiden van soorten die tolerant zijn voor verstoring (d.w.z. snel groeiende, kortlevende soorten). Daarnaast stellen we effectkenmerken samen, dit zijn kenmerken die van invloed zijn op ecosysteemfuncties (bijv. mobiliteit, graafdiepte), die kunnen worden gecombineerd om belangrijke ecologische processen zoals bioturbatie en bio-irrigatie te schatten.

Op basis van deze uitgebreide compilatie van macrobenthische soorten en hun respectieve kenmerken onderzoeken we de relaties tussen omgevingsvoorspellers en kenmerken met behulp van multivariate benaderingen zoals RLQ- en fourth-corner-analyses. Deze relaties werden vervolgens opgeschaald naar de gehele plat met behulp van Eigenschappen Verspreiding Modellen, gebaseerd op kunstmatige neurale netwerken. Met onze modellen kunnen we bentische eigenschappen, levensgeschiedenisstrategiegroepen en ecologische processen (bioturbatie en bioirrigatie) in kaart brengen over het noordwestelijke plat (**Hoofdstuk 5**). Onze belangrijkste bevindingen tonen aan dat zuurstof op de bodem en het type substraat sterke voorspellers zijn voor de ruimtelijke verspreiding van eigenschappen. Met name kustgebieden met een hoge toevoer van organisch materiaal naar de bodem, zoals in de buurt van de Donau delta, vertonen bentische gemeenschappen met grotere graafdieptes en hogere ventilatie- en pompsnelheden in modderig zandsubstraat. Het potentieel voor bioturbatie en bioirrigatie is groter in de buurt van de kust en neemt vervolgens sterk af naarmate de diepte toeneemt. Op grotere diepten (ongeveer 40-60 m) herbergen gebieden met gemengd en grof substraat epifaunale sessiele gemeenschappen met een hoger potentieel voor biodepositie (d.w.z. filtering van zwevende deeltjes en uitscheiding op de zeebodem). Aan de rand van het plat worden permanent hypoxische wateren gekenmerkt door een zeer laag potentieel voor bioturbatie en bio-irrigatie van sedimenten en een zeer lage bentische biodiversiteit. Alleen opportunistische soorten met specifieke eigenschappen (zoals homogamie) kunnen deze barre omstandigheden het hoofd bieden. Het in kaart brengen van kwetsbaarheden (hier: levensgeschiedenisstrategieën) en functies (hier: ecologische processen) kan mariene beheerstrategieën ondersteunen en helpen bij het nemen van beslissingen die gericht zijn op het behoud van de mariene biodiversiteit.

Voortbouwend op dit doctoraatswerk (**Hoofdstuk 6**) zou verder onderzoek veranderingen in de ruimtelijke verspreiding van eigenschappen kunnen voorspellen met behulp van onze ontwikkelde eigenschappen verspreiding modellen en klimaatveranderingsscenario's als nieuwe voorspellers

(uitbreiding van hypoxie, stijging van de temperatuur). De mogelijke verandering in de verspreiding van macrobenthos-eigenschappen en de bodemomstandigheden zal een impact hebben op het functioneren van het ecosysteem, die kan worden beoordeeld aan de hand van een nieuw diagenetisch model dat informatie over biota-kenmerken bevat. Momenteel negeren of vereenvoudigen de meeste oceanmodellen de variabiliteit van macrobenthosfuncties en hun potentiële impact op bentisch-pelagische fluxen en biogeochemische budgetten. Dankzij de kaarten van ecologische processen (bioturbatie, bioirrigatie) uit dit doctoraatswerk kan de variabiliteit van macrobenthische activiteiten worden opgenomen in de parametrisering van een diagenetisch model, waardoor ruimtelijk biogene meng- en biodiffusiecoëfficiënten worden gemoduleerd. Een bentisch model dat wordt beperkt door de biodiversiteit op de zeebodem zal een belangrijke stap vormen voor de ontwikkeling van oceanmodellen die rekening houden met de impact van milieuveranderingen op het bentische leven en het vermogen daarvan om belangrijke ecosysteemfuncties te vervullen, zoals denitrificatie en koolstofvastlegging. Dit proefschrift presenteert een op kenmerken gebaseerde benadering voor het in kaart brengen van de functionele bentische biodiversiteit op het noordwestelijke continentaal plat van de Zwarte Zee, waarbij een verband wordt gelegd tussen lokale waarnemingen op soort-niveau en grootschalige toepassingen in het kader van ecosysteembeheer. Door modellen voor de verspreiding van eigenschappen te koppelen aan mechanistische biogeochemische modellen kunnen het milieu, de biodiversiteit en de functies op ecosystemniveau met elkaar in verband worden gebracht. Deze aanpak is niet alleen toepasbaar op de Zwarte Zee, maar ook op andere kustsystemen met een andere soortensamenstelling. Door kaarten van bentische functies en kwetsbaarheden te verstrekken, draagt dit doctoraatswerk rechtstreeks bij tot het behoud van de goede ecologische toestand van de Zwarte Zee en biedt het een solide wetenschappelijke basis voor het toekomstige ontwerp van beschermde mariene gebieden en het behoud van de mariene biodiversiteit.

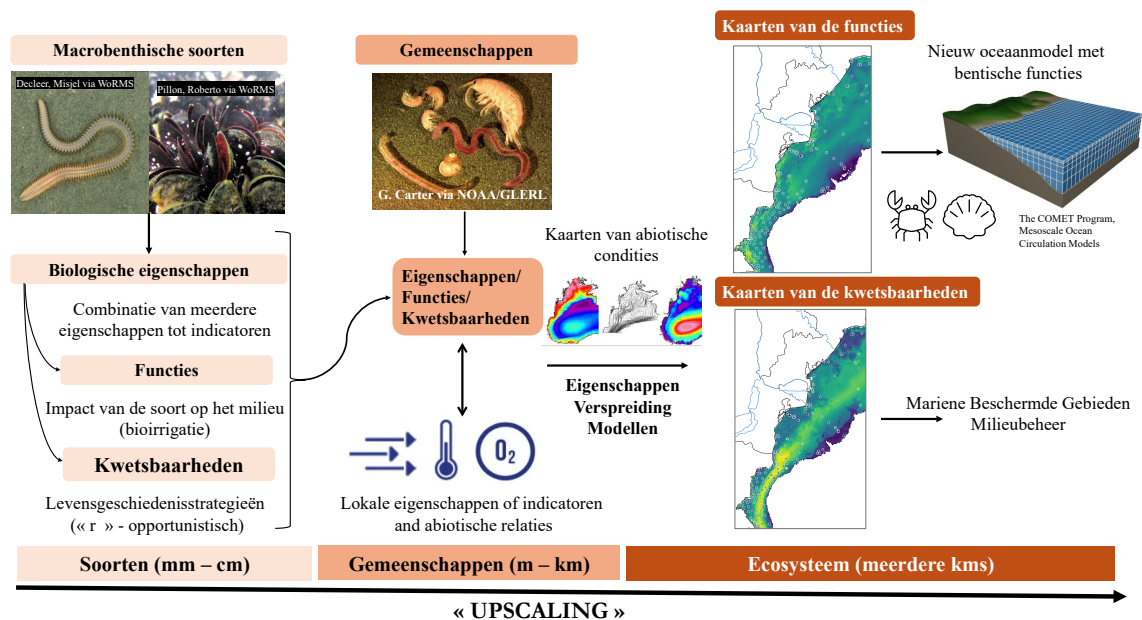


Figure 0.2: Grafische samenvatting van het proefschrift: van bentische soorten tot kaarten van bentische kwetsbaarheden en functies via een op eigenschappen gebaseerde modelleringsaanpak.

Translated from English to Dutch with DeepL.

RÉSUMÉ

Les communautés biologiques présentes au fond de la mer, le macrobenthos, jouent un rôle central dans les écosystèmes marins en influençant le couplage benthique-pélagique et les cycles biogéochimiques dans les eaux côtières peu profondes. Grâce à des processus écologiques tels que la bioturbation (remaniement des sédiments) et la bioirrigation (ventilation des terriers), le macrobenthos redistribue les ressources alimentaires et crée des micro-niches pour d'autres organismes. Les espèces macrobenthiques sont donc considérées comme des espèces ingénieures qui modifient significativement leur environnement. En raison de leur durée de vie relativement longue et de leur mobilité limitée, les espèces macrobenthiques intègrent la qualité de l'eau et des sédiments sur le long terme et sont fréquemment utilisées comme bioindicateurs de l'état des écosystèmes marins. Ainsi, le macrobenthos est important pour le maintien de la biodiversité et le fonctionnement des écosystèmes, ce qui en fait une cible prioritaire à protéger. Nous assistons actuellement à un déclin constant de la biodiversité marine, qui porte gravement atteinte au fonctionnement et à la stabilité des écosystèmes ainsi qu'au bien-être humain. Pour faire face à ce problème, la stratégie de gestion écosystémique s'est imposée comme un outil central pour la conservation de la biodiversité marine. Cela signifie entre autres que nous avons besoin de cartes de la biodiversité pour concevoir des aires marines protégées basées sur des considérations écologiques et environnementales.

Malgré la présence de bases de données internationales compilant des millions d'observations des espèces du macrobenthos, la biodiversité benthique reste mal cartographiée à grande échelle, ce qui limite la création d'aires marines protégées, contrairement au domaine terrestre où la cartographie de la biodiversité est beaucoup plus développée. En effet, le développement d'outils de cartographie et de modélisation prédictive pour le macrobenthos est particulièrement difficile en raison de la grande hétérogénéité de l'écosystème benthique et du caractère fragmentaire des informations disponibles sur les habitats benthiques. L'absence d'information sur la distribution spatiale des fonctions du macrobenthos (par exemple, la bioturbation, la bioirrigation) est également l'une des principales raisons pour lesquelles les modèles océaniques ne considèrent pas l'impact du macrobenthos sur le couplage benthique-pélagique et les cycles biogéochimiques à l'échelle du plateau continental.

Cette thèse de doctorat vise à combler ce manque de cartes en développant des outils de modélisation prédictive capables de cartographier les caractéristiques du macrobenthos à l'échelle du plateau continental. Ce travail aborde plus précisément la question suivante: **«Comment pouvons-nous extrapoler les observations ponctuelles de la biodiversité benthique afin de cartographier les fonctions et les vulnérabilités du macrobenthos à l'échelle du plateau continental ?»** (Figure 0.3). Une approche fonctionnelle de la biodiversité a été choisie, ce qui implique que les communautés benthiques sont définies par les principales caractéristiques de leurs espèces (leurs traits) à la place de leur composition taxonomique. Une approche basée sur les traits est préférable car les traits sont plus étroitement liés aux conditions environnementales que les espèces elles-mêmes.

La méthodologie est appliquée et développée pour le plateau continental Nord-Ouest de la mer Noire, une région fortement touchée par l'eutrophisation et les pressions humaines. La mer Noire est un cas idéal pour étudier les relations entre les traits du macrobenthos et la diversité des conditions abiotiques du fond marin, en particulier le gradient d'oxygène depuis les eaux côtières bien oxygénées et les eaux profondes appauvries en oxygène.

Notre méthodologie (**Chapitre 2**) combine l'échantillonnage sur le terrain, la biodiversité fonctionnelle, les analyses statistiques multivariées ainsi que des modèles statistiques (modèles de distribution des traits) et mécanistiques (modèle océanique). Plus précisément, nous avons compilé des données in-situ sur les espèces macrobenthiques (occurrences, biomasses et abondances) ainsi que le type de substrat à partir de 237 stations couvrant des habitats benthiques différents sur le plateau nord-ouest de la mer Noire. Les descripteurs abiotiques sont fournis par un modèle couplé biogéochimique-physique, à haute résolution temporelle et spatiale, utilisé en mode opérationnel par le Copernicus Marine Service (CMEMS). Nous avons utilisé une douzaine de facteurs biogéochimiques (par exemple, l'oxygène et le flux de carbone organique vers les sédiments) et physiques (par exemple, la température et les courants de fond) comme prédicteurs de la distribution des traits. Nous avons également compilé les traits ayant un effet sur les conditions environnementales (par exemple, la profondeur de terrier et le type de mobilité) et les traits qui répondent aux pressions environnementales (par exemple, le type de progéniture et l'âge à maturité sexuelle). Les analyses statistiques nous ont permis d'identifier les variables environnementales qui expliquent la distribution spatiale des traits ou des indicateurs basés sur la combinaison de plusieurs traits. Nous avons combiné l'analyse RLQ et la statistique du 4ème coin, avec l'approche proposée par Dray et al. (2014). Brièvement, l'analyse RLQ est une méthode multivariée qui ordonne les trois matrices: R (conditions abiotiques), L (les espèces) et Q (les traits) afin de résumer la structure conjointe entre les trois matrices tandis que la statistique du 4ème coin teste les relations individuelles entre les traits et l'environnement. À partir de ces informations, nous avons développé des modèles de distribution des traits afin d'établir le lien entre le trait et sa « niche » (c'est-à-dire les conditions environnementales auxquelles il est adapté) et de prédire sa distribution spatiale à l'aide de cartes des conditions environnementales fournies par des modèles numériques océaniques. Une approche basée sur les traits est donc un moyen adéquat de relier la multitude d'informations fournies par les modèles océaniques à la biodiversité fonctionnelle.

Premièrement, nous nous sommes intéressés à l'état de la mer Noire en elle-même, car elle offre une occasion unique d'étudier les relations à long terme entre le macrobenthos et les conditions environnementales (**Chapitre 3**). Bien que la mer Noire ait fait l'objet de nombreuses études en ce qui concerne les communautés macrozoobenthiques, il y a toujours un manque d'études à grande échelle sur l'ensemble du plateau examinant les relations entre les espèces et les conditions abiotiques. Cette situation se complique encore davantage lorsqu'il s'agit de comparer l'état des communautés macrobenthiques sur plusieurs décennies en raison des différences dans les protocoles d'échantillonnage et des problèmes d'identification taxonomiques. Par exemple, le benthos a connu un déclin de sa biodiversité lié à l'eutrophisation et à l'hypoxie des fonds marins depuis la fin des années 60. Cependant, aujourd'hui, le statut des communautés benthiques actuelles reste incertain. Dans cette thèse de doctorat, nous avons étudié les changements à long terme de l'état des communautés macrozoobenthiques entre le milieu des années 90 (période post-eutrophisation) et le début des années 2010. Grâce à une approche analytique multivariée appropriée, nous avons liés les changements dans les communautés benthiques uniquement à cause de la réduction de la pollution organique.

Nos principaux résultats montrent des signes de rétablissement avec une augmentation de la biodiversité et de la proportion d'espèces vulnérables à l'enrichissement organique (par exemple, certains bivalves et crustacés). Ces changements sont directement liés à une diminution de l'eutrophisation par les apports des rivières ainsi que de l'hypoxie sur le fond. Cependant, certains bivalves à longue durée de vie typiques de cette région ne montrent toujours pas de signe de rétablissement notable, et la prolifération d'espèces envahissantes suggère que le processus de rétablissement n'est pas encore complètement achevé.

De plus, nous proposons de combler le manque de données sur les caractéristiques des espèces du macrobenthos de la mer Noire en fournissant un jeu de données, en libre accès, sur les traits biologiques du macrobenthos échantillonné lors de précédentes campagnes sur le terrain. Pour ce travail de doctorat, nous avons compilé un ensemble complet de données sur les traits de plus de 120 espèces macrozoobenthiques (**Chapitre 4**). Les traits de réponse, qui sont les traits qui réagissent aux pressions environnementales (par exemple, l'âge à la maturité sexuelle, le type de progéniture), peuvent être combinés pour déterminer les stratégies d'histoire de vie. Les stratégies d'histoire de vie sont un ensemble de caractéristiques qui apparaissent ensemble et optimisent le fitness d'un organisme dans un environnement particulier. Ces stratégies renseignent sur la capacité des espèces à résister aux pressions externes en triant les espèces vulnérables (c'est-à-dire à croissance lente et à longue durée de vie) de celles qui tolèrent les perturbations (c'est-à-dire à croissance rapide et à courte durée de vie). De plus, nous avons compilé des traits d'effet, qui sont les traits ayant un impact sur les fonctions de l'écosystème (par exemple, la mobilité, la profondeur de terrier), qui peuvent être combinés ensemble pour estimer des processus écologiques tels que la bioturbation et la bioirrigation.

À partir de cette compilation exhaustive des espèces macrobenthiques et de leurs traits respectifs, nous avons exploré les relations entre les variables environnementales et les traits à l'aide d'approches multivariées telles que la méthode RLQ et la statistique du 4ème coin. Ces relations ont ensuite été extrapolées à l'ensemble du plateau continental à l'aide de modèles de distribution des traits, basés sur des réseaux neuronaux artificiels. Nos modèles de distribution de traits nous ont permis de cartographier les traits de réponse et d'effet qui sont liés à l'environnement, les groupes de stratégies d'histoire de vie et les processus écologiques (par exemple, la bioturbation et la bioirrigation) à l'échelle du plateau continental Nord-Ouest de la mer Noire (**Chapitre 5**). Nos principaux résultats montrent que la concentration en oxygène et le type de substrat sont des prédicteurs importants de la distribution spatiale des traits. Plus précisément, les zones côtières où l'apport de matière organique vers le fond est élevé, comme près du delta du Danube, présentent des communautés benthiques avec des profondeurs d'enfouissement plus importantes et des activités de ventilation de terriers plus élevées dans les substrats sableux boueux. Les potentiels de bioturbation et de bioirrigation sont plus élevés près de la côte, puis diminuent drastiquement avec la profondeur. À des profondeurs plus élevées (environ 40 à 60 m), les zones avec un substrat mixte et grossier abritent des communautés d'épifaune sessile avec un potentiel de biodéposition plus élevé (c'est-à-dire filtration des particules en suspension et excrétion de ces particules sur le fond marin). Au bord du plateau continental, les milieux pauvres en oxygène se caractérisent par un très faible potentiel de bioturbation et de bioirrigation des sédiments et par une biodiversité benthique très faible. Seulement quelques espèces opportunistes présentant des caractéristiques spécifiques (telles que l'homogamie) peuvent faire face à ces conditions presque anoxiques (sans oxygène). La cartographie des vulnérabilités (ici les stratégies d'histoire de vie) et des fonctions (ici les processus écologiques) peut soutenir les stratégies de gestion marine et aider à la prise de décisions visant à préserver la biodiversité marine.

En s'appuyant sur ce travail de doctorat (**Chapitre 6**), des recherches supplémentaires permettraient de prédire les changements dans la distribution spatiale des traits à l'aide des outils que nous avons développés et de scénarios de changement climatique comme nouveaux prédicteurs (extension de l'hypoxie, augmentation de la température). La modification de la distribution des traits du macrobenthos et des conditions environnementales du fond marin aura un impact sur le fonctionnement de l'écosystème, qui pourrait être évalué à l'aide d'un modèle diagenétique amélioré intégrant des informations sur les caractéristiques du benthos. Actuellement, la plupart des modèles océaniques ignorent ou simplifient la variabilité des fonctions du macrobenthos et leur impact potentiel sur les flux benthiques-pélagiques et les bilans biogéochimiques. Grâce aux cartes des processus écologiques issues de ce travail de doctorat, la variabilité des activités macrobenthiques pourront être incluse dans la paramétrisation de modèle diagenétique, modulant spatialement les coefficients de mélange biogénique et de bio-diffusion selon les cartes des fonctions du benthos. Un modèle benthique contraint par la biodiversité des fonds marins constituerait une étape importante pour le développement de modèles océaniques qui prennent en compte l'impact des changements environnementaux sur la vie benthique et sa capacité à assurer des fonctions écosystémiques clés telles que la dénitrification et la séquestration du carbone.

Cette thèse présente une approche basée sur les caractéristiques pour cartographier la biodiversité fonctionnelle benthique sur le plateau continental Nord-Ouest de la mer Noire, en établissant un lien entre les observations locales au niveau des espèces et les applications à grande échelle dans le cadre de la gestion écosystémique. Le couplage de modèles de distribution des traits et de modèles biogéochimiques mécanistes permet de relier l'environnement, la biodiversité et les fonctions à l'échelle de l'écosystème. Au-delà de son application à la mer Noire, cette approche est transposable à d'autres systèmes côtiers présentant des compositions d'espèces différentes. En fournissant des cartes des fonctions et des vulnérabilités benthiques, ce travail de doctorat soutient directement la préservation du bon état écologique de la mer Noire et fournit une base scientifique solide pour la conception future d'aires marines protégées et la conservation de la biodiversité marine.

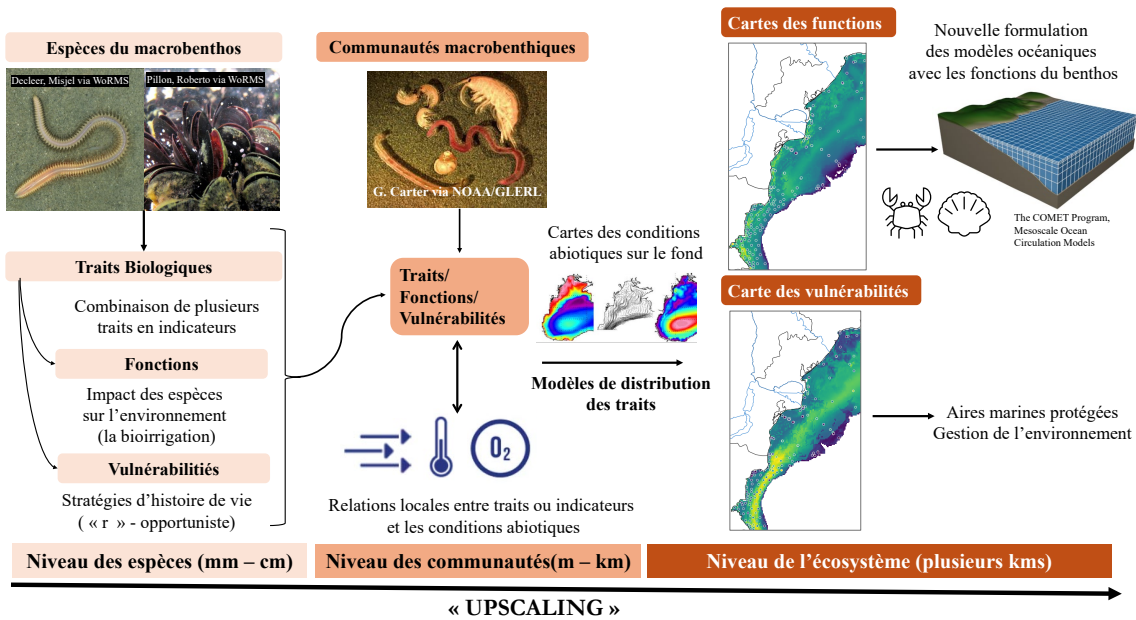


Figure 0.3: Résumé graphique de la thèse de doctorat : des espèces benthiques aux cartes des vulnérabilités et des fonctions benthiques grâce à une approche basée sur les traits.

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GLOSSARY

3D ocean model a numerical model (run on a computer) to carry out calculations to predict ocean behaviour across three dimensions—latitude, longitude, and depth—over time (3D). These models solve physical equations that describe the dynamics of the ocean (processes) in response to driving forces and can predict its evolution over time. The models are improved by assimilating actual satellite or *in situ* measurements. The definition has been taken from the glossary of CMEMS.

Benthic-pelagic coupling exchanges of particulate and dissolved substances (e.g. nutrients, gases) between sediment and the overlying water column.

Biodeposition benthic-pelagic transfer through suspension feeding (Graf & Rosenberg, 1997).

Biodiversity the variety of life and can be qualified by a taxonomic, genetic or functional approach (Pavoine et al., 2009).

Bioindicators include biological processes, species, or communities and are used to assess the quality of the environment and how it changes over time (Holt & Miller, 2011).

Bioirrigation one of the two processes of bioturbation through ventilation (Kristensen et al., 2012).

Biomixing one of the two processes of bioturbation through particles reworking (Kristensen et al., 2012).

Bioturbation all transport processes carried out by animals that directly or indirectly affect sediment matrices. These processes include both particle reworking (biomixing) and burrow ventilation (bioirrigation) (Kristensen et al., 2012).

Denitrification the microbial production of N_2 from fixed N (Knowles, 1982) : $2NO_3^- + 10e^- + 10H^+ \rightarrow N_2 + 6H_2O$.

Early Diagenesis refer to the combination of biological, chemical, and physical processes which change the quantity and composition of the organic matter in the upper several hundred meters of marine sediments. During early diagenesis, between 30% and 99% of the organic matter deposited on the sediment surface is remineralised by sediment organisms (Henrichs, 1992).

Ecological Processes changes in energy and matter over time and space through biological activity, within the ecosystem, due to abiotic (physical and chemical) and biotic (organism) factors and interactions (Reiss et al., 2009).

Ecosystem Engineer organisms that modify habitat by causing changes in biotic and abiotic conditions, influencing biodiversity and ecosystem functions (C. G. Jones et al., 1997).

Ecosystem Functions the ecological processes that control the fluxes of energy, nutrients and organic matter through an ecosystem (Cardinale et al., 2012).

Ecosystem Services the benefits human populations derive, directly or indirectly, from ecosystem functions (Costanza et al., 1997). According to the Millennium Ecosystem Assessment, these services can be classified into four main categories: provisioning, regulating, supporting and cultural services.

Eutrophication the increased rate of primary production and accumulation of organic matter which may cause bottom hypoxia (Rabalais et al., 2002).

Euxinic no oxygen and high concentrations of hydrogen sulphide.

Fourth-corner matrix a matrix describing trait–environment associations (Legendre et al., 1997).

Hypoxia lack of dissolved oxygen with respect to the requirements of aerobic respiring species (Breitburg et al., 2018).

Life-history tactic a tactic (synonym to strategy) is defined as a set of coadapted traits designed, by natural selection, to solve particular ecological problems (Stearns, 1976).

Marine Protected Area geographic areas designated, regulated and/or managed to achieve specific conservation objectives (Day et al., 2012).

RLQ analysis multivariate ordination technique which permits the simultaneous ordination of the three tables: abiotic conditions (matrix R), species (matrix L) and traits (matrix Q) (Dolédec et al., 1996).

Trait morphological, physiological or phenological features defined at the species level that can be related to their performance (Violle et al., 2007; McGill et al., 2006; Díaz & Cabido, 2001). Traits can be subdivided into several categories called modalities.

Trait Distribution Model predictive models that link traits to environmental factors to predict the spatial distribution of traits across an ecosystem (Kjørboe et al., 2018).

ACRONYMS

AFDW Ash-Free Dry Weight.

AMBI AZTI Marine Biotic Index.

ANNs Artificial Neural Networks.

BAMHBI Biogeochemical Model for Hypoxic and Benthic Influenced areas.

BCA Between-Class Analysis.

BEF Biodiversity - Ecosystem Functions.

BGA Between-Group Analysis.

BGC-ARGO Biogeochemical ARGO.

BGCoIA Between-Group Co-Inertia Analysis.

BIC Bayesian Information Criterion.

BS-MFC Black Sea-Monitoring Forecasting Centre.

BTA Biological Trait Approach.

CIL Cold Intermediate Layer.

CMEMS Copernicus Marine Environment Monitoring Service.

CoA Correspondence Analysis.

CoIA Co-Inertia Analysis.

CR Correlation Ratio.

CWM Community Weighted-Mean.

DIC Dissolved Inorganic Carbon.

DIN Dissolved Inorganic Nitrogen.

EBM Ecosystem-Based Management.

EG Ecological Group.

FCA Fuzzy Correspondence Analysis.

FDiv Functional Divergence.

FEVe Functional Evenness.

FRic Functional Richness.

GES Good Environmental Status.

HD Habitat Directive.

IPBES Intergovernmental Science Policy Platform on Biodiversity and Ecosystem Services.

MAR Modèle Atmosphérique Régional.

MAST Modelling for Aquatic Systems.

MCA Multiple Correspondence Analysis.

MPAs Marine Protected Areas.

MSFD Marine Strategy Framework Directive.

NEMO Nucleus for European Modelling of the Ocean.

NSCA Non-Symmetric Correspondence Analysis.

NWS North-Western Shelf.

ODP Oxygen Depth Penetration.

ODU Oxygen Demand Units.

OM Organic Matter.

OPCAIV Orthogonal Principal Component Analysis on Instrumental Variables.

p.s.u practical salinity unit.

PAR Photosynthetic Active Radiation.

PCA Principal Component Analysis.

POC Particulate Organic Carbon.

POM Particulate Organic Matter.

POSE Precocial-Opportunist-Survivor-Episodic.

SDMs Species Distribution Models.

SPI Sediment Profile Imaging.

SSP Shared Socioeconomic Pathway.

SST Sea Surface Temperature.

TDMs Trait Distribution Models.

VPA Variance Partitioning Analysis.

WoRMS World Register of Marine Species.

WW Wet Weight.

ZPF Zernov's Phyllophora Field.

GENERAL INTRODUCTION

1.1 General background

The effects of earthworms on the structure of the substrate in which they live were first presented in the final book of Charles Darwin: “*The Formation of Vegetable Mould through the Action of Worms, with Observations on their Habits*”. Since this pioneering work at the end of the 19th century, the large-scale consequences of the activities of small animals have been extensively studied across a wide range of ecosystems (Kristensen et al., 2012). In aquatic systems, **macrozoobenthos**, defined as benthic invertebrates larger than 0.5 – 1 mm, play a key role in biogeochemical cycles by mediating benthic-pelagic fluxes between the water column and sediment (R. C. Aller, 1994). Through their activities (e.g. burrow ventilation and maintenance), they act as **ecosystem engineers** by modifying the gradients of reactions mediated by bacteria and redistributing food for other organisms (C. G. Jones et al., 1997). In return, macrozoobenthic communities respond to organic matter supply from the water column, hydrodynamics regime, and temperature (R. C. Aller, 1994; Snelgrove & Butman, 1994). Due to their relatively fixed position and long lifespan, macrobenthos are widely used as **bioindicators** of the health of marine ecosystems (Borja et al., 2000). Therefore, macrobenthic species are important for the maintenance of biodiversity and ecosystem functioning, making them key biodiversity conservation targets (Olsen et al., 2013). The protection of marine benthic biodiversity is currently even more critical as our planet is currently facing the 6th extinction of biodiversity mostly driven by human pressures (Barnosky et al., 2011). This rapid decline in biodiversity raises serious concerns about the ability of marine systems to sustain crucial ecosystem functions (Solan et al., 2004a; Emmerson et al., 2001) and calls for stronger policy decisions aimed at maintaining the health of benthic habitats (Solan et al., 2020, 2023).

Decisions aimed at conserving marine living communities must be based on an *Ecosystem-Based Management (EBM)* strategy (Long et al., 2015). This means that planning and management should be based on ecosystem boundaries rather than political or jurisdictional frontiers (Olsen et al., 2013). Within this context, maps of marine biodiversity provide the necessary spatial context to achieve a sound design of *Marine Protected Areas (MPAs)* based on ecological and ecosystem considerations. Despite the existence of international databases compiling millions of records of benthic species (e.g. OBIS), there is still a critical lack of spatial information on the distribution of benthic species, their ecological functions and their vulnerabilities to environmental pressures. The distribution of marine living organisms is, indeed, strongly discontinuous in geographic space, particularly due to the high heterogeneity of the environment. This complicates the task of modelling the spatial distribution of benthic species using simplified interpolation techniques. As a result, the implementation of *MPAs* lags behind their terrestrial counterparts (Olsen et al., 2013) due to the lack of maps for effective conservation planning. Additionally, the absence of maps depicting the functions of the benthos at a larger scale, partly explains why benthic life characteristics are often not explicitly incorporated in many current ocean models or are simply parameterised (Kostylev, 2012). This limitation seriously hinders our the capacity of model to accurately evaluate the importance of benthic biota in providing key ecosystem functions, such as carbon sequestration and nutrient recycling. This PhD thesis aims to fill this gap by developing modelling tools able to map the functions and vulnerabilities of macrobenthic communities at the shelf scale. Specifically, it addresses the following question:

“How can we upscale local observations of benthic biodiversity to map benthic functions and vulnerabilities at the shelf scale?”.

Two aspects of macrobenthic biodiversity are considered. First, we focus on benthic characteristics that respond to environmental pressures, with the objective of producing maps of benthic vulnerabilities that can inform *EBM* strategy. Second, we examine benthic characteristics that influence ecosystem functioning, to estimate the contribution of macrobenthos to key ecosystem processes (e.g. biomixing, bioirrigation) and to add the variability of macrobenthos into benthic ocean model. To achieve this, this PhD thesis adopts a functional approach to biodiversity to represent the functions and vulnerabilities of macrobenthos. This means that living communities are described by the main characteristics of their species (their traits) rather than their taxonomic composition (Díaz & Cabido, 2001). The choice of a functional approach over a taxonomic one is motivated by the Southwood’s habitat templet theory Southwood (1977), which suggests that the traits of a given community may be more related to environmental characteristics than species composition itself. In this PhD study, statistical analyses of traits and environmental data (here RLQ combined with the fourth-corner approach) will help identify the environmental variables that explain the spatial variability of traits. From this information, *Trait Distribution Models (TDMs)* (as defined in Violle et al. (2014) as model predicting the geographic distribution of the diversity of traits), will be developed to establish the link between traits and the environmental conditions.

Then, *TDMs* will predict the spatial distribution of traits using maps of environmental conditions provided by an ocean numerical model. A trait-based approach is thus an adequate way of linking the wealth of information provided by ocean models (which are considered as poor predictors of biodiversity) to functional biodiversity considerations. A trait-based framework will also facilitate comparisons across regions and may allow the methodology developed here to be applied to other coastal ecosystems with different taxonomic compositions (Poff, 1997). The methodology is developed and applied to the *North-Western Shelf (NWS)* of the Black Sea which serves as a perfect case study for understanding the relationship between the traits of macrobenthos and the high diversity of bottom abiotic conditions, specifically the oxygen gradient that ranges from well-oxygenated coastal waters to oxygen-depleted deep waters (Wijsman et al., 1999; Friedrich et al., 2002).

1.2 A trait-based approach

1.2.1 General description of trait-based ecology

The concept behind trait-based ecology is that **traits** are morphological, physiological, or phenological features defined at the species level that can be related to their performance (Díaz & Cabido, 2001; Violle et al., 2007). Trait-based ecology aims to understand and to predict how traits of species determine community structure and the provision of ecosystem functions and services (Díaz & Cabido, 2001; Lavorel & Garnier, 2002). The distribution of traits in geographic space and time can be predicted using **Trait Distribution Model (TDM)** equivalent of the species distribution model (i.e. models that aim to predict species diversity and composition (Ferrier & Guisan, 2006; J. Miller, 2010)). *TDMs* are predictive models that link traits to environmental factors in order to predict the spatial distribution of traits across an ecosystem (Violle et al., 2014). In this way, *TDMs* can be used to extrapolate the expected value of traits in locations where observations are missing, based on the environmental conditions present there, therefore bridging the gap between punctual observations and spatial maps (Kjørboe et al., 2018). *TDMs* are generally less complicated to parametrise than *Species Distribution Models (SDMs)* and often produce more generalisable results that can be applied to other systems (Zakharova et al., 2019), because they focus on the ecological roles and functions of organisms rather than their taxonomic classification. It offers an alternative to classical species-centric approaches and has the potential to describe complex ecosystems in simple ways and to assess the effects of environmental change on ecosystem structure and function (Kjørboe et al., 2018).

1.2.2 Framework of response and effect trait

In the framework of **response-effect trait**, traits can be classified into two main categories: **response** (that respond to environmental conditions) traits and **effect** (with an impact on the environmental conditions) (**Figure 1.1**). The distinction between effects and responses can be very subtle, because some traits fall under both categories.

For example, the burrowing characteristics of species can be representative of their fitness and are also indicators of their bioturbative effects (Beauchard, 2023).

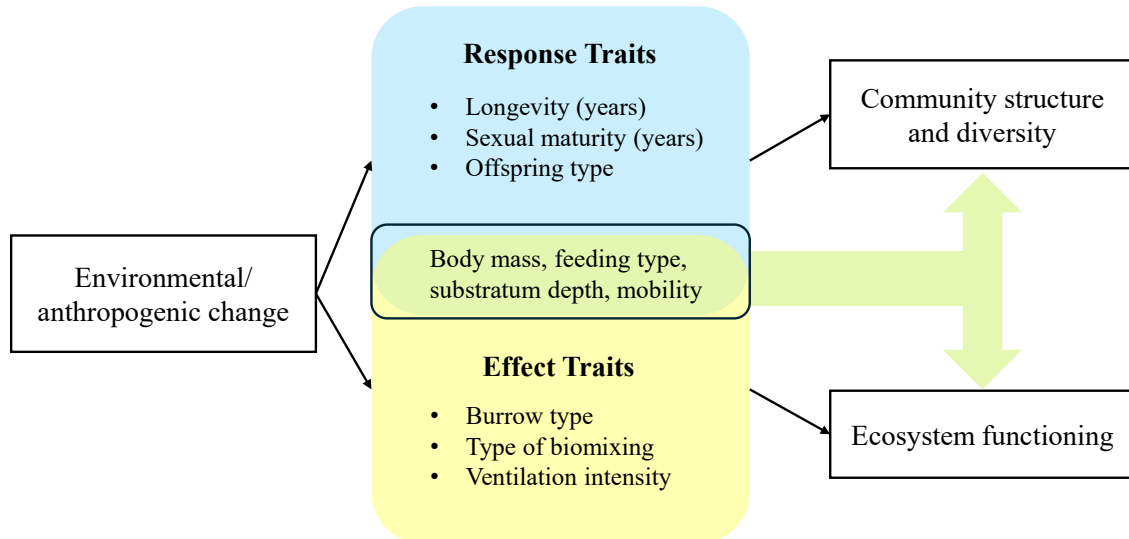


Figure 1.1: Graphical representation of the **response/effect trait framework**. The compilation of the response (blue) and effect traits (yellow) of the species within a community enables the study of how environmental or anthropogenic changes impact the structure and diversity of the community, as well as the resulting consequences for ecosystem functioning. A trait can be a response or effect trait at the same time (in green). The figure has been adapted from Lavorel and Garnier (2002).

On the one hand, **response traits** are the results of evolutionary processes through environmental filtering (i.e. abiotic constraints) and biotic interactions that push organisms toward specific life strategies as selected combinations of traits (Southwood, 1977; Stearns, 1976; C. R. Townsend & Hildrew, 1994). These traits cover the major functions that express fitness components (e.g. growth, survival, reproduction) in response to abiotic and biotic environmental stressors. Key responses traits are related to reproduction such as reproductive frequency (seasonal or continuous) and annual fecundity are fundamental traits (Beauchard et al., 2017) because they inform on the potential of recovery of the species (Hinz et al., 2021). Other key response trait is mobility as it is a crucial trait for the avoidance of physical disturbance and translate foraging potentials (Beauchard et al., 2022) as well as lifespan related to the resilience of an organism in the presence of disturbance (de Juan et al., 2022). Response traits can be used successfully to predict community disassembly in response to environmental changes (Lavorel et al., 1997) and human disturbances (e.g. Villnäs et al. (2011); Mouillot et al. (2013); Villnäs et al. (2013)). Therefore, response traits can be viewed as those that explain why a community is found in a specific habitat and not in another one (Beauchard et al., 2017).

On the other hand, the roles of benthic species are important in regulating and supporting ecosystem processes (Snelgrove et al., 2018), and these functions are determined by the **effect traits** of the species (Solan et al., 2004a; Bremner, 2008). Effect traits (e.g. mobility, burrowing depth) have an impact on the ecosystem and are proxies for the effect of macrofauna on ecosystem functioning (Braeckman et al., 2010; Bremner, 2008; Queirós et al., 2013; Solan et al., 2004a; Mermillod-Blondin et al., 2004; Thrush et al., 2006). Evidence suggests that the functional traits of individual species within a community are more important than species richness in shaping biodiversity-ecosystem functioning relationships (Cardinale et al., 2012). Consequently, trait-based approaches are commonly used to assess the effects of macrozoobenthos on ecosystem functioning (Bremner et al., 2003). Effect traits are important in estimating ecological processes because direct *in-situ* bioturbation measurements are costly, scarce, and insufficient for large-scale assessments (Gogina et al., 2022; Solan et al., 2004b). Additionally, there is a critical lack of experimental studies to measure the vertical transport of particles and to estimate the bioturbation capacity at the species or community level (Solan et al., 2004b). Experimental measurements of bioturbation typically involve the use of tracers (e.g. radionuclides, luminophores, phaeopigments) and *Sediment Profile Imaging (SPI)*, directly measuring the depth of sediment mixing from vertical sediment profiles (Solan et al., 2004b; Rhoads & Cande, 1971; Teal et al., 2010). The combination of multiple effect traits is needed to estimate the contribution of a macrobenthic species to a given ecological process (e.g. biomixing, biodeposition), precursors of key ecosystem functions such as nutrient recycling and carbon sequestration (Bremner, 2008; Díaz & Cabido, 2001) and ecosystem services (**Figure 1.2**). Single or multiple traits can influence several ecosystem functions, or one ecosystem function could be affected by several traits (Lam-Gordillo et al., 2020). For example, the trait body mass influences several ecological processes, which can then affect single or multiple functions in the ecosystem (e.g. productivity, nutrient recycling and habitat creation) and, depending on the ecosystem functions involved, one or more ecosystem services (**Figure 1.2**).

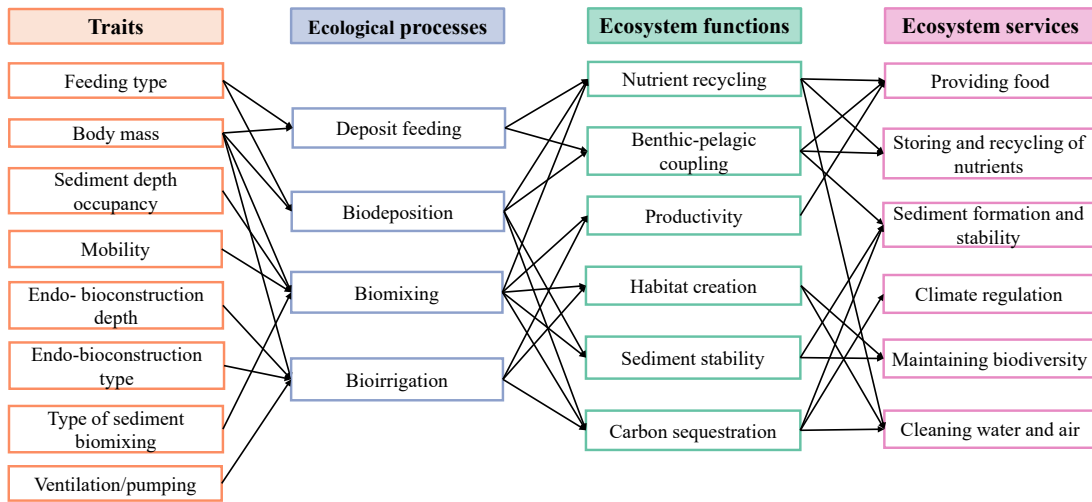


Figure 1.2: The role of macrobenthic species in multiple ecosystem functions and services can be estimated from the effect traits. The interconnections between ecosystem processes (the precursors to functions), functions, and services complicate our understanding of seafloor ecosystems. The figure has been adapted from Snelgrove et al. (2014).

1.3 Macrozoobenthic community structure and diversity

The composition of macrozoobenthic communities in space and time is influenced by dispersal limitation, environmental filtering and biotic interactions (**Figure 1.3**), which are the three main filters that shape and maintain species assemblages (Weiher et al., 1998). Dispersion is the first filter that limits the geographic range of species within their fundamental niche (Green et al., 2022). Environmental filtering refers to abiotic constraints that restrict species within an ecological niche, while limiting similarity considers that biotic forces (e.g. competition, predation) prevent species from being too similar in the same niche. In addition, drift (i.e. stochastic changes in species abundance) and speciation (i.e. creation of new species) are two major processes explaining patterns in the composition and diversity of species (Vellend, 2010). This PhD work focusses on environmental filtering and how abiotic conditions shape benthic community assemblages.

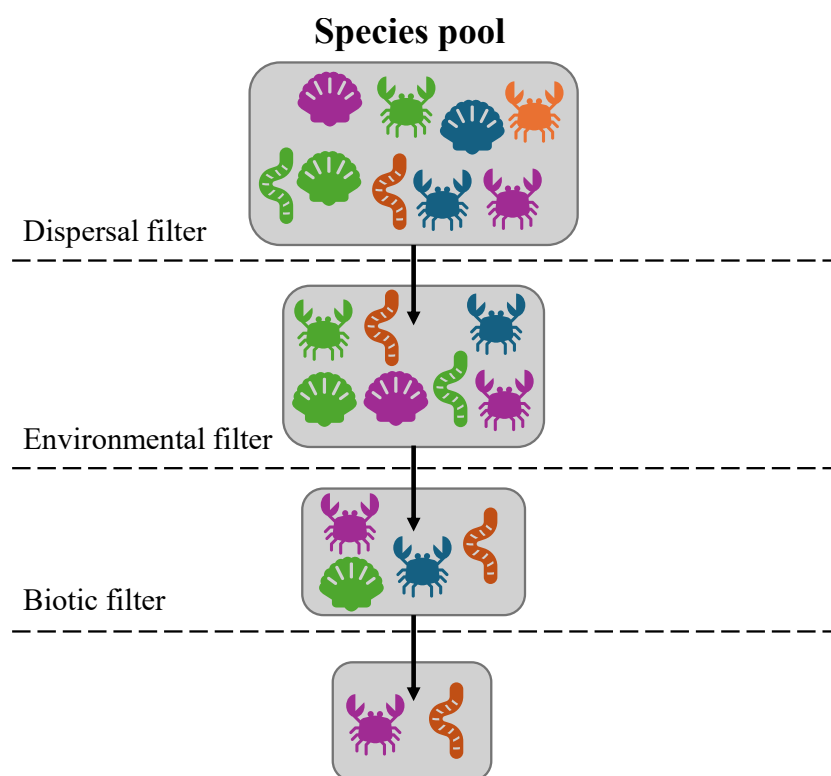


Figure 1.3: Conceptual view of the series of hierarchical filters that reduce a pool of potential local species to the ones that are observed. The figure has been redrawn from Cadotte and Tucker (2017).

Major abiotic factors determining the diversity and composition of benthic communities are the flux of *Organic Matter (OM)* to the bottom (Rosenberg, 1995; Pearson & Rosenberg, 1978), food quality (Dauwe et al., 1998; Wieking & Kröncke, 2005), oxygen levels, hydrodynamics (Snelgrove & Butman, 1994), temperature and grain size of sediment (Gray, 2002). Macrozoobenthic species have varying tolerances to environmental stress and integrate water-sediment quality conditions through their relatively long lifespan and low mobility (Dauer, 1993). In this way, the state of macrozoobenthic communities is a widely used **bioindicator** of long-term ecological changes in marine ecosystems (Gray & Pearson, 1982; Borja et al., 2000).

The **Pearson-Rosenberg model** provides a fundamental framework for understanding how benthic communities and sedimentary structure qualitatively respond to shifting environmental conditions (e.g. organic pollution, low oxygen or physical disturbance) (Pearson & Rosenberg, 1978). The successional sequence illustrated in **Figure 1.4** follows the same pathway with distance from, or time since, perturbation and is reversible, transitioning from a highly diverse and active macrobenthic community left to an anoxic (no oxygen) with no macrobenthic life at the right as perturbation intensifies (Solan et al., 2020).

Under well-oxygenated, unimpacted conditions, diverse macrobenthic communities are present, including large, long-lived, and deep-burrowing taxa (**Figure 1.4**). In more stable and unpolluted habitats, species with K-selected life-history strategies prevail (e.g. large long-lived bivalves). With increasing organic loading, communities progressively shift from K-selected, sensitive species to opportunistic, r-selected taxa that dominate in organic-enriched sediments (Pearson & Rosenberg, 1978). r-strategists have rapid development, high reproductive rates, and short lifespans, allowing them to quickly exploit fluctuating or disturbed environments where organic matter is abundant (Grassle & Grassle, 1974). In contrast, K-strategists are slow-growing, produce fewer offspring, and are better adapted to stable conditions with limited resources. This transition reflects increasing environmental stress, as elevated organic matter inputs enhance microbial respiration, leading to oxygen depletion (hypoxia) and the production of toxic by-products such as sulphides and ammonia (Diaz & Rosenberg, 1995; Pearson & Rosenberg, 1978; Hyland et al., 2005).

With increasing disturbance, the community shifts into a state of pronounced disequilibrium, where sensitive species disappear and opportunistic species dominate, resulting in very low biodiversity. The peak of opportunistic species corresponds to the maximum abundance and the second maximum of biomass (**Figure 1.4**). Concurrently, there is a decrease in the depth at which macrofauna occurs, related to the reduction of the oxygenated layer due to an excess of organic loading (**Figure 1.4**). In general, large and deep burrowers are observed in normal, well-oxygenated conditions compared to small and surface deposit feeders under polluted conditions (Nilsson & Rosenberg, 1997; Dauer et al., 1992; Levin et al., 2009).

Under severe and prolonged enrichment, communities become highly simplified and may be dominated by one or a few opportunistic species. In extreme cases, anoxic conditions and the accumulation of reduced toxic compounds such as hydrogen sulphide (H_2S), result in the loss of macrozoobenthic fauna (Nilsson & Rosenberg, 1994). Changes in the structure and functioning of the benthic community with shifting environmental conditions have major consequences for early diagenetic pathways due to a reduction in macrozoobenthos activity and loss of diversity (Levin et al., 2009; Middelburg & Levin, 2009; Pascal et al., 2024; Villnäs et al., 2012; Van Colen et al., 2012).

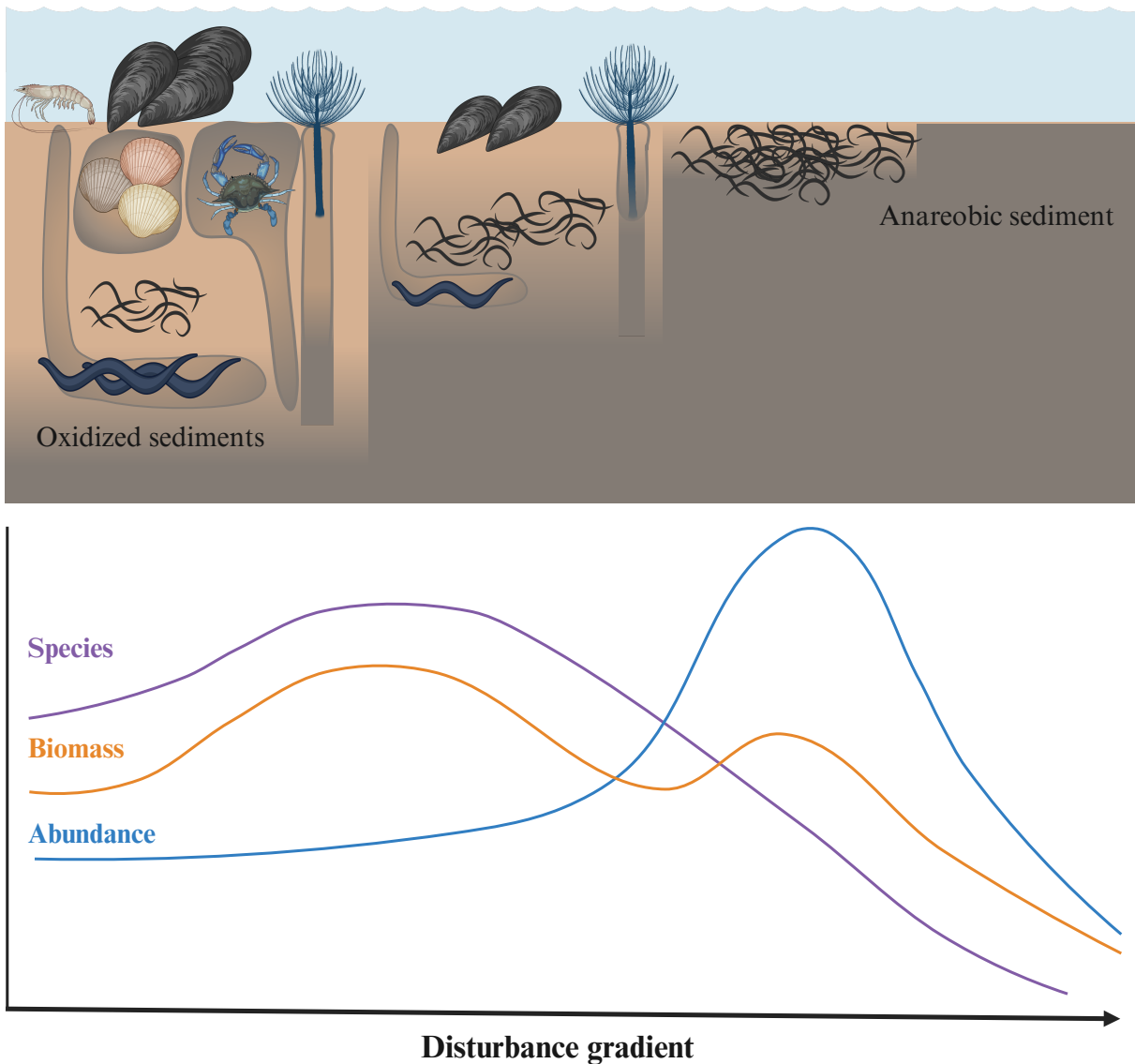


Figure 1.4: In the top line: the sequential changes in benthic community structure and sedimentary conditions along an increasing gradient of organic pollution from left to right (Pearson & Rosenberg, 1978). Sediment colour reflects redox status (from light brown oxidised, darker shades reduced and dark brown anoxic). In the bottom line, changes in the number of species, abundance and biomass of macrozoobenthic communities. The figure has been inspired by Solan et al. (2020) and redrawn from Nilsson and Rosenberg (1997, 2000). The figure has been created using BioRender.

1.4 Macrozoobenthos and ecosystem functioning

Marine sediments cover the largest habitat area on Earth (Gray & Elliott, 2009) and host diverse biodiversity that regulates biogeochemical cycles, climate-active gases, and carbon sequestration (Solan et al., 2020). This habitat plays an essential role in global biogeochemical cycles by linking rapid oceanic, atmospheric, and terrestrial cycles to slow geological rock cycles (Falkowski et al., 1998; Van Cappellen, 2003). Through numerous **benthic-pelagic** fluxes, sediments regulate marine productivity and contribute to global ocean cycles and biogeochemical budgets (Canfield, 1993). The coupling between the benthic and pelagic parts of the ocean is particularly tight in shallow coastal ecosystems, where a significant part of the primary production ($\sim 15 - 50\%$) escapes degradation in the water column and reaches the sediment in a relatively fresh state (Heip et al., 1995; Canfield, 1993; Wollast, 1998). In general, what is returned to the water column is fundamentally different from what was deposited due to the biological, chemical, and physical processes that occurred in the sediment, called “**early diagenesis**” (Bernier, 1980; Boudreau, 1997). *OM* delivered to the sediments is mineralised by aerobic and anaerobic microbial sediment processes, and then inorganic reduced compounds (e.g. CO_2 , nutrients) are released into the overlying water column or buried in the deeper layers of sediments (Smith, 1974). In the sediment, organic matter is mineralised following a well-defined **diagenetic sequence (Figure 1.5A)**, with electron acceptors used in order of energy yield: oxygen, nitrate, manganese, iron, sulphate, and methanogenesis (Froelich et al., 1979; Claypool & Kaplan, 1974) (**Figure 1.5A**). Within the first millimetres of sediment, oxygen is consumed through oxidic mineralisation of *OM* or re-oxidation of the reduced products (such as Fe^{2+} , Mn^{2+} , or H_2S) from anaerobic respiration occurring deeper in sediment (Arndt et al., 2013; Glud, 2008) (**Figure 1.5A**). As a consequence, the penetration of oxygen in the sediment is very shallow (only a few millimetres) in undisturbed coastal sediments (Revsbech, Jørgensen, & Blackburn, 1980).

Early diagenesis processes in marine sediments are strongly influenced by the activities of **macrozoobenthos** (e.g. R. C. Aller (1994); Snelgrove and Butman (1994); Kristensen (1991)). Indeed, the macrobenthos plays a crucial role in regulating organic matter turnover, nutrient cycling, and sediment-water exchanges, which are vital for ecosystem functioning in marine environments (e.g. R. C. Aller (1994); Boudreau (1986); Kristensen et al. (2012)) (**Figure 1.5B**). Through their activities (e.g. burrowing, feeding, locomotion), macrobenthos act as ecosystem engineers by altering microbially driven biogeochemical reactions through changes in carbon and nutrient availability for microbes or by changing abiotic conditions that influence reaction rates (Gutiérrez & Jones, 2006). Sediment porosity, permeability, particle size, cohesion, and organic content, as well as their spatial heterogeneity, can be altered through gut passage and particle selectivity, creating niches for smaller organisms such as bacteria, micro- and meio benthos (R. C. Aller, 1984; van Nugteren et al., 2009; Waldbusser et al., 2004). The influence of biological mixing on benthic–pelagic fluxes depends on the characteristics of the species, their abundance and biomass, and the structure of the sediment (Glud, 2008; Mermillod-Blondin et al., 2005).

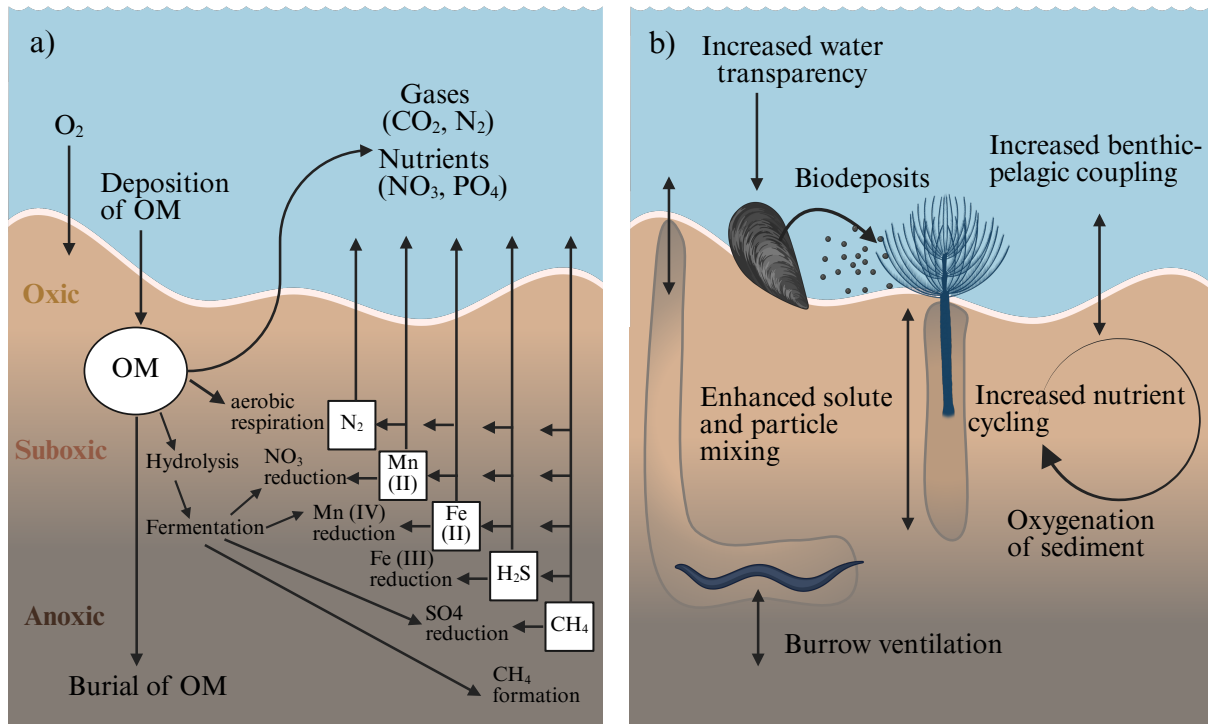


Figure 1.5: Classical view of benthic-pelagic coupling on the left, and the impact of macrobenthos on benthic-pelagic coupling on the right. This figure has been inspired by Bianchi et al. (2021) and Snelgrove et al. (2018). The figure has been created using BioRender.

All animal transport processes that directly or indirectly affect sediment properties are defined as **bioturbation** (Kristensen et al., 2012). Bioturbation affects the upward and downward transport of particles (e.g. OM, particulate metals) through **biomixing** and the exchanges of solutes (e.g. O₂, NO₃⁻, NH₄⁺) with the overlying water through advective and diffusive **bioirrigation** (Figure 1.6). Other key ecological processes include biodeposition and deposit feeding by macrobenthic species. In the following section, an overview of the main ecological processes used in this PhD work is given.

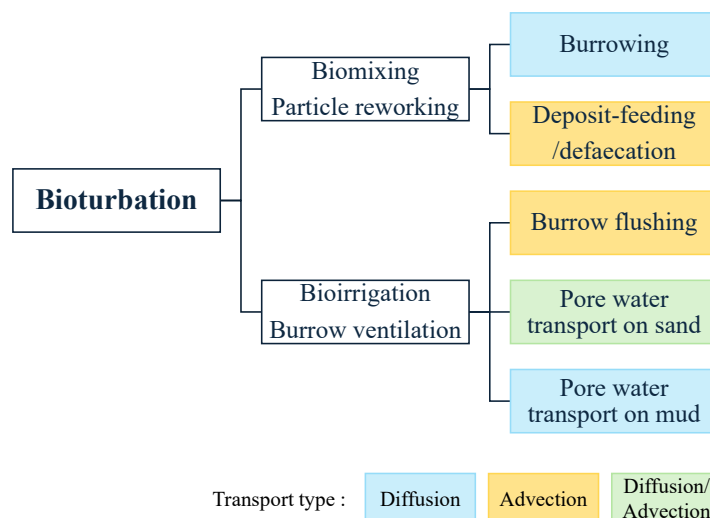


Figure 1.6: Overview of bioturbation modes, including biomixing and bioirrigation, and their respective transport processes. The different types of transport are indicated by colours: diffusion is represented in blue, advection in yellow, and both diffusion/advection in green. The figure has been adapted from Kristensen et al. (2012).

1.4.1 Biomixing

Biomixing (i.e. reworking of particles) is carried out by benthic organisms through activities such as burrows construction and maintenance, as well as ingestion and defecation (**Figure 1.6**), which causes the sediment matrix to mix at rates much faster than diffusion alone (Kristensen et al., 2012). Indeed, burrows increase sediment surface area available for diffusive exchanges up to 500% and stimulate exchanges between the sediment and the water column (Forster & Graf, 1992; Hylleberg & Henriksen, 1980). Below the oxygenated layers, oxygen is only present in the proximity of animal burrows (Revsbech, Sorensen, et al., 1980), which become an extension of the oxic-anoxic interface (Glud et al., 2016). For example, *Corophium volutator* can increase the volume of oxic sediment by 150%, enhancing the aerobic decomposition of organic matter (Kristensen, 1991). The mucus lining of the burrows is enriched in labile organic carbon, intensifying mineralisation and re-oxidation processes (Kristensen, 2000; Dale et al., 2019).

Based on their behaviours, lifestyles and feeding strategies, particles reworking animals can be classified into functional groups (François et al., 1997; Kristensen et al., 2012): biodiffusors, upward conveyors, downward conveyors, regenerators and gallery-diffusors (**Figure 1.7**). The different modes of reworking lead to distinct effects on the vertical distribution of both *OM* and inorganic particles within sediments, resulting in different impacts on sediment biogeochemistry (Mermillod-Blondin et al., 2005; Kristensen, 2001). On the one hand, upward conveyors can increase the net carbon mineralisation rate by transporting deeper anoxic sediment back to the oxic zone near the sediment-water interface, intensifying local oxygen consumption (R. C. Aller & Aller, 1998; Kristensen, 2001).

Under these conditions, anaerobic processes such as sulphate reduction and denitrification (Jørgensen, 1977a) may occur at the sediment surface. On the other hand, downward conveyors mix labile organic matter into deeper sediment layers, decreasing aerobic respiration (Bernier & Westrich, 1985) and increasing by several fold the net carbon burial rate (R. C. Aller, 1982).

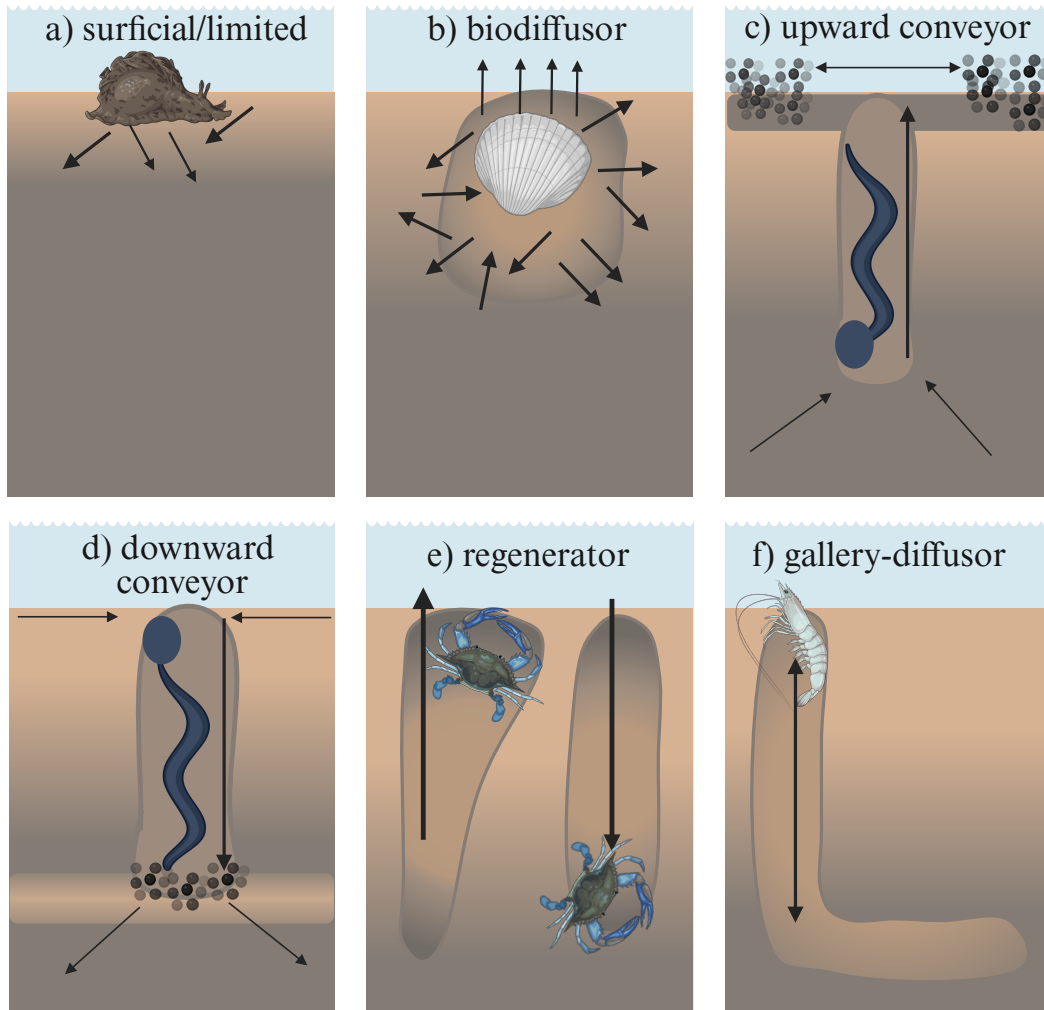


Figure 1.7: Schematic representation of the major biomixing mechanisms of macrobenthos: a) an epifaunal mobile species with limited bioturbation capacity, b) a biodiffusor redistributing sediment in all directions, c) an upward conveyor depositing faecal material at the surface d) downward conveyor depositing faecal material under the surface and e) a regenerator digging out sediments at depth and moving it to the surface or conversely and f) a gallery-diffusor. Sediment colour reflects redox status (from light brown oxidised, darker shades reduced and dark brown anoxic). The figure has been adapted from François et al. (1997) and Kristensen et al. (2012). The figure has been created using BioRender.

1.4.2 Bioirrigation

Bioirrigation refers to the process by which macrofauna ventilate their burrows intermittently (i.e. not continuously), actively flushing water and solutes in and out of the surrounding sediment (Kristensen et al., 2012). By irrigating their burrows with oxygenated water, they flush accumulated reduced metabolites (e.g. ammonium, sulphide) from the sediment into the water column and replenish oxygen and other electron acceptors (e.g. sulphate and nitrates) into the sediment (Kristensen, 1984; R. C. Aller, 1984). The efficiency of bioirrigation depends on body mass, ventilation behaviour, and feeding strategies of macrofauna (Christensen et al., 2000; Wrede et al., 2018). For example, suspension feeders irrigate their burrows more intensely for feeding purposes and therefore have a greater impact on nutrient fluxes compared to deposit feeders, which irrigate intermittently (D. C. Miller et al., 1992; Kristensen & Kostka, 2005). Bioirrigation enhances the penetration of oxygen and nutrients into deeper sediment layers through burrows, thus promoting aerobic microbial processes (R. C. Aller & Yingst, 1978; Archer & Devol, 1992), nutrient turnover, and increasing microbial activity (Kristensen, 2000; Snelgrove et al., 2018; Volkenborn et al., 2012). Porewater irrigation may increase O_2 fluxes several times (3 to 4 times) compared to those calculated from the gradient-driven diffusive flux (Archer & Devol, 1992). Bioirrigation increases the oxygenation of sediments and leads to nitrification in a narrow zone along the burrows (Renz & Forster, 2014), enhancing overall nitrogen cycling (Kristensen et al., 2014). The oxidised forms of nitrogen can then undergo denitrification, being reduced to N_2 gas or transferred into the pelagic organic nitrogen pool (Hulth et al., 2005). Oxygenation of sediments promotes the binding of phosphates to iron oxides, increasing phosphorus sequestration (Kraal et al., 2012). Bioirrigation is also indirectly influenced by the morphology and composition of burrow walls, which regulate the dynamics of pore water exchange (Kristensen et al., 2012). For example, burrow architecture plays a crucial role in determining how solutes and pore water move within sediments (Kristensen & Kostka, 2005). In permeable sandy sediments, burrow irrigation is typically driven by advection, especially in blind-ending burrows (Kristensen et al., 2012). In contrast, in fine-grained, less permeable muddy sediments, open-ended burrows are more common, and pore water exchange occurs primarily through radial diffusion (Kristensen & Kostka, 2005).

1.4.3 Biodeposition

Suspension feeders can enrich sediment by filtering suspended particles from the overlying water and subsequently excreting non-digestible materials such as faeces and pseudo-faeces (i.e. biodeposits) on the seafloor (Graf & Rosenberg, 1997). In this process, termed **biodeposition**, carbon and nitrogen-enriched deposits provide food for other organisms and stimulate microbial production (Graf & Rosenberg, 1997; Newell, 2004; Norkko et al., 2001). Biodeposition can locally dominate the delivery of carbon to the sea floor and significantly contribute to benthic-pelagic fluxes (Herman et al., 1999).

1.4.4 Deposit feeding

Deposit feeders can alter muddy bottoms by reworking sediment and hindering the settlement of suspension feeders or epifaunal organisms (De Wilde, 1991). Deposit feeders can be either sub-surface deposit (within the sediment) or surface deposit (from the sediment surface).

1.4.5 Impact of macrofaunal activities on ecosystem functioning

At the ecosystem scale, macrofaunal activities have a great impact on **ecosystem functioning**, particularly on the recycling and storage of organic carbon and nutrients in shallow coastal habitats (e.g. R. C. Aller (1982); Braeckman et al. (2010); Kristensen et al. (2012); Snelgrove et al. (2018); Middelburg (2018)). The global volume of bioturbated sediment has been estimated at nearly 30 000 km³, highlighting the important role of benthic invertebrates in global biogeochemical processes (Teal et al., 2008). The impact of macrobenthos on the benthic-pelagic coupling has already been proven by several studies linking carbon remineralisation, oxygen and nutrient efflux to macrobenthos activities (e.g. Braeckman et al. (2010); Lohrer et al. (2004); Mermillod-Blondin et al. (2005); O'Meara et al. (2020); Waldbusser et al. (2004)). Sediment oxygen consumption, for example, can increase by 85-200% in the presence of macrofauna (Glud, 2008; Webb & Eyre, 2004). Positive relationships have also been observed between species richness and ammonium release to the water column (Bulling et al., 2010; Henriksen et al., 1983). Ammonium is produced during the microbial decomposition of organic matter and through macrofaunal excretion, and its efflux to the overlying water is often amplified several times by bioturbation (Kristensen, 1984, 1991). However, the impact of macrofauna on nitrogen cycling is species-specific and their behaviour determines if sediments act as a source or sink for nutrients (Olsgard et al., 2008). Denitrification is generally stimulated by bioturbation (Henriksen et al., 1983; Kristensen et al., 1985), as enhanced nitrification along burrow walls and increased solute transport through ventilation both increase nitrate availability within sediments. Experimental studies have shown that burrowing activities of macrofauna can enhance denitrification rates by as much as 400% (Gilbert et al., 1998; O'Meara et al., 2020; Webb & Eyre, 2004). Benthic activity can facilitate the burial and storage of organic carbon in sediments, contributing to the sequestration of CO₂ and the **mitigation of global warming** (Atwood et al., 2020). Areas with a higher potential for bioturbation by infaunal benthic organisms are more likely to have a greater capacity for **blue carbon sequestration** (Solan et al., 2020). In contrast, sediments with high levels of bioirrigation activity, where oxygen penetrates deeper, tend to sequester less blue carbon due to the increased aerobic microbial degradation of organic matter (Huang et al., 2025). In addition, by filtering the water column, removing particulates and reducing turbidity, suspension feeders can **improve water clarity** in shallow coastal ecosystems (National Research Council, 2010). As an example, oysters could filter the entire water column in 3-6 days in Chesapeake Bay before human disturbances (Newell, 1988). Macrozoobenthos also **provide habitats** through burrows, tubes, and shells that create microhabitats, with conditions different from the surrounding sediment, for microorganisms, meiofauna, and juvenile fish.

They serve as an important food source for higher trophic levels and therefore **support marine food webs** (Herman et al., 1999). Macrobenthos also play a role in enhancing sediment cohesion and stability. For example, tube-building polychaetes and bivalves reefs create structures that increase sediment cohesion, **reduce sediment erodibility**, and create microhabitats by trapping particles and altering water flow (J. M. H. Murray et al., 2002). Macrobenthos play a key role in recycling nitrogen, phosphorus, and carbon because bioturbators help promote oxic conditions in the sediment, enhancing *OM* turnover and **nutrient recycling** rates. Benthic **denitrification** is a key ecosystem function, as it is one of the main nitrogen removals in the ocean (Devol, 2008) and can contribute to reducing nitrogen concentrations in the overlying water column and decreasing eutrophication. By stimulating nutrient fluxes to the water column, macrobenthos can **modulate primary productivity** in the overlying water column (Kristensen et al., 2014; Olsgard et al., 2008) and these regenerated nutrients can support pelagic primary production limited by the availability of inorganic nitrogen (Sandwell et al., 2009). In addition, benthic bioturbators may facilitate **remediation** of bad sediment conditions by decreasing the concentrations of toxins such as sulphides and ammonia in sediment porewater. Thus, benthic communities can improve their resilience to hostile conditions, ensuring good functioning of benthic ecosystems (Lam-Gordillo et al., 2022).

1.4.6 Integrating the effect of macrobenthos in benthic models

Despite the high importance of macrobenthos on ecosystem functioning, life on the seafloor is still not well represented in Earth System Models with in general an oversimplification of the benthic compartment in 3D ocean models (Snelgrove et al., 2018; Radtke et al., 2019; Siedlecki et al., 2025). Although both pelagic and benthic models have both progressed substantially in recent years, their coupling is much less advanced, and it appears that the choice of a lower boundary condition in water column models is more a matter of convenience than based on a careful representation of the essentials of sediment–water exchange (Soetaert et al., 2000). The main reason for the oversimplification of the benthic compartment is the excessive computing time and prohibitive memory requirements of simulations coupling vertically resolved full diagenetic models with 3D ocean models for multi-year integration (Lessin et al., 2018). As a result, the benthic compartment is usually poorly represented in 3D ocean models and consists in either using very crude parameterisations derived in the best case from punctual observations and applied at the shelf scale or considering that what falls into the bottom is directly remineralised (Grégoire et al., 2023). In addition, current models do not consider regional heterogeneity (e.g. differences in carbon remineralisation and burial from the coast to the shelf) and only consider biogeochemical processes within sediments in the vertical dimension (Froelich et al., 1979). Model validation is further complicated by the scarcity of observations in both space and time (Siedlecki et al., 2025). Among the oversimplifications that affect benthic models lies the ignorance of the biota characteristics in their formulation.

Although, for instance, the transport and mixing caused by bioturbators have been identified as the most influential processes of ubiquitous importance for benthic-pelagic coupling (Holstein & Wirtz, 2009). With a few exceptions (e.g. Butenschön et al. (2016); W. Zhang and Wirtz (2017)), these models omit explicit descriptions of macrozoobenthos but may include parameterisations related to some aspects of their activity (Grégoire et al., 2023). In the case of macrofauna, their contribution to sediment particle displacement and mixing (bioturbation) is, in the best case, parameterised via non-local exchange or local biodiffusive modelling formalisms (Meysman et al., 2003) and to enhanced solute transport (bioirrigation) via model formulations of varying complexity (Meysman et al., 2006). We urgently need to incorporate biological processes into global geochemical models to improve projections of the impact of global environmental change on functioning and deliver better societal decisions on ecosystem management (Snelgrove et al., 2018; Bianchi et al., 2021; Godbold & Solan, 2013).

1.5 The Black Sea: a unique ecosystem under multiple pressures

1.5.1 General description

The **Black Sea** is the largest semi-closed basin in the world with a surface area of 4.2×10^5 km² and a volume of 5.3×10^5 km³ (Özsoy & Ünlüata, 1997). This basin is in the northern part of the subtropical zone and in the temperate climatic zone. As a result, atmospheric fluxes over the Black Sea surface are influenced by the Azores and Siberia anticyclones, as well as by atmospheric patterns over adjacent lands in Europe and Asia (Staneva & Stanev, 1998). The Black Sea has limited exchanges in the north-east with the Azov Sea through the Kerch Strait and in the south-west with the Marmara and Mediterranean Seas through the Bosphorus Strait (**Figure 1.8**). The Black Sea is an **estuarine basin** with high freshwater flows, including river run-off and precipitations, compared to its total volume (Stanev, 2005). The bottom temperature of deep layers remains constant at 8–9 °C year-round, while the temperature in the first 50 metres can vary rapidly over seasons. The upper layer has a salinity of approximately 17.5–18.5 *practical salinity unit (p.s.u)*, about half that of the Mediterranean's waters, compared to the deep water salinity of 22–23 *p.s.u* (J. W. Murray et al., 1991).

The Black Sea exhibits classical **thermal stratification** in a temperate climate. Only the well-oxygenated euphotic surface layer is influenced by atmospheric forcings and exhibits a pronounced seasonal variability (Stanev, 1990). There are two main thermal processes: **winter renewal of the Cold Intermediate Layer (CIL)** and the **formation of the seasonal thermocline** in summer (Tuzhilkin, 2008). From January to mid-March, strong winds ventilate the upper 50 metres of the water column, maintaining cold (less than 9°C) and oxygen-rich waters (300–350 μM). After March, surface warming initiates the seasonal thermocline, isolating bottom layers and potentially leading to bottom hypoxia in summer as bottom layers are no longer ventilated. In autumn, stronger winds break down the seasonal thermocline and mix the water column (Tuzhilkin, 2008).

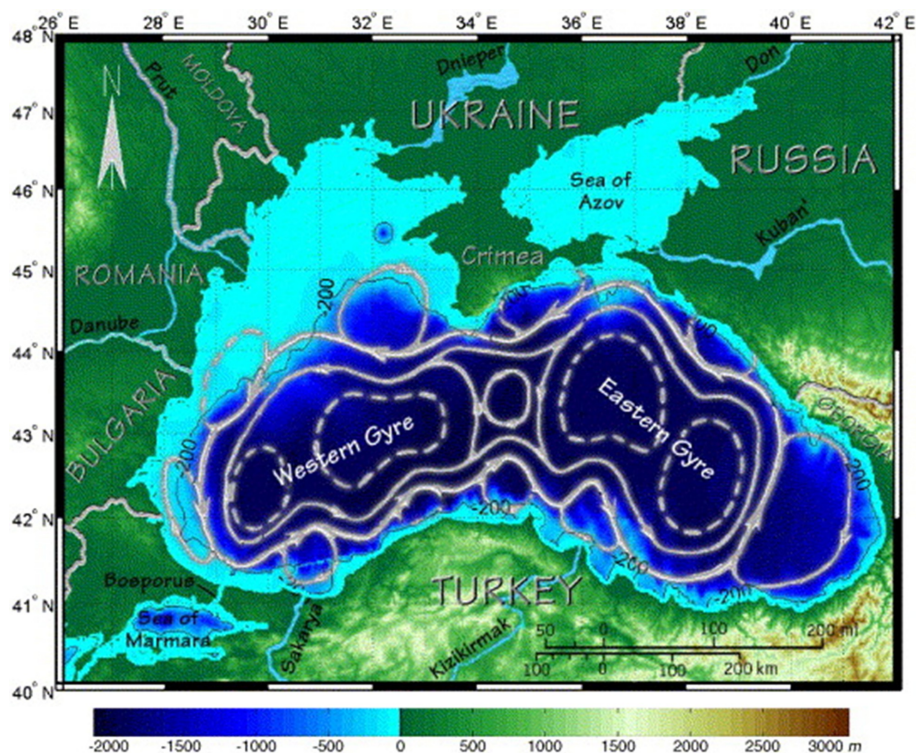


Figure 1.8: The Black Sea geography, bottom topography and circulations. The surface circulation of the main basin consists of two large cyclonic (anti-clockwise) central gyres, defining the eastern and western basins (Stanev et al., 2002), encircled by a cyclonic boundary current known as the **Rim Current**. The figure was taken from Grégoire et al. (2008).

1.5.2 Black Sea in multiple layers

The Black Sea as we know it today was formed approximately 8 000 years ago when a previously enclosed freshwater lake became connected to the saltier Mediterranean Sea. This connection occurred as a result of rising sea levels during the Holocene (Ryan, 2007). The flow of fresh water into the upper layer and the incursion at depth (± 500 m) of saline Mediterranean water (36 p.s.u) leads to a **permanent haline stratification** at a depth of 150-200 metres (Özsoy et al., 1993). In the Bosphorus Strait, there is an exchange with the Mediterranean Sea occurring in two layers: the fresh water at the surface flows out, while the denser salty water, approximately half of the surface current that flows out, flows in and sinks directly towards the deeper levels (**Figure 1.9**). The permanent pycnocline limits exchanges between the oxic surface and the depleted in oxygen deep waters, creating a unique chemical and biological environment (Konovalov et al., 2005). This permanent stratification leads to a depletion of oxygen in the bottom layer due to the degradation of sinking organic matter and the production of sulphur hydrogen in the anoxic sediments (i.e. no oxygen), which escapes into the overlying water column (J. W. Murray & Yakushev, 2006). The concentration of H_2S increases with depth and reaches $350\text{-}400 \mu\text{M}$ near the sea bed, as a result of the decomposition of organic matter by sulphate-reducing bacteria in the absence of oxygen (Jørgensen, 1977b).

Consequently, the Black Sea is the largest sulphidic basin in the world with 87% of its total volume being anoxic and containing massive amounts of hydrogen sulphide (Caspers, 1957). The oxygenated area suitable for the development of aerobic life is therefore a very thin layer. In the surface layer, oxygen levels are primarily controlled by primary production, gas exchange, and its solubility influenced by temperature and salinity (Konovalov & Murray, 2001). The thickness of the oxic zone varies from 70-100 metres in the central basin and 120-200 metres near the shelves (Yakushev et al., 2007). The well-oxygenated surface layer is separated from the euxinic (i.e. without oxygen and high concentrations of hydrogen sulphide) deep waters by the suboxic zone between the isopycnal depths of $\sigma_t = 15.40$ and $16.20 \text{ kg} \times \text{m}^{-3}$ (J. W. Murray et al., 1995, 1989). In the suboxic layer, oxygen is nearly depleted (below $10 \mu\text{M}$) and sulphates replace oxygen as an organic matter oxidiser, leading to the production of toxic hydrogen sulphide. This transition zone is critical for the cycling of elements such as nitrogen, manganese, iron, and sulphur, and is a site for important redox reactions (J. W. Murray & Yakushev, 2006).

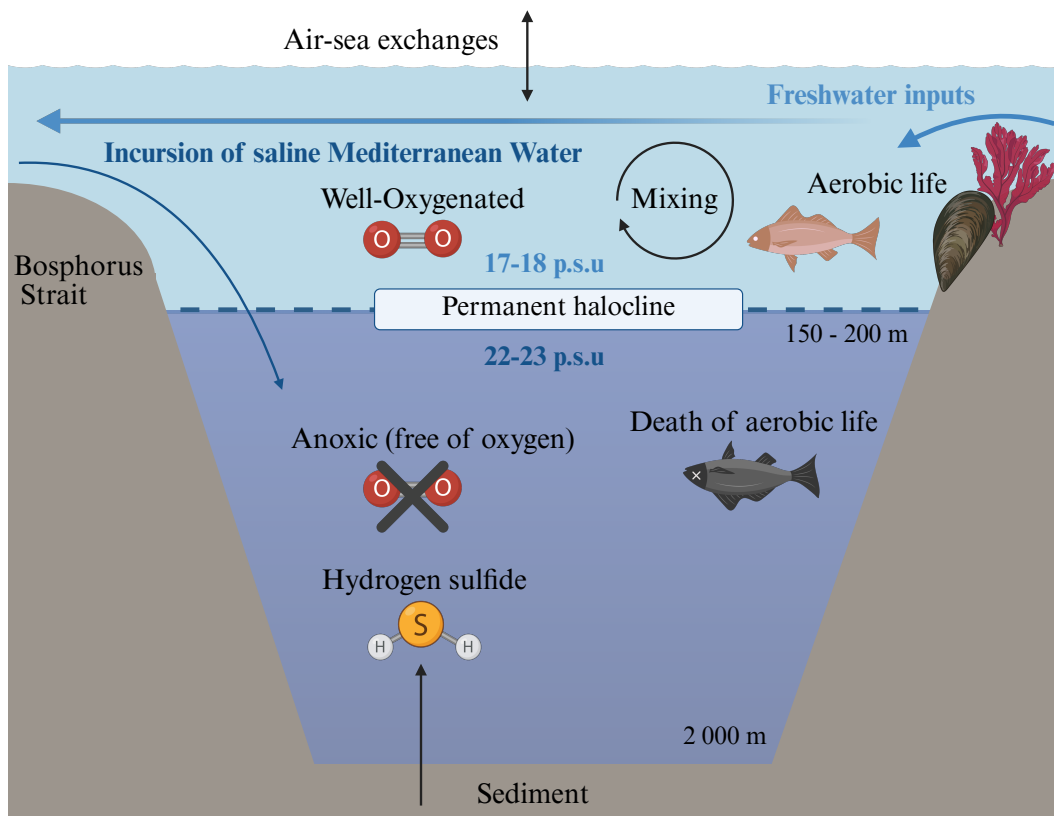


Figure 1.9: Simplified representation of the two layers system in the Black Sea, redrawn from Özsoy and Ünlüata (1997). The oxic surface (in light blue) and deep euxinic (in dark blue) layers are separated by a permanent halocline. As a result, aerobic life is concentrated in the thin surface layer. The figure is not at scale and was made with BioRender.

1.5.3 The northwestern shelf of the Black Sea

The Black Sea can be divided into two major parts: a wide northwestern continental shelf with an average depth of 50 m connected to the deep basin (more than 2 000 m). The northwestern continental shelf covers an area of 70 000 km² and this region is bordered in the south by the city of Varna, in Bulgaria, and in the east by Sevastopol, in Ukraine (Wijsman et al., 1999). This sector is characterised by low salinity (less than 17 *p.s.u*) and high nutrient concentration caused by run-offs from large rivers such as the Danube, the Dniester and the Dnieper. The Black Sea receives a total of freshwater inputs of 300 km³ per year, from Central and Eastern Europe, and its catchment area covers 1.8 × 10⁶ km² which is about 4.5 times larger than the surface of the Black Sea (Staneva & Stanev, 1998). The northwestern sector receives about 80% of the total freshwater input (Mikhailov et al., 2008).

The wide bottom of the northwestern shelf is covered by a diverse fauna and flora that inhabit contrasting benthic habitats (Zaitsev & Mamaev, 1997). The majority of oxic benthic life is concentrated on the *NWS* since the bottom water of the deep basin is anoxic (Zaitsev, 2008). The period prior to the mid-sixties is considered as a relatively undisturbed **pristine period** by anthropogenic factors with more than 20 biological associations and communities studied (Băcescu et al., 1971). The northwestern continental shelf was home to large beds of mussels and the red algae *Phyllophora* spp., which both provided habitats for hundreds of species of invertebrates (Vinogradov & Zakutsky, 1967). In the late 1960s, human activities (e.g. urban expansion, agricultural and industrial development) led to a sharp increase in nitrate and phosphate loads, causing **eutrophication** and subsequent **bottom hypoxia** (Garnier et al., 2002; Zaitsev, 1992). Bottom hypoxia altered macrobenthos structure and composition, leading to mass mortalities of an estimated 60 million tons of benthic macrofauna (Zaitsev, 1992). Benthic biodiversity has decreased by almost a factor of three and shallow communities have shifted towards a dominance of opportunistic species, at the expense of the most vulnerable species to hypoxia. From the early nineties, benthic communities seem to be slightly improving but their recovery remains precarious and macrozoobenthos remains vulnerable to anthropogenic impacts (e.g. beam trawling and dredging activities), invasive species and climate change (Mee et al., 2005).

A brief overview of the most extensively investigated biocoenoses is given in the following text and illustrated in (**Figure 1.10**). Since the 1960s, these communities have been changing rapidly: certain species benefit, whereas others are put at risk by climate change and growing human pressures. Biocoenoses are designated according to the dominant macrobenthic species they contain (Zaitsev, 2008).

- The **Zernov's *Phyllophora* Field (ZPF)**, located in the north-east sector of the *NWS*, was historically one of the most emblematic benthic habitats of the Black Sea. Covered by the red algae *Phyllophora* spp., *ZPF* hosted more than 100 species of invertebrates between 30 and 60 m. Before the onset of eutrophication, the *ZPF* covered an area of 11 000 km², with a total algal biomass estimated at 6 – 10 million tons.

Eutrophication limited light penetration, leading to a sharp decrease in surface covered (50 km²) and total biomass (300 000 tons) at the early nineties (Zaitsev, 1992). The decline of filter feeders exacerbated the drop in the amount of light to the bottom, negatively impacting the *ZPF* (Zaitsev & Mamaev, 1997). Much of the original algal cover was replaced by meadows of *Dipolydora quadrilobata* tubes, leading to substantial changes in the structure of the community.

- **Sandy bottoms** extend from approximately 18 to 30 m in depth. Historically, shallow sandy bottoms supported the biocoenosis of *Lentidium mediterraneum* and the clam *Chamelea gallina*. However, this community has been largely replaced by the invasive bivalve *Mya arenaria*, which became dominant in many sandy areas during the late twentieth century (Zaitsev & Mamaev, 1997).
- The influence of the **Danube River** is predominant for the distribution of macrobenthos over the *NWS* (Teacă et al., 2019a; Wijsman et al., 1999). Benthic habitats in the plume of rivers are characterised by high inputs of fine suspended matter, which is permanently deposited on the seabed, high turbidity, low transparency in the water column, and lower oxygen content. Before the eutrophication period, this area supported a characteristic community composed of *Spisula*, *Abra*, and *Acanthocardia* (Băcescu et al., 1971). The present **muddy habitats** are now largely dominated by species with high ecological plasticity, capable of adjusting their feeding strategies or life history traits to persist under fluctuating conditions such as opportunistic polychaetes (e.g. *Melinna palmata* and invasive bivalves *Anadara kagoshimensis* and *M. arenaria* (Teacă et al., 2019a, 2020).
- At higher depths, between approximately 30 and 55 metres, **biogenic coarse** sediments host the **mussel biocoenosis**. These coarse sediments represent remnants of ancient littoral bars, composed of dead shells accumulated at this level after a sea level transgression approximately 30 000 years ago (see references in Teacă et al. (2019a)). *Mytilus* biogenic reefs host a diversity of epibenthic species, including active predators such as polychaetes and nemertean, as well as grazing crustaceans, hydroids and sponges (Zaitsev & Mamaev, 1997; Zaitsev, 2008; Teacă et al., 2019a). Biogenic reefs have been severely affected by eutrophication, predation by the invasive whelk *Rapana venosa* and bottom trawling. Since the early 2000s, a new community has emerged, characterised by the coexistence of *Mytilus* and the invasive polychaete *Dipolydora quadrilobata* (Teacă et al., 2020). This new habitat is now one of the richest in terms of diversity and abundance (Teacă et al., 2019a, 2020).
- Between mussel reefs and deeper communities, a transitional area with a mix of *Mytilus* and *Modiolula* is observed between 55-70 metres (Gomoiu, 1997).

- In the **offshore circalittoral zone**, the environmental conditions are relatively stable and benthic communities are less impacted by human disturbances compared to shallower water biocoenoses (Begun et al., 2010). The biocoenosis of the horse-mussel *Modiolula phaseolina*, found at depths from 55-60 to 120 metres, is one of the most extensive in the Black Sea. On mixed and coarse sediment (shells debris in a muddy matrix), a variety of species thrive, such as sponges, tunicates, polychaetes, and some crustaceans (Begun et al., 2010). These deeper communities are mostly threatened by the intrusion of euxinic waters from the deep basin and the shallowing of the oxycline. The biomass and average size of *M. phaseolina* have declined twofold compared to the mid-20th century, indicating recent community degradation (Friedrich et al., 2014).
- At the shelf break, the intrusion of euxinic waters affects negatively the health of macrofauna communities, and only some polychaetes and nematodes can survive below 100- 150 metres (Zaitsev, 1992). This area is expected to expand because of global changes (section 1.5.4)

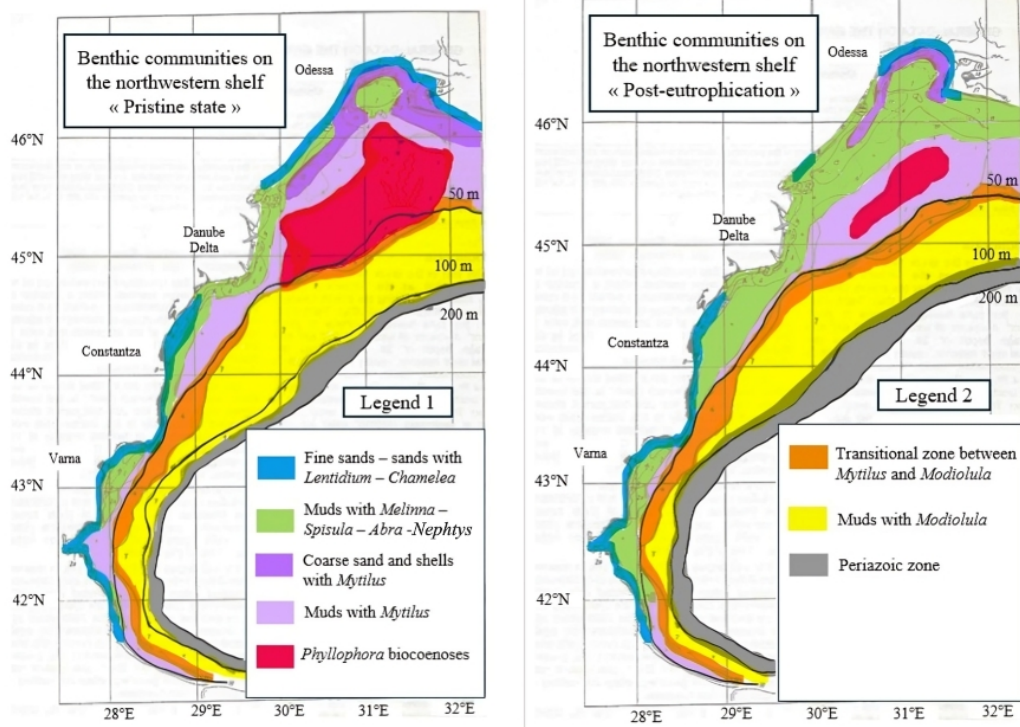


Figure 1.10: Simplified representation of **benthic communities** on the northwestern shelf of the Black Sea. On the left side is the pristine state before the 1960s (Băcescu et al., 1971), and on the right side is the post-eutrophication state in the nineties (Gomoiu, 1992; Zaitsev & Mamaev, 1997). The legend is at the bottom right of both figures. Black lines indicate the isobaths at 50, 100 and 200 metres. The template has been taken from (Gomoiu, 1997).

1.5.4 The Black Sea under multiple pressures

The northwestern continental shelf of the Black Sea is heavily stressed by hypoxia driven by eutrophication, along with other stressors such as pollution and the introduction of invasive species. These factors have led to a drastic simplification of both pelagic and benthic communities (Diaz & Rosenberg, 1995; Zaitsev & Mamaev, 1997). The Black Sea is particularly vulnerable to anthropogenic disturbances and natural variability due to its landlocked character (Grégoire & Soetaert, 2010). During the past decades, macrozoobenthic communities have been strongly impacted by climate change and human activities (**Figure 1.11**). In the following section, a summary of the main pressures on the NWS is given.

- **Bottom trawling**, initially for sprat in the 1970s–1980s and later for the invasive whelk *R. venosa*, caused major disturbances to mussel communities and their associated epifauna. This led to a shift towards less diverse assemblages dominated by infaunal polychaetes and clams (Nenciu et al., 2023; Zaitsev & Mamaev, 1997; Todorova et al., 2021). As an indirect impact, strong siltation further decreased epibenthic organisms such as bivalves and crustaceans, with additional impacts from by-catch of slow-growing species such as *Spisula subtruncata* and *Pitar rudis* (Teacă et al., 2019b).
- **Expansion of shallow seasonal hypoxia**: As mentioned previously, lack of oxygen poses a major threat to benthic communities and their functioning. In the 1960s, coastal bottom hypoxia covered a total of 2 000 – 3 000 km², which increased to 15 000 km² in the 1980s at depths ranging from 7 to 40 m (Zaitsev, 1992). The extent of coastal hypoxia is expected to grow in the context of climate change and increasing anthropogenic disturbance (Breitburg et al., 2018; Diaz & Rosenberg, 2008).
- **Invasive species**: Unintentional introduction of new species has significantly transformed some communities in the Black Sea (Gomoiu, 1984; Gomoiu et al., 2002; Zolotarev, 1996). In a century, more than 60 alien species were estimated in the Black Sea benthic's communities (Shalovenkov, 2020). In the last 20 years, the rate of new alien species has increased significantly compared to the previous century (Shalovenkov, 2020). This increase may be related to warmer temperatures and increased sea traffic, which transport invasive species in ballast water and fouling (Shiganova, 2010).
- **Global warming** has multiple effects on the ventilation processes and dynamics of the Black Sea (Capet et al., 2020; Stanev et al., 2019). Global warming leads to a reduction in the ventilation process, the warming trend in the *CIL* of approximately 0.05 °C/year was more than double the rate of previous decades, bringing its temperature close to that of deeper waters (Stanev et al., 2019). Warmer winters now limit the formation of the *CIL* to every two or three years instead of annually (Stanev et al., 2019).

Sea Surface Temperature (SST) has been rising since the 1990s (Shapiro et al., 2010), with reported trends ranging from 0.4 – 0.9 °C per decade (Nardelli et al., 2010; Ginzburg et al., 2004; Korotaev et al., 2024). Warming is also recorded at depth, with an increase of almost 0.4 °C over the last 20 years at 150 m (Korotaev et al., 2024). Higher temperatures are expected to expand the duration of the stratified period in summer and reduce oxygen solubility, amplifying deoxygenation (Capet et al., 2013)

- The haline structure of the basin has changed significantly in the past decade, with a noticeable **increase in salinity** in the pycnocline layer (Kubryakov et al., 2024). This rise weakens haline stratification and may alter nutrient entrainment in the upper Black Sea (Kubryakov et al., 2021). Global warming may be one of the reasons for the rise in surface salinity due to wind and mechanical mixing. As a result, warmer and saltier conditions (**Mediterranisation**) can favour invasive species while reducing habitat suitability for native cold-water species (Sezgin et al., 2010).
- Global warming also affects the dynamics of the hydro sulphuric layer and is considered one of the main reasons for the progressive **rise of euxinic waters** towards the surface (Oguz, 2005), in combination with eutrophication (McQuatters-Gollop et al., 2008). During the past 60 years, the total oxygen inventory in the Black Sea has decreased by 44%, and the depth of the oxygen layer has decreased from 140 m to 90 m, resulting in compression of habitable oxygenated space with potential severe ecological and economic consequences (Capet et al., 2012, 2016).
- The northwestern shelf of the Black Sea is increasingly affected by **pollution** from industrial and agricultural sources (Pokazeev et al., 2021). The expansion and industrialisation of agriculture are increasing nitrate and phosphate discharges, driving eutrophication and the development of bottom hypoxia (Mee et al., 2005). These effects are exacerbated by the semi-enclosed, land-locked nature of the basin and the long water renewal time, which limit dilution and dispersal. Therefore, pollutants such as oil, sewage, pesticides, heavy metals, microplastics, and marine litter accumulate and persist, further deteriorating benthic habitats and ecosystem functioning (Pokazeev et al., 2021).

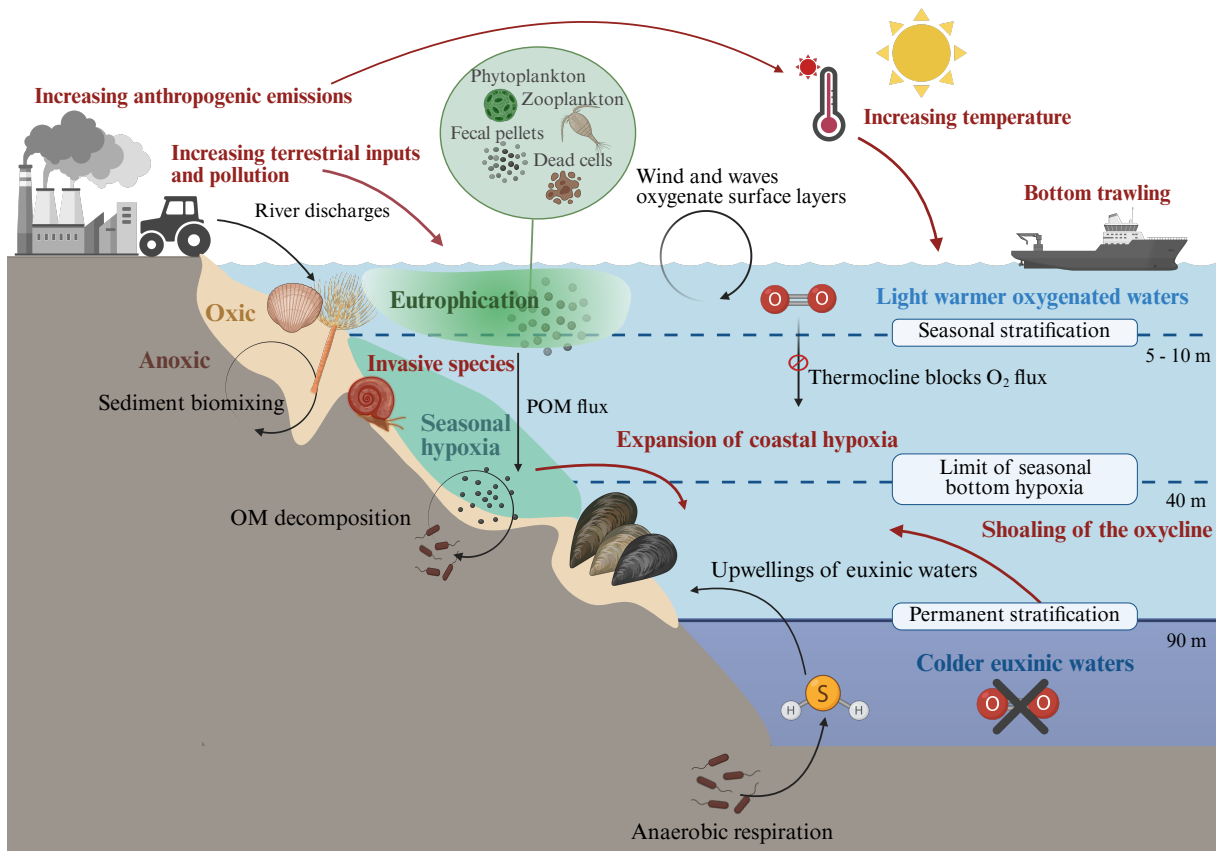


Figure 1.11: Graphical summary of the **main pressures** on benthic ecosystems (indicated by dark red arrows). The figure has been made with BioRender.

1.6 PhD thesis objectives

1.6.1 Questions and research gaps

This present PhD thesis targets the development of modelling tools able to map the distribution of benthic traits at the scale of the ecosystem from the analysis of their local distribution of the scale of the community. To achieve this goal, this PhD work combined a numerical ocean model, field sampling data, trait compilation, multivariate analysis, and *TDMs* to link punctual observations of biodiversity to the maps of vulnerabilities and functions of macrobenthos on the shelf scale (**Figure 1.12**). This methodology will answer the following main question of research:

“How can we upscale local observations of benthic biodiversity to map benthic functions and vulnerabilities at the shelf scale relevant for policy management and ocean model?”.

As a sub-question, we are interested in the Black Sea itself as it offers unique opportunities to investigate long term changes of benthic biodiversity in relation to environmental gradient. As discussed previously, the northwestern shelf of the Black Sea has been affected by eutrophication and bottom hypoxia since the sixties.

However, the nature of the current benthic communities remains questionable and difficult to compare the macrobenthos communities state between different periods due to the difference in sampling protocol and taxonomic issues. In this PhD thesis, we propose to elucidate this gap by providing a comprehensive study of the long-term changes of macrobenthos's communities in relation to abiotic gradient since the mid-nineties. In addition, despite the growing interest of trait-based approaches in marine ecology, there is still a clear lack of studies in major marine systems, such as the Black Sea, and gathering trait information is time-consuming, unwell documented and scarce for many benthic taxa (Lam-Gordillo et al., 2020; Tyler et al., 2012; Beauchard, 2023). Here, we propose to fulfil the knowledge gap in macrozoobenthic traits for the northwestern shelf of the Black Sea by compiling an extensive trait data set.

The few existing studies that have correlated benthic traits to the environment at local scales only used information on sediment composition or surface satellite temperature and chlorophyll to define the habitat (Kostylev & Hannah, 2007; Paganelli et al., 2012). These studies were mostly limited to local research (Bremner et al., 2006a; Paganelli et al., 2012). Currently, there is a lack of initiatives that attempted to establish statistical models for formulating the trait-environment relationship in the marine realm. In this PhD thesis, we propose to use the environmental information provided by a numerical ocean model to connect benthic traits with a wide range of biogeochemical (e.g. bottom oxygen concentration, content of *Particulate Organic Carbon (POC)* in the sediment) and physical variables (e.g. bottom shear stress, bottom temperature) by developing *TDMs* based on neural networks. *TDMs* have been increasingly used over the past 20 years, but inconsistencies in terminology, neglect of intraspecific trait variability and limited trait databases hinder their broader application (Funk et al., 2017). *TDMs* are more developed for plants, as trait-based approaches were originally developed and discussed for plants in terrestrial ecosystems (Zakharova et al., 2019; Barton et al., 2013). In the review of Zakharova et al. (2019), fifty percent of all studies focus on terrestrial vegetation, with marine ecosystems primarily represented by fish (Beukhof et al., 2019; Fisher et al., 2010; Frainer et al., 2017) and plankton populations (Litchman & Klausmeier, 2008; Litchman et al., 2013; Brun et al., 2016, 2019). This highlights the urgent need to develop this kind of trait-based modelling approaches for benthic species as well. Here, numerical ocean models will predict the distribution of environmental variables and *TDMs* will be developed to establish the link with benthic traits and map their distribution at the shelf scale.

Based on these gaps in research for the Black Sea and biological trait studies in general, the main question of research:

“How can we upscale local observations of benthic biodiversity to map benthic functions and vulnerabilities at the shelf scale?” can be refined into four more specific ones (**Figure 1.12**): first question is the **Chapter 3** (corresponding to the first paper; Chevalier et al. (2024)), second question is the **Chapter 4** (corresponding to the second paper; Chevalier et al. (2025a)), third and fourth questions are the **Chapter 5** (corresponding to the third paper; Chevalier et al. (2025b)).

1. What are the relationships between macrobenthic species and abiotic conditions at the local scale?
 - 1.1. What are the changes in environmental conditions 1990s and 2010s?
 - 1.2. Are the species compositional changes associated with these changes, particularly regarding organic pollution ?
2. What are the biological traits of the macrobenthos' species?
 - 2.1. How can we define the vulnerabilities of the macrobenthos?
 - 2.2. How can we define the functions of the macrobenthos?
3. What are the relationships between (multi)-traits and abiotic conditions at the local scale?
4. How can we upscale this local information to map the distribution of these traits at the shelf scale?

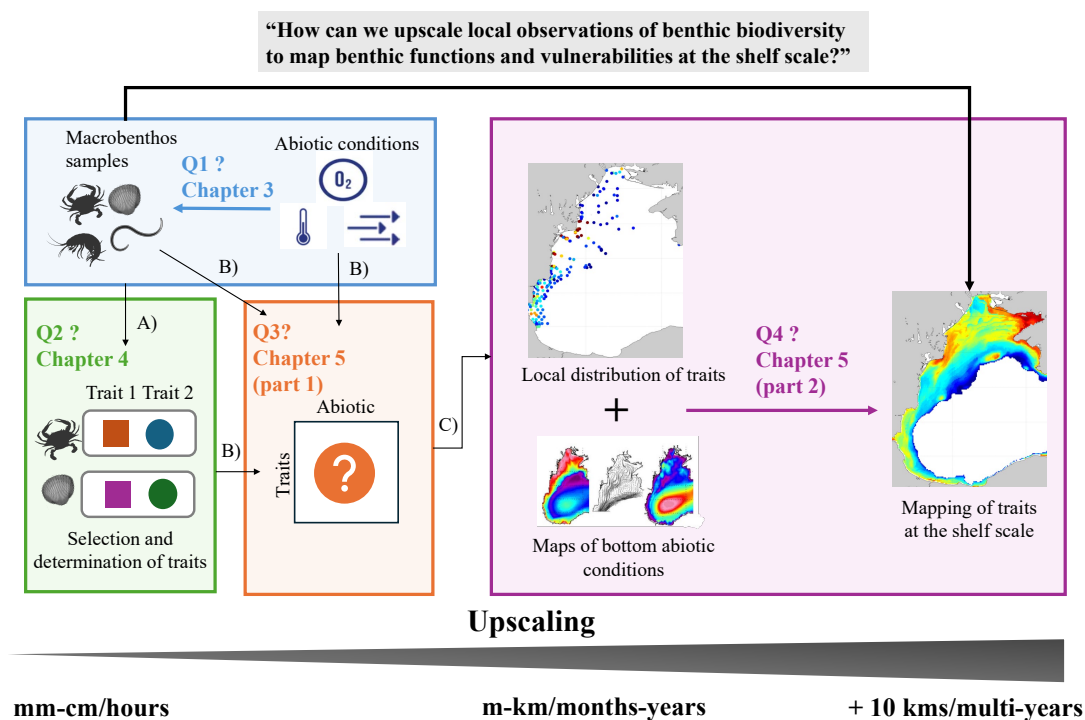


Figure 1.12: Connections between the different sub-questions and associated chapters of this PhD thesis. A) the compilation of traits was done on a subset of macrobenthic species, B) traits, species and abiotic data were combined together and C) the trait-abiotic relationships guided the development of *TDMs*, used to predict the spatial distribution of traits based on environmental predictors and local distributions of traits. Upscaling from the species level (mm-cm at the scale of hours) to the maps of benthic functions and vulnerabilities to inform ecosystem-based management at the scale of several kilometres and multi-years.

1.6.2 Outline of this PhD thesis

This PhD thesis is organised as follows:

- General Materials and Methods (**Chapter 2**): This chapter presents the general materials and methods used in this PhD work. Briefly, I give an overview of the species, abiotic and traits data. More details about field sampling, multivariate statistical approaches (e.g. RLQ and fourth-corner methods), *TDMs* and ocean numerical model are also given.
- Linking species and environmental conditions in the past and present (**Chapter 3**): This chapter explores the long-term relationship between organic pollution and macrobenthic communities over the past 25 years (1995-2017) on the Black Sea *NWS*. This chapter corresponds to my first paper as the first author, published in *Marine Pollution Bulletin* (Chevalier et al., 2024) and provides answers for the first sub-question (**Q1**) of this PhD work.
- Trait data compilation (**Chapter 4**): This chapter presents the data compilation work and gives some insights on how to use the data. We present an open-access dataset that combined 27 functional traits of more than 120 benthic species with environmental variables derived from an ocean biogeochemical model. We selected traits for their relevance to ecosystem functioning (e.g. sediment dwelling depth and mobility) and sensitivity to environmental pressures (e.g. body size, reproductive strategy). This chapter corresponds to my second paper published in *Scientific Data* (Chevalier et al., 2025a) and aims to answer the second sub-question (**Q2**) of this PhD work.
- Mapping of benthic traits (**Chapter 5**): This chapter presents the study of the relationships between abiotic predictors and traits or multiple trait-based indicators. Using multivariate statistical approaches (e.g. RLQ and fourth-corner analyses), we identified relationships between environmental drivers and traits. The trait-abiotic conditions relationships guided the development of *TDMs*, used to predict the spatial distribution of traits based on environmental predictors. We provided maps of key ecological indicators derived from traits (e.g. biomixing, bioirrigation, biodeposition), which can be used as proxy of ecosystem functions delivered by the macrobenthos. We also map vulnerabilities of macrobenthos's communities with characteristic life-history strategies. This chapter corresponds to the third publication published *Scientific Reports* (Chevalier et al., 2025b) and aims to answer the sub-questions 3 (**Q3**) and 4 (**Q4**).
- General discussion (**Chapter 6**): This chapter presents a final general discussion of the main results of this work and possible improvements. Perspectives of this PhD thesis are also discussed such as the predictions of the functional biodiversity in a future climate or the integration of the variability of seafloor functions into ocean model to better estimate ecosystem functions at the shelf scale.

1.6.3 Supporting publications

A complete list of publications and communications related to this thesis is provided in **Appendix A**. The following are the three main publications on which this thesis is based. A summary of the contributions of the authors is provided in **Table 1.1**. This PhD work is financially supported by the FNRS (Fonds de la Recherche Scientifique de Belgique, Communauté Française de Belgique) through funding a FRIA grant to SC. SC and MG acknowledged the support of the EU H2020 BRIDGE-BS project under grant agreement No. 101000240 and EU HE NECCTON project under grant agreement No. 101081273. The following publications benefited from computational resources made available on Lucia, the Tier-1 supercomputer of the Walloon Region, infrastructure funded by the Walloon Region under the grant agreement No. 1910247.

- **Paper I:** Chevalier, S., Beauchard, B., Teacă, A., Soetaert, K., & Grégoire, M. (2024). Partial recovery of macrozoobenthos on the northwestern shelf of the Black Sea. *Marine Pollution Bulletin*, 207,116857. <https://doi.org/10.1016/j.marpolbul.2024.116857>.
- **Paper II:** Chevalier, S., Beauchard, O., Teacă, A., Begun, T., Todorova, V., Vandembulcke, L., K. Soetaert., & Grégoire, M. (2025a). A macrozoobenthic data set of the Black Sea northwestern shelf. *Scientific data*, 12(1), 957. <https://doi.org/10.1038/s41597-025-05311-2>. Dataset associated on *Figshare* (<https://doi.org/10.6084/m9.figshare.27888843.v1>).
- **Paper III:** Chevalier, S., Beauchard, O., Vandembulcke, L., Teacă, A., Begun, T., Todorova, V., K. Soetaert., & Grégoire, M. Trait modelling to predict benthic functions and vulnerabilities across Black Sea seascapes. *Scientific Reports*, 15(1), 39076. <https://doi.org/10.1038/s41598-025-24508-4>.

Table 1.1: Authors contribution per published paper: Adrian Teaca (AT), Karline Soetaert (KS), Luc Vandembulcke (LV) Marilaure Grégoire (MG), Olivier Beauchard (OB), Séverine Chevalier (SC), Tatiana Begun (TB) and Valentina Todorova (VT)

Contributions	Paper I Chapter 3	Paper II Chapter 4	Paper III Chapter 5
Study design	SC,OB	SC,OB	SC,LV,OB
Data collection	AT	AT,OB,TB,VT	AT,OB,TB,VT
Data analysis	SC,MG,OB	SC,OB	SC,OB,LV
Manuscript preparation	SC,MG,OB	SC,OB	SC,MG
Manuscript revision	All authors	All authors	All authors
Supervision	MG and KS	MG and KS	MG and KS

GENERAL MATERIALS AND METHODS

This PhD work combines field sampling, the determination of response and effect traits, multivariate analysis, Trait Distribution Models based on neural networks and an ocean numerical model (**Figure 2.1**). Our data set, including species occurrences, trait information, abiotic variables, and code for building ecological indicators from traits (e.g. biomixing and bioirrigation potentials), are publicly available in the repository *Figshare*. In this chapter, a brief overview of the data and tools used in this thesis is given.

Briefly, species data (i.e. abundance and biomass data) were gathered from field sampling conducted on the northwestern shelf between 1995 and 2018. Substratum type was also identified through field sampling, while other abiotic predictors (e.g. bottom temperature, bottom shear stress, oxygen) were obtained from an ocean model (the biogeochemical model BAMHBI described in Grégoire et al. (2026)). In **Chapter 3**, relationships between macrozoobenthic species and abiotic conditions were determined using co-inertia analysis (2-tables ordination). Then, we compiled traits for a subset of 127 taxa (out of a total of 215 taxa) for which trait information was available in **Chapter 4**. The links between the three tables (abiotic, species and traits) were explored using multivariate statistics (RLQ combined with fourth-corner Dray et al. (2014)) in the first part of **Chapter 5**. Finally (second part of **Chapter 5**), we developed Trait Distribution Models based on artificial neural networks to create full coverage maps of traits. These maps were constructed using the local relationships between traits and abiotic descriptors, as well as maps of bottom abiotic conditions obtained from the ocean model (**Figure 2.1**).

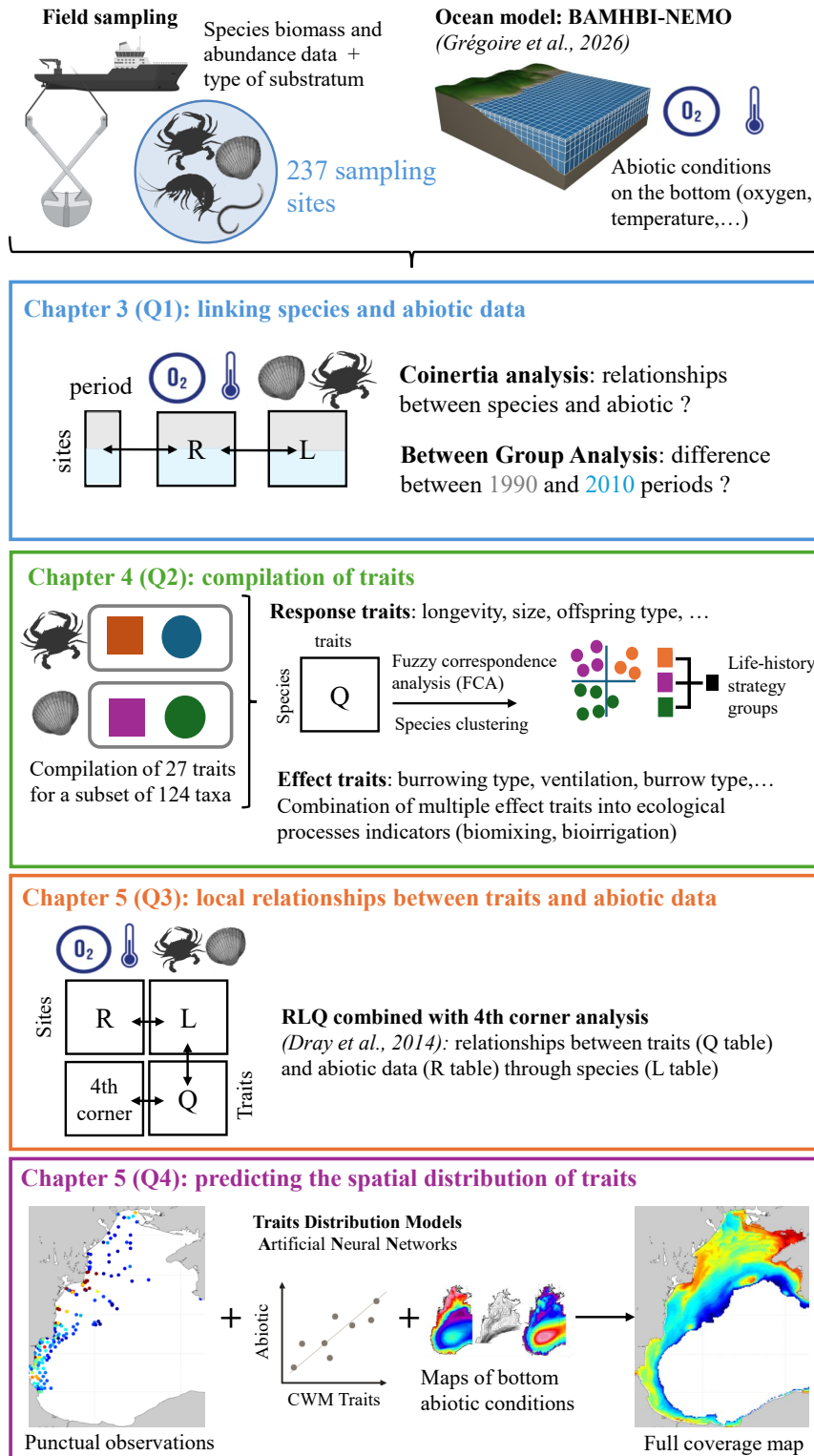


Figure 2.1: Summary of the data and techniques used throughout this PhD thesis. Colours correspond to one of the research questions stated in **Chapter 1**.

2.1 Field sampling

2.1.1 Campaigns over the northwestern shelf

Macrozoobenthic species (i.e. invertebrates < 0.5 mm) have been collected between 1995 and 2017 for a total of 237 stations sampled in six different campaigns over the northwestern shelf of the Black Sea. The spatial and temporal coverage of the sampling effort is provided in **Figure 2.2**. The list of campaigns with their respective date and number of sampling sites is provided in **Table 4.1**. From 5 to 27 August 1995, 27 stations were sampled on the northwestern continental shelf of the Black Sea in the framework of the European project EROS-2000 (Wijsman et al., 1999; Gomoiu, 1997). The depth of the sampling stations varied from 12 m to 69 m. For the most recent period (i.e. between 2008 and 2017), the sampling, processing and identification of the macrobenthos samples was done by the Romanian National Institute of Marine Geology and Geoecology (GeoEcoMar) for SESAME (2008), HYPOX (2010 and 2011) as well as for EMBLAS (2016-2017). The SESAME (for the Southern European Seas Assessing and Modelling Ecosystem Changes) campaign took place between the 7th of April and 11th of September 2008. A total of 8 stations were sampled during the SESAME campaign, with an increasing depth gradient beginning in the southern part of the Delta of the Danube to deeper stations at depths greater than 140 m. The depth varies between 14 and 146 m.

Four stations were sampled in April 2008, and the last four were sampled in September 2008. The HYPOX campaign took place between 14th May 2010 and 10th April 2011. A total of 81 stations were sampled along a transect from the Danube Delta to deeper, less oxygenated water (**Figure 2.2**). The HYPOX campaign covers the southern part of the northwestern continental shelf. Four cruises with the R/V MARE NIGRUM within HYPOX were carried out and covered the entire Romanian continental shelf at water depths ranging from 10 to 200 m. The first cruise took place in May 2010 with 32 stations, the second cruise took place in July 2010, with a total of 4 stations, the third cruise took place in September 2010 with a total of 15 stations, and the last cruise took place in April 2011 with a total of 30 stations. The EMBLAS campaigns were first carried out in May 2016 and, in July 2017, aboard the Mare Nigrum R/V. The first campaign took place between 17th May and 21st May and covered the northern part of the continental shelf (north-east to the Delta of the Danube). The depth ranged between 13 and 52 m, and 15 stations were sampled. The second campaign took place on the 10th of July, and the depth varied between 16 and 70 m. A total of 7 stations were sampled. For the Bulgarian coast, *in-situ* samples were collected in the framework of *Marine Strategy Framework Directive (MSFD)* between October and December 2017 by the Bulgarian Institute of Oceanology (IO-BAS). More details on the sampling protocol can be found in **Chapter 4**.

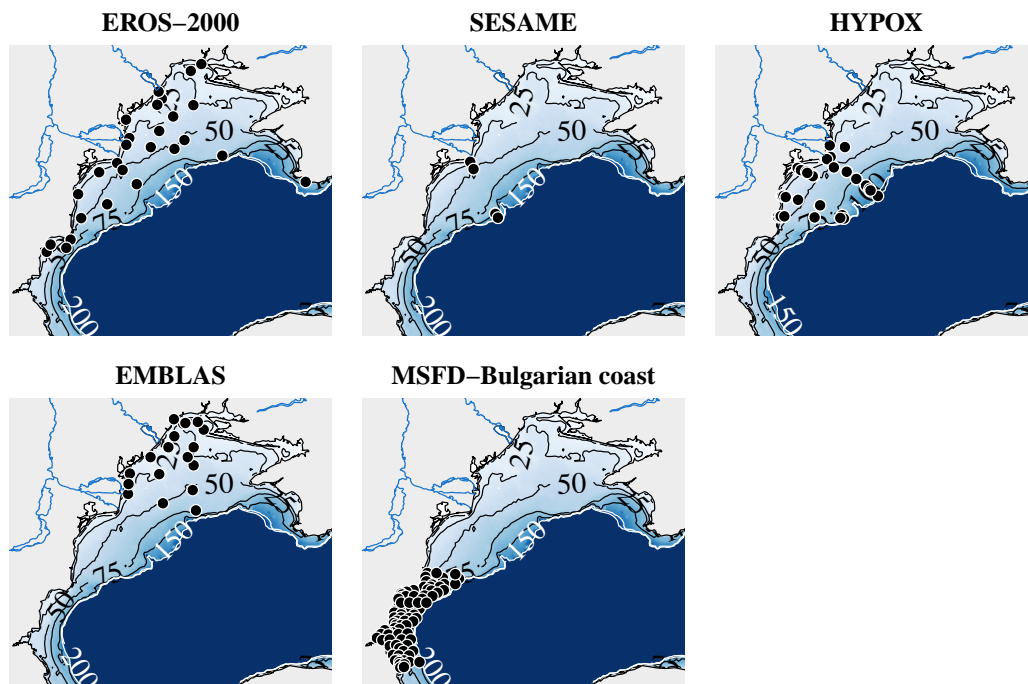


Figure 2.2: Mapping of the sampling sites per campaign at the scale of the northwestern shelf of the Black Sea from 1995 to 2017: EROS-2000 (1995), SESAME (2008), HYPOX (2010-2011), EMBLAS (2016-2017) and MSFD-Bulgarian (2017). Contours of the bathymetry are indicated with continuous black and white lines and in different shades of blue. Black dots are sampling stations for the considered campaign.

2.1.2 Type of substratum

Samples from *in-situ* field sampling also include the type of substratum information as well as the depth of the sampling. The type of sediment was estimated from either direct observation, sediments characteristics analysis or derived from map of seabed substrate. In more details, for EROS-2000 campaign, sediment characteristics (e.g. medium grain size, % of shells) were provided in Wijsman et al. (1999), for HYPOX and SESAME campaigns, sediment types (e.g. sandy mud, coarse mixed sediments) were recorded after sampling following the nomenclature from Todorova and Konsulova (2005) and for EMBLAS campaign, sediment type was estimated based on a map produced by EMODnet. Given the different sources of sediment description between campaigns, we chose to classify sediment type based on “EUNIS-like” classes applicable in the Black Sea (Populus et al., 2017; Vasquez et al., 2021). Hence, we divided sediment types into three categories: “Sand to Muddy Sand”, “Mud to Sandy Mud” and “Mixed to Coarse sediments”. The latter class is justified in the Black Sea as the origin of sand is mostly biogenic, formed by small fragments of shells in a mud matrix (Teacă et al., 2020). The spatial distribution of the type of substratum at each sampling site is provided in **Figure 2.3**.

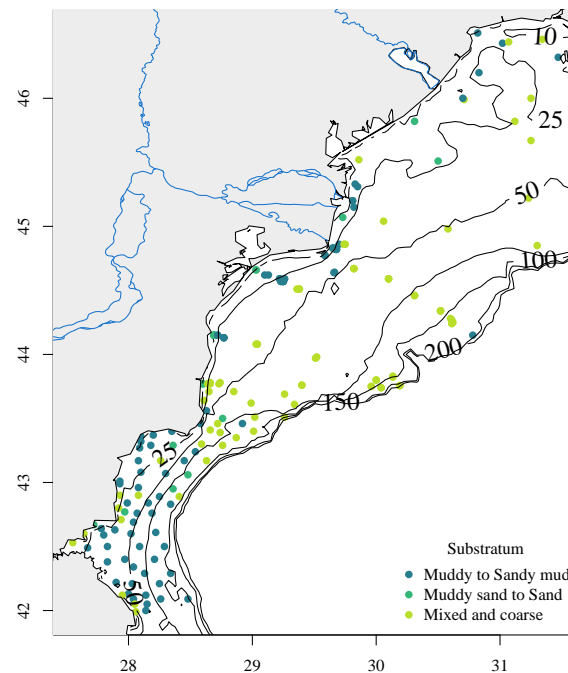


Figure 2.3: Spatial distribution of the type of substratum over the northwestern shelf of the Black Sea. Each dot correspond to a sampling site. At the bottom left, colour code for the three chosen type of substratum. Contours of the bathymetry are indicated with continuous black lines and major rivers are in blue.

2.1.3 Species data

During the EROS-2000 campaign, the fauna was sampled with a Reineck box corer (60×30×30 cm) and then, samples were stained with Congo Red (the sodium salt of 3,3'-([1,1'-biphenyl]-4,4'-diyl)bis(4-aminonaphthalene-1-sulfonic acid) and preserved in buffered formaldehyde. In the laboratory, the samples were washed through 1.0 mm mesh and all the material retained was examined by binocular microscope. Macrozoobenthic animals were identified, if possible, to the lowest taxonomic level (i.e. species level). Abundance was expressed as number of individual organisms per m^2 and wet weight was determined and reported as g per m^2 . Animals with shells are weighed with their shells and the water should be drained off bivalves before weighing. Since 2005, the Black Sea Commission recommended to use the standardised macrozoobenthos protocol described in Todorova and Konsulova (2005). This protocol was used for all the samples collected between 2008 and 2017. Briefly, samples were collected with a Van Veen grab with surface of 0.135 m^2 and washed through a 0.5 mm mesh size sieve. Organisms retained on the sieve were fixed with formaldehyde 4%, buffered with seawater, and finally stored in plastic jars. In the laboratory, the organisms were counted and identified to the lowest possible taxonomic level. Individual organism density was expressed in number of individuals per m^2 and biomass density, including shells, was expressed in wet weight g per m^2 .

Our species data set recorded 5069 occurrences at 237 sampling stations belonging to 215 taxa from 11 phyla. Specifically, there were 260 occurrences at the phylum level, 214 occurrences at the class level, 16 occurrences at the order level, 256 occurrences at the genus level and 4323 occurrences at the species level. In this PhD thesis, we focus exclusively on macrobenthos (size above 0.5 mm), and we do not consider the influence of the meiofauna even if their contribution on ecosystem processes can be significant especially in hypoxic environments (Bonaglia et al., 2014). Almost half of the total occurrences belong to the phylum Annelida, with most of the occurrences falling under the class Polychaeta (**Figure 2.4**). The phyla Arthropoda and Mollusca each account for 20% of the total occurrences (**Figure 2.4**). Most of the Arthropoda are Malacostraca (989 occurrences out of 1070), while most of the Mollusca are Bivalvia (852 occurrences out of 1049) (**Figure 2.4**). The most frequent taxa are (at least present at 25% of the sampling sites) are listed in **Table 2.1**.

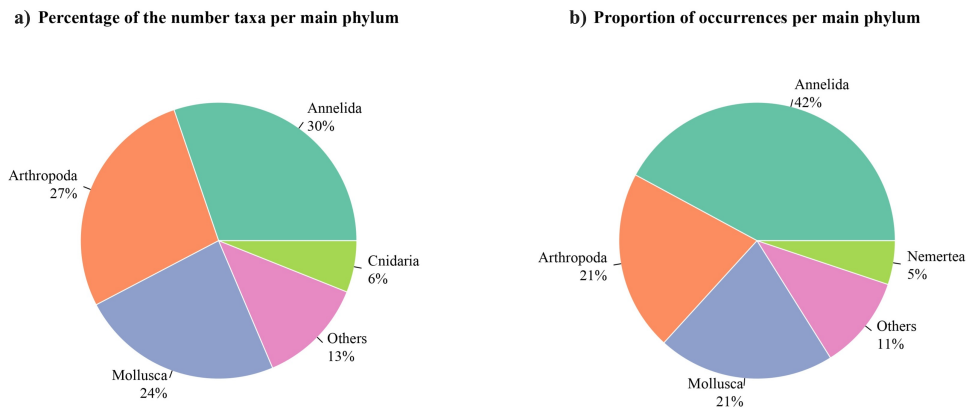


Figure 2.4: a) Percentage of the number of taxa per main phylum and b) proportion of occurrences per main phylum for the whole data set (237 sampling sites). Phylum with less than 5% of the number of occurrences or taxa are merged into “Others” category.

Table 2.1: List of the **most frequent taxa** (at least present at 25% of the sampling sites) per major group (bivalves, crustaceans and polychaetes). Taxa column corresponds to the smallest taxonomic units.

Group	Taxa	Percentage (%)
Bivalves	<i>Abra alba</i>	43
	<i>Spisula subtruncata</i>	35
	<i>Mytilus galloprovincialis</i>	32
	<i>Abra nitida</i>	31
	<i>Parvicardium simile</i>	26
	<i>Pitar rudis</i>	26
Crustaceans	<i>Ampelisca</i> sp.	57
	<i>Iphinoe elisae</i>	49
	<i>Perioculodes longimanus</i>	40
	<i>Phthisica marina</i>	32
	<i>Apseudopsis ostroumovi</i>	27
Polychaetes	<i>Nephtys hombergii</i>	75
	<i>Heteromastus filiformis</i>	73
	<i>Prionospio maciolekae</i>	57
	<i>Melinna palmata</i>	46
	<i>Terebellides stroemii</i>	43
	<i>Phyllodoce mucosa</i>	39
	<i>Polydora cornuta</i>	32
	<i>Capitella</i> sp.	32
	<i>Dipoydora quadrilobata</i>	30
Other	Oligochaetes (worms)	90
	<i>Phoronis psammophila</i> (phoronid)	54
	<i>Sagartia undata</i> (sea anemone)	35
	<i>Amphiura stepanovi</i> (sea brittle)	32

2.2 Abiotic data from a coupled physical-biogeochemical model

Numerical ocean models are widely used tools that predict changes in the physical and biogeochemical status of the ocean over time scales of several decades. In this PhD work, we used the **ocean model *Biogeochemical Model for Hypoxic and Benthic Influenced areas (BAMHBI)***, run in an operational mode for the Black Sea, to derive abiotic conditions of the bottom ocean. *BAMHBI* is an ocean model with a 28-variable pelagic component (including the carbonate system) and a 6-variable benthic component. *BAMHBI* models trophic interactions across the food web, spanning from bacteria to mesozooplankton throughout both the pelagic and benthic systems (**Figure 2.5**). This model has been developed to simulate oxygen deficient environments (it explicitly represents processes in the anoxic layer) and the generation of euxinic waters, and thus is particularly relevant for the Black Sea (Grégoire & Soetaert, 2010; Grégoire et al., 2008; Capet et al., 2016; Grégoire et al., 2026). *BAMHBI* is an intermediate complexity model that represents the cycles of C, N, Si, O₂, P through the pelagic and benthic compartments.

BAMHBI describes the degradation of *OM* using a succession of oxidants. Bacteria preferentially use oxygen and, in anaerobic conditions, first nitrate and then other oxidants (not explicitly described) (Grégoire et al., 2026). The model involves five optional modules that explicitly model 1) the Chl-a content of each phytoplankton functional type, 2) two zooplankton gelatinous groups, 3) the number of aggregates, 4) the carbonates and 5) the benthic compartment (Grégoire et al., 2026).

BAMHBI is a stand-alone biogeochemical model that can be coupled to any hydrodynamical model (Grégoire et al., 2026). Currently, *BAMHBI* is online coupled with the hydrodynamic model *Nucleus for European Modelling of the Ocean (NEMO)* (<https://www.nemo-ocean.eu/>) within the framework of the Copernicus Marine Service (CMEMS) to forecast (daily production of a 10-day forecast) and to hindcast (back to 1950) the biogeochemistry of the Black Sea (Ciliberti et al., 2021). For more details on the *BAMHBI* model, the lecturer can refer to the most recent paper published in *Geoscientific Model Development* (Grégoire et al., 2026).

Our list of abiotic predictors includes **physical** (bottom salinity, bottom temperature, bottom shear stress) and **chemical** variables (bottom oxygen concentration, *POC*, vertically integrated organic carbon concentration in the sediment (slow and fast decay), photosynthetically active radiation (PAR) and the flux of *POC* to the bottom). The list of abiotic descriptors from the model and their respective units is provided in **Table 4.2**. The spatial distributions of a selection of abiotic variables (temperature, shear stress, oxygen and the slow decaying stock of organic carbon in sediment) are provided in **Figures 2.6, 2.7, 2.8 and 2.9**.

2.2. ABIOTIC DATA FROM A COUPLED PHYSICAL-BIOGEOCHEMICAL MODEL

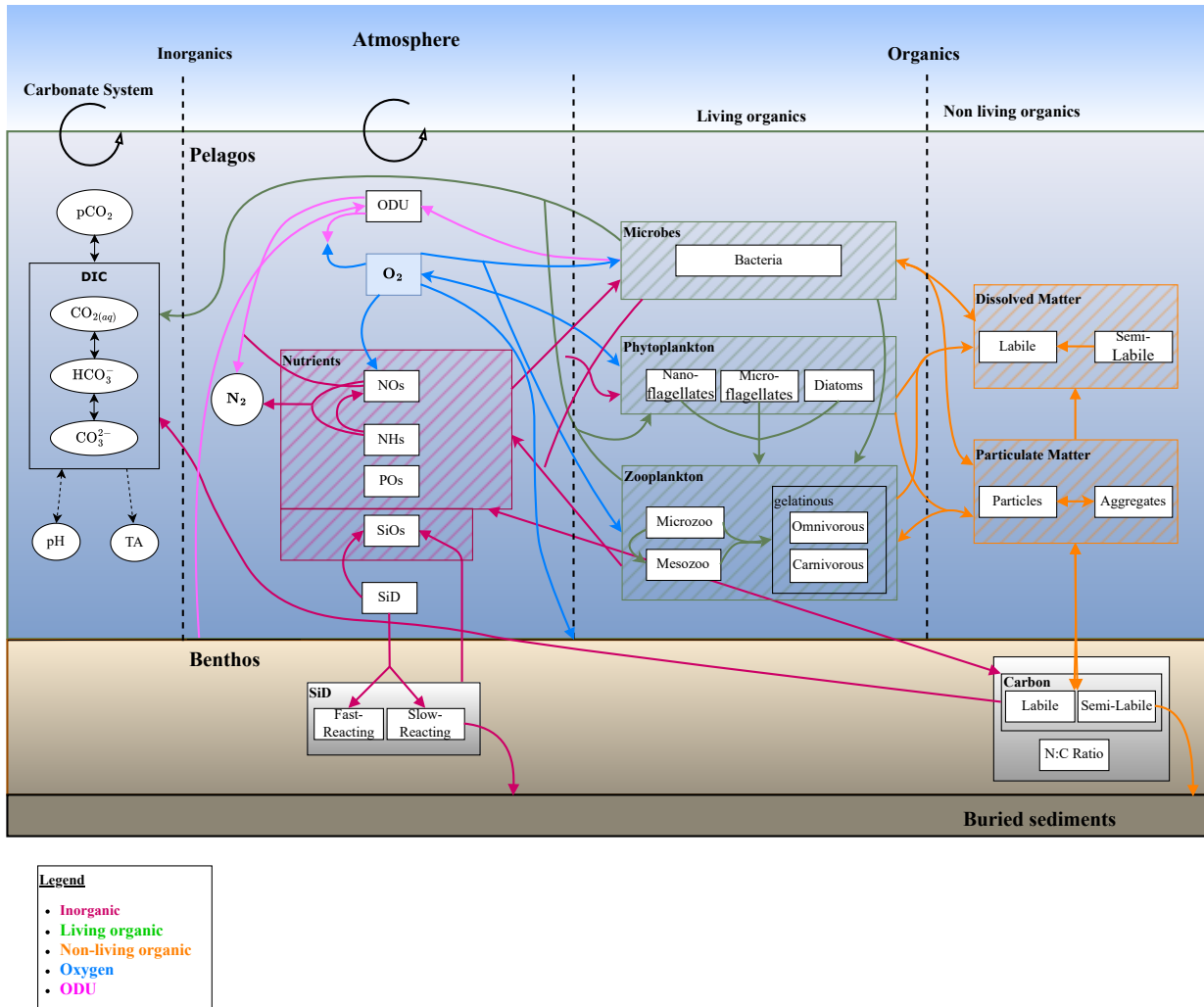


Figure 2.5: Schematic representation of the *BAMHBI* biogeochemical model showing the model state variables (in rectangular boxes) and interactions between them in the pelagic and benthic region differentiating the inorganic, living and non-living organic parts which are made of carbon, nitrogen and phosphorus, and silicon. It also shows the oxygen (O_2) and oxygen demand unit (ODU) fluxes. Optional modules are framed in black: carbonate, benthic, gelatinous, aggregation modules. An additional optional module explicitly computes the Chl-a content of each Plankton Functional Type. The figure was taken from Grégoire et al. (2026).

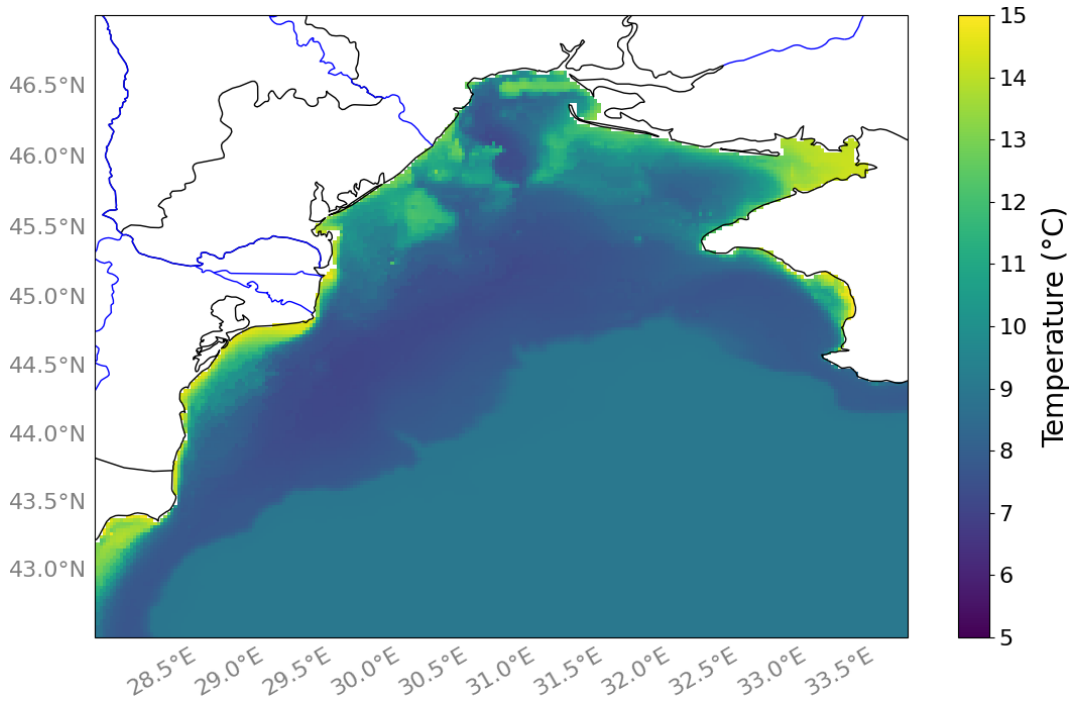


Figure 2.6: Spatial distribution of the **bottom temperature** (in °C, TEMP), from the *BAMHBI* model over the northwestern shelf of the Black Sea. Annual mean from a climatology for the period 2008-2018. Colour bar at the right. Main rivers are in blue.

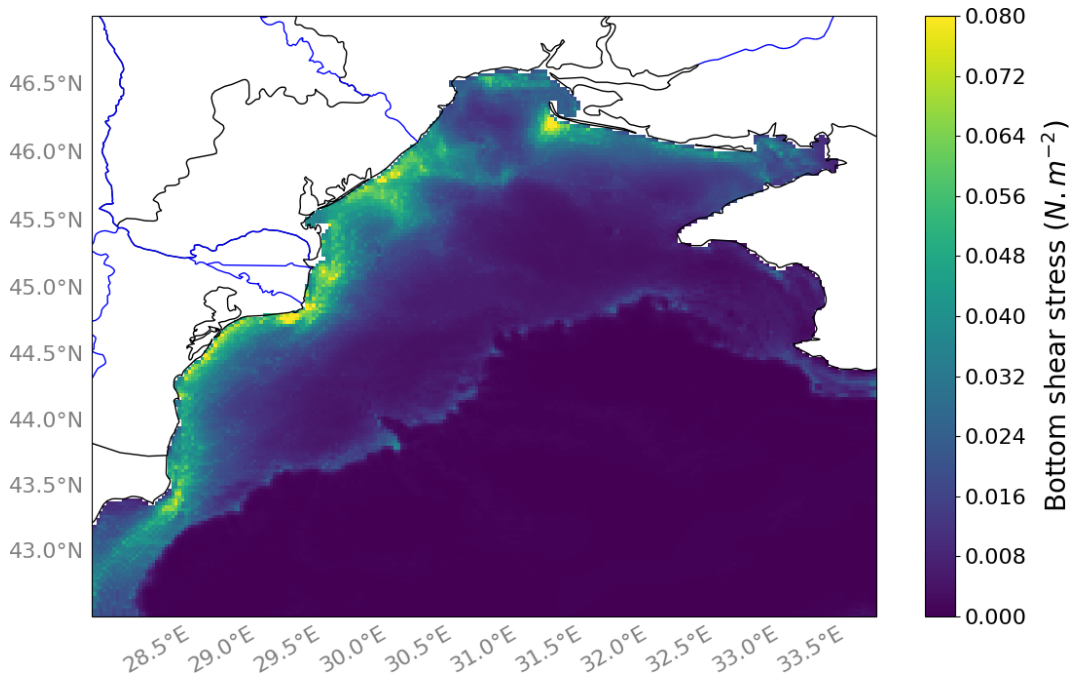


Figure 2.7: Spatial distribution of the **bottom shear stress** (in $N \times m^{-2}$, SHEAR), from the *BAMHBI* model over the northwestern shelf of the Black Sea. Annual mean from a climatology for the period 2008-2018. Colour bar at the right. Main rivers are in blue.

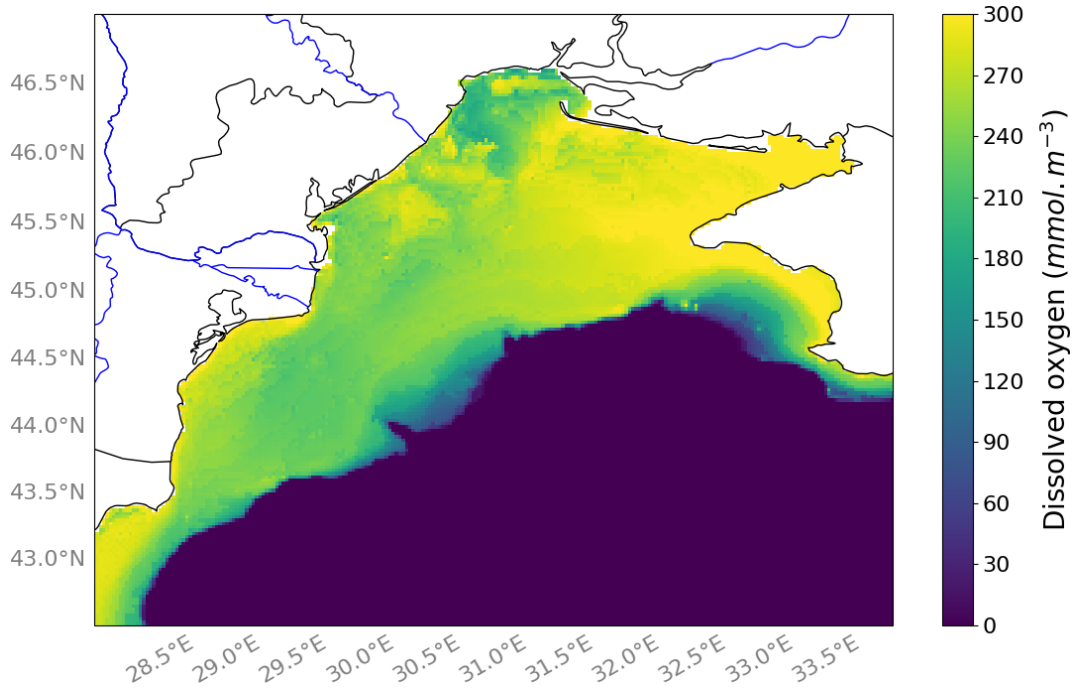


Figure 2.8: Spatial distribution of the bottom concentration of **dissolved oxygen** (in $mmol \times m^{-3}$, DOX), from the *BAMHBI* model over the northwestern shelf of the Black Sea. Annual mean from a climatology for the period 2008-2018. Colour bar at the right. Main rivers are in blue.

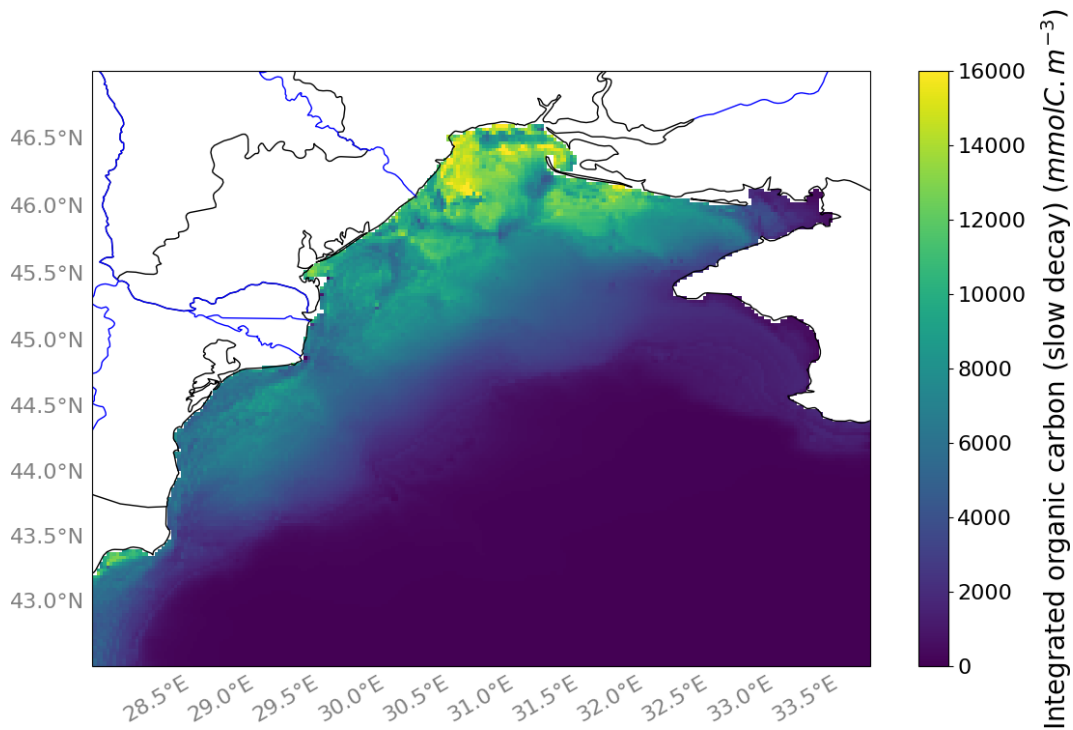


Figure 2.9: Spatial distribution of the **slow decaying stock of organic carbon** in sediment (in $mmol \times m^{-3}$, SSED), from the *BAMHBI* model over the northwestern shelf of the Black Sea. Annual mean from a climatology for the period 2008-2018. Colour bar at the right. Main rivers are in blue.

It is important to note that many abiotic factors co-vary, which may challenge the understanding of the mechanisms underlying observed relationships between species (and their traits) and environmental variables (e.g. Bolam et al. (2016)). In our data set, most of the abiotic descriptors vary along the depth-gradient, including oxygen concentration, temperature, and *POC*, complicating the understanding of the relationships between traits and abiotic predictors. The correlations between the selected abiotic predictors are given in **Figure 2.10**.

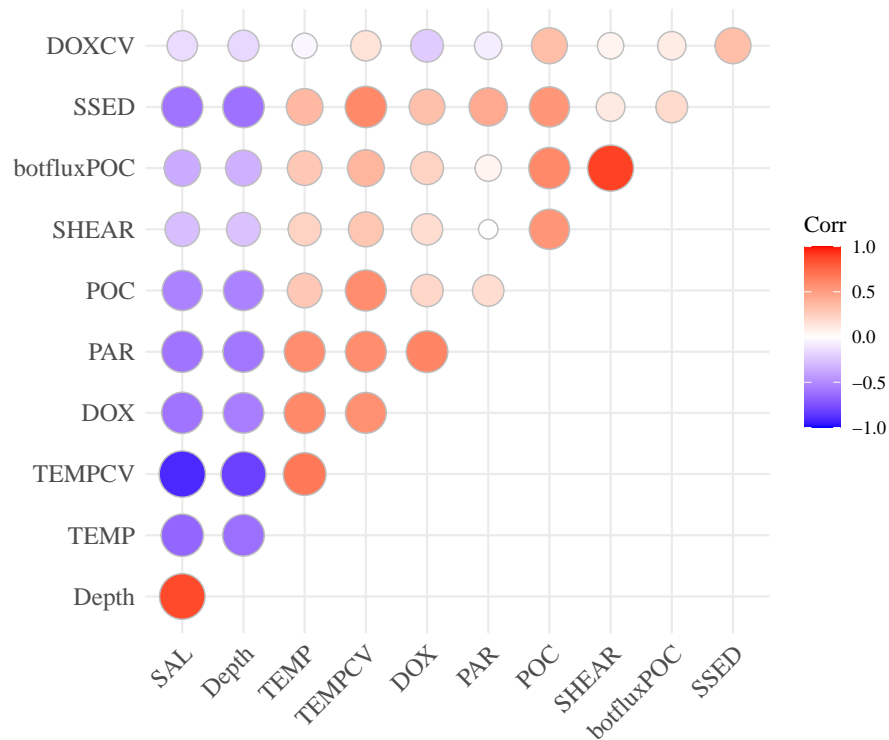


Figure 2.10: Kendall correlations between the set of abiotic predictors derived from the biogeochemical model *BAMHBI* coupled with *NEMO*. The list of abbreviation can be retrieved in **Table 4.2**. Negative correlations are in blue and positive in red and the size of the dots is proportional to the intensity of the correlation. The colour bar is at the right

In addition to this problem of collinearity, there is a mismatch in data collection scales with species data (grab samples) and abiotic data from a model collected on different spatial (metres versus several kilometres), making it difficult to link traits to environmental gradients and processes (Green et al., 2022). It is well known that the seabed can be highly variable at the scale of metres and this variability is currently not reflected in ocean models. For instance, the presence of hard substratum on the soft bottom can host a completely different benthic community from the one just next to the structure.

In addition to the difference in spatial scale, there is also a mismatch in temporal scale between the punctual observations of macrobenthos' communities and abiotic conditions. Seasonal variability was not explicitly considered, despite evidence that bioturbation and benthic community structure can vary strongly throughout the year (Queirós et al., 2015). For example, shallower burrowing depths during summer have been linked to bottom-water hypoxia, while seasonal hypoxic events may also reduce species densities through mass mortality of sessile organisms and avoidance behaviour in mobile taxa (Levin et al., 2009). Seasonal changes in organic carbon inputs can further influence bioturbation intensity, with reported differences of up to one order of magnitude in bioturbation diffusivity between seasons (W. Zhang & Wirtz, 2017). Nevertheless, previous studies suggest that spatial variation in the bioturbation potential of the community generally exceeds temporal variation, partially mitigating this limitation (Gogina et al., 2017, 2020).

2.3 Trait data

The selection of biological traits is important and is strongly dependent on the subject of interest (Beauchard et al., 2017). In this PhD work, we are interested in the response of species to environmental factors as well as the effect of species on the environment. For answering these two questions, two sets of traits were derived: the response traits and the effect traits (Lavorel & Garnier, 2002). The complete list of biological traits is provided in **Table 4.3**. These traits can be combined to derive ecological indicators that are useful for the evaluation of ecosystem health (Beauchard et al., 2017; Miatta et al., 2021; Villnäs et al., 2018). These multiple traits based indicators are a more practical and economical way to monitor the health of the benthic communities (Beauchard et al., 2017) and are complementary to traditional biodiversity and habitat-based indicators (Miatta et al., 2021). In this PhD work, we combined multiple response traits into a group of life-history strategies in a similar way than Beauchard et al. (2022) and Quell et al. (2021). We also combined multiple effect traits into proxies of ecological processes (e.g. biomixing and bioirrigation) in a similar way than Beauchard et al. (2023).

The use of biological traits itself has several inherent limitations and should therefore be applied with caution. Firstly, traits remain the same across the life span of each individual (Queirós et al., 2013). Traits defined at the species level do not account for intra-species variability or the ecological plasticity of some taxa, which can mask the true variability between organisms (de Juan et al., 2022; Miatta et al., 2021). Many benthic marine organisms are poorly documented, and their trait affinities are often inferred from related species within the same genus or higher taxonomic levels, potentially introducing bias into the analysis (Beauchard et al., 2017; Piló et al., 2016). Finally, it is challenging to compile homogeneous trait datasets, as different studies can use varying traits or modalities even for the same area, making it difficult to make comparisons between studies (Festjens et al., 2023).

2.3.1 Response Traits

In this PhD work, we chose our set of response traits to derive life-history strategies such as traits related to reproductive strategy or to the offsprings. Life-history theory seeks to explain intra- and interspecific variation in survival, growth, and reproductive traits of organisms as responses to environmental pressures (Pianka, 1970; Stearns, 1977). A central concept of life-history theory is that organisms face trade-offs arising from energetic, physiological, developmental, or genetic constraints and that these trade-offs shape the patterns observed in nature (Alonzo & Kindsvater, 2008). These traits are not independent; rather, they are interrelated through correlations that reflect underlying trade-offs, complicating the assessment of selective pressures acting on individual traits (Litchman & Klausmeier, 2008). A trade-off arises when a trait that improves performance in one function reduces performance in another, such as the classic trade-off between colonising ability and competitive ability (Verberk et al., 2013; Roff, 1993). For example, offspring size is negatively correlated with fecundity due to the constraints of energetic allocation (Brockelman, 1975). The use of response traits often leads to a limited number (generally three or four) of recurrent life strategies across taxa (Beauchard, 2023). The grouping of species into typological life-history strategies based on multiple traits has been proposed as a way to address the complexity arising from multiple interacting trade-offs (Statzner et al., 1997).

In this PhD thesis, we chose the most recent life history strategy framework developed by Kindsvater et al. (2016) and was built on life history continuum models from Winemiller and Rose (1992) and Grime (1977). The ***Precocial-Opportunist-Survivor-Episodic (POSE)*** concept connects four life history strategies to the ability of communities to withstand mortalities at juvenile and adult stages (**Figure 2.11**). The **Opportunist** and **Precocial** categories correspond to high adult mortality rates and respectively low and high juvenile survival. **Survivor** and **Episodic** strategies are two categories long-lived species but with low and high juvenile mortality, respectively (Kindsvater et al., 2016). Our set of biological traits allow us to determine both adult and juvenile mortality (**Table 4.3**) and to assign to each species one of the life-history group based on their response traits. Briefly, a single-table ordination (*Fuzzy Correspondence Analysis (FCA)*) was applied on the response traits table to reduce the dimensionality of the data (Chevenet et al., 1994). Hierarchical clustering was then applied on the scores extracted from the ordination to derive groups of species that share a similar combination of traits (see **Chapter 5** for more details).

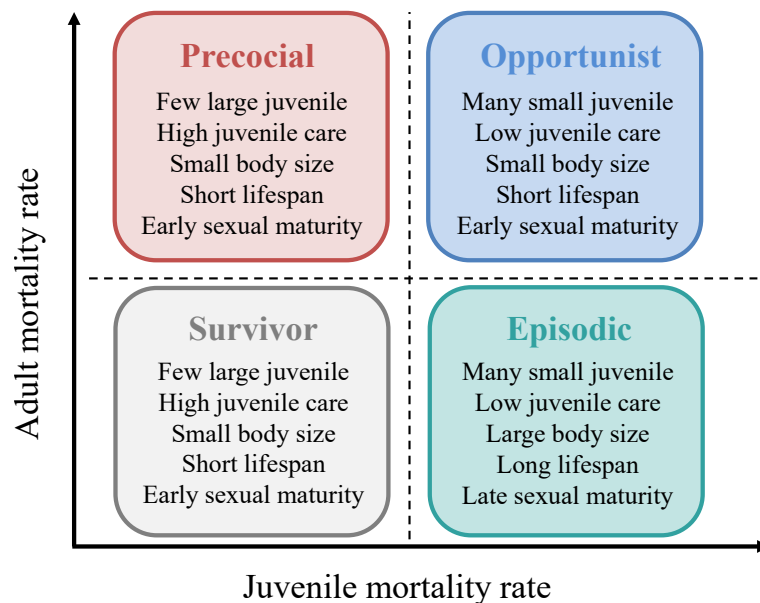


Figure 2.11: The *POSE* concept defining four life-history strategies along adult and juvenile mortality rates. The figure has been adapted from Kindsvater et al. (2016).

2.3.2 Effect traits

We also selected effect traits with an impact on the environment to estimate ecological processes. The contribution of a macrobenthos species to a given ecological process (e.g. biomixing, bio irrigation, biodeposition), supporting key ecosystem functions (e.g. *OM* remineralisation, nutrient recycling) (Bremner, 2008; Díaz & Cabido, 2001) is assessed by a combination of effect traits (i.e. with an effect on ecosystem; Lavorel and Garnier (2002)) weighted by its biomass (**Figure 1.2**). The list of multi-trait based indicators of ecological processes is provided in **Table 5.2**.

2.4 Multivariate statistics

Throughout this PhD study, we used multivariate statistics to analyse, interpret, and reduce the dimensionality of complex data sets where multiple outcomes are interdependent. To assess the response of traits to environmental gradients, we must analyse three tables simultaneously: L (species distribution across samples), R (environmental characteristics of samples), and Q (species traits) (**Figure 2.12**). The first step was to investigate the relationship between the species (L) and abiotic (R) tables before testing the relationships between the environment (R) and traits (Q) through species composition (L). This first step corresponds to (**Chapter 3**). Firstly, a **co-inertia** analysis was applied on the species and abiotic data (Dolédec & Chessel, 1994), a multivariate statistical method used to identify and measure the relationships (co-structures) between two datasets containing the same samples (Dray et al., 2003).

In addition, **between-group analysis** was applied to test if there are any differences in terms of abiotic and species composition between the two main periods of the dataset (respectively 1995 and 2008-2017 considered as two groups (Dolédec & Chessel, 1987, 1989)). The two techniques (between-group analysis and co-inertia analysis) can be combined together in a single procedure, Between-Group Co-inertia Analysis (BGCoiA) to link species and abiotic data and emphasise a temporal effect between the two periods (Franquet et al., 1995). Because there were no sampling sites along the Bulgarian coast in the mid-nineties, we removed the sampling sites from MSFD-Bulgarian campaign to compare both periods.

After the co-inertia analysis between R and L, the next logical step was to investigate relationships between the three tables: R (abiotic data), L (species data) and Q (traits data). In this PhD work, we use the statistical technique combining **RLQ** and **fourth-corner analysis** proposed in (Dray et al., 2014) to determine the environmental variables that are significantly related to a given trait (**Chapter 5**). This combination of RLQ and the fourth-corner method is now considered the most powerful method for *Biological Trait Approach (BTA)* (Beauchard et al., 2017). Both methods are based on the analysis of the fourth-corner matrix, which combines biological traits and environmental variables weighted by species abundance (Dray et al., 2014). RLQ analysis extends co-inertia analysis to ordinate three tables simultaneously (Dray et al., 2003) and aims to connect the species-traits table with the environmental table (Dolédec et al., 1996). The fourth-corner method assesses the bivariate associations between traits and abiotic factors, meaning one trait and one environmental variable at a time (Legendre et al., 1997). Each method has its own drawback: on one hand, RLQ analysis is only a global test that does not allow identifying which environmental variables is acting on which combination of trait and the interpretation of the graphical outputs can be difficult (Dray et al., 2014). On the other hand, 4th corner analysis involves a high number of tests, no consideration of the covariation among traits or among environmental variables and do not provide information about samples and species (Dray et al., 2014). Dray et al. (2014) have been integrate these two approaches into a unified framework to improve ecological data analysis and counterbalance these drawbacks. By testing directly the association between RLQ axes and traits/environmental variables improves the interpretation of RLQ and fourth-corner results (Dray et al., 2014).

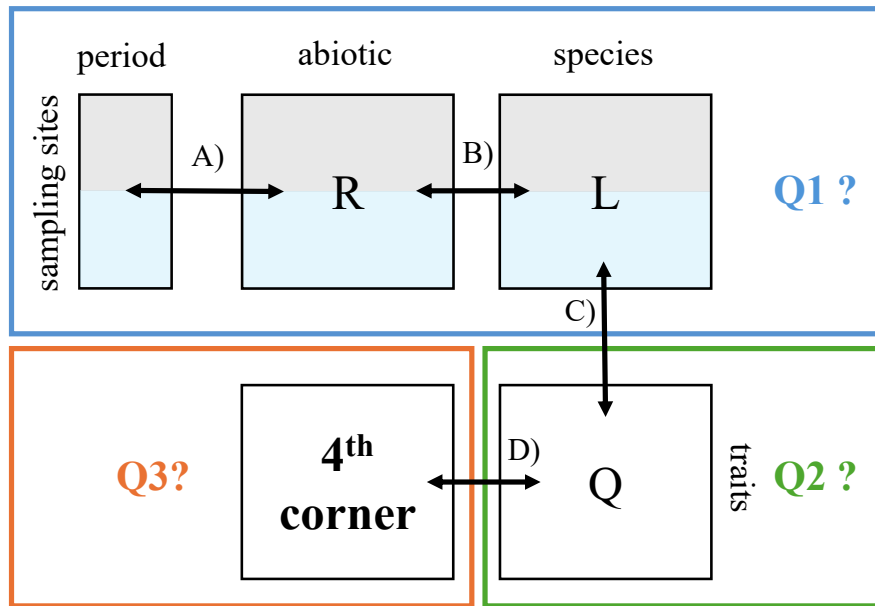


Figure 2.12: Schematic representation of the data set. Block arrows indicate how matrices (squares) match each other (i.e. either by the same number of rows or columns). **(A)** The sampling stations \times periods partition matches the R (stations \times abiotic descriptors) and L matrices (stations \times species) for the two main periods (i.e. 1990s in grey and 2010s in light blue). **(B)** R and L matrices can be combined to investigate the environmental determinants of taxonomic composition through co-inertia analysis in **Chapter 3** answering the question 1 of this PhD thesis. For 127 species, the compilation of biological traits was done in **Chapter 4** answering the question 2 of this PhD thesis. **(C)** The L and Q matrices are matched by their 127 species in common; by extension, this enables the combination of the RLQ and Fourth-corner methods to relate habitat descriptors and faunal functionalities **(D)** in **Chapter 5** answering the question 3 of this PhD thesis.

2.5 Trait Distribution Modelling

TDMs developed in **Chapter 5** are used to predict the spatial distribution of traits over the scale of the shelf from the punctual observations of traits at the community level and the maps of bottom abiotic conditions from an ocean model (**Figure 2.13**). The multivariate statistics analysis (i.e. RLQ combined with the fourth-corner analysis) help to select the best set of abiotic predictors for the list of traits significantly correlated with the environment. As a first step, we calculate the *Community Weighted-Mean (CWM)* traits per sampling site based on species abundances (i.e. number of individuals) and their trait values (**Figure 2.13A**). *CWM* is the most commonly used community-level trait metric and calculates the average trait value of a community where the trait value of each species is multiplied by its relative abundance (Ricotta & Moretti, 2011; Lavorel et al., 2008). The *CWM* trait originally comes from the ‘mass-ratio hypothesis’ for plants, which states that species with the highest biomass and the traits that they carry will have the strongest impact on ecosystem functioning (Grime, 1998). Then, we modelled the *CWMs* of each trait (response variable) with the set of environmental variables as predictors using *Artificial Neural Networks (ANNs)* (**Figure 2.13B**).

ANNs allowed us to generate projections of the distribution of *CWM* traits at the shelf scale in regions where information was missing based on maps of environmental conditions (**Figure 2.13C**).

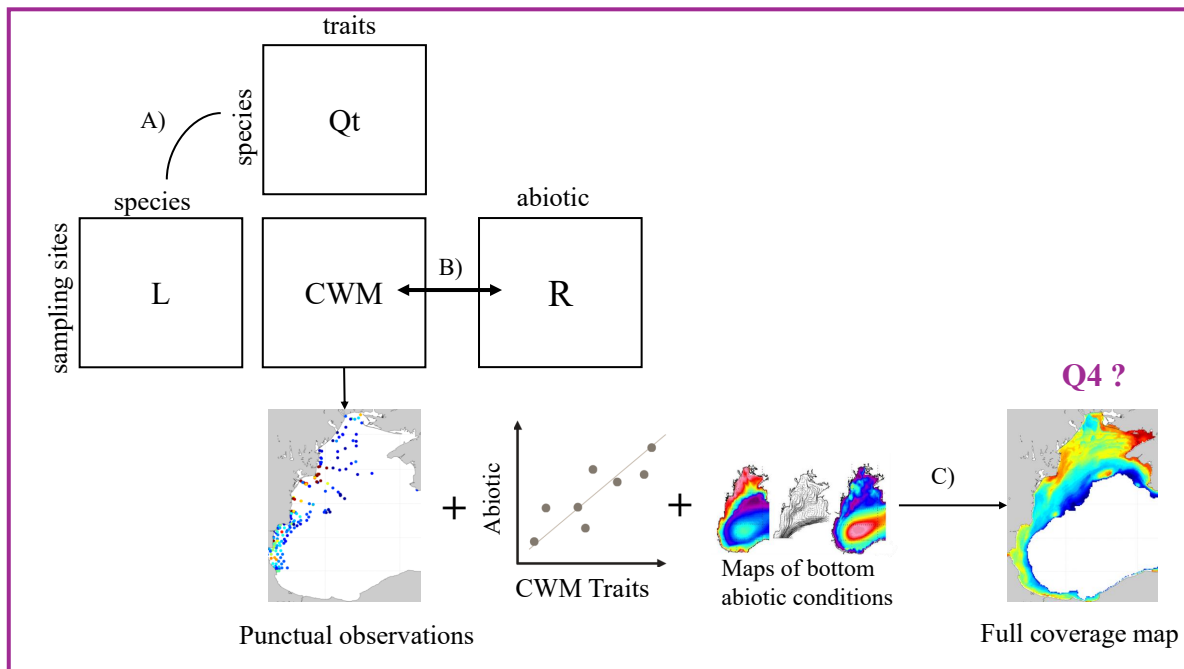


Figure 2.13: Overview of the methodological approach taken in **Chapter 5** for answering the question 4 of this PhD thesis. A) *CWM* trait values (matrix sites \times traits) are computed by crossing the table L (matrix sites \times species) with the transposed of the table Q (matrix species \times traits). B) *CWM* trait values across abiotic conditions were modelled using neural networks and C) the neural networks are used to project the *CWM* traits over the northwestern shelf based on the maps of bottom abiotic conditions.

However, the use of *CWM* has also some disadvantages because it ignores biological variations within communities as a result, overoptimistic correlations have been reported (Peres-Neto et al., 2017). Indeed, simulation analyses evidenced that correlations between *CWMs* and abiotic variable could lead to inflated Type I error (i.e. false positives) and low statistical power, because the *CWMs* are not truly independent from each other (Peres-Neto et al., 2017; J. E. Miller et al., 2019; ter Braak et al., 2018). On the contrary, multivariate methods (e.g. RLQ and fourth-corner analysis), as described previously, use all available information (traits, species abundances and abiotic variables) directly without any aggregation, thereby making use of the full variation within the data (Kleyer et al., 2012). Today, the most sophisticated and unbiased regression method has been developed according to multi-level modelling and enables all options of mixed modelling with linear as well as non-linear versions (ter Braak, 2019). Multi-level models maintain the entire structure of the data and jointly model all species abundances simultaneously in relation to environment, while incorporating species traits to control the effect of environment on species abundances (Jamil et al., 2013).

However, our large trait datasets can make these models computationally intensive and the use of *CWMs* is more suitable for large-scale and big data analyses, as in **Chapter 5**. *CWMs* offer the advantage of summarising underlying species dynamics and interactions, thereby reducing overall complexity (Funk et al., 2017). In this PhD work, we used firstly multivariate methods such as RLQ and fourth-corner which provide unbiased traits-environment relationships for interpretability and then we chose *ANNs* for predicting spatial patterns of *CWMs* at the shelf scale.

We chose *ANNs* as *TDMs* because it can help to deal with the non-linearities in the relationships between traits and abiotic data. In addition, *ANNs* are powerful tools for models, especially when the underlying data relationships are unknown (Lek & Guégan, 1999) because these models do not require any assumptions on the data used. However, this is a “black box” approach and therefore, it is difficult to interpret ecological relationships (M. C. Jones & Cheung, 2015) and it is relevant to couple this approach with multivariate statistics to better understand the underlying relationships between traits and environmental conditions. neural network is a method of machine learning inspired by the structure and functions of biological neural networks (Guisan et al., 2017). *ANNs* are made up of interconnected units of a large number of neurones. Each neuron receives input signals, processes them through weighted connections and an activation function, and generates an output signal that is transmitted to subsequent neurons in the network (Z. Zhang, 2018). In this work, a **feedforward** neural network architecture was employed, which means that information flows unidirectionally from the input layer through the hidden layers to the output layer, without feedback connections. The network consists of an input layer that contains source nodes that receive input features, one hidden layer responsible for learning intermediate representations, and an output layer that produces the final network response (**Figure 2.14**).

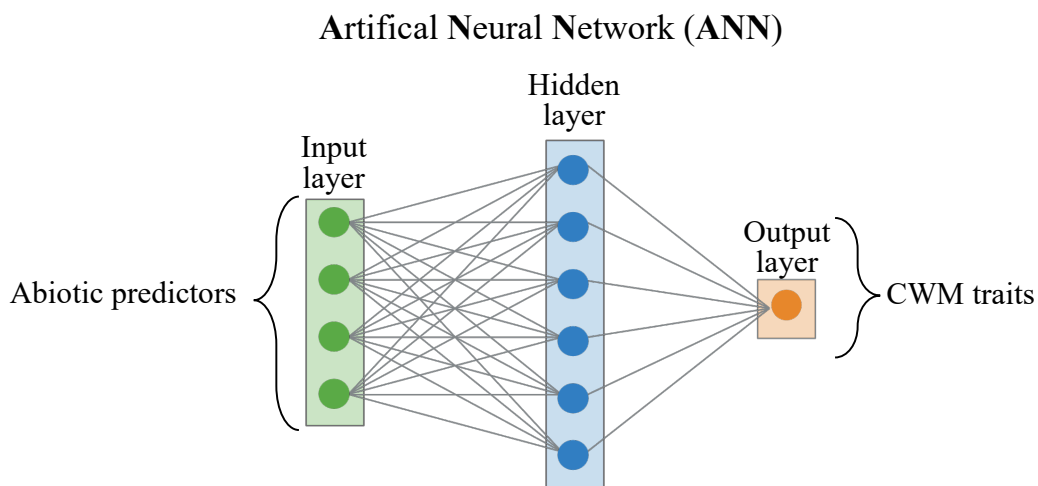


Figure 2.14: Schematic representation of an *ANNs* with an input (in green), hidden (in blue) and output (in orange) layers. The input layer corresponds to the abiotic predictors and the output layer to the response variable (here *CWM* Traits). Each dot corresponds to a neuron.

LINKING SPECIES AND ABIOTIC CONDITIONS IN THE PAST AND PRESENT

Foreword

This chapter explores the relationships between the abiotic and species data. We are also interested in the changes between the 1990s considered as the post-eutrophication period with bad status and the most recent period (2008-2017) to test if there are any changes in species composition related to environmental changes. A graphical summary of this chapter is provided in **Figure 3.1**. Supplementary materials associated with **Chapter 3** are available in **Appendix B**.

This Chapter has been published as : Chevalier, S., Beauchard, B., Teacă, A., Soetaert, K., & Grégoire, M. (2024). Partial recovery of macrozoobenthos on the northwestern shelf of the Black Sea. *Marine Pollution Bulletin*, 207, 116857. <https://doi.org/10.1016/j.marpolbul.2024.116857>.

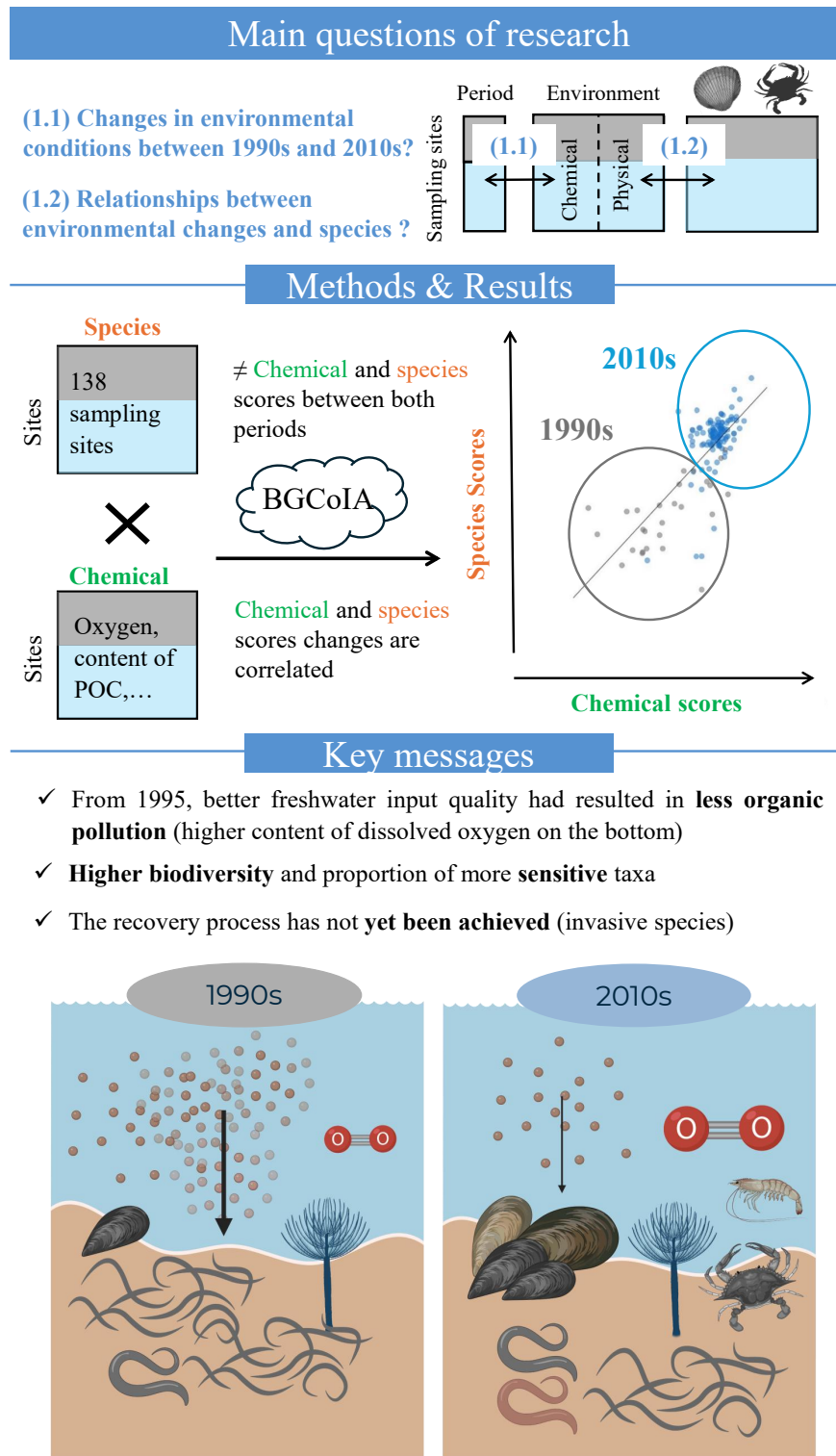


Figure 3.1: Graphical Abstract of the **Paper I** (Chevalier et al., 2024): **Chapter 3** aims to answer the first question of this thesis by linking species and abiotic data through a coinertia analysis (CoIA). The difference between the period 1990s and 2010s is investigated using a Between Group Analysis (BG). BGC*o*IA corresponds to a combination of these two approaches.

3.1 Abstract

The northwestern shelf of the Black Sea has been affected by eutrophication and bottom hypoxia since the sixties. Consequently, the macrozoobenthos has suffered a well-established decline in biodiversity. However, the nature of the current benthic communities remains questionable. From 1995 to 2017, we compiled species and abiotic data for 138 sites over the shelf. Through an appropriate multivariate analytical approach, we identified benthic community changes solely due to organic pollution variations. Our results show signs of recovery with an increase in biodiversity and proportion of species vulnerable to organic enrichment. These changes were related to a decrease in riverine loads and subsequent eutrophication. However, some long-lived species typical of the area still did not exhibit noticeable recovery, which suggests that either the recovery process has not yet been achieved or some environmental conditions are still not met to warrant a sea floor ecosystem state substantially healthy.

3.2 Introduction

3.2.1 The Black Sea and eutrophication

Macrozoobenthos play critical roles in the sea floor ecosystem functioning through pelagic-benthic coupling and bioturbation (Snelgrove, 1998). Given their relatively sedentary nature and variable life span, macrozoobenthic species can integrate various water-sediment quality conditions and have long been recognised as valuable indicators of ecosystem health (Dauer, 1993). Main habitat drivers in the marine benthos are either physical (Snelgrove & Butman, 1994) or chemical (Pearson & Rosenberg, 1978). In absence of pollution, physical aspects are primary drivers in many benthic habitats (J. Y. Aller, 1989; Hall, 1994; Snelgrove & Butman, 1994). However, due to recurrent problems related to freshwater inputs into coastal areas appearing in the second half of the twentieth century, organic and xenobiotic disturbances on the benthos was a major focus in early works of marine benthic ecology (Pearson & Rosenberg, 1978; Rhoads & Morse, 1971; Rosenberg, 1977).

In the Black Sea, depth influences the vertical gradients of chemical (e.g. dissolved oxygen, and trophic resources) and physical factors (e.g. temperature and salinity) and therefore determines the distribution of macrozoobenthic communities (Zaitsev, 2008; Todorova & Konsulova, 2005). Bottom sediment type is one the main factors that drive the composition of benthic communities (Stevens et al., 2019; Teacă et al., 2020; Zaitsev, 2008; Surugiu, 2011) as well as oxygen availability (Friedrich et al., 2014) and nutrient discharges from rivers (Teacă et al., 2020; Wijsman et al., 1999). Until the mid-sixties, macrozoobenthic communities were in a pristine state, relatively undisturbed by human factors. More than twenty biological associations were described, including large beds of mussels (*Mytilus galloprovincialis*) providing habitats for a diversity of species (Băcescu et al., 1971).

In the late 1960s to early 1970s, increasing human activities (e.g. agricultural and industrial development) led to higher nitrates and phosphates loads from rivers (Zaitsev, 1992; Gomoiu, 1992; Mee et al., 2005). Consequently, high levels of nutrients in coastal waters caused eutrophic conditions and its subsequent hypoxia (i.e. shortage of dissolved oxygen with respect to the requirements of aerobic respiring species (Breitburg et al., 2018)). On the northwestern shelf of the Black Sea, bottom seasonal hypoxia develops during the summer-fall period when the seasonal stratification inhibits ventilation of bottom waters where large consumption of oxygen occurs through respiration and oxidation of organic matter (Capet et al., 2013; Wijsman et al., 1999; Friedrich et al., 2002). The greater impact was observed near the Danube Delta (Garnier et al., 2002; Zaitsev, 1992) and at depth down to 45 m in the worst case (Capet et al., 2013). Bottom hypoxia led to a 2.9-fold reduction in macrozoobenthos abundance and biodiversity (Alexandrov & Zaitsev, 1998; Zaitsev, 1992). The number of epibenthic species decreased (e.g. crustaceans, gastropods, some bivalves), contrary to the infaunal tube dwelling worms, which flourished under strong disturbed conditions (e.g. *Polydora cornuta* and *Lagis neapolitana*; Losovskaya, 1978 cited in Surugiu (2005b)). This situation was worsened by human pressures (e.g. overfishing, heavy metals pollution, ship traffic) and invasive species (e.g. the rapa whelk *Rapana venosa*). The decline of the macrobenthos at the end of the 1980s and 1990s is well reported in the literature (e.g. Gomoiu (1992, 1997, 1999); Zaitsev (1992); Zaitsev and Mamaev (1997)).

In the early nineties, the collapse of the Soviet Union and the introduction of conservation programs led to a decrease in nutrient loads by a factor of 2 to 4 (Mee et al., 2005). Pelagic waters showed rapid signs of recovery for phytoplankton communities with a decrease in the intensity of algal blooms (Konsulov et al., 1998). However, the fate of the macrozoobenthic community is less evident as it takes longer time to recover due to the persistence of nutrients in the sediments even a decade after a eutrophication event (Capet et al., 2013). Since the beginning of the 2000s, several studies have been conducted and has evidenced some signs of partial recovery of the macrozoobenthos. For instance, positive signs include higher biodiversity (Friedrich et al., 2014; Stevens et al., 2019; Revkov et al., 2018), an increased abundance ratio of slow growing species to opportunistic ones (Begun et al., 2010; Revkov et al., 2018; Teacă et al., 2006, 2019a, 2020) and the reappearance of species abundant in the period of ecological stability (Mee et al., 2005; Teacă et al., 2019a). However, most of these studies are spatially and temporally localised and/or restricted to certain taxa with the polychaete being the most intensively studied group (Surugiu, 2005b, 2005a; Surugiu et al., 2010; Surugiu, 2011; Boltachova et al., 2021; Surugiu & Feunteun, 2008; Şahin & Çınar, 2012). In addition, only a few studies investigated the link between environmental factors and macrofaunal composition with a limited set of mixed physical and chemical descriptors (Friedrich et al., 2014; Stevens et al., 2019; Teacă et al., 2019a, 2020; Wijsman et al., 1999).

3.2.2 Aims of the study

In this paper, we investigate how the modifications of environmental conditions during the last two decades have affected the structure of macrozoobenthic communities over the entire northwestern shelf of the Black Sea by considering the historical sources of organic input. We assess species compositional changes due to the chemical factors the most related to organic disturbance (e.g. dissolved oxygen, content of organic carbon), solely and independently on natural physical forces (e.g. temperature, salinity). To this aim, we used an appropriate multivariate approach to mask the effects of undesired forces and to counteract experimental constraints (e.g. number of samples, spatial coverage). We hypothesised that a reduction in organic disturbance had led to a recovery of the less organic-tolerant species over the last decades and unravelled the specific effects of organic disturbance on benthic community structure.

3.3 Materials and Methods

3.3.1 Study area

The Black Sea is a large semi-enclosed basin with restricted exchanges with the Marmara and Mediterranean Seas through the Bosphorus strait (**Figure 3.2**). Due to its landlocked character, the Black Sea is very sensitive to organic pollution. Large riverine inputs of fresh water in surface and the intrusion of saline Mediterranean waters into the bottom layer lead to a strong stratification at 150-200 m depth. This permanent pycnocline limits the exchanges between the well-oxygenated surface layer and the depleted-oxygen deep water. Therefore, most of the aerobic respiring benthic life is concentrated at depth down to 150 m on a wide northwestern shelf connected to the deep sea (**Figure 3.2**). The shelf is very shallow with an average depth of 50 m, and it exhibits a strong seasonality directly influenced by atmospheric fluxes. The northwestern shelf is characterised by low salinity and high nutrients concentration caused by run-off from the main rivers (**Figure 3.2**). The Danube, which accounts for 75% of freshwater inputs per year, strongly influences the spatial and temporal distribution of nutrients and sediment grain-size over the shelf. The direction and extension of the Danube's plume are highly variable and depend on water discharge intensity and the direction of the wind. From fall until early spring, the plume moves southwards in response to northly wind while when the Danube's discharge intensifies in spring and the northly wind weaken, the plume moves northwards and forms an anticyclonic gyre (Grégoire & Friedrich, 2004). The anticyclonic gyre keeps water rich in nutrients and suspended particles close to the shore in the north-east sector of the shelf (Friedrich et al., 2002; Wijsman et al., 1999). The shoreline south of Constanta receive a scarce input of sediments from the Danube and the sediments are mainly of organic origin, derived from mollusc shells (Panin & Jipa, 2002).

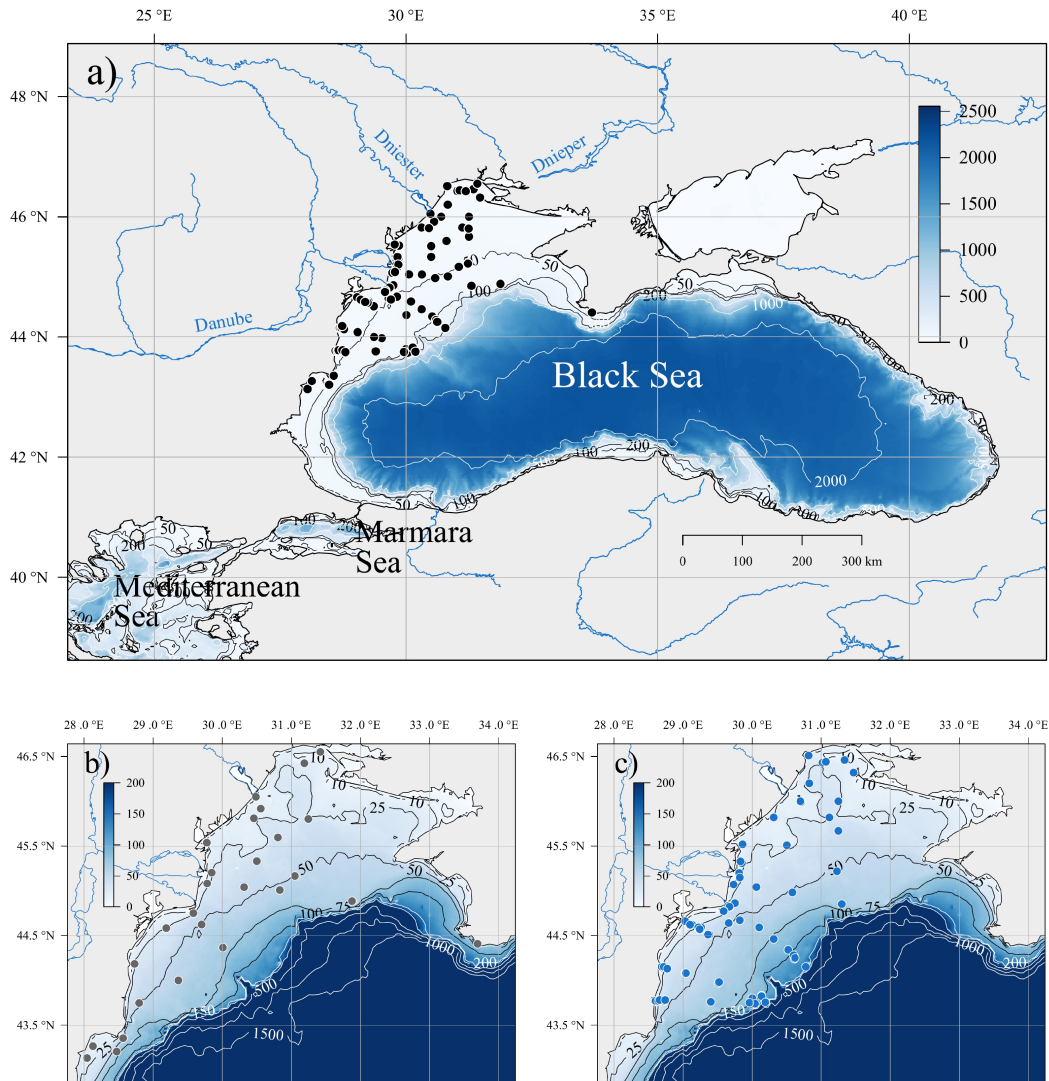


Figure 3.2: a) The Black Sea in the regional context of the study with the most rivers as the main sources of freshwater inputs; black dots represent the stations of both sampling periods 1995 and 2008-2017; b-c) northwestern shelf up to 200 m: grey dots (b) represent 1995 sampling stations and blue dots (c) are 2008-2017 sampling stations.

3.3.2 Faunal data

3.3.2.1 Sampling survey and design

We compiled a macrozoobenthos dataset that covers the whole northwestern shelf of the Black Sea (**Figure 3.2**). A total of 138 sites were sampled over 4 different campaigns between 1995 and 2017: EROS-2000 (1995), SESAME (2008), HYPOX (2010 and 2011) and EMBLAS (2016-2017). A total of 163 taxa were collected. We divided the investigated period into two parts: a first period (1995) just after the reduction of eutrophication in the early 1990s and a second one more than ten years later (2008-2017). Over 1995-2017, the sampling design shows some limitations. Firstly, the 1995 sampling effort was spatially (i.e. only 27 sites) and temporally (i.e. only in August) lower than during the 2008-2017 period. Nevertheless, the distribution of samples in 1995 covers the whole shelf (**Figure 3.2**) while a large part of the data in 2008-2017 was collected at the end of the bloom and during the stratification period between the end of spring and early fall (May-September). More details about the sampling design are provided in the Supplementary Materials of this paper. During the EROS-2000 campaign (1995), the macrofauna was sampled with a Reineck box corer (60 x 30 x 30 cm) and washed through a 1.0 mm mesh sieve. More details on the sampling protocol for the first period are provided in Gomoiu (1997); Wijsman et al. (1999). Since 2005, the Black Sea Commission has recommended the use of the standardised macrozoobenthos protocol described in Todorova and Konsulova (2005). This methodology was used for all samples collected between 2008 and 2017. Briefly, macrofauna was sampled with a Van Veen grab with a surface of 0.135 m² and washed through a 0.5 mm mesh sieve.

3.3.2.2 Identification and World Register of Marine Species (WoRMS) corrections

There are uncertainties in species identification well-known for the Black Sea (Radashevsky & Selifonova, 2013; Surugiu, 2005a, 2012; Teacă & Gomoiu, 2007). We performed taxonomic adjustments following the literature and expert knowledge before comparing macrofaunal samples between both periods. Then, the compiled dataset was checked to adopt a common nomenclature of species names aligned with the World Register of Marine Species (<https://www.marinespecies.org/>). When possible, all the taxa were identified at the species level. When an identification at species level was not possible due to damage or unsolved taxonomic problems, the lowest reliable taxonomic level was given. For each taxon, the species level was kept when it accounted for more than 90% of all records while the remaining records were discarded; if not, the genus level was considered. Taxonomic corrections are available in Supplementary Materials.

3.3.2.3 The constraint of heterogeneous sampling designs and need for presence-absence

The difference in the sampling protocol between EROS-2000 and the other campaigns may have affected the comparison of the data. Indeed, the smaller mesh sieve size may have influenced the reliability of density and biomass estimates in the macrofaunal sampling from coastal waters (Bachelet, 1990; Schlacher & Wooldridge, 1996).

However, the influence on diversity has been found to be much reduced or no significant. For example, Bishop and Hartley (1986) and Thompson et al. (2003) evidenced that few additional species were retained through a 0.5 mm sieve compared to 1.0 mm in the coastal area. The difference in sampling gear may also affect the results, as the Van Veen grab may underestimate the biomass of deep-burrowing taxa in sandy sediments compared to the box corer with higher penetration depth (Beukema, 1974). In muddy sediments or very coarse sand, Van Veen grab depth may be increased to 10 cm and gives satisfactory estimation in fauna biomass (Beukema, 1974). Indeed, previous studies found that no significant differences between fauna sampled on silty sediments with a Van Veen grab and box corer (e.g. Heip et al. (1985) cited in Wijsman et al. (2022), Souza and Barros (2015)). In the Black Sea, Van Veen grab is considered as an efficient sampler for the range of soft sediments encountered (Todorova & Konsulova, 2005), mostly biogenic with shells debris or sandy mud (Stevens et al., 2019; Teacă et al., 2020; Vasquez et al., 2021). In this paper, to limit the effect of different sampling protocols, the analysis was done on the species presence/absence. In addition, we compared the list of meiobenthos species from EROS-2000 (i.e. smaller than 1 mm) with that from the other campaigns. Only one species (*Phoronis psammophila*) was identified as meiofauna in EROS-2000 and was added to the list of macrofaunal species in 1995.

3.3.2.4 Taxonomic richness pattern

In **Appendix B**, we provide an account of taxonomic richness per each sampling station (i.e. the number of taxa at one sampling station) with a spatial description of the study area (**Figure B.1**). In an informative way, we assess the mean specific richness per period in **Table B.1**.

3.3.3 Abiotic descriptors

3.3.3.1 In-situ measurements

For each station, *in-situ* data include water depth and information on sediment type. The type of sediment was estimated from either direct observation, sediments characteristics analysis or derived from map of seabed substrate. In more details, for EROS-2000 campaign, sediment characteristics (e.g. medium grain size, % of shells) are provided in Wijsman et al. (1999), for HYPOX and SESAME campaigns, sediment types (e.g. sandy mud, coarse mixed sediments) were recorded after sampling following the nomenclature from Todorova and Konsulova (2005) and for EMBLAS campaign, sediment type was estimated based on map produced by EMODnet Seabed Habitats (more details are provided online: <https://emblasproject.org/>). Given the different sources of sediment description between campaigns, we chose to classify sediment type based on “EUNIS-like” classes applicable in the Black Sea (Vasquez et al., 2021; Populus et al., 2017). Hence, we divided sediment types into three categories: “Sand to muddy Sand”, “Mud to sandy Mud” and “Mixed to Coarse sediments”. The latter class is justified in the Black Sea as the origin of sand is mostly biogenic, formed by small fragments of shells in a mud matrix (Teacă et al., 2020).

Indeed, biogenic materials are made mainly from shells of formerly abundant mussels *M. galloprovincialis*, between 30-50 m, and *Modiolula phaseolina* from 55-60 m and extending to 120 m (Băcescu et al., 1971; Gomoiu & Tiganus, 1977; Begun et al., 2010).

3.3.3.2 Biogeochemical and physical data

We used environmental data from a model reanalysis delivered by the *Copernicus Marine Environment Monitoring Service (CMEMS)* Black Sea Marine Forecasting centre (Ciliberti et al., 2021). These data were produced by the hydrodynamical-biogeochemical coupling of *Nucleus for European Modelling of the Ocean (NEMO)* 4.2 with *Biogeochemical Model for Hypoxic and Benthic Influenced areas (BAMHBI)*, run in reanalysis mode at a resolution of 2.5 km with 59 vertical levels. *BAMHBI* simulates oxygen, nitrogen, phosphorus, silicate and carbon cycling mediated by several plankton functional types and the microbial loop. It explicitly represents processes in anoxic conditions and the benthic remineralisation and resuspension processes (see details in Grégoire et al. (2008); Grégoire and Soetaert (2010); Capet et al. (2016)). Details on the implementation of the Black Sea Marine Forecasting Center can be found in Ciliberti et al. (2021). Abiotic conditions were extracted at each sampling station as quarterly mean centred around the date of sampling. The simulated bottom values of temperature (TEMP), salinity (SAL), photosynthetic active radiation (PAR), shear stress (SHEAR), dissolved oxygen concentration (DOX), particulate organic carbon content (POC), the fast and slowly degrading sedimentary carbon (respectively FSED and SSED), were selected as potential variables that can explain changes in species composition. Coefficient of variation (i.e. the ratio between the quarterly standard deviation and the corresponding mean value) was evaluated for temperature and oxygen concentration to assess thermal and oxygen seasonality. A summary with simulated environmental variables is provided in **Table 3.1**. In this study, we ignored the effect of anthropogenic stressors such as bottom trawling and dredging. Despite the negative effects of these pressures on faunal abundance, natural gradients in species composition can persist in case of sufficiently strong natural abiotic forces (Beauchard et al., 2022).

Table 3.1: List of abiotic descriptors derived from the coupled hydrodynamical-biogeochemical *NEMO 4.2 – BAMHBI* model with their respective abbreviation and unit. Abiotic variables are classified into two categories: physical and chemical variables.

Abiotic descriptors	Abbreviation	Unit
Physical variables		
Temperature	TEMP	°C
Coefficient of variation for the temperature	Thermal seasonality	-
Salinity	SAL	<i>practical salinity unit (p.s.u)</i>
Bottom shear stress	SHEAR	$N m^{-2}$
Chemical variables		
Photosynthetically active radiation	PAR	$W m^{-2}$
Particulate organic carbon content	POC	$mmol C m^{-3}$
Vertically integrated organic carbon concentration (fast decay)	FSED	$mmol C m^{-3}$
Vertically integrated organic carbon concentration (slow decay)	SSED	$mmol C m^{-2}$
Dissolved oxygen concentration	DOX	$\mu mol l^{-1}$
Coefficient of variation for dissolved oxygen concentration	DOX seasonality	-

3.3.3.3 Analysis of nutrient loads

We analyse time series of riverine loads derived from data in Ludwig et al. (2009) to detect potential changes in nutrient discharges (nitrates, phosphates, ammonium, particulate/dissolved organic carbon and nitrogen). These data are those used to force the biogeochemical model. We identified periods with distinct averages over time in the framework of the regime shift analysis described in Zeileis et al. (2003). Firstly, we tested the presence of at least one significant change point (i.e. a 2-segment model with different average) based on a *F*-Statistic tests Zeileis et al. (2003) and references therein. Then, the locations of the most likely change points were identified considering from one to five points. Finally, *Bayesian Information Criterion (BIC)* was used to choose until which point the addition of a new change point in the model was still relevant (Raftery, 1986a, 1986b). The model with the lowest *BIC* criterion was selected. More details on the optimal number of breakpoints and their corresponding date are provided in **Appendix B**.

3.3.3.4 Spatial and temporal pattern of low-oxygen conditions

We derived the number of days of low bottom oxygen conditions in one year from a 10-year climatology centred around the two periods of the data collection (i.e. 1990-2000 and 2008-2018). In this paper, we chose a threshold value at $120 \mu mol O_2/l$ below which negative effects on biodiversity appear (Steckbauer et al., 2011; Vaquer-Sunyer & Duarte, 2008).

Sampling stations with low-oxygen conditions were mapped and the number of days with low-oxygen conditions in one year per sampling station are provided in Supplementary Materials.

3.3.4 Linking organic pollution and species compositional changes

The analysed dataset was composed of a sampling stations \times abiotic descriptors matrix that matches a sampling stations \times taxa. The two matrices were equally partitioned in the two periods, 1995 and 2008-2017. Before statistical analysis, to avoid non-linear relationships between species assemblages and environmental descriptors, each quantitative environmental descriptors were firstly converted into qualitative ones as factor with multiple levels (Beauchard et al., 2022). For example, temperature, expressed in $^{\circ}\text{C}$, was converted into a factor with 5 levels from very cold ($< 8^{\circ}\text{C}$), cold ($8-10^{\circ}\text{C}$), intermediate ($10-12^{\circ}\text{C}$), warm ($12-14^{\circ}\text{C}$) to very warm ($> 14^{\circ}\text{C}$). Details about the new qualitative environmental descriptors are provided in **Appendix B**.

As not all abiotic descriptors were expected to change over time, especially the physical ones (i.e. temperature, salinity, depth, shear stress, and sediment type), everything else related to oxygen was assumed to do. We first verified this with *Variance Partitioning Analysis (VPA)* (Legendre et al., 2012), and then, to address the effect of organic disturbance on assemblage composition of the assemblage independently of the persisting physical effects, we detrended the chemical descriptors from the effects of the physical ones. Hence, through detrending, observed compositional changes were necessarily and solely due to chemical aspects, with physical aspects everywhere being equal. Then, the chemical descriptors matrix was processed by *Multiple Correspondence Analysis (MCA)* (Tenenhaus & Young, 1985), adapted to categorical data, and fauna by *Non-Symmetric Correspondence Analysis (NSCA)* (Gimaret-Carpentier et al., 1998). *NSCA* attributes a unique weight to species so that sampling stations be separated only due to compositional changes rather than number of occurrences, which also minimises the influence of rare species. The two ordinations were detrended from the effects of the physical descriptors matrix through *Orthogonal Principal Component Analysis on Instrumental Variables (OPCAIV)* (Rao, 1964; Thioulouse, 2011).

Then, the two detrended matrices were combined in a single procedure to emphasise the period effect through *Between-Group Co-Inertia Analysis (BGCoIA)* (technically described in Franquet et al. (1995) and examples are provided in Thioulouse (2011) and Thioulouse et al. (2018)). *BGCoIA* builds multivariate axes that maximise the covariances between chemical descriptors and taxa (*Co-Inertia Analysis (CoIA)*, Dolédec and Chessel (1994); Dray et al. (2003)), chemical and faunal matrices being preliminary processed by *Between-Group Analysis (BGA)* considering each period (1995 and 2008-2017) as a group (Dolédec & Chessel, 1987, 1989). Through *BGCoIA*, *BGA* ensures that chemical and faunal variables that covary the most along *CoIA* axes are those that respond the most to organic disturbance (i.e. between-period effect). As the effect of organic disturbance includes only two periods, *BGCoIA* returns only a single axis defined by the two period-centroids within the multivariate space and that accounts for the overall variance in period change.

Hence, a single axis was considered for interpretation by positioning scores of habitats and faunal variables, and sampling station scores from the two matrices passively projected onto the *BGCoIA* axis.

The covariation between chemical conditions and community composition was assessed by Pearson's correlation between the sites scores from the two matrices. The effect was also tested for each series of scores by Wilcoxon's test to verify that both chemistry and fauna were affected by a temporal effect. Important chemical descriptors were identified by their Pearson's correlations with the *BGCoIA* axis, after *p*-value correction for multiple testing through the false discovery rate (Benjamini & Hochberg, 1995).

Finally, we assessed the functional response of the fauna to organic disturbance by characterising species according to their saprobic affinity. We used *AZTI Marine Biotic Index (AMBI)* classification (Borja et al. (2000); available online: <https://ambi.azti.es>); when information was not available at the species level, we used a higher taxonomic level. The concept is similar to earlier freshwater developments (Sládeček, 1973) and applied to stream invertebrates whereby species saprobity is related to biological oxygen demand (Metcalf, 1989). The taxa were classified into five groups of growing tolerance to organic enrichment: group I, very sensitive species to organic enrichment only present under unpolluted conditions; group II, species indifferent to organic enrichment; group III, tolerant species to an excess in organic matter; group IV, second-order opportunistic species from slight to pronounced unbalanced conditions; group V, first-order opportunistic species tolerating strongly unbalanced conditions (Hily, 1984; Glémarec, 1986; Grall & Glémarec, 1997). In **Appendix B**, we compared the proportion of saprobic groups between 1995 and 2008-2017 (**Figure B.3**). Relationships between saprobic groups and the *BGCoIA* axis were tested with the Fourth-corner method (Dray & Legendre, 2008; Dray et al., 2014). The groups were considered as a qualitative trait (5 group-modalities) matching the *BGCoIA* axis by species positions and correlated to sample scores through the samples \times species matrix (i.e. in accordance with the RLQ approach; (Dray & Legendre, 2008)). The test was applied at two levels: an overall test of group distribution (global *F*-statistic) and separate tests per binomial group-variable (Pearson's *r*-correlation and within-group homogeneity); the tests were run with 49 999 random permutations of model 6 from Dray and Legendre (2008). A graphical summary of the statistical procedure is provided in (**Figure 3.3**).

All analyses were done with R 4.3.3; we used the packages “vegan” for variation partitioning (Oksanen et al., 2013), “ade4” for community ordination (Chessel et al., 2004; Dray & Dufour, 2007), “strucchange” for river time series analysis (Zeileis et al., 2002) and ggplot2 for graphs (Wickham, 2016).

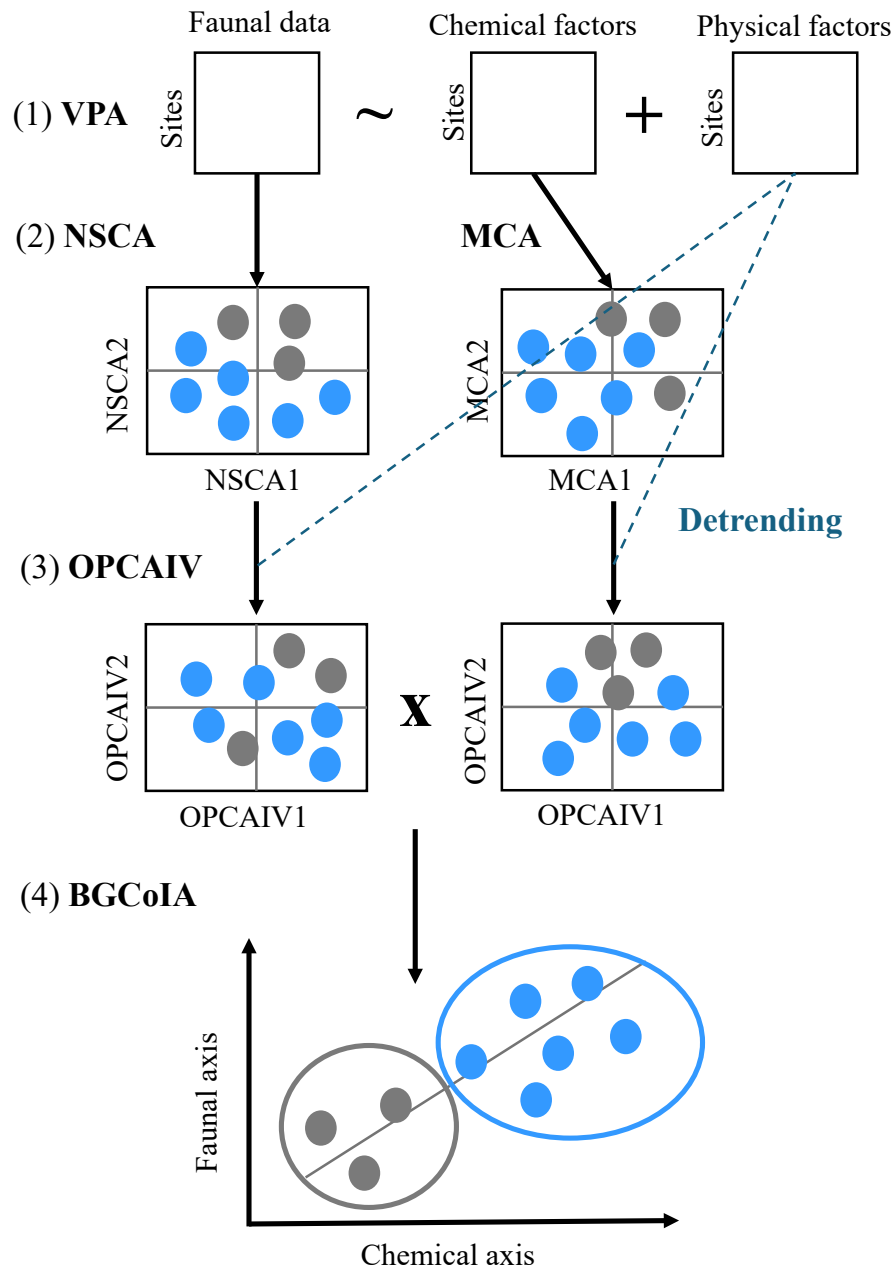


Figure 3.3: Graphical summary of the main steps of the statistical procedure: 1) separate physical and chemical factors effect on faunal assemblages through *VPA*, 2) Separate ordination with respectively *NSCA* applied on faunal data and *MCA* applied on chemical factors. Each dot represents one sampling site projected on faunal and chemical hyperspaces (here for simplicity only the two first axes are displayed), grey dots for 1995 and blue dots for 2008-2017, c) detrending part of the variance of the faunal and chemical ordination explained by physics through *OPCAIV*, d) determine changes in species-environment relationships by analysing series of pairs of tables (here 1995 and 2008-2017) through *BGCoIA*.

3.4 Results

3.4.1 Abiotic descriptors

3.4.1.1 Analysis of nutrient loads

Regime shift analysis of river loads revealed a decline in inorganic nutrients, particulate and dissolved organic carbon and nitrogen from 1990 to 2018 (**Figure 3.4**). The analysis showed one breakthrough point (i.e. point where a significant change occurs) for all the variables in 1995-1996 except for the phosphates with two breakpoints in 1994 and 1999 (**Figure 3.4**).

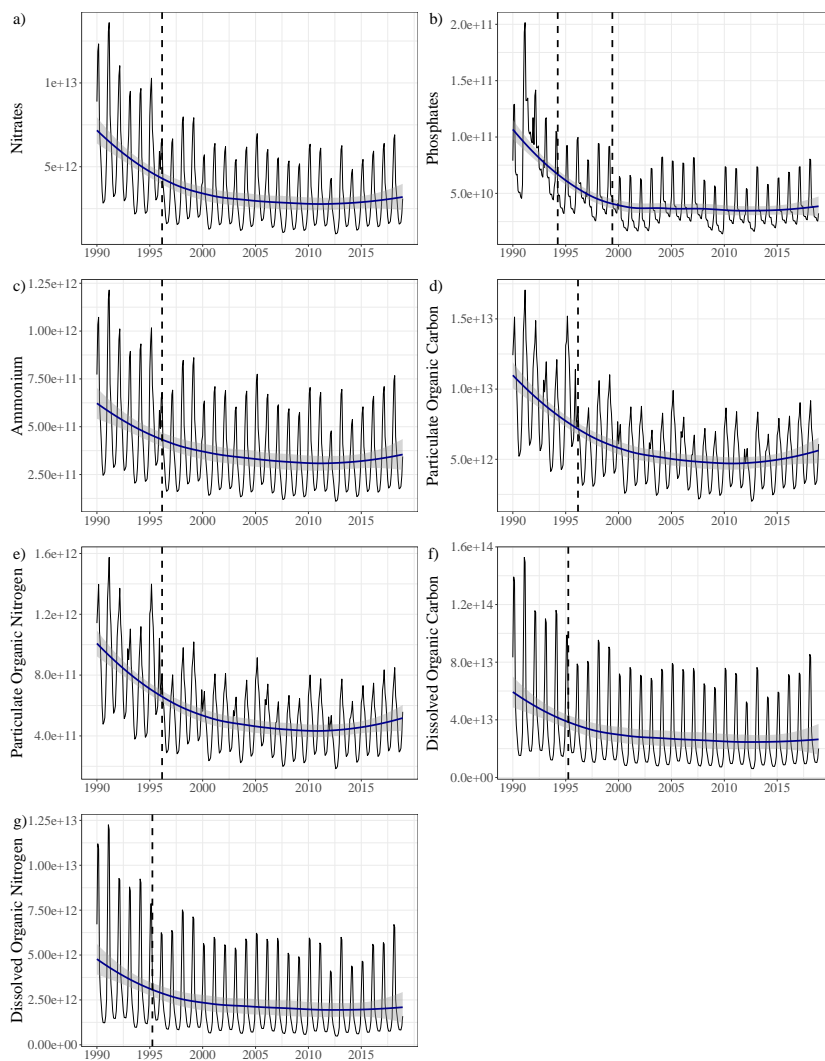


Figure 3.4: Regime shift analysis of the inputs of biogeochemical components (in mmol/month) into the Black Sea northwestern shelf by the three major rivers (i.e. Danube, Dniester and Dnieper) from 1990 to 2018: general trend in dark blue line, confidence interval in grey and date of the breakpoint in dashed black line.

3.4.1.2 Spatial and temporal pattern of low oxygen conditions

In the nineties, low-oxygen stations were found in the northeastern part of the shelf and extend to the south of the Danube's Delta and at depth down to 50 m (**Figure 3.5a**). In the worst case, low-oxygen conditions could persist for 6 months near the mouth of the Dniester River (**Figure 3.5a**). In 2008-2018, some low-oxygen stations were observed in the northeastern sector close to the mouth of the Dniester River (**Figure 3.5b**). The number of days of low-oxygen conditions was reduced near the mouth of the Dniester River from almost half a year to less than 3 months in 2008-2018 (**Figure 3.5b**). Sampling stations deeper than 100 m were permanently depleted in oxygen through the whole year (**Figure 3.5b**). There was no equivalent in 1995 (maximum depth sampling at 70 m) so we could not directly compare the two periods in the deeper sector.

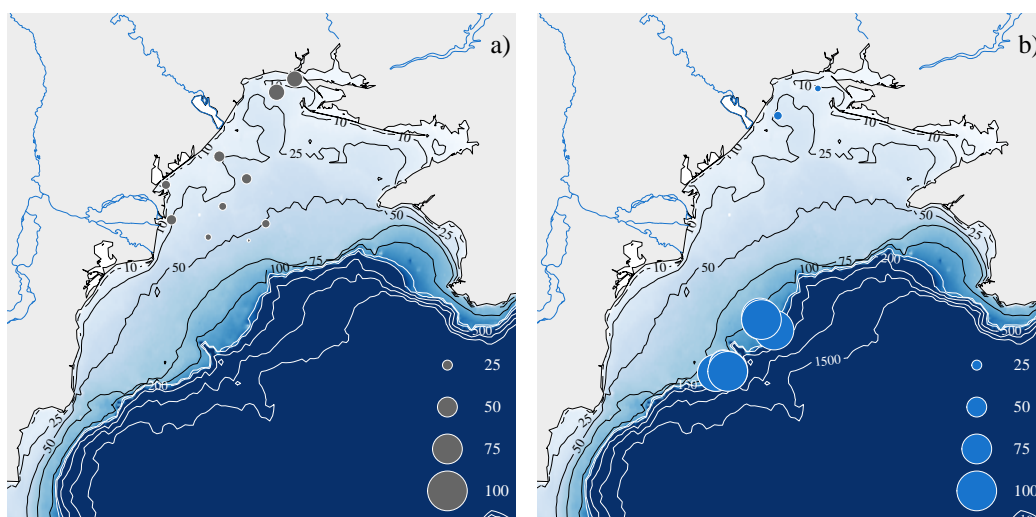


Figure 3.5: Spatial pattern of low bottom oxygen conditions over the study area. a) Grey dots, 1990-2000 sampling stations and b) blue dots, 2008-2018 stations. Legend, number of days with low oxygen concentration within a year (i.e. $< 120 \mu\text{mol O}_2/\text{l}$) expressed as a percentage (i.e. 100% means 365 days a year with low oxygen conditions).

3.4.2 Linking organic pollution and species changes

3.4.2.1 Separate physical and chemical contributions

Variation partitioning revealed that physical and chemical descriptors explained together 22% of faunal compositional variance ($R^2 = 0.46$, adjusted $R^2 = 0.22$); physics and chemistry, 13% ($R^2 = 0.26$, adjusted $R^2 = 0.13$) and 15% ($R^2 = 0.30$, adjusted $R^2 = 0.15$), respectively, with a confounding effect of 6% (adjusted $R^2 = 0.06$). So, we detrended chemical and faunal components from the effect of physics to explore the 9% of faunal variance solely explained by chemical variance.

3.4.2.2 Temporal effect on chemical factors and fauna

BGA patterns were found highly significant (p -value < 0.001 , for fauna and chemistry), evidencing a temporal effect on chemical and faunal components independently of physical drivers. Through *BGCoIA*, both fauna and chemistry were found significantly covariant along the unique axis (**Figure 3.6a**). As a general tendency, negative axis scores were associated with the first period (in grey; 1995) and positive scores are related to the second period (in blue; 2008-2017). However, some sampling stations exhibited higher discordance between fauna and chemistry scores and did not follow the general trend. For the first period, some sampling stations from the isobath 25 m, in the northeastern part of the shelf, showed high inconsistencies between chemistry and fauna scores (**Figure 3.6b**). For 2008-2017, high discordances were observed at depth below 100 m and in the northeastern part closed to the mouth of the Dnieper and Dniester Rivers (**Figure 3.6c**).

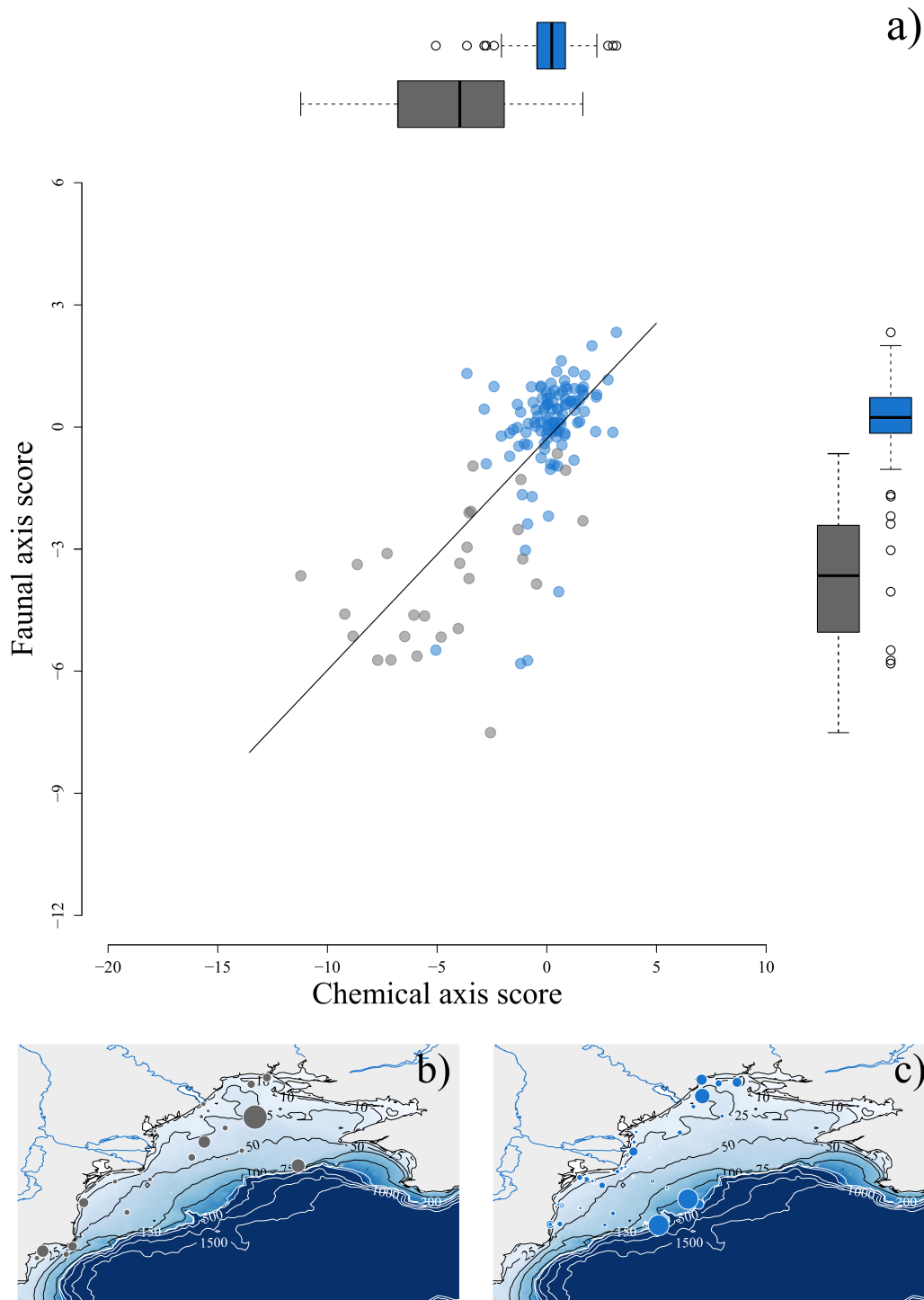


Figure 3.6: Between-Group Co-Inertia Analysis. a) Relationship between chemical and faunal scores of the unique axis defined by the two periods (grey, 1995; blue, 2008-2017; $r = 0.73$, p -value < 0.001). Both series of scores respond significantly to the period effect (Wilcoxon test, p -value < 0.001 in both cases). b) and c) Residuals of the model per sampling site in 1995 (b, grey) and 2008-2017 (c, blue). Dot size proportional to absolute residual value, with and higher values indicating higher inconsistency between faunal and chemical descriptors.

3.4.2.3 Relationships between chemical descriptors and faunal composition

From the first to the second period, the first *BGCoIA* axis exhibited a clear trend in faunal composition (**Figure 3.7**) and taxonomic richness (**Figure 3.8**), driven by a limited set of chemical descriptors that covaried positively and negatively as displayed in (**Figure 3.7a**). Among the thirty binary chemical descriptors, eighteen were significantly correlated to the *BGCoIA* axis **Table 3.2**. These included mostly oxygen and its seasonality as well as the vertically integrated content of organic carbon in the sediment (i.e. slow decay; SSED and fast decay; FSED). *Photosynthetic Active Radiation (PAR)* also had an influence on faunal composition in spite of relatively low correlations with the first *BGCoIA* axis ($r = -0.25$ and 0.22 , for respectively low and high *PAR* modalities).

Table 3.2: List of significantly correlated chemical descriptors with the Between-Group Co-Inertia axis (complementary to (**Figure 3.7**) and their respective Pearson's r -correlation. List of chemical descriptors by increasing Pearson's r -correlation. Only significant correlations are displayed (p -value < 0.05 after correction for multiple testing).

Chemical descriptors	r
Very high SSED	0.64
Very low DOX	-0.59
Very high FSED	-0.47
High DOX seasonality	-0.43
High SSED	-0.38
Very high DOX seasonality	-0.33
Very high POC	-0.33
Low DOX	-0.25
Low PAR	-0.25
Intermediate DOX seasonality	-0.23
Low DOX seasonality	0.2
High PAR	0.22
Low FSED	0.26
Intermediate SSED	0.26
High DOX	0.33
Very low DOX seasonality	0.37
Low SSED	0.38
Very high DOX	0.46

As a general trend, stations in 1995 were characterised by oxygen impoverishment and high oxygen variability (in grey, **Figure 3.7a**). The sediments in 1995 were highly enriched in organic matter and the quantity of light reaching the bottom was low (in grey, **Figure 3.7a**). On the contrary, stations in 2008-2017, benefited more oxic conditions with less seasonal variability (in blue, **Figure 3.7a**). The amount of organic content in the sediment was lower and the amount of light reaching the bottom was higher compared to 1995 (in blue **Figure 3.7a**). In terms of species assemblage in 1995 (left side of the axis, in grey, **Figure 3.7b-c**, very negative scores were associated with annelid species opportunistic or tolerant to organic enrichment such as *P. cornuta* (EG IV), *Melinna palmata* (EG III), *Alitta succinea* (EG III), Clitellata (EG V) and *Capitella* spp. (EG V). From left to central positions of the axis, more intolerant species were observed such as bivalves (e.g. *M. phaseolina* (EG I) and *M. galloprovincialis* (EG III)) and crustaceans (e.g. *Ampelisca* spp. (EG I), *Iphinoe elisae* (EG I), *Phtisica marina* (EG I)). While a shift from the 1995 to the 2008-2017 sample positions was quite perceptible along the *BGCoIA* axis (**Figure 3.7a-b**), this was less the case between positions of species functional groups (**Figure 3.7c-d**). In the 2008-2017 period, species intolerant to organic enrichment were especially represented by crustaceans such as *Apherusa chierighinii* (EG I), *Eudorella truncatula* (EG I), *Microdeutopus versiculatus* (EG I), *Perioculodes longimanus* (EG II) and *Upogebia pusilla* (EG I), the gastropod *Pusillina lineolata* (EG I), and long-lived bivalves such as *Acanthocardia paucicostata* (EG I), *Parvicardium simile* (EG I), *Pitar rudis* (EG II) and *Spisula subtruncata* (EG I). Among others, active carnivores also contributed to the trend such as nemerteans and epibenthic polychaetes (e.g. *Eumida sanguinea* (EG II), *Sphaerosyllis pontica* (EG II), *Syllides longocirratu*s (EG II)) and the sea spider *Callipallene phantoma* (EG I). Other opportunistic taxa among bivalves (*Abra* spp. (EG III) and the invasive ark clam *Anadara kagoshimensis* (EG IV)) and polychaetes (e.g. *Heteromastus filiformis* (EG IV), *Polycirrus jubatus* (EG IV) and the invasive worm *Dipolydora quadrilobata* (EG IV)) were also found characteristic of the recent period.

The Fourth-corner test did not indicate significant differences between the saprobic group positions along the *BGCoIA* axis (p -value > 0.05) nor significant correlations of individual groups; only groups I and II, least tolerant to organic enrichment, exhibited more homogeneous positions on the central-right part of the axis (respectively p -value= 0.00018 for group 1 and p -value = 0.03566 for group 2), contrary to the more tolerant groups III and IV that spread over the entire range, especially in the most disturbed conditions (**Figure 3.7c-d**).

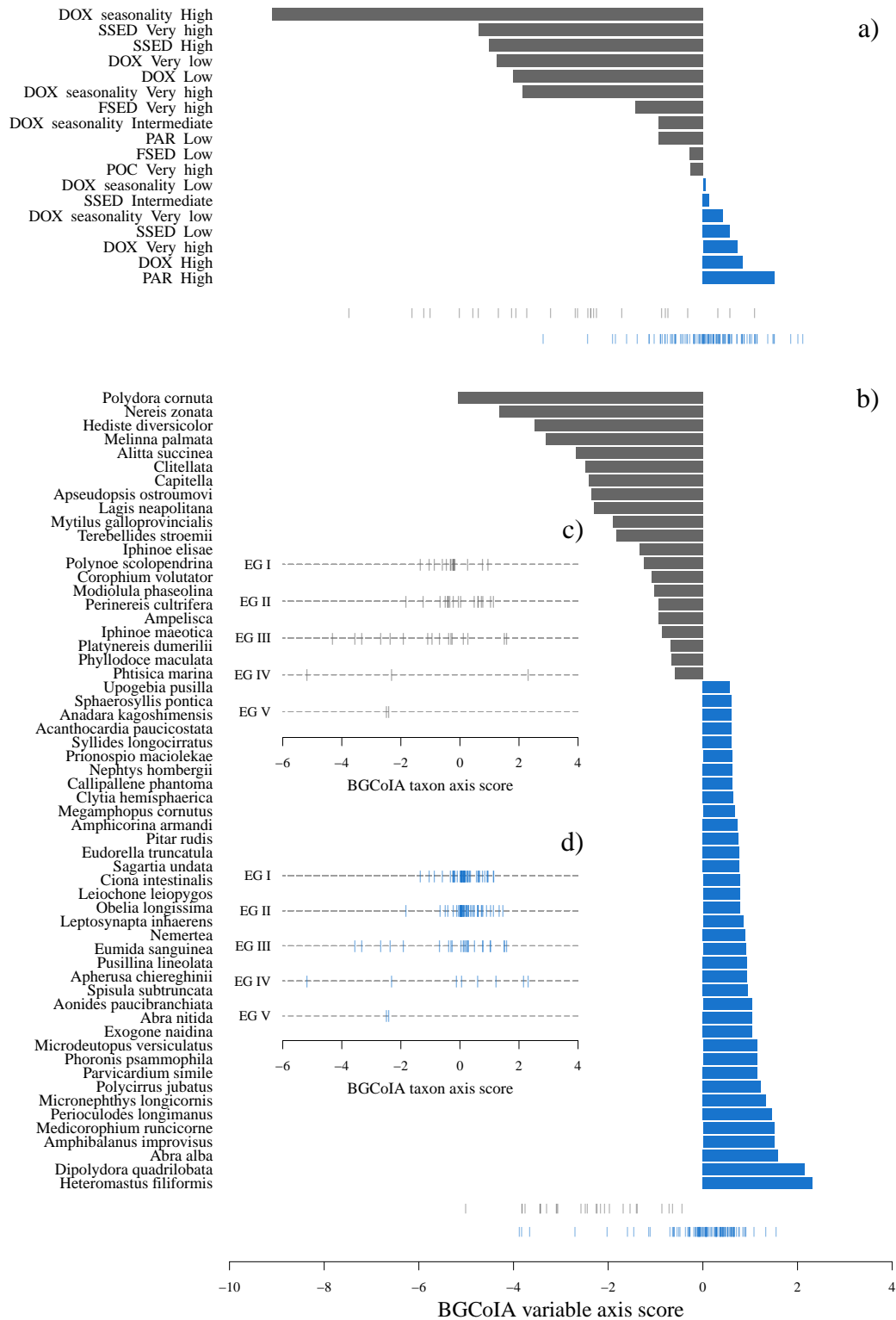


Figure 3.7: Between-Group Co-Inertia Analysis, variable scores along the axis; vertical segments, sampling stations; grey, 1995; blue, 2008-2017. a) Habitat descriptors significantly correlated to the axis (consistently with Table 3.2). b) Taxa; for clarity, only taxa with higher scores (more than 0.5 in absolute value) are displayed. All taxa (vertical segments), grouped per ecological group of organic tolerance along the axis in c) grey, for 1995 period and d) blue, for 2008-2017 period.

Next to faunal compositional changes, taxonomic richness was found to increase along the *BGCoIA* trend (**Figure 3.8a**). The taxonomic richness was preliminarily detrended from physical descriptors to only consider the effect of organic disturbance; the main drivers can be identified in (**Figure 3.7a**). In 1995, all habitats had relatively low richness, especially in the northeast sector (**Figure 3.8b**). Spatially, the increase in richness was not general as it remained low in areas closed to the mouth of the Danube and at the edge of the shelf (**Figure 3.8c**). Besides, abiotic conditions among the stations of the first period were more variable (i.e. heterogeneous) than those during the recent one, explaining the greater contribution of variables pointing to the 1995 period (**Figure 3.7a**); this was also reflected in **Figure 3.6a**. In contrast, the recent period exhibited more homogeneous faunal and abiotic scores except for a few sampling sites with very low faunal scores (**Figure 3.6a**). Regarding taxonomic richness, it is the opposite with higher variance in 2008-2017 compared to 1995 (**Figure 3.8c**).

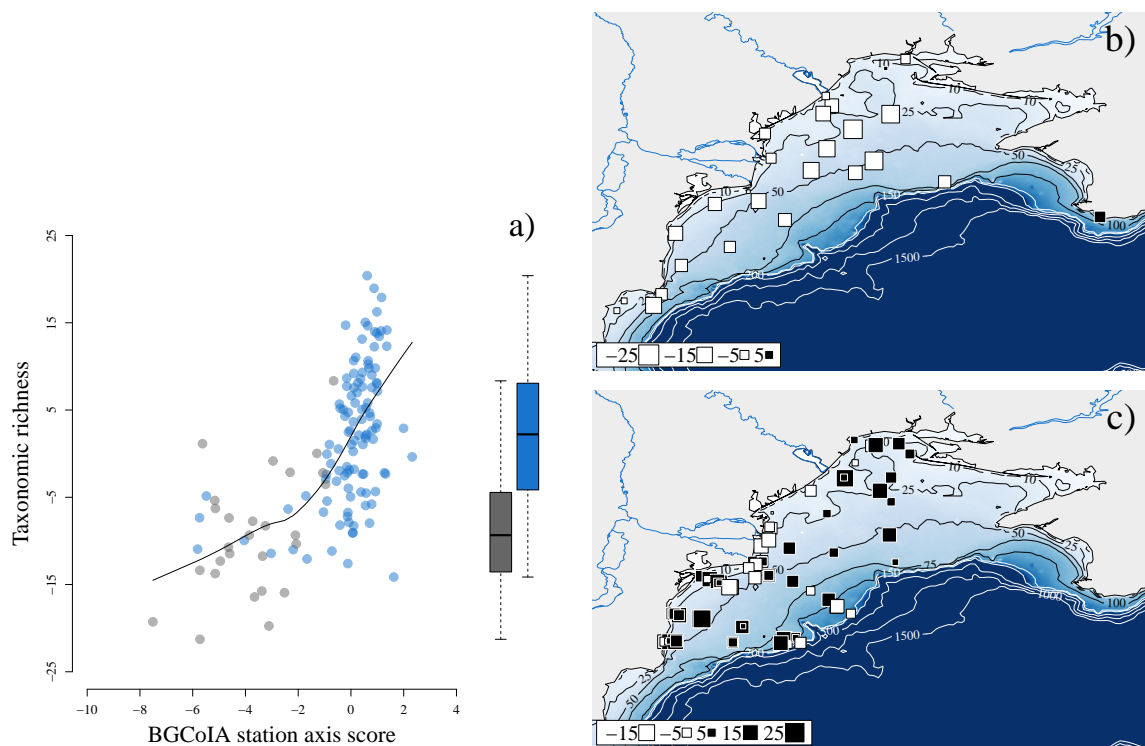


Figure 3.8: Taxonomic richness pattern. a) Trend in taxonomic richness per sampling station (grey, 1995; blue, 2008-2017) along the *BGCoIA* axis. Black curve, Lowess non-parametric fitting illustrating the recovery trend from left (1995) to right (2008-2017). Boxplot, significant difference between the two distributions (Wilcoxon test, p -value < 0.001). Spatial distribution of detrended taxonomic richness in b) for 1995 period and c) for 2008-2017 period. Legend in the bottom-left corner: white squares, low scores; black squares, high scores; square size is proportional to the deviation from the mean taxonomic richness.

3.5 Discussion

3.5.1 Benthic compositional changes

In this study, we compiled one of the most extensive macrozoobenthic data set of the Black Sea to provide a comprehensive analysis of community compositional variations over the northwestern shelf in the recent history that had undergone a massive eutrophication. Our analytical approach enabled to counteract some experimental constraints inherent to the study of biodiversity at the regional scale where several causes often result in a same effect. By detrending chemical determinants of the benthic communities from physical ones, our results shade light on macrozoobenthic changes solely due to organic disturbance. In general, hypoxia negatively impacts macrozoobenthic biodiversity (Breitburg et al., 2018; Diaz & Rosenberg, 1995; Levin et al., 2009; Diaz & Rosenberg, 2008) and can lead to species loss (Vaquer-Sunyer & Duarte, 2008). This is clearly explicit through our results, and two aspects deserve to be discussed in this context.

On the one hand, the strong covariation between chemistry and fauna indicates that specific organic conditions determine specific species assemblages (**Figure 3.6a**). Thus, changes in species assemblages over time, accompanied by a species enrichment, are consistent responses to a reduction in organic disturbance. This supports the direct effect of the freshwater input mitigation over the last thirty years due to a decrease in nutrients and organic loads from the three main tributaries of the Black Sea northwestern shelf (**Figure 3.4**). The regime shift observed in the mid-1990s matches the abrupt stop of collectivistic farming at the end of the Soviet Union (Mee et al., 2005). As a direct consequence, eutrophication and its subsequent hypoxia have been reduced where riverine inputs are released. Indeed, through water circulation, most of the discharges from the Danube are carried to the northeastern part of the shelf (Grégoire & Friedrich, 2004). In this sector, we evidenced a smaller extent and shorter duration of bottom low-oxygen conditions over the last thirty years (**Figure 3.5**). Yet, hypoxia is still present in this area due to lower ventilation consequent from water warming (Capet et al., 2013).

In the 1990s, sediments were characterised by high organic content and low dissolved oxygen, with strong seasonal variability and low amount of light reaching the bottom due to the strong absorption of light by suspended materials (**Figure 3.7a**). Following the mitigation of eutrophic conditions, lower amount of organic matter reached the shelf, and water transparency increased as well as oxygen content (**Figure 3.7b**). As a result, oxygen may have become less seasonally variable given the reduction in low-bottom oxygen conditions in the northeastern sector of the shelf (**Figure 3.5b**). It is relevant to note that high *Particulate Organic Carbon (POC)* in the sediment is not exclusively related to low-oxygen conditions as very deep anoxic sites are characterised by low content of organic carbon in the sediment. In normoxic conditions, various levels of can be encountered (from very low to high) without necessarily indicating bad chemical conditions.

That may explain why modalities of *POC* are anecdotally associated with the first axis of the *BGCoIA* (i.e. only very high *POC* negatively correlated with the first axis; **Table 3.2**).

The improvement of water quality led to higher taxonomic richness (**Figure 3.8a**). In 1995, all habitats had relatively low richness except for some sampling sites outside the influence of the Danube (i.e. southern sector and offshore sites), that benefited from more oxic conditions (**Figure 3.5a**). Over 2008-2017, taxonomic richness has increased only in habitats with beneficial conditions resulting in a higher variability of taxonomic richness between favourable and organic polluted habitats (**Figure 3.8a,c**). In addition, these changes benefited species intolerant to organic enrichment and mainly positioned from the central to the right part of the *BGCoIA* axis (**Figure 3.7d**). Such species include typical long-lived bivalves (e.g. *A. paucicostata*, *S. subtruncata*, *P. simile*, *P. rudis*). The presence of *P. rudis* is positive as it is listed in the Order No. 488/2020 as endangered species of the Romanian Black Sea (available online: Black Sea commission). Also, mobile crustaceans are considered sensitive to low-oxygen conditions but able to actively escape from hypoxia and to recolonise more oxic areas (Vaquer-Sunyer & Duarte, 2008); most of them belonged to EG I (e.g. *A. chiereghinii*, *E. truncatula*, *M. versiculatus* and *U. pusilla*) or EG II (e.g. *P. longimanus*). The mud shrimp *U. pusilla*, with high oxygen requirements and which disappeared from many areas of the Black Sea during the 1990s (Todorova & Konsulova, 2000; Zaitsev & Mamaev, 1997), is especially representative of the recovery. In addition, the number of active predators, such as nemerteans and epibenthic polychaetes, increased over the investigated period. Indeed, predators are generally more vulnerable to hypoxia than other feeding groups (Levin et al., 2009) and need higher metabolism to feed on prey (Sperling et al., 2013). In 2008-2017, more oxic conditions may potentially led to higher predator contribution. Previous study also evidenced an increase in the number of active predators likely related to an improvement in crustacean status (Teacă et al., 2020).

On the other hand, the weaker covariation between chemical conditions and fauna in the 1990s (**Figure 3.6**, in grey) could explain the opportunistic nature of the remaining fauna during that period, especially in the northeastern sector of the shelf, more affected by organic enrichment and low oxygen conditions (**Figure 3.5a**). At that time, the fauna was mainly represented by opportunistic polychaetes and oligochaetes (e.g. Clitellata and *Capitella* spp.). Although these organisms are known to particularly proliferate in highly disturbed conditions (Pearson & Rosenberg, 1978; Vaquer-Sunyer & Duarte, 2008), they are not specific to them, hence explaining the observed lack of covariation between chemistry and fauna in the 1990s. Also, the assemblages from the *Mytilus* biocenosis, mainly observed at 25 m in the 1990s (Wijsman et al., 1999), were hardly recognisable with some epibenthic polychaetes attached to shells or living in their interstices pointing towards negative values on the *BGCoIA* axis (e.g. *Platynereis dumerlii*, *Perinereis cultrifera* and *Nereis zonata*). Similarly, the corresponding *BGCoIA* chemical and faunal scores were highly discordant for these stations (**Figure 3.6a,c**).

Next to the positive signs of recovery highlighted in our analysis, it is worth noting that the most contributing species towards positive scores are far from being vulnerable (e.g. *Abra* spp., *D. quadrilobata*, *H. filiformis*, *P. jubatus* in **Figure 3.7b,d**) and taxonomic richness remains low near the Danube Delta and at the edge of the shelf (**Figure 3.8c**). Indeed, the plume-area near the Danube mouth still receives high inputs of suspended matter leading to low transparency of water, high organic matter content in the sediment and low oxygen concentrations (Teacă et al., 2019a, 2020). Also, stations at the edge of the shelf (100-150 m), far from the coast and less affected by freshwater inputs, are permanently depleted in oxygen due to the intrusion of euxinic waters from the deep-sea basin (**Figure 3.5b**). Therefore, only a few opportunistic species with no particularly specific environmental requirements can cope with these harsh conditions (Kolyuchkina et al., 2022), as also shown by the low chemical-faunal covariation of these stations (**Figure 3.6a,c**).

3.5.2 Analytical constraints and uncontrolled processes

Our large-scale analysis provides a synthetic update on the Black Sea northwestern sea floor ecosystem health. Although the highlighted pattern indicates areas where the biota recovered, the constraining nature of our faunal data and the biogeographical context may have limited analytical performance. Indeed, the unavailability of abundance data due to sieve mesh size incompatibility prevented an efficient use of the saprobic groups as some sensitive species may have likely been in lower densities than tolerant ones under strong organic conditions, while reversing the ratio toward recovery. As shown by Warwick et al. (2010), abundance data (i.e. individual organism or biomass density, or combination of both) may have enabled more consistent associations between sampling periods and saprobic groups. The clear *BGCoIA* pattern is the strongest point in our results to evidence the influence of organic disturbance on community composition, while *AMBI* group distributions are less convincing. Hence, these results strengthen the conclusions of (Warwick et al., 2010) regarding the weakness of presence-absence data in the use of *AMBI* groups. Nevertheless, for the first time, we provided the use of the *AMBI* typology as a trait through the Fourth-corner method as it should be applied in a multivariate context when manipulating species attributes to prevent the detection of false positives (Peres-Neto et al., 2017).

Also, the introduction of non-native species has been a critical issue in the Black Sea, especially due to the expansion of maritime international trade (Gomoiu & Skolka, 1996; Gomoiu et al., 2002; Çinar et al., 2014; Skolka & Preda, 2010). Over the investigated period, there may have been a certain number of invasive species newly established, and among them, certain may have succeeded more than some native ones during the restoration phase when recolonising vacant niches. In the recent period, the main benthic invaders were the balanus *Amphibalanus improvisus*, one of the most common species on the Black Sea introduced in the middle of the 19th century (Gomoiu & Skolka, 1996), the ark clam *A. kagoshimensis* (syn. *Scapharca inaequalvis*), firstly identified in Gomoiu (1984) and the deposit-feeder polychaete *D. quadrilobata*, firstly identified in 2003 along the Romanian coast (Surugiu, 2012).

D. quadrilobata is successfully established on the Romanian shelf and it is one of the main engineering species on the circalittoral mixed sediments with *Mytilus* biogenic reefs (Teacă et al., 2019a, 2020) and circalittoral offshore mixed sediments with *Modiolula* (Begun et al., 2010). Hence, newly settled species may have contributed slightly to the increase in taxonomic richness as an effect confounded with recovery from organic disturbance, not as a causal response.

In addition, there are still confusions in taxonomic identification that may have led to misinterpretation of faunal changes. For instance, it is a well-known fact that the identification of *Polydora* species and related genera (e.g. *Dipolydora*) are not so clear in the Black Sea (Surugiu, 2012). Thus, it is a possibility that *D. quadrilobata* prospered even in the nineties but the taxonomic expertise, at that time, was deficient which led to confusions. There are also possibilities that species in the nineties are still found in the recent period but under another name. For instance, in 1995, *Eunereis longissima* may be misidentified as *N. zonata* (Surugiu et al., 2018). That can explain why *N. zonata* was only reported in 1995, not because of the chemical conditions at this time, but because its identification changes in the most recent period.

3.5.3 Conclusions and perspectives

In this study, we provided the first analytical overview of the Black Sea northwestern shelf benthic ecosystem state from the recent history. We efficiently combine newly generated chemical descriptors of water and sediment quality with faunal data to testify sea floor ecosystem recovery from eutrophication over the shelf. Although the episode of pronounced hypoxia in the Black Sea made benthic compositional changes expected, we tested and supported the subjacent hypothesis of recovery through a constraining methodological approach penalised by variance removal (i.e. data detrending). By this way, we provided a useful method, replicable in other areas to investigate changes in species composition due to specific causes where natural or human-mediated forces act in concert. We evidenced an increase in taxonomic richness, contribution of more sensitive species as well as active predators related to a decrease in organic pollution over the last twenty years. However, regarding the environmental reality and taxonomic uncertainties, this recovery success should be considered with caution. Firstly, the quality of freshwater inputs remains questionable in some sectors of the shelf where causes of hypoxia are persisting. Secondly, several long-lived species with a typical slow recovery following disturbance, and formerly abundant in the Black Sea, still have low occurrence frequencies. Among these species, large bivalves as those encountered here (e.g. *S. subtruncata*, *M. galloprovincialis*, *M. phaseolina*) sometimes require several decades to achieve a minimum of reproductive successes to ensure successful recovery (Beukema et al., 2001). The fact that these species did not dominate the recovery trend evidences a partial recovery, and do not exclude the concomitant effects of other human-mediated pressures such as bottom trawling that persists on the shelf (Egekvist et al., 2021). Conclusively, our findings advocate for a benthic follow up of the area like in other European waters where ecosystem health should comply with the *Marine Strategy Framework Directive (MSFD)* (Council of the European Union, 2008).

TRAIT DATA COMPILATION

Foreword:

Chapter 3 evidenced significant relationships between abiotic conditions and species data over the northwestern shelf of the Black Sea. Once the link between abiotic conditions and species has been investigated, it is possible to move on to the study of the link between abiotic conditions and the respective traits associated with the species. This chapter presents the data compilation work and gives some insights on how to use the data. We present an open-access dataset that combines 27 functional traits of more than 120 benthic species with environmental variables derived from an ocean biogeochemical model. We selected traits for their relevance to ecosystem functioning (e.g. sediment dwelling depth and mobility) and sensitivity to environmental pressures (e.g. body size, reproductive strategy). Prior to this PhD work, there was a lack of effort in compiling traits in the Black Sea, and to the best of my knowledge, this PhD study is the first to provide an open-access compilation of traits for the Black Sea. A graphical summary of this chapter is provided in **Figure 4.1**. Supplementary materials associated with **Chapter 4** are available in **Appendix C**.

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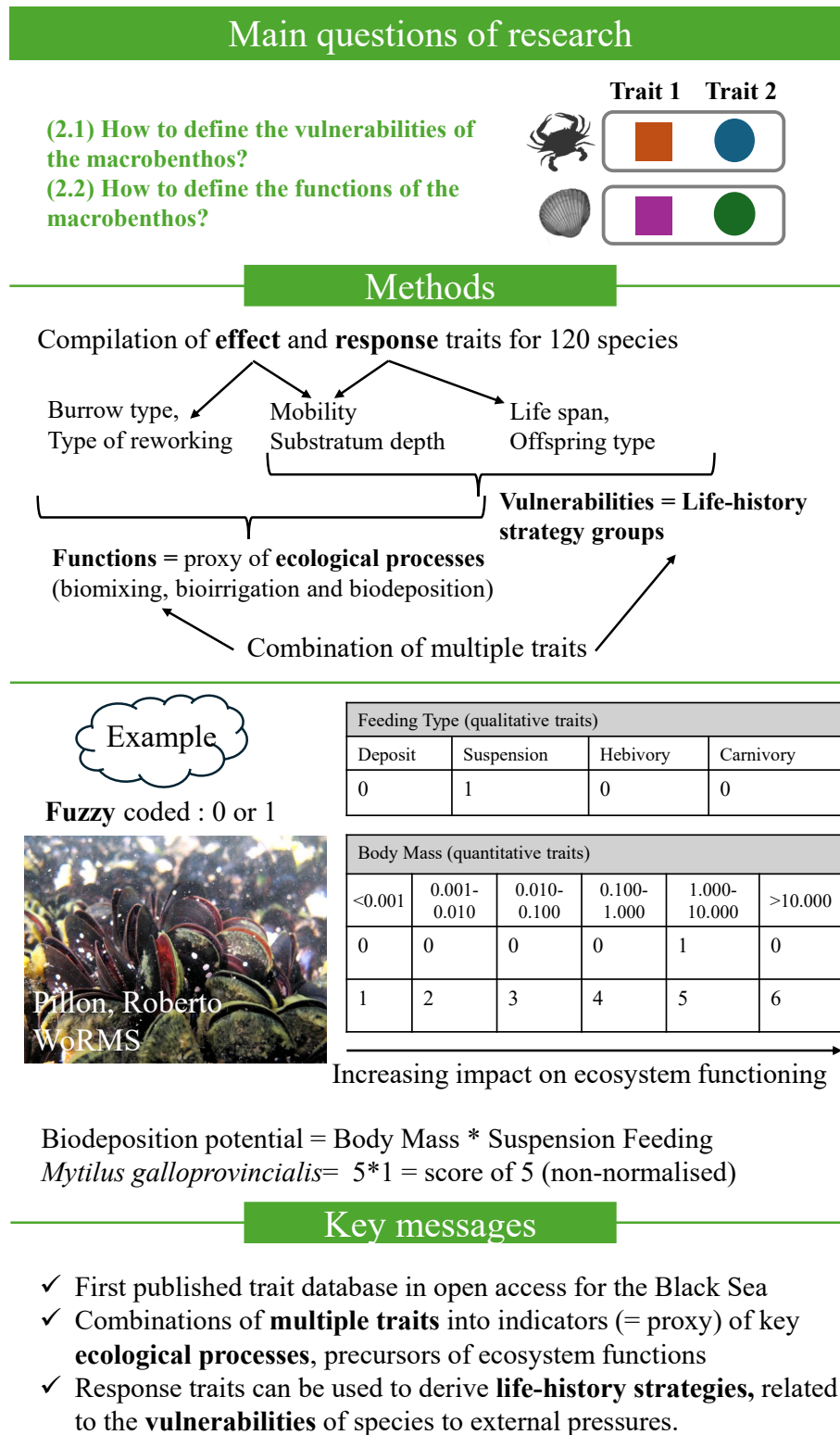


Figure 4.1: Graphical Abstract of the Paper II (Chevalier et al., 2025a): **Chapter 4** aims to answer the second question of this PhD thesis by providing an extensive compilation of traits.

4.1 Abstract

Benthic ecological data are crucial to study and manage ecosystems. On the one hand, abiotic and species data provide complementary information to identify habitats. On the other hand, trait data, describing taxon characteristics, are required to predict anthropogenic impacts on marine ecosystems. Indeed, species traits are now widely used to understand natural selection in communities or to highlight ecosystem functions. While trait data are in growing demand, compiling them is challenging, time-consuming and there are no properly established procedures for major marine ecosystems. Here, we share a data set comprising macrozoobenthic occurrences for 215 taxa over the Black Sea northwestern shelf, between 1995 and 2017, and 27 traits documented for 127 taxa that related to life cycle and ecosystem function. In addition, we provide an abiotic data set of physical and chemical variables generated by a model or compiled from *in-situ* data. This data set aims to fill the functional knowledge gap in the Black Sea and offers research opportunities to future studies covering ecosystem functions, biodiversity conservation, and management.

4.2 Background & Summary

The macrozoobenthos, which lives in soft-bottom sediment, is a key component of the marine ecosystem because it regulates the fluxes of energy and matter between the sediment and the water column (e.g. Kristensen et al. (2012); Kristensen and Kostka (2005); Snelgrove et al. (2018); Snelgrove (1998)). In shallow coastal waters, their behaviour and activity have a great impact on climate regulating and supporting ecosystem services via their effect on processes like carbon sequestration or denitrification (Braeckman et al., 2010; Queirós et al., 2013; Snelgrove, 1998). In addition to their important roles in ecosystem functioning, macrozoobenthic species are widely used as reliable bio-indicators to monitor ecosystem health because of their wide distribution, relative fixed position, and sensitivity to external stressors (Borja et al., 2000, 2003; Grall & Glémarec, 1997; Gray & Pearson, 1982; Pearson & Rosenberg, 1978). In the framework of the *MSFD* (Council of the European Union, 2008) and the EU *Habitat Directive (HD)* (Council of the European Communities, 1992) programs, acquisition of macrozoobenthos data is critical for assessing the state and changes in marine ecosystems in response to external pressures. To enhance our understanding of the interactions between organisms and their environment, benthic faunal data need to be combined with habitat descriptors (e.g. substrate type, oxygen levels). Thus, habitat descriptors as well as species occurrences are crucial to identify biotopes, enabling more targeted and effective management strategies (Council of the European Communities, 1992; Council of the European Union, 2008).

In recent decades, trait-based approaches have become useful tools for better predicting benthic community assemblages and ecosystem functional changes (Beauchard et al., 2017; de Juan et al., 2022; Degen et al., 2018; Martini et al., 2021). Traits are any morphological, physiological or phenological features, described at the species level (Díaz & Cabido, 2001; Violle et al., 2007; McGill et al., 2006). Traits can define species vulnerability to disturbances (Lavorel et al., 1997; Mouillot et al., 2013) (i.e. response traits Lavorel and Garnier (2002)) and estimate role of the species on ecosystem functions (Díaz et al., 2013; Hooper et al., 2005) (i.e. effect traits Lavorel and Garnier (2002)). Thus, trait data are necessary for gaining mechanistic insights into space occupation (e.g. environmental filtering) and better predict the roles of benthic communities on ecosystem functioning (Beauchard et al., 2017; Bremner et al., 2003; Bremner, 2008).

Here, we present an extensive data set combining species, biological traits and abiotic descriptors to gain a holistic understanding of marine benthic health. We compile occurrences for 215 macrozoobenthic taxa, sampled over the northwestern continental shelf of the Black Sea. This part of the Black Sea has been heavily disturbed by eutrophication and organic enrichment, leading to strong bottom hypoxia, and other anthropogenic pressures such as bottom trawling, which generally alters benthic ecosystem functions (Begun et al., 2022; Capet et al., 2013; Gomoiu, 1992; Mee et al., 2005; Nenciu et al., 2023; Teacă et al., 2019a, 2020; Zaitsev, 1992).

Our data set includes a wide panel of response and effect traits that enable various kinds of investigations on functional diversity (Díaz & Cabido, 2001; Villéger et al., 2008; Petchey & Gaston, 2006), development of ecological indicators through the simultaneous combination of response traits (Beauchard et al., 2017, 2023; de Juan et al., 2022; Villnäs et al., 2018) and estimation of key ecological processes (e.g. bioturbation (Beauchard et al., 2023; Queirós et al., 2013; Solan et al., 2004a) and bioirrigation (Beauchard et al., 2023; Wrede et al., 2018; Renz et al., 2018), precursors of ecosystem functions (Lam-Gordillo et al., 2020; Snelgrove et al., 2014). We also provide abiotic data to investigate community response to environmental drivers (e.g. RLQ analysis; Dolédec et al. (1996); Dray et al. (2014)). Abiotic data can be used as predictors for ecological processes spatial distribution modelling (Bolam et al., 2023; Gogina et al., 2017, 2020; Weinert et al., 2022). The mapping of these indicators can help to identify hotspots of functional diversity and ecosystem processes, supporting ecosystem-based management (Bianchi et al., 2021; Cogan et al., 2009; Lam-Gordillo et al., 2020; Miatta et al., 2021; Snelgrove et al., 2014).

4.3 Methods

4.3.1 Sampling station records

We compiled a macrozoobenthic data set covering the Black Sea northwestern shelf from various projects: EROS-2000 (SESAME, HYPOX), EMBLAS Joint Black Sea Surveys 2016-2019 (<https://emblasproject.org/>) and MSFD for Bulgarian coast in 2017 from the Institute of Oceanology - BAS. Macrozoobenthos data (i.e. biomass and individual densities) are archived and freely accessible from Marine Data Archive for EROS-2000 (Wijsman, 1995) and PANGEA database for SESAME (Teacă et al., 2015a, 2015b) and HYPOX (Gomoiu, 2012a, 2012b). The data from EMBLAS were collected, processed and analysed by GeoEcoMar, in Romania. Within the EMBLAS project (Teacă et al., 2022), the Ukrainian Scientific Centre of Ecology of the Sea (UkrSCES) processed and analysed samples from the same stations. Macrozoobenthos data from MSFD-Bulgarian were collected, processed and analysed by the Institute of Oceanology, Bulgarian Academy of Sciences (IO-BAS) in the framework of Black Sea monitoring program (Todorova & Stamatova, 2017). A total of 237 stations were sampled between 1995 and 2017 (**Figure 4.2**). A summary with the number of stations per campaign and their respective date is provided in **Table 4.1**. The investigated period was characterised by important environmental changes with the progressive reduction of eutrophication due to the decrease of the nitrate and phosphate river discharges in the 1990s (Gomoiu, 1992) and the warming of the water that significantly reduces the formation of the cold intermediate waters since 2008 (Stanev et al., 2019; Capet et al., 2020, 2014; Miladinova et al., 2018). A more detailed study of environmental changes and how it affected species compositional changes between 1995 (Wijsman, 1995) and 2008-2017 (Teacă et al., 2015a, 2015b; Gomoiu, 2012a, 2012b; Teacă et al., 2022) is provided in Chevalier et al. (2024).

Table 4.1: List of campaigns with their respective date of sampling and number of stations.

Campaign	Date of sampling	Number of stations
EROS-2000	5 to 27 August 1995	27
SESAME	7 to 8 April 2008	4
	9 to 11 September 2008	4
HYPOX	14 to 24 May 2010	32
	22 July 2010	4
	5 to 9 September 2010	15
	2 to 10 April 2011	30
EMBLAS	17 to 21 May 2016	15
	10 July 2017	7
MSFD Bulgarian coast	6 to 28 October 2017	84
	15 to 17 October 2017	15
	5 December	5
Total	-	237

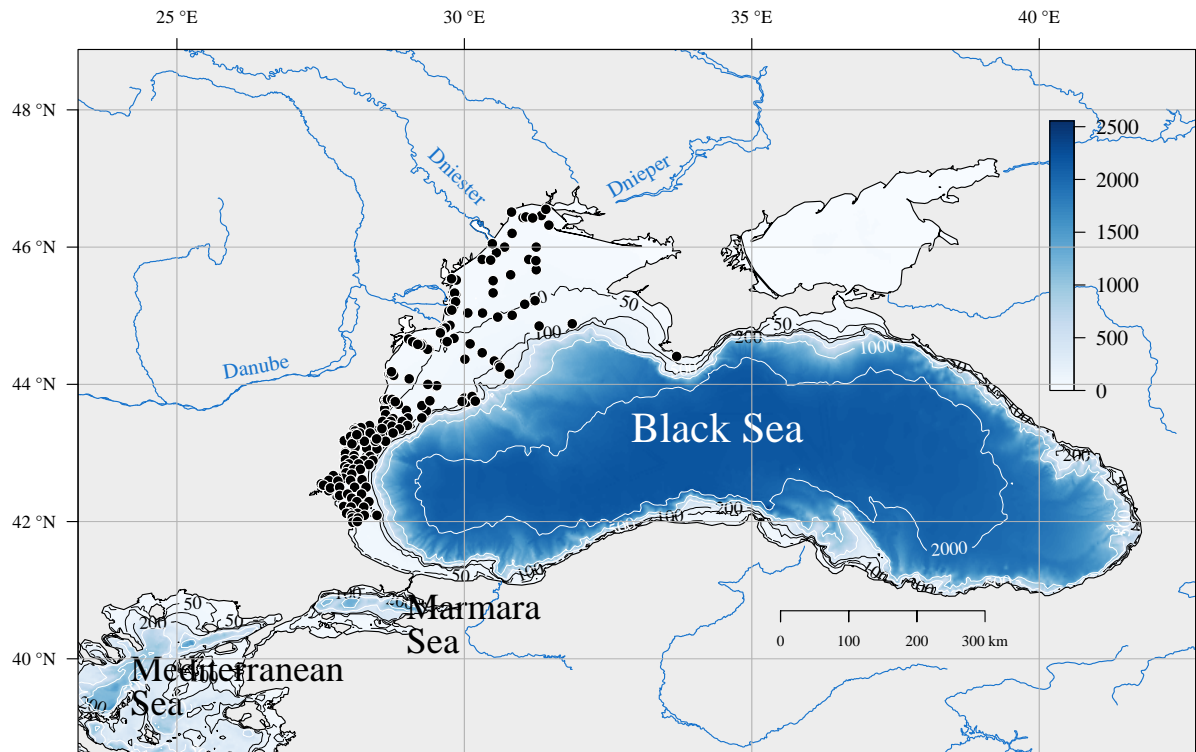


Figure 4.2: The Black Sea in the regional context of the study. Named rivers as the main sources of freshwater inputs; black dots represent the sampling stations between 1995 and 2017. Contour lines and colour bar denote water depth in meters.

4.3.2 Abiotic data

Abiotic conditions were provided from in-situ data and a 3D ocean numerical model run in operational mode in the frame of the *CMEMS*. More details about abiotic data can be found in Chevalier et al. (2024).

4.3.2.1 *In situ* data

For each sampling station, sampling depth was provided as well as substrate type information. Several substratum classifications were used throughout the campaigns. For the EROS-2000 campaign, sediment characteristics averaged over the top 21 cm were provided (e.g. porosity, medium grain size, and silt content, defined as the sediment fraction $< 63 \mu\text{m}$). For SESAME, HYPOX and MSFD on the Bulgarian coast, sediment type was directly derived from Van Veen grab samples. For EMBLAS data, substrate information was taken from EMODnet (“EUNIS-like” classification for the Black Sea (Populus et al., 2017)). To compare substrate information, we determined a single substratum code divided into three main categories: “muddy to sandy mud”, “muddy sand to sand” and “mixed and coarse sediment” (adapted from EMODnet Geology, in Populus et al. (2017); Vasquez et al. (2021)).

4.3.2.2 Model data

We computed abiotic data from a model reanalysis delivered by the *CMEMS* Black Sea Marine Forecasting center (Ciliberti et al., 2021). These data were produced by the coupled hydrodynamical-biogeochemical *NEMO 4.2 BAMHBI* model run in reanalysis mode at a resolution of 2.5 km with 59 vertical levels. *BAMHBI* is implemented in the three-dimensional hydrodynamic model *NEMO* (<https://www.nemo-ocean.eu>) and is currently run in operational mode in the Black Sea. The model simulates oxygen, nitrogen, phosphorus, silicate and carbon cycling mediated by several plankton functional types and the microbial loop. It explicitly represents processes in anoxic conditions and the benthic remineralisation and resuspension processes. More details on the model can be found in Capet et al. (2016); Grégoire and Soetaert (2010); Grégoire et al. (2008). A summary with the abiotic descriptors, their respective abbreviation and unit is provided in **Table 4.2**.

Table 4.2: List of abiotic descriptors derived from the coupled hydrodynamical-biogeochemical *NEMO 4.2 BAMHBI* model with their respective abbreviation and unit. Abiotic variables are classified into two categories: physical and chemical variables.

Abiotic descriptors	Abbreviation	Unit
Physical variables		
Bottom temperature	TEMP	$^{\circ}\text{C}$
Bottom salinity	SAL	<i>p.s.u</i>
Bottom shear stress	SHEAR	N m^{-2}
Chemical variables		
Photosynthetically active radiation	PAR	W m^{-2}
Particulate organic carbon content	POC	mmol C m^{-3}
Vertically integrated organic carbon concentration (fast decay)	FSED	mmol C m^{-3}
Vertically integrated organic carbon concentration (slow decay)	SSED	mmol C m^{-2}
Dissolved oxygen concentration	DOX	$\mu\text{mol l}^{-1}$
Particulate organic carbon flux to the bottom	botfluxPOC	$\text{mmol C m}^{-2} \text{day}^{-1}$

4.3.3 Taxonomic occurrence data

4.3.3.1 Data acquisition

During the EROS-2000 campaign, the fauna was sampled with a Reineck box corer (60 × 30 × 30 cm), and 8.6 cm i.d. cores were taken from the box corer 10-12 cm down into the sediment. On the ship, samples were stained with Congo Red and preserved in buffered formaldehyde. In the laboratory, the samples were washed through 1.0, 0.25- and 0.125-mm mesh sieves to remove the mud and to separate macro- and meio-zoobenthos. All the material retained by the three sieves was examined by binocular microscope. Macrozoobenthic animals were identified, if possible, to the lowest taxonomic level (i.e. species level). Abundance was expressed as number of individual organisms per m² (hereafter called individual density) and wet weight, including shells, was determined and reported as g per m² (hereafter called biomass density). More details on the sampling protocol and data are provided in Gomoiu (1997) and Wijsman et al. (1999). Since 2005, the Black Sea Commission recommended to use the standardised macrozoobenthos protocol described in Todorova and Konsulova (2005). This protocol was used for all the samples collected between 2008 and 2017. Briefly, samples were collected with a Van Veen grab with surface of 0.135 m² and washed through a 0.5 mm mesh size sieve. Organisms retained on the sieve were fixed with formaldehyde 4%, buffered with seawater, and finally stored in plastic jars. In the laboratory, the organisms were counted and identified to the lowest possible taxonomic level. When an identification at species level was not possible due to damage or unsolved taxonomic problems, the lowest reliable taxonomic level was given. Individual organism density was expressed in number of individuals per m² and biomass density, including shells, was expressed in wet weight g per m². More details on the sampling protocol between 2008 and 2017 are provided in Todorova and Konsulova (2005).

4.3.3.2 Taxonomic correction

Firstly, the faunal data set was checked to adopt a common nomenclature of species name aligned with the *World Register of Marine Species (WoRMS)* (accessible online: <https://www.marinespecies.org/>). We updated some taxa according to expert knowledge and the newest taxonomic research over the Black Sea. A taxonomic list with original macrozoobenthos name and our corrections is provided in the data set. In a nutshell, we applied those rules below:

- Taxa not relevant (i.e. not belonging to macrozoobenthos) to be included were removed (e.g. Chironomidae (insect)).
- Taxa relevant to be included in macrozoobenthos list were added (e.g. *Phoronis psammophila* was misclassified as meiobenthos in Gomoiu (1997) and Wijsman et al. (1999) and was added to our updated taxonomic list).
- If an identification was too unclear, we removed the taxon from the taxonomic list (e.g. identified species misidentified with juveniles from another species or presence too controversial).

- Revision of taxonomic name according to the most recent literature and expert knowledge (e.g. *Apherusa bispinosa* referred in fact to *Apherusa chiereghinii*).
- Some species were misidentified (e.g. *Polydora cornuta* misidentified as *Polydora ciliata* in 1995 (Surugiu, 2012) so *P. ciliata* occurrences were merged with the correct name *P. cornuta*).

Then, for each taxon, the species level was kept when it accounts for more than 90% of all records while the remaining records were discarded (less than 1% of the records). If for a certain taxon, less than 90% of the records was done at species level, the genus level was considered (e.g. the 3 generic records of *Abra* spp. were discarded, keeping the 103, 74 and 5 records of *Abra alba*, *Abra nitida* and *Abra segmentum* respectively).

4.3.4 Trait data

4.3.4.1 Biological trait documentation

127 taxa were documented for 27 traits that express either their response to environmental forces or their role in ecosystem functioning **Table 4.3**. The literature compiled comprised 590 sources, including 544 articles, 23 books and book chapters, 13 academic theses, 8 reports and 2 websites. Each trait was composed of several binary variables as modalities representing either intervals along a gradient (e.g. lifespan, in years, includes the modalities < 1, 1 – 3, 3 – 10 and > 10) or qualitative states (e.g. feeding type includes 4 modalities: Deposit feeding, Suspension feeding, Herbivory/Grazing and Carnivory/Scavenging). When a taxon had clearly an affinity for one or more modalities, it was attributed 1 for the corresponding modalities, 0 everywhere else.

When the information was not available at the species level, information at the genus level was considered. In some cases, we used the information at the family level when genera were synonyms. Exceptionally, opossum shrimps, partly identified below the order Mysida (that included, next to Mysida, the dominant species *Gastrosaccus sanctus* and *Paramysis pontica*, and marginally *Siriella jaltensis*), were all considered at the order level. Thus, we used information on *G. sanctus* and *Paramysis* spp. to derive traits for the order Mysida. Indeed, like in many small crustaceans (especially amphipods, isopods and hooded shrimps), opossum shrimp dwelling modes, developmental and reproductive biology are extremely homogeneous within the order (Pandian, 2016).

As shown in **Figure 4.3**, only a few trait modalities are marginally observed in the data set. Modality dominance in the overall fauna is mostly encountered for age-related traits (e.g. life span, age at maturity) and ecosystem functions (e.g. bioconstructions). This is especially the case for endo-bioconstructions, such as blinded-ended (T21.M2) or open-ended burrows (T21.M3), which are observed in a smaller proportion of records compared to the absence of such structures (T21.M1). Indeed, the construction of burrows requires a certain degree of specialisation in terms of sediment engineering and physiology (i.e. burrow construction and survival through ventilation), and are not dominant modalities in a community, as shown in a larger species pool (Beauchard et al., 2023).

4.3.4.2 Representativeness of functional information

Next to trait description, **Table 4.3** indicates trait expression. Traits expressing mainly response are primarily life history descriptors, typical of life cycle aspects (T1 to T13). Other traits expressing response such as body mass, body length, mobility and substratum depth occupancy (T14 to T17) are more related to dwelling mode with possible implications in species interactions, can also express ecosystem function. Most of the remaining traits (T18 to T27) mostly account for ecosystem functions such as bioturbation (Kristensen et al., 2012) (e.g. sediment biomixing and bioirrigation), sediment stabilisation and habitat creation (e.g. epi- and endo-bioconstruction that provide refuges); these effect traits are more specific to ecosystem engineering whereby the activity of an engineer species indirectly affects the fitness of co-occurring species through modifications of the physical environment (C. G. Jones et al., 1997); the relevance of these effect traits is supported in Beauchard et al. (2023).

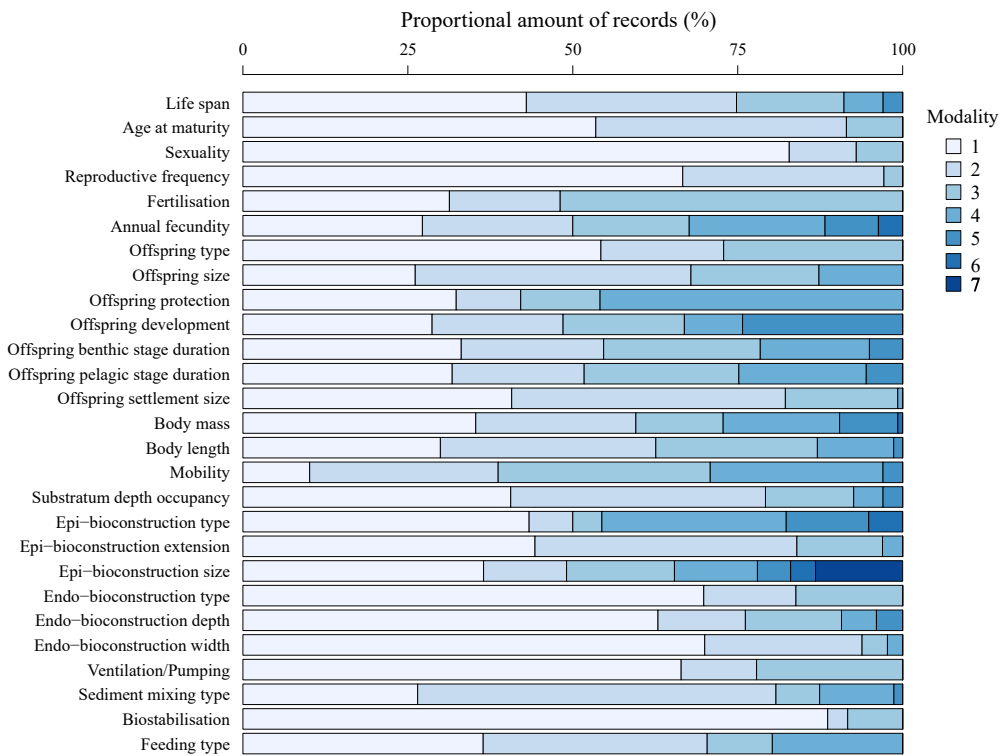


Figure 4.3: Relative amounts of records within biological traits. Modality codes correspond to those in **Table 4.3**.

Table 4.3: Detailed description of the biological traits. Each trait is labelled with a "TX.MY" notation where X denotes the trait and Y its modality. "Expression" indicates whether a trait is mainly involved in fitness (response), ecosystem function (effect) or both.

Trait Label	Trait	Modality	Description	Expression	
				R	E
T1.M1	Life span (years)	< 1	Time necessary to achieve a life cycle during which at least one reproductive success is ensured; informs also on growth rate.	+	
T1.M2		1–3			
T1.M3		3–10			
T1.M4		10–20			
T1.M5		> 20			
T2.M1	Age at maturity (years)	< 1	Time after which reproductive success can be expected; informs also on growth rate.	+	
T2.M2		1–3			
T2.M3		> 3			
T3.M1	Sexuality	Gonochorism	Gonochorism, constraint of reproductive completion, unlike homogamy. Protandry, growth constraint.	+	
T3.M2		Homogamy			
T3.M3		Protandry			
T4.M1	Reproductive frequency	Sexual seasonal	Degree of reproductive resilience.	+	
T4.M2		Sexual continuous			
T4.M3		Asexual			
T5.M1	Fertilisation	Broadcasting	Informs on proximity between genitors.	+	
T5.M2		Spermcasting			
T5.M3		Pairing			
T6.M1	Annual fecundity (number of offsprings)	< 10e2	Potential of annual demographic recruitment. Generally correlated to offspring mortality.	+	
T6.M2		10e2–10e3			
T6.M3		10e3–10e4			
T6.M4		10e4–10e5			
T6.M5		10e5–10e6			
T6.M6		> 10e6			
T7.M1	Offspring type	Egg	Offspring once released and independent from the parents. Expresses offspring survival.	+	
T7.M2		Larva			
T7.M3		Juvenile			
T8.M1	Offspring size (mm)	< 0.1	Reproductive allocation per capita.	+	
T8.M2		0.1–0.5			
T8.M3		0.5–1.5			
T8.M4		> 1.5			
T9.M1	Offspring protection	None	Expresses parental cares and offspring survival.	+	
T9.M2		Gel			
T9.M3		Capsule			
T9.M4		Bearing/Brooding			
T10.M1	Offspring development	Planktotrophic	Informs on developmental complexity, embryonic vulnerability and adult reproductive effort.	+	
T10.M2		Lecithotrophic			
T10.M3		Mixed planktotrophic			
T10.M4		Mixed lecithotrophic			
T10.M5		Internal			
T11.M1	Offspring benthic stage duration (days)	Null	Critical time on the sea floor necessary to achieve offspring development.	+	
T11.M2		< 15			
T11.M3		15–30			
T11.M4		30–60			
T11.M5		> 60			
T12.M1	Offspring pelagic stage duration (days)	Null	Critical time in the water column necessary to achieve offspring development.	+	
T12.M2		< 15			
T12.M3		15–30			
T12.M4		30–60			

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T12.M5		> 60			
T13.M1	Offspring settlement size (mm)	< 0.5	Early body size. See below.	+	
T13.M2		0.5–1.5			
T13.M3		1.5–5			
T13.M4		> 5			
T14.M1	Body mass (g AFDM)	< 0.001	Body size, amount of living tissues. Expresses well metabolic demand.	+	+
T14.M2		0.001–0.010			
T14.M3		0.010–0.100			
T14.M4		0.100–1.000			
T14.M5		1.000–10.000			
T14.M6		> 10.000			
T15.M1	Body length (cm)	< 1	Length of the main body part, not necessarily correlated to body mass. Strongly involved in space occupation, also in vulnerability to predation.	+	+
T15.M2		1–3			
T15.M3		3–10			
T15.M4		10–20			
T15.M5		> 20			
T16.M1	Mobility	Immobile	Dwelling mode, metabolic demand and species interactive potential.	+	+
T16.M2		Limited			
T16.M3		Slow			
T16.M4		Fast			
T16.M5		Very fast			
T17.M1	Substratum depth occupancy (cm)	0	Occupied layers of the sediment. Informs on ability to escape predation or disturbance.	+	+
T17.M2		0–5			
T17.M3		5–15			
T17.M4		15–30			
T17.M5		> 30			
T18.M1	Epi-bioconstruction type	None	Biogenic structure type generated on the sediment.		+
T18.M2		Mat			
T18.M3		Mound			
T18.M4		Shell			
T18.M5		Tube/Tubular protusion			
T18.M6		Protuberance/Lobe			
T19.M1	Epi-bioconstruction extension	None	Complexity of epi-bioconstruction as previously described.		+
T19.M2		Simple			
T19.M3		Flattened			
T19.M4		Complex			
T20.M1	Epi-bioconstruction size (cm)	None	Largest dimension of epi-bioconstruction as previously described.		+
T20.M2		< 1			
T20.M3		1–3			
T20.M4		3–10			
T20.M5		10–20			
T20.M6		20–50			
T20.M7		> 50			
T21.M1	Endo-bioconstruction type	None	Biogenic structure built in the sediment; Closed, only one opening. Opened, U- or Y- shaped burrow, possibly more than two openings.		+
T21.M2		Blind-ended			
T21.M3		Open-ended			
T22.M1	Endo-bioconstruction depth (cm)	None/Surficial	Size of previously described.		+
T22.M2		0–5			
T22.M3		5–15			
T22.M4		15–30			
T22.M5		> 30			
T23.M1	Endo-bioconstruction width	None/Surficial	Width of previously described. May not be necessarily related to body size, also morphology.		+
T23.M2		Narrow (< 5mm)			
T23.M3		Intermediate (5–10 mm)			

T23.M4		Wide (> 10mm)		
T24.M1	Ventilation / Pumping	None	Ability to flush water into a burrow.	+
T24.M2		Low		
T24.M3		High		
T25.M1	Sediment mixing type	None	Type of sediment particle displacement. Diffusion, random, local. Conveying, vertical, non-local. Regeneration, instantaneous downward transfer (e.g., large burrow collapse).	+
T25.M2		Diffusion		
T25.M3		Upward conveying		
T25.M4		Downward conveying		
T25.M5		Regeneration		
T26.M1	Biostabilisation	None	Ability to prevent sediment erosion.	+
T26.M2		Low		
T26.M3		High		
T27.M1	Feeding type	Deposit feeding	Informs on food resource origine and how organic matter is processed.	+
T27.M2		Suspension feeding		
T27.M3		Herbivory/Grazing		
T27.M4		Carnivory/Scavenging		

4.4 Data Records

4.4.1 Data Availability

All the data are accessible from the repository Figshare. Data sets can be freely viewed and downloaded from Figshare and users are required to cite this data paper in any resulting works (license CC-BY). Data are downloadable as six .xlsx files : (1) sampling station records, (2) model data, (3) substrate classification, (4) taxonomic correction, (5) occurrences and (6) Trait Data Black Sea; and two .csv files (UTF-8, delimited by comma): (1') traits and (2') labels. The contents of all files are detailed below. In addition, all the relevant contents relative to Supplementary information are available in one .zip file "SuppMat", also accessible from the repository Figshare.

4.4.2 Sampling station records

Each sampling station is assigned an ID number, listed in "stations.xlsx" (1). For each locality, the following is provided: spatial coordinates (latitude, longitude), sampling date (year-month-day), campaign name (see **Table 4.1**), dredge equipment used (either box corer or Van Veen grab), surface sampled (either 0.09 or 0.135 m²), mesh sieve size (either 0.5 or 1 mm). We also provide *in-situ* data (i.e. depth of sampling and substratum code).

4.4.3 Abiotic data

Model data are given in "model data.xlsx" (2) and we provide original substratum description and code classification in "substratum classification.xlsx" (3). For each variable derived from the model, its average, standard deviation, minimum and maximum are computed at trimestral scale (centred around the sampling time) for each sampling station.

Model variables can be sub-divided into two major categories: habitat descriptors related to physics (e.g. temperature, salinity) and those related to chemistry (e.g. dissolved oxygen, organic matter content). For very shallow sampling stations (i.e. below 11 m), the model cannot predict sufficiently good values due to the proximity with the land. Thus, for 12 sampling stations from the data set, all abiotic data are set as non-available (“NA”).

4.4.4 Taxonomic occurrence data

A taxonomic list with original taxa name and our corrections is provided in the file “taxonomic correction.xlsx”(4). The faunal occurrence list contains 5069 records of 215 taxa from 11 phyla among the 237 sampling stations. 260 occurrences are identified at the phylum level, 214 occurrences at the class level, 16 occurrences at the order level, 256 occurrences at the genus level and 4323 occurrences at the species level. The faunal occurrences are stored in a long format list after an appropriate selection of the taxonomic level in “occurrences.xlsx” (5). The structure of the list of occurrences is as follows:

- Sampling station information extracted from the file “stations.xlsx” (column “station” refers to station ID number, column “longitude”, column “latitude”, column “year”, column “month”, column “sampling gear”, column “sampling surface” and column “sieve”).
- Biomass (in wet weight g per m²) and individual densities (number of individual organisms per m²) for each taxonomic occurrence. The selected taxonomic level, at which the occurrence is kept, is provided in column “taxon”.

4.4.5 Trait data

Trait data are organised in three files (“Trait data Black Sea.xlsx” (6), “traits.csv” (1) and “labels.csv”(2)). The first .xlsx-file contains a taxa × traits modality matrix (Data sheet); each taxon is preceded by its respective taxonomic attributes and code for each modality of trait is provided in Labels sheet. Each specific trait is referenced for each specific taxa in sheet References. The two other files contain utility data frame to manipulate the trait data matrix (“traits.csv”) and labels (“labels.csv”). In C, an example for data ordination is given, using these two “csv”-files.

4.5 Technical Validation

We provide traits for 127 taxa out of a total of 215 (59%). The data subsets documented for traits represented 88 and 86% of total individual organism densities for the 1995 campaign and the rest of the data set (2008-2017), respectively. Considered in presence-absence, this represents 78% of total taxon presences of the entire data set since the most frequent taxa are documented. Per sampling station, this does not cause critical underrepresentation of the documented fauna (**Figure 4.4**): the larger the total number of taxa, the larger the number of documented taxa.

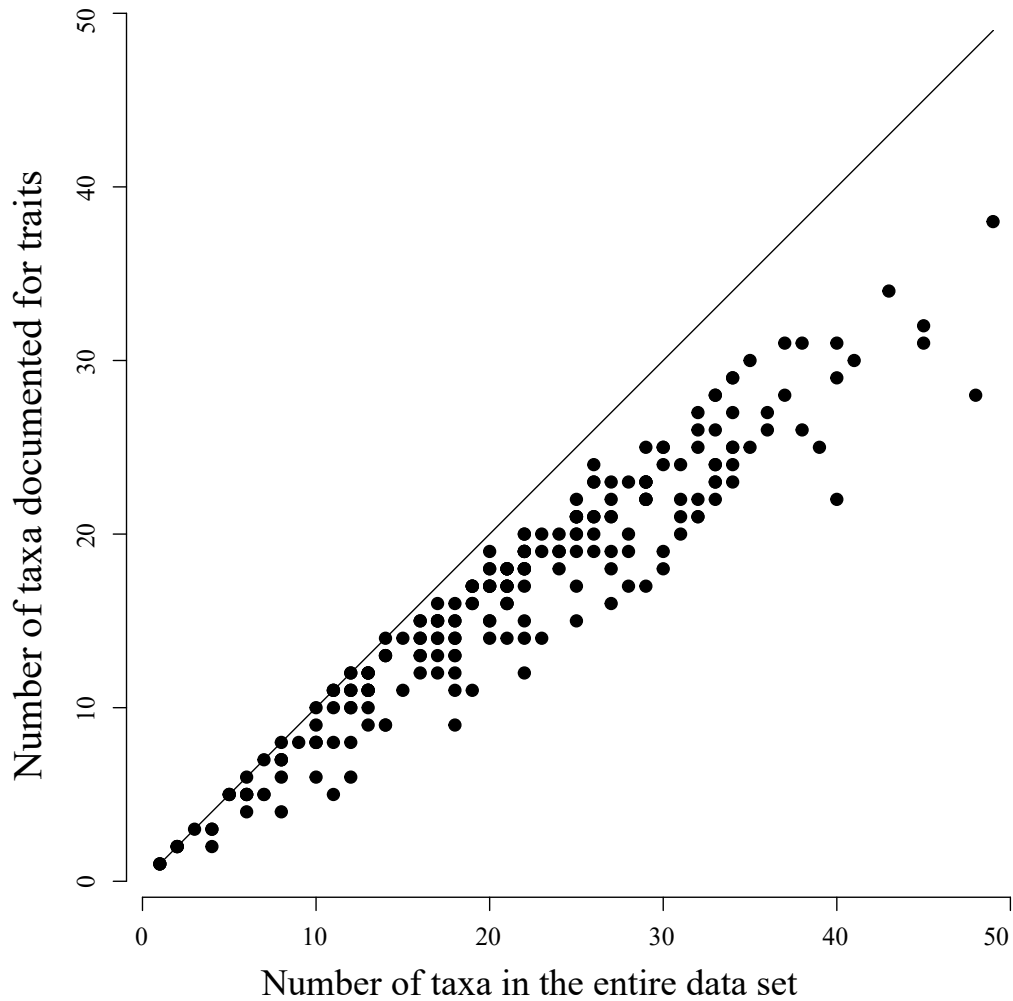


Figure 4.4: Representativeness of the taxonomic subset documented for biological traits. Despite limited numbers of taxa documented for traits (y-axis values lying below the 1:1 line), the documentation rate remains proportional to the total taxonomic richness (x-axis); $r = 0.96$, $p < 0.001$

Furthermore, we test the structural representativeness of the assemblage of the taxonomic subset documented for traits. First, for any case study, trait data are potentially relevant if species communities exhibit significant compositional variations along spatio-temporal gradients. It is a necessary condition to expect variations in species biology. Therefore, we verified significant spatio-temporal variations in community composition in the entire data set before comparing it to the subset documented for traits. With that aim, we perform a Co-Inertia Analysis (Dolédec & Chessel, 1994; Dray et al., 2003) (CoIA; **Figure 4.5**) of the full taxonomic data matrix (referred here after as the entire taxocenosis) and on the submatrix containing the taxa for which the traits have been estimated (referred here after as the trait-estimated taxocenosis). Before performing the CoIA, each taxa matrix is transformed in presence-absence matrix and is ordinated by Principal Coordinate Analysis (Jaccard's dissimilarity index (Jaccard, 1901)). The multivariate correlation (Escoufier, 1973) is strong and highly significant between the entire taxocenosis and the trait-estimated one ($RV = 0.98$, $p < 0.001$). This supports the assumption that trait composition remains unaltered in a random taxonomic subset (Beauchard et al., 2017; Bremner, 2008), especially if this later includes taxa of high occurrence frequency.

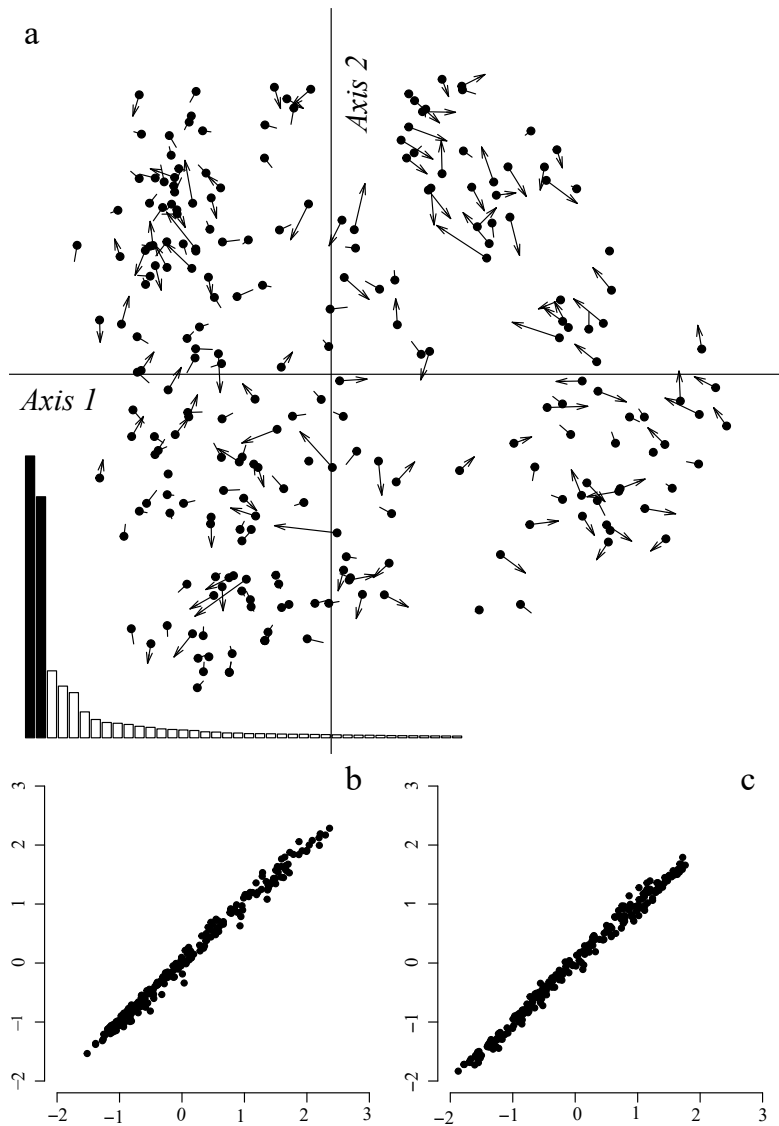


Figure 4.5: Structural representativeness of the taxonomic subset documented for biological traits: Co-Inertia Analysis (CoIA) between the 237 sampling stations \times 215 taxa matrix (full taxocenosis) and the matrix limited to the 127 taxa documented for traits. a) CoIA returns axes that maximise the co-structure between the two separate multivariate ordinations: black dots, sampling stations with the full taxocenosis; arrow tips, corresponding stations for the subset documented for traits; the longer the arrows, the higher the discrepancy between the ordinations of the two sets of station scores; barplot, eigenvalues showing two main axes (in black). b) Relationship between the first axis score of sampling stations from the ordination with the full taxocenosis (black dots in (a)) and the ordination with the subset documented for traits (arrow tips in (a)). c) Relationship between the second axis score of sampling stations from each ordination. The very short arrows in (a) evidence that the subset of taxa documented for traits well captures the overall community structure, the two axis scores in (b) and (c) being extremely close to each other. Hence, this suggests that the functional information of the entire taxonomic pool is likely well represented within the subset, especially due to frequent taxa for which trait information are not missing. Multivariate analyses performed with the ade4 package (Chessel et al., 2004; Dray & Dufour, 2007).

4.5.1 Abiotic data

Abiotic data are delivered by a numerical model used in an operational mode by the Black Sea-Monitoring Forecasting Centre (<https://marine.copernicus.eu/about/producers/bs-mfc>). The model is validated according to the skills assessment plan established in *CMEMS* using historical and *Biogeochemical ARGO (BGC-ARGO)* data sets (accessible online: <https://catalogue.marine.copernicus.eu/documents/QUID/CMEMS-BLK-QUID-007-005.pdf>). Briefly, the quality of the data is assured in *CMEMS* through a dedicated working group, while *Black Sea-Monitoring Forecasting Centre (BS-MFC)* implements state-of-the-art metrics and supports the centralised product quality dashboard (dashboard accessible online: <https://pqd.mercator-ocean.fr/>) (Ciliberti et al., 2021). *BS-MFC* product quality is based on GODAE/OceanPredict and MERSEA/MyOcean standards for the evaluation of product accuracy (Ciliberti et al., 2021).

4.5.2 Taxonomic occurrence data

Our data set is based on macrozoobenthic data previously validated and published in articles (Friedrich et al., 2014; Gomoiu, 1997; Todorova et al., 2020; Wijsman et al., 1999) and official scientific reports (available online: for SESAME project at <https://cordis.europa.eu/project/id/36949/reporting>, for HYPOX project at <https://cordis.europa.eu/project/id/226213/reporting> and for EMBLAS project at <https://emblasproject.org/publications-and-reports>). Since 2005, a standardised protocol has been used for quantitative sampling and treatment of macrozoobenthos in the Black Sea (Todorova & Konsulova, 2005). Besides, we provide taxonomic corrections to match the most recent literature and species names are aligned based on *WoRMS* database.

4.5.3 Trait data

We provide a full list of sources used to compile trait data sets in the References section. Our selected traits have been published in peer-reviewed papers (Beauchard et al., 2022, 2023; Quell et al., 2021) for other marine ecosystems.

4.6 Usage Notes

Our data set offers various research and analytical opportunities in the field of species community ecology. A particular opportunity lies in the temporal dimension of the data set that combines effects of space and organic disturbance on community composition from the 1990s to the recent period (Chevalier et al., 2024). The difference in sampling protocol between 1995 and the most recent period 2008-2017 may affect the comparison of the data (e.g. difference in mesh sieve size, sampling gear). A more detailed review of the constraint of the heterogeneous sampling designs is provided in Chevalier et al. (2024).

Briefly, we recommended to use presence-absence species data to limit the effect of different sampling protocols and to remove the Bulgarian data from *MSFD* as the Bulgarian coast was not sampled in 1995. The three objects, namely habitat descriptors, taxon distributions, and traits enable either individual or combined matrix exploratory analyses. Here, we provide some research directions based on the most widely used analytical approaches in species community ecology. A schematic representation of several analytical possibilities is provided in **Figure 4.6**.

Each table can be processed by a single ordination method to solely explore stations, habitat descriptors, species or trait compositional structure. The combination of two matrices represents another exploratory level as typically done for identifying habitats (i.e. abiotic descriptors and fauna through co-inertia or canonical analysis (Dolédec & Chessel, 1994; Dray et al., 2003). In this respect, abiotic descriptors can be converted into qualitative factors with multiple levels to circumvent the constraint of non-linear relationships between variables, in a similar way used in Beauchard et al. (2022); Chevalier et al. (2024). When focusing on stations, the data set offers the possibility to associate the temporal effect, either on individual or combined matrices through *Between-Class Analysis (BCA)* (Dolédec & Chessel, 1987; Thioulouse, 2011). Ultimately, *CoIA* can be extended to produce a simultaneous ordination of abiotic characteristics (**matrix R**), species (**matrix L**) and traits (**matrix Q**) data in the framework of RLQ analysis (Dolédec et al., 1996; Dray et al., 2014; Dray & Legendre, 2008). RLQ analysis can be also used to investigate the impact of human pressures (e.g. bottom trawling, pollution) on benthic trait composition (Beauchard et al., 2021; Hinz et al., 2021). Finally, various functional diversity indices can be computed from L and Q matrices to express trait information in a synthetic way at the species community level (Mouillot et al., 2013; Villéger et al., 2008).

In general, a fuzzy coding procedure is used to describe the affinity of a taxa for different modalities of a given trait (Chevenet et al., 1994). Indeed, fuzzy coding is the most appropriate transformation since a taxon can exhibit a non-null affinity for several modalities (Bolam et al., 2017; Usseglio-Polatera et al., 2000). For instance, a deposit feeding species receives a score profile of 1/0/0/0 while another species performing deposit and suspension feeding is scored 1/1/0/0; following fuzzy coding, these profiles become 1.0/0.0/0.0/0.0 and 0.5/0.5/0.0/0.0, respectively, hence both summing to 1 and ensuring equal species weights. An example of trait data transformation and manipulation is provided in **Appendix C**; we also give some guidelines to synthesise new traits based on the provided ones. The combination of multiple traits into a single ecological indicator is a useful tool to better understand functioning and vulnerabilities of the ecosystem (Beauchard et al., 2017; Miatta et al., 2021; Villnäs et al., 2018). Hence, our data set enables the application of a wide panel of ecological data explorations of the northwest part of the Black Sea and could be a strong support to help the implementation of monitoring and conservation programmes in the area (e.g. European *MSFD*; Council of the European Communities (1992)).

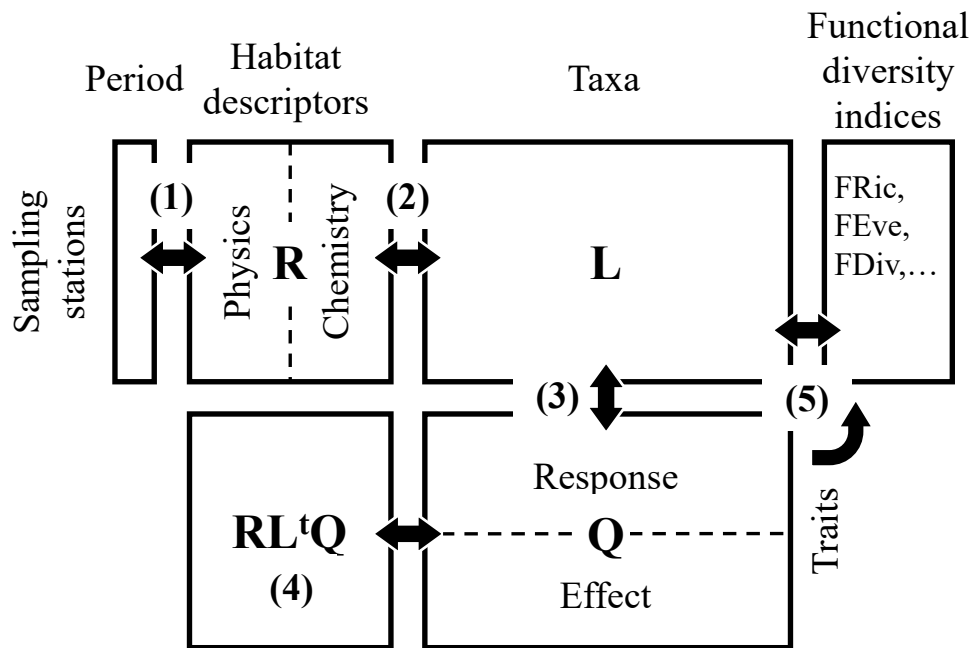


Figure 4.6: Schematic representation of the data set. Block arrows indicate how matrices (rectangles) match each other (i.e. either by the same number of rows or columns). (1) The sampling stations \times periods partition matches the R (habitat descriptors, R) and L matrices (fauna, L) for the two main periods (i.e. 1990s and 2010s). (2) R and L matrices can be combined to investigate the environmental determinants of taxonomic composition. (3) The L and Q matrices are matched by their 127 species in common; by extension, this enables the combination of the RLQ and Fourth-corner methods to relate habitat descriptors and faunal functionalities (4). Finally, functional diversity indices represent an ultimate step in community ecology (5); they can be directly derived from the combination of matrices L and Q, or indirectly by combining L with multivariate axes from Q or RL^tQ matrix ordination.

TRAIT MODELLING TO PREDICT BENTHIC FUNCTIONS AND VULNERABILITIES OVER THE NORTHWESTERN SHELF OF THE BLACK SEA

Foreword:

The compiled dataset in **Chapter 4** allows us to study the link between traits (Q) and abiotic conditions (R) through the species table (L). Even though the compilation of traits was done on a subset of all the taxa sampled and we are missing some taxa because no information was found for these species, we have highlighted that the list of taxa with documented traits considers the major species structuring communities on the northwestern shelf of the Black Sea.

The first part of this chapter focused on studying the local relationships between the three tables R (abiotic conditions), L (species) and Q (traits) using multivariate statistics and answering the third questions of this PhD thesis. Then, the list of traits and abiotic conditions with significant relationships helped to build trait distribution models based on neural networks for answering the fourth and final question of this PhD thesis. The second part of this chapter aimed to obtain maps of traits at the shelf scale based on the local relationship between traits and abiotic conditions and the maps of abiotic conditions from an ocean 3D model. Two graphical summaries of this chapter are provided in **Figure 5.1** for the first part and **Figure 5.2** for the second part. Supplementary materials associated with **Chapter 5** are available in **Appendices D, E and F**.

This Chapter is modified from a paper published as : Chevalier, S., Beauchard, O., Vandembulcke, L., Teacă, A., Begun, T., Todorova, V., K. Soetaert., & Grégoire, M. Trait modelling to predict benthic functions and vulnerabilities across Black Sea seascapes. *Scientific Reports*, 15(1), 39076. <https://doi.org/10.1038/s41598-025-24508-4>.

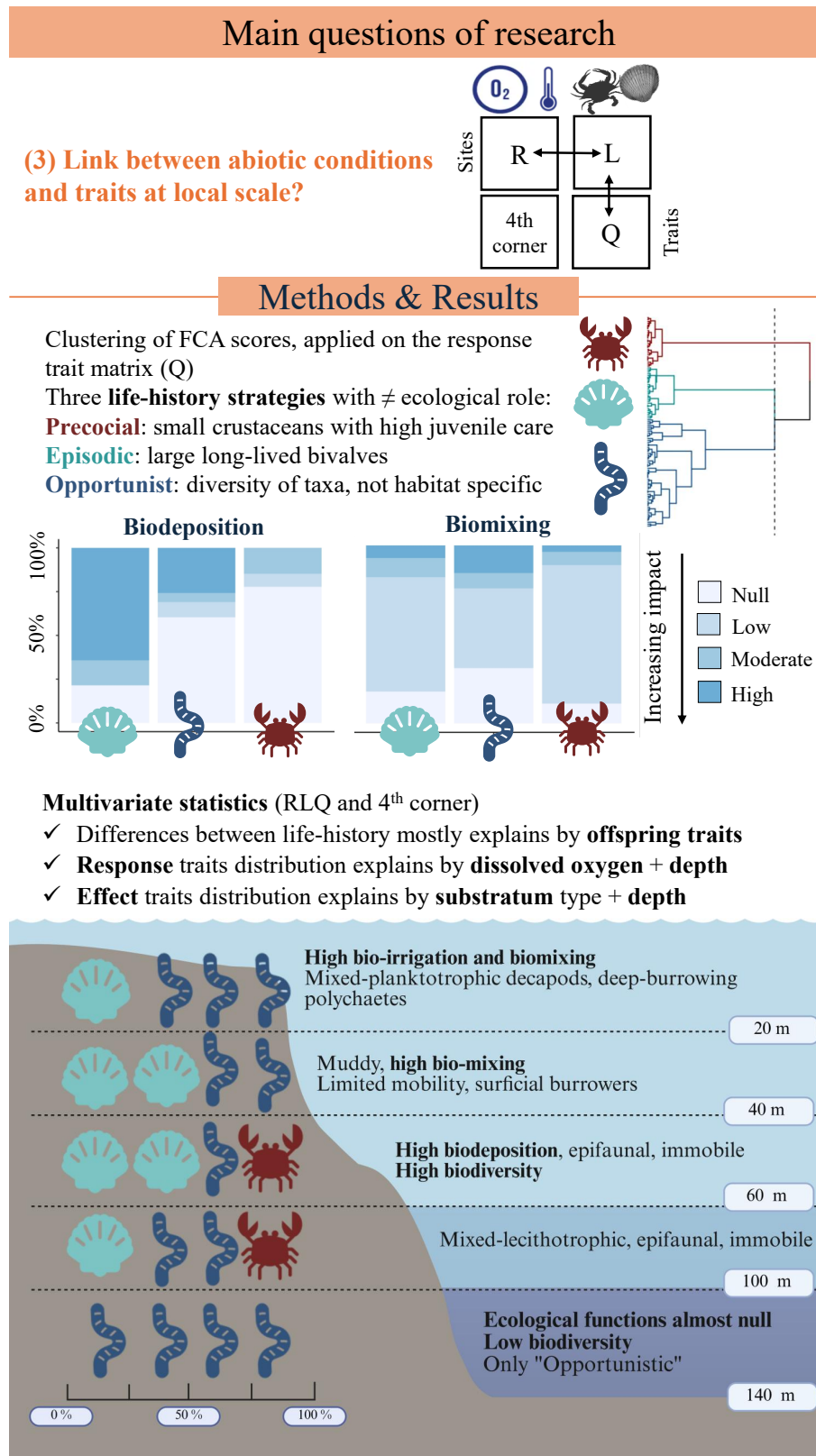


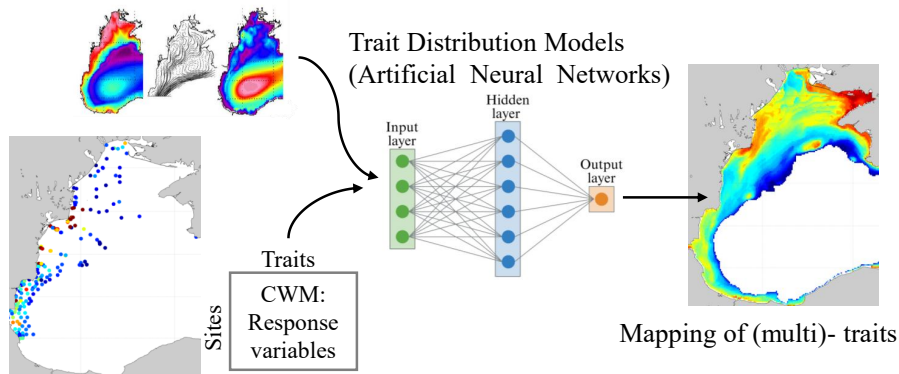
Figure 5.1: Graphical Abstract of the first part of the Paper III (Chevalier et al., 2025b): this first part of **Chapter 5** aims to answer the third question of this PhD thesis.

Main questions of research

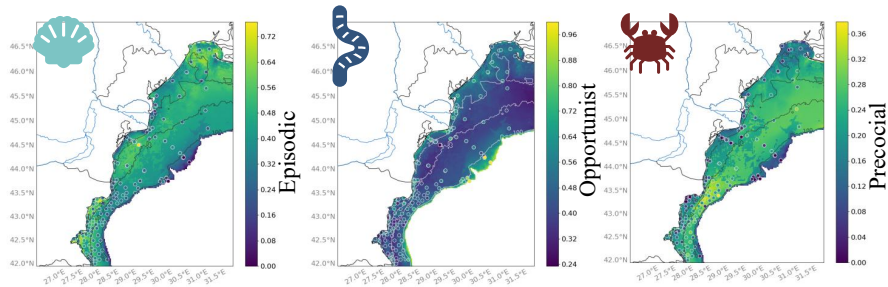
(4) How to predict the spatial distribution of traits at shelf scale ?

Methods & Results

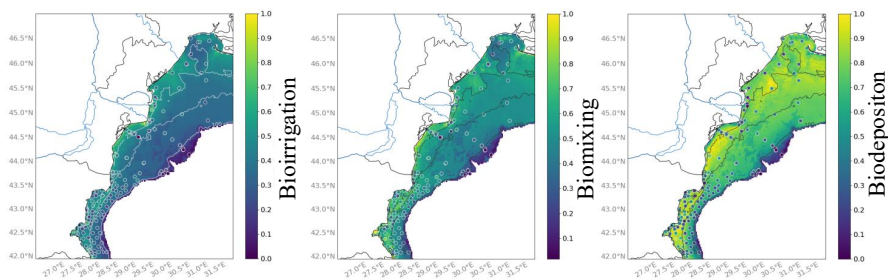
Predictors: bottom abiotic maps



Mapping of vulnerabilities of benthic communities



Mapping of ecological processes



Key messages

- ✓ Need maps of benthic vulnerabilities and functions for **sustainability development**
- ✓ Add the role of macrobenthos into ocean models thanks to the maps of functions for **ecosystem functioning** and services **evaluation**

Figure 5.2: Graphical Abstract of the second part of the Paper III (Chevalier et al., 2025b): this second part of **Chapter 5** aims to answer the fourth question of this PhD thesis.

5.1 Abstract

Benthic biodiversity is of global significance for the provision of ecosystem services and the mediation of global biogeochemical cycles. The lack of detailed spatial distributions of the functions and vulnerabilities of the benthos critically prevents us from protecting benthic biodiversity and its functioning in the context of increasing human perturbations and climate change. Here, we propose a multidisciplinary approach to bridging *in-situ* benthic data to the maps of macrobenthic functions and vulnerabilities at the scale of the northwestern shelf of the Black Sea. Our findings show that oxygen availability is a key driver of the functional trait composition of macrozoobenthic communities. Shallower well-oxygenated areas support high biomixing and bioirrigation on muddier-sandier substrata and high biodeposition on coarser substrata associated with mussel reef communities. In contrast, at depleted oxygen areas at the edge of the shelf, macrobenthic communities are functionally impoverished with only a combination of a few typical opportunistic traits, and those communities have a negligible impact on ecosystem functions. Mapping of vulnerabilities and functions of benthic communities can support marine management strategies aligned with the United Nations Sustainable Development Goal 14: Life Below Water.

5.2 Introduction

The macrozoobenthic community that inhabits sediments has an important impact on sediment geochemistry and on global biogeochemical cycles. Through their movement, feeding, respiration and burrowing activities, benthic animals rework the sediment and move water (ventilation) in a process termed bioturbation (Kristensen et al., 2012). Bioturbation impacts the upward and downward transport of particles through biomixing and increases the exchange of solutes with the overlying water through bioirrigation (Kristensen et al., 2012; Meysman et al., 2006). Through these ecological processes, macrofauna alter the physical and chemical properties of sediments, modify microbially driven reactions, influence the resources for other biota and promote the oxygenation of the sediment, leading to greater mineralisation and nutrient turnover (R. C. Aller, 1994; Kristensen, 2000; Mermillod-Blondin, 2011). Filter-feeders can also play a role in enriching the sediment by filtering particles from the overlying water and excreting non-digestible materials such as faeces and pseudofaeces (i.e. biodeposits) at the sediment-water interface (Norkko et al., 2001). In this process, termed biodeposition, carbon and nitrogen-enriched deposits provide food for other organisms and stimulate microbial production (Lopez & Levinton, 1987).

At the scale of the ecosystem, bioturbation by benthic macrofaunal communities has a great impact on marine ecosystem functions such as the recycling and storage of organic carbon and other elements (e.g. removal of nitrogen through denitrification) in shallow coastal habitats (e.g. R. C. Aller (1982); Braeckman et al. (2010); Kristensen et al. (2012); Meysman et al. (2008)).

The benthos is thus a dominant actor in biogeochemical cycles and ecosystem functioning and is currently threatened by global environmental changes and human activities at sea. Climate change is foreseen to cause major changes in marine habitats, through increased water temperature, ocean acidification and the expansion of oxygen minimum zones, all of which will alter the health of benthic life, threatening its functioning and its ability to underpin the delivery of ecosystem services (Díaz et al., 2006). The removal of ecosystem engineers, such as bioturbators, could induce large changes in the structure of the habitat, with cascading effects on local biodiversity and impairing the macrozoobenthos contribution to ecosystem functioning (Bianchi et al., 2021; Lohrer et al., 2004; Olsgard et al., 2008).

The vulnerabilities of benthos to disturbances and their role in ecosystem functioning can be estimated by the traits of the species. Species traits are the morphological, phenological and physiological characteristics of a species (Díaz & Cabido, 2001, 2001; McGill et al., 2006; Violle et al., 2007). According to the Southwood's habitat templet theory (Southwood, 1977), species traits are strongly related to the environment more than the species composition. This trait-environment link can be used to develop trait distribution models to upscale very sparse measurements in the benthic system across ecosystem scale. Benthic functional maps are essential for conducting integrated ecosystem assessments (Cogan et al., 2009) and bridging *in situ* local data to the scale of ecosystem-based management (Snelgrove et al., 2014; Bianchi et al., 2021).

However, the mapping of ocean biology and, in particular, of benthic biology, is still far behind that of terrestrial systems (Brown et al., 2012; Tittensor et al., 2010). In terrestrial ecology, the development of correlative bioclimatic models or niche-based ecological models that assess the distribution of a given species based on the distribution of the environmental variables to which it is suited has been accelerated by the necessity of assessing the effect of climate change on biodiversity (Morin & Thuiller, 2009). This development has also been favoured by the significant progress in our ability to observe the Earth's surface with increased spatial and temporal resolution (Franklin, 2013; Guisan & Zimmermann, 2000). However, benthic system is still clearly undersampled, especially for biological and environmental data (Costello et al., 2010; M. Townsend et al., 2014). Benthic sampling is strongly heterogeneous and it is very rare to have the necessary data to develop mapping tools (i.e. biological and environmental data sampled at the same place and time to assess the species niche (Rodríguez et al., 2007)). This lack of data has greatly prevented regular mapping of the benthic system. There are still very few maps of the benthos species, functions and biodiversity (Brown et al., 2012; Gerovasileiou et al., 2019). This limitation hampers sound management and modelling of the benthic system in relation to climate change and biodiversity protection (Gignoux-Wolfsohn et al., 2024). As a result, current ocean numerical models ignore the variability of life on the seafloor and its impact on biogeochemical cycles and ecosystem functioning (Bianchi et al., 2021; Snelgrove et al., 2014). In recent years, there has been a change in the paradigm concerning the availability of environmental data for the ocean (Le Traon et al., 2019).

Since 2014, *CMEMS* has provided daily 10-day forecasts and 20-30 years of reanalysis of the physical and biogeochemical states of the global and Arctic oceans and European regional seas (marine.copernicus.eu). These modelled environmental data combined with in situ data can remedy the lack of available environmental maps to support the development of mapping methods for the benthos.

In this paper, we study the relationships between traits of macrozoobenthos and environmental conditions. Relevant functional traits are selected, and their connections with environmental conditions, both *in-situ* measured and modelled, are analysed. This analysis is done for individual traits and for combinations of traits that define empirical indicators of ecological processes or vulnerabilities. The trait-environment relationships are investigated at the local scale and are subsequently scaled up to obtain maps at the ecosystem scale of individual traits and derived indicators of ecological functions. These maps can be used to support management strategies (e.g. definition of *Marine Protected Areas (MPAs)*) and the inclusion of benthic life characteristics in ocean numerical models. This study was performed on the northwestern shelf of the Black Sea, which features an important gradient in species trait composition and environmental conditions due to the large river discharge and the strong gradient in bottom-water oxygen concentrations.

5.3 Methods

5.3.1 Study area

The northwestern shelf of the Black Sea is the most productive region of the basin thanks to the very large delivery of rivers nutrients per unit of shelf area (Capet et al., 2016). The pelagic and benthic systems are tightly coupled, with approximately 30% of the shelf-produced primary production degraded in the sediment for 40% of the shelf oxygen consumption (Capet et al., 2013). The shelf acts as a filter for land and river inputs, with about half of the riverine nitrogen input to the shelf lost in the shelf sediment by denitrification and burial (Friedrich et al., 2002). Although it is recognised that the northwestern shelf plays a crucial role in ecosystem functioning and biogeochemical budgets at the basin scale, the impact of benthic life variability on shelf functions is still poorly understood. The gradients of bathymetry, bottom substratum, and environmental conditions from the coastal area to the shelf break are expected to affect the distribution of benthic functions. For instance, in Chevalier et al. (2024), we evidenced that the strong gradient of bottom dissolved oxygen from oxic shallow waters to deep euxinic waters (i.e. no oxygen and a high content of hydrogen sulphide), affects the macrozoobenthic biodiversity.

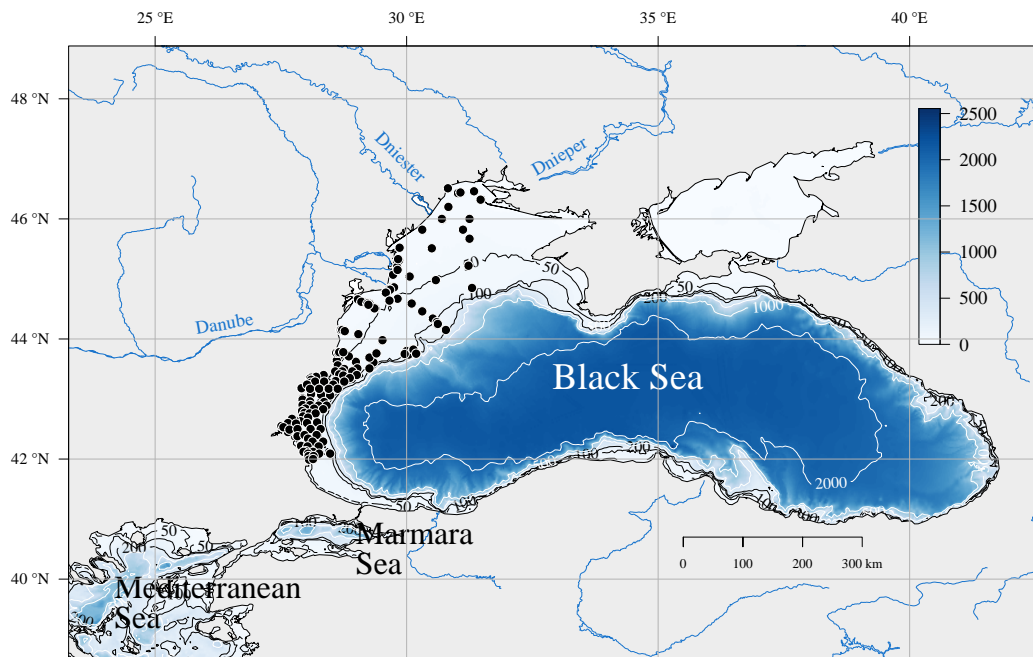


Figure 5.3: The Black Sea. Named rivers as the main sources of freshwater inputs; black dots represent the sampling stations from 2008 to 2017. Contour lines and colour bar denote water depth in meters.

5.3.2 Data collection

5.3.2.1 Species collection

We compiled a set of macrobenthos data collected between 2008 and 2017 from 210 sampling stations across the entire northwestern shelf (**Figure 5.3**). The spatial distribution of the sampling sites covers regions of contrasted environmental conditions and consequently, a variety sea floor functionalities may be expected over the shelf. The data were collected according to the standardized macrozoobenthos protocol described in Todorova and Konsulova (2005) during all these campaigns. Macrobenthic species were sampled with a Van Veen grab with a surface area of 0.135 m² and washed through a 0.5 mm mesh sieve. In the laboratory, the organisms were counted and identified to the lowest possible taxonomic level. Individual organism densities were expressed in number of individuals per m² and were $\log_{10}(x + 1)$ transformed to down-scale large values. More details about the species collection data can be found in Chevalier et al. (2024, 2025a).

5.3.2.2 Abiotic data

We selected physical and bio-geochemical descriptors based on those expected to influence macrozoobenthos distribution (**Table 4.2**). We combined *in situ* data with data from a model reanalysis delivered by the *CMEMS* Black Sea Marine Forecasting Center (Ciliberti et al., 2021). These data were produced by the coupled hydrodynamical - biogeochemical *NEMO 4.2 BAMHBI* model run in reanalysis mode at a resolution of 2.5 km with 59 vertical levels (Grégoire et al., 2026).

We derived the mean, maximum, minimum, and standard deviation for each abiotic variable from a 10-year climatology covering the period of the data collection (i.e. 2008-2018). Coefficients of variation (i.e. the ratio between the standard deviation and the corresponding mean value) were evaluated for temperature and oxygen concentration to assess thermal and oxygen seasonality. Then, we extracted model values at each sampling station (i.e. matrix sampling station \times abiotic conditions). Our list of abiotic predictors (**Table 4.2**) focuses on a variety of environmental variables expected to influence the distribution of species traits and their derived indicators.

5.3.2.3 Trait compilation

Among the 206 taxa, we used a subset of 124 taxa for which documentation was available (Chevalier et al., 2025a). We selected effect traits with an impact on sediment mixing and nutrients cycling (**Table 5.1**) and response traits important for the resilience of benthic communities (see the traits with a response expression in **Table 4.3**). Each trait was divided into several modalities representing either intervals along a gradient (e.g. lifespan, in years) or qualitative states (e.g. feeding type). When a taxon had clear affinity for one or more modalities of a given trait, it was attributed 1 for the corresponding modalities, 0 everywhere else. When the information was not available at the species level, information at the genus level was considered. More details about trait compilation can be found in Chevalier et al. (2025a).

5.3.3 Building ecological indicators

5.3.3.1 Indicators based on effect traits

The contribution of each species to an ecological process was estimated via indicators, meaning a trait derived from a formula incorporating several traits (Beauchard et al., 2017). Here, we chose the indicators defined in Beauchard et al. (2023) for four ecological processes that play a key role in the mediation of the benthic-pelagic coupling: biomixing (Equation 5.1), bioirrigation (Equation 5.2), biodeposition (Equation 5.3) and deposit feeding (Equation 5.4). Each indicator is a combination of traits that influence ecological processes. Within a trait, a score was assigned to each of the modalities, with an increasing score indicating an expected greater effect on ecosystem functioning (**Table 5.1**). Then, for each species, the indicators were calculated as the product of their modality scores following the respective formula given in **Table 5.2**. The resulting trait-based indicator was rescaled between 0 and 1 for each species. An example is provided in **Appendix D**. More details on indicators can be found in Beauchard et al. (2023).

Table 5.1: The selected effect traits are listed with their units, abbreviations, modalities and scores that indicate the contribution of the modality to the intensity of the selected ecological processes defined in **Table 5.2**. Description of the traits can be found in **Table 4.3**.

Trait [Unit] (Abbreviation)	Modality	Score
Body Mass [g AFDM] (<i>BM</i>)	< 0.001	0.17
	0.001–0.010	0.33
	0.010–0.100	0.5
	0.100–1.000	0.67
	1.000–10.000	0.83
	> 10.000	1
Feeding type (<i>FT</i>)	Deposit feeding (De)	0.25
	Suspension feeding (Su)	0.25
	Herbivory/Grazing (HeGr)	0.25
	Carnivory/Scavenging (CaSc)	0.25
Substratum depth occupancy [cm] (<i>SD</i>)	0	0.2
	0–5	0.4
	5–15	0.6
	15–30	0.8
	> 30	1
Mobility (<i>MB</i>)	Immobile	0.2
	Limited	0.4
	Slow	0.6
	Fast	0.8
	Very fast	1
Sediment mixing type (<i>AF</i>)	None	0
	Diffusion	0.25
	Upward conveying	0.25
	Downward conveying	0.25
	Regeneration	0.25
Ventilation/Pumping (<i>VP</i>)	Null	0.33
	Low	0.66
	High	1
Endo-bioconstruction type (<i>BT</i>)	None	0.33
	Blind-ended	0.66
	Open-ended	1
Endo-bioconstruction depth [cm] (<i>maxBD</i>)	None/Surficial	0
	0–5	0.25
	5–15	0.5
	15–30	0.75
	> 30	1

Table 5.2: Indicator of ecological process at the species level with their definition and formula. Details on each trait used in the formula can be found in **Table 5.1**. The biomixing potential (mi) is the sum of the four main categories of biomixing: diffusion, upward and downward conveying, and regeneration.

Indicator	Definition	Formula
Biomixing (mi)	One of the two processes of bioturbation through particle reworking (Kristensen et al., 2012)	(5.1) $mi = AF \times BM \times MB \times SD$
Bioirrigation (ir)	One of the two processes of bioturbation through ventilation (Kristensen et al., 2012)	(5.2) $ir = VP \times BM \times BT \times maxBD$
Biodeposition (de)	Benthic-pelagic transfer through suspension feeding (Graf & Rosenberg, 1997)	(5.3) $de = BM \times Su$
Deposit feeding (Dep)	Ingestion of large volumes of sediment (Lopez & Levinton, 1987)	(5.4) $Dep = BM \times De$

5.3.3.2 Indicators based on response traits

Life history strategies involve species that have similar abilities to withstand mortality at the juvenile and adult stages according to the POSE concept, which differentiates four main groups: **Precocial**, **Opportunist**, **Survivor** and **Episodic** (Kindsvater et al., 2016; Winemiller & Rose, 1992). This framework extends the classic two-strategy life-history framework of r-K strategies for respectively for fast and slow life histories (Pearson & Rosenberg, 1978; Pianka, 1970). Briefly, a first axis separates fast (precocial and opportunist) and slow (survivor and episodic) life histories, and the second axis separates high (opportunist and episodic) and low (precocial and survivor) juvenile mortality rates (**Figure 2.11**). In our study, each species was assigned to one of the four life history strategies groups, based on their trait modalities (**Table 4.3**). Firstly, *Fuzzy Correspondence Analysis (FCA)* was applied to the trait matrix. *FCA* is an ordination method that uses eigen analysis to investigate differences between samples, based on the biological traits exhibited by species present in the assemblages (Bremner et al., 2006b). Then, life history strategy groups were derived using hierarchical clustering. Ward's aggregation criterion (Murtagh & Legendre, 2014) was applied to species scores from the two first axes of the *FCA*. More details on the methodology can be found in Beauchard et al. (2022).

5.3.4 From species to community scale

5.3.4.1 Community weighted-mean traits

The *Community Weighted-Mean (CWM)* is a weighted average of the trait modalities of all the species at the scale of a community (Lavorel et al., 2008). For response traits, we used relative mean values (i.e. species densities relative to the total density of the community).

For effect traits, we kept the absolute means of a trait, as an expected greater effect on ecosystem functions is expected with higher individual densities. We computed the CWM for selected traits with the “FD” library in R (Laliberté et al., 2014).

5.3.4.2 Community-level indicators

The biomixing, bioirrigation, biodeposition or deposit feeding potentials of the community, were obtained by summing across the species of the community the potential of individual species (i.e. individual density of the species multiplied by the species score). The community level potential is then scaled to be between 0 and 1. For the life history group, species densities were expressed as the ratio of density to the total density of a site. Then, relative species densities are summed up by group within each community.

5.3.5 Local trait-environment relationships

Our data set is composed of three matrixes: a matrix (R) that contains the sampling stations (rows) and the abiotic descriptors (columns); a matrix (L) that contains the sampling stations (rows) and the taxa matrix (columns); and a matrix (Q) that contains the taxa (rows) and the traits or indicators (columns). We used an RLQ analysis in combination with the fourth-corner approach to identify the environmental variables that are significantly correlated with the traits or indicators (Dray et al., 2003; Dray & Dufour, 2007; Dray & Legendre, 2008; Legendre et al., 1997).

Before the RLQ analysis, an exploratory analysis was performed to select the most relevant environmental descriptors and traits. Traits and abiotic variables were tested according to the fourth-corner method combined with RLQ (Dray et al., 2014). This preliminary procedure is used to increase the global significance of trait-environment relationships. We used the function “fourthcorner2” from ade4 package in R (Dray & Dufour, 2007). The fourthcorner2 function is a multivariate statistic equal to the sum of eigenvalues of RLQ analysis. It measures the link between two variables by a square correlation coefficient (Legendre et al., 1997). For the preliminary test, we used the model 6 (i.e. combination of the outputs of models 2 and 4), proposed in Dray and Legendre (2008) and 9 999 random permutations for preliminary tests. The results of these preliminary tests are presented in **Appendix E**.

We kept the abiotic factors likely correlated with the trait matrix and the traits likely correlated with the abiotic data. After an appropriate selection of traits and environmental descriptors, we applied an RLQ analysis with the new tables (reduced) R and Q. As a first step, we applied separate univariate analyses on the three tables: *MCA* on environmental table R (using qualitative abiotic data as in Chevalier et al. (2024) and Beauchard et al. (2022) ; *Correspondence Analysis (CoA)* on the species table L; and *FCA* on the fuzzy-coded trait data table (Q) or *Principal Component Analysis (PCA)* if traits are quantitative. Separate analyses of traits and environmental variables were weighted respectively by the species and site weights derived from the *CoA* (Dray & Dufour, 2007).

The three separate analyses were subsequently combined in the RLQ analysis framework, which maximises the covariation between environmental factors and traits (Dray et al., 2014) (**Figure 5.4**). The degree of associations between trait and environmental tables was quantified by RLQ inertia equal to the fourth-corner statistic from Dray and Legendre (2008). The significance of the total inertia of the RLQ analysis was tested by 49 999 row permutations of table L to break with the link with Table R (i.e. Model 2) and column permutations to break the link with table Q (i.e. Model 4). Relationships were significant when the p -values of both model tests were lower than the fixed threshold of 0.05 (Ter Braak et al., 2012). A p -value adjustment was performed following the false discovery rate method (Benjamini & Hochberg, 1995). Then, we tested the correlation between the RLQ axes and individual traits or abiotic conditions, as proposed in Dray et al. (2014). We used 49 999 random permutations. Directly testing the associations between RLQ axes and traits/environmental variables improved the interpretability of the RLQ results. The RLQ and fourth-corner analyses were performed using the ade4 package for R software (Chessel et al., 2004; Dray & Dufour, 2007).

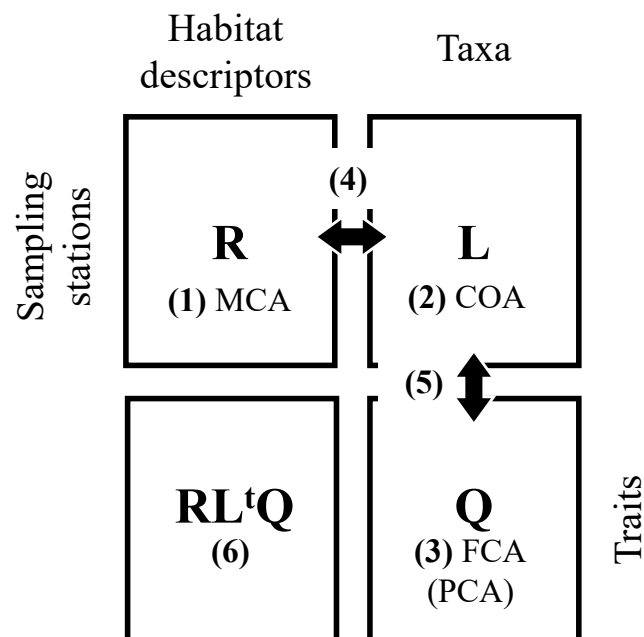


Figure 5.4: Schematic representation of the data set. Block arrows indicate how matrices (rectangles) match each other (i.e. either by the same number of rows or columns). Single ordination techniques are applied on the three tables (from 1 to 3), double arrows (4 and 5) are for the ordination of two tables. These five multivariate analyses are summarized through RLQ and Fourth-corner analysis (6). Figure adapted from Dolédec et al. (1996)

5.3.6 From community to ecosystem scale

Upscaling from the community level to the ecosystem was done using *Trait Distribution Models (TDMs)*. The trait-environment relationships were identified at the local scale, and then, using environmental conditions delivered by a reanalysis (*CMEMS* here), the traits were mapped at shelf scales using a neural network. The methodology used to investigate local trait-environment relationships is first described below, followed by the neural network and spatial prediction. We provided a summary of our methodology for both ecological processes and life history strategies in **Figures 5.5** and **5.6**, respectively.

5.3.6.1 Trait Distribution Modelling

Based on the general trend derived from multivariate analysis, we adapted *TDMs* to detect local relationships between community weighted-mean indicator values and a suite of abiotic descriptors measured at the same sampling station. In our study, we chose to use a neural network for our *TDMs*, in response to a growing interest in machine learning tools in ecology and their ability to resolve complex spatial and temporal non-linear patterns in oceanography (C. Dong et al., 2022; Lou et al., 2023; Rubbens et al., 2023; Song et al., 2022). We used the scikit-learn library for the Python programming language (Pedregosa et al., 2011). Our *TDMs* do not take into account biotic interactions such as competition and predation, which are important for refining predictions of species distributions and abundances (Green et al., 2022).

To generate spatial maps of traits, based on discrete data at a limited number of stations, we used a feedforward neural network comprising an input layer, 2 hidden layers with 12 neurons each, using a leaky ReLU activation function, and an output layer, leading to 601 trainable parameters. The method was implemented using Keras software with the Torch backend. The hyperparameters were tuned by splitting the station data into a training set (50% of the data), a validation set (30%) and a test set (20%) and repeating the training 50 times. The score was the mean squared error between the real output and the one predicted by the neural network. Once trained, the network was then applied to each pixel of the domain to infer the traits. The input data were the bottom type (mud, sand, or gravel), the depth, and the climatologic mean, standard-deviation, minimum and maximum for the variables listed in **Table 4.2**, obtained from the Copernicus Marine Black Sea Reanalysis and computed using the coupled *NEMO 4.2 BAMHBI* physical-biogeochemical model. All the inputs were normalised. More details about the error metrics associated with the constructed neural networks can be found in **Appendix F**. Maps were produced with the Cartopy package (v0.24.1) in the open-source software Python 3.12. The Python software is available from <https://www.python.org/downloads/> and the Cartopy package is available at <https://scitools.org.uk/cartopy>.

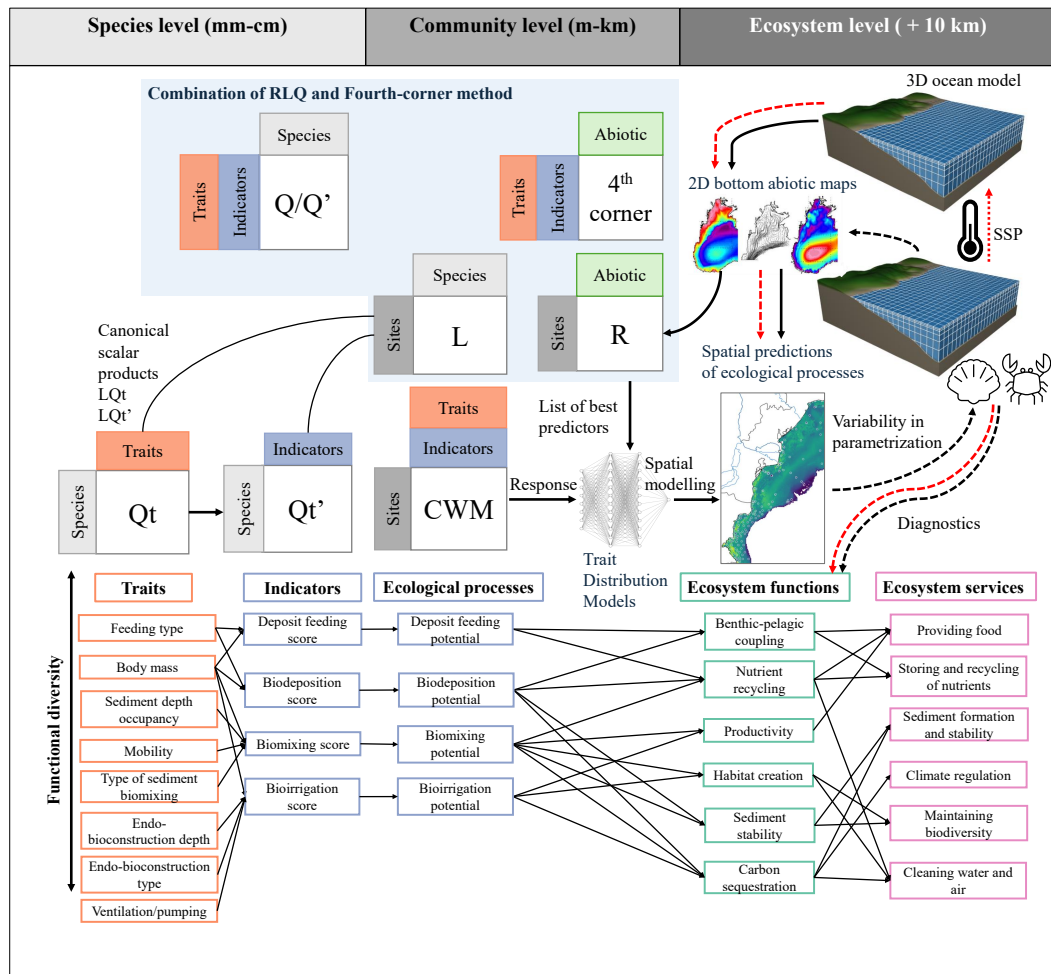


Figure 5.5: Upscaling from the species level to ecosystem services. The list of effect traits considered in this study related to ecological processes at the community scale and to ecosystem functions and ecosystem services at a larger scale. This paper mainly focuses on the link from species traits to the spatial distribution of ecological processes at the shelf scale (i.e. maps). Here, black dashed arrows indicate the potential question of research, and red dashed arrows indicate projections in the context of global change. Figure adapted from Snelgrove et al. (2014). Model grid from comet.

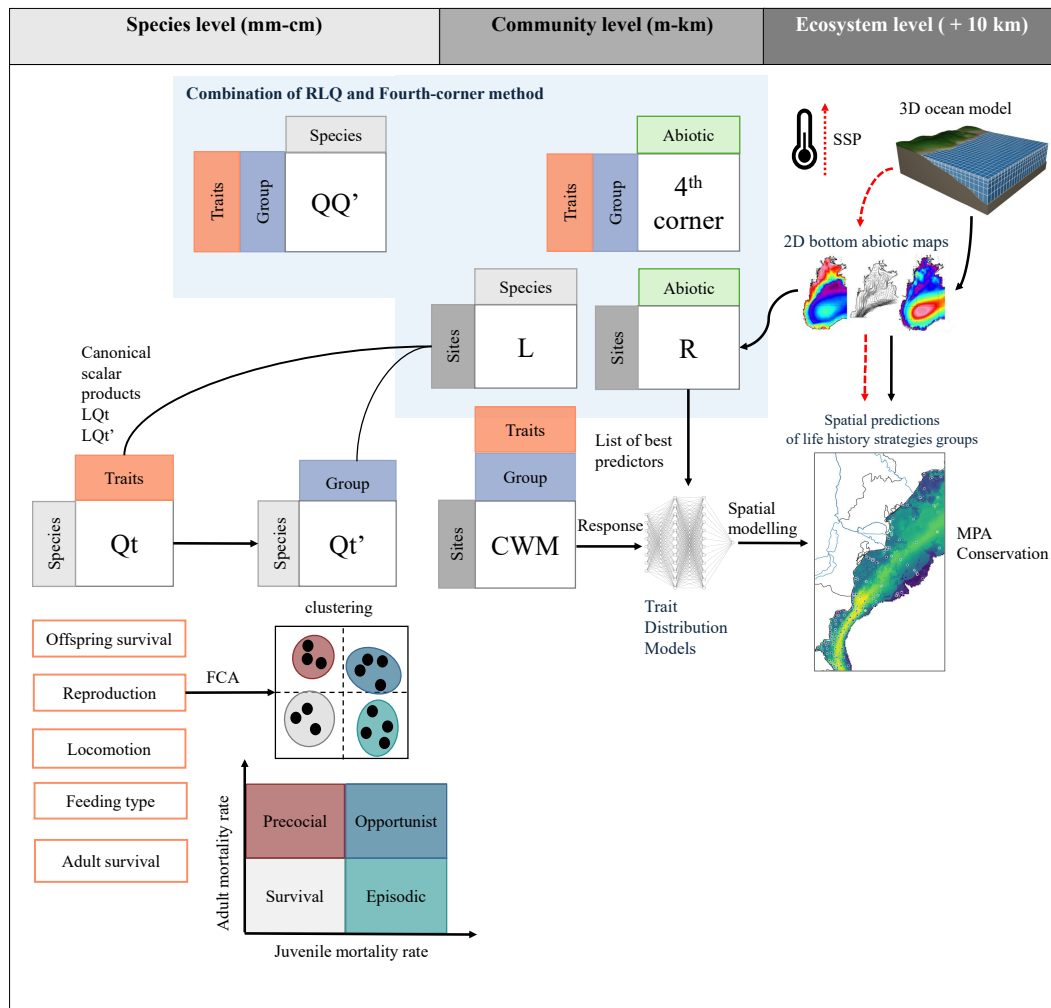


Figure 5.6: Upscaling from the species level to the maps of life history strategies for conservation and MPAs management. The definition of life history groups is based on the concept *Precocial-Opportunist-Survivor-Episodic (POSE)* (Kindsvater et al., 2016). Here, the red dashed arrows represent projections in the context of global change. Figure adapted from Snelgrove et al. (2014). Model grid from comet.

5.4 Results

5.4.1 Functional diversity

5.4.1.1 Life history groups

Our study identifies three dominant life history strategies based on response traits: “P”- precocial with short life cycle and high offspring survival probability, “O”- opportunist with a relatively short life span and low offspring survival probability and “E”- episodic with a longer longevity for achieving a minimum of reproductive success. **Figure 5.7** presents the ordination of species based on their response traits. Three main groups of life history strategies are derived: “O”- opportunistic (dark blue), “E”-episodic (dark cyan) and “P”-precocial (dark red). The first axis separates the “P” and “E” strategies, and the second axis separates the “O”-strategists from the “P” and “E”-strategists (**Figure 5.7**).

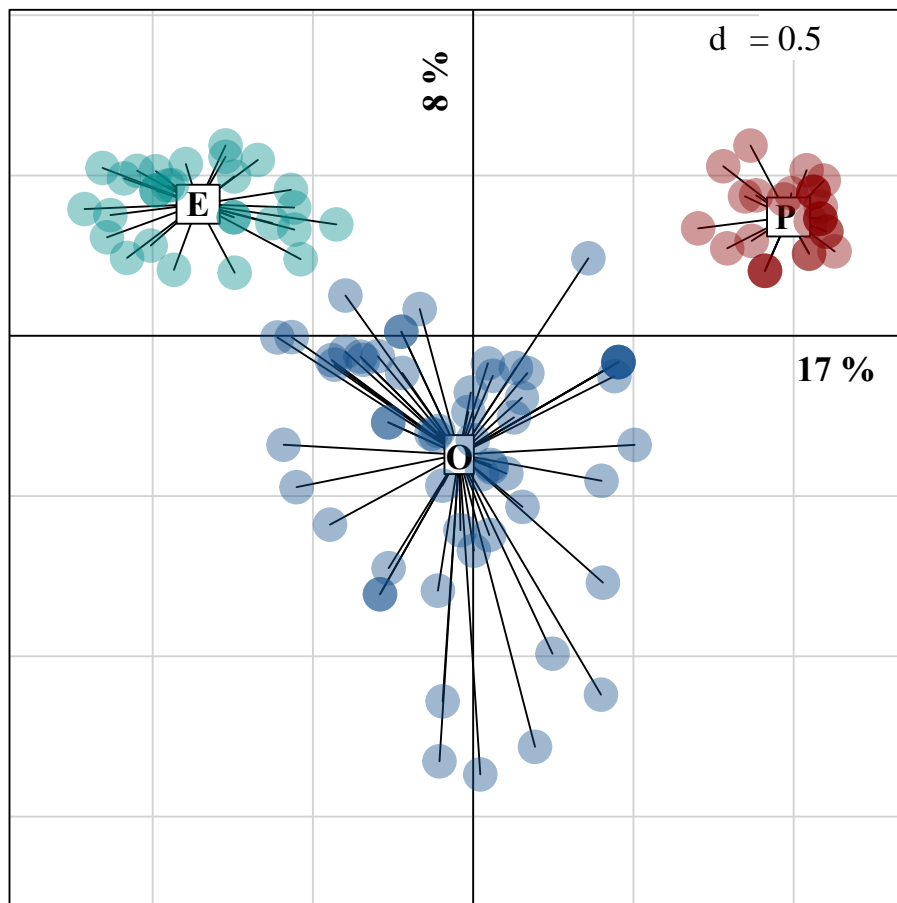


Figure 5.7: Ordination of species based on their response traits using *FCA*. Species scores on the first axis (x) and second axis (y) of the *FCA*. Species are represented by dots; coloured dots correspond to a life history strategy group (dark red: “**Precocial**”, dark blue: “**Opportunist**” and dark cyan = “**Episodic**”). Size of the grid in the top-right corner. The first two axes of the *FCA* explained 25% of the total inertia, with 17% for the first and 8% for the second axes.

The traits contributing most to the axes were identified by their correlation ratios (*Correlation Ratio* (*CR*), **Table 5.3**). Traits linked to reproductive (e.g. annual fecundity) and offspring characteristics (e.g. offspring size, protection, development) discriminated the most species according to their differences in traits.

Table 5.3: *CR* between the first and second axes of the *FCA* and functional traits. *CR* returns the proportion of explained variance of the axis species scores by the discrimination of the trait modalities. High correlation ratios along the first axis and/or second axis meaning that the trait is an important factor determining the difference between species based on their biological attributes.

Traits	CR-axis 1	CR-axis 2
Life span	0.44	0.07
Age at maturity	0.34	0.04
Sexuality	0.00	0.34
Reproductive frequency	0.33	0.03
Fertilisation	0.62	0.50
Annual fecundity	0.72	0.25
Offspring type	0.72	0.52
Offspring size	0.75	0.41
Offspring protection	0.66	0.12
Offspring development	0.87	0.74
Offspring benthic stage duration	0.63	0.21
Offspring pelagic stage duration	0.67	0.22
Offspring settlement size	0.44	0.02
Body mass	0.30	0.08
Body length	0.44	0.08
Mobility	0.27	0.29
Substratum depth occupancy	0.11	0.06
Feeding type	0.24	0.04

Episodic species are generally large, long-lived, and have late sexual maturity (Figure 0.2). These taxa lack parental care and often reproduce through seasonal broadcasting of numerous small, planktotrophic pelagic eggs, which are unprotected and have high mortality rates (**Figure 5.8**). Bivalves are representative of this group with key species of the benthic communities of the Black Sea (e.g. *Mytilus galloprovincialis*, *Modiolula phaseolina*, *Abra alba*, *Spisula subtruncata*, *Chamelea gallina* and *Donax semistriatus*) but also invasive species (e.g. *Mya arenaria* and *Anadara kagoshimensis*). Other phylogenetically distant taxa, such as echinoderms (e.g. *Amphiura stepanovi*) and carnivorous polychaetes (e.g. *Glycera tridactyla*, *Nephtys hombergii*, *Eumida sanguinea* and *Pholoe inornata*) are also merged into this group (**Figure 5.9**). Within the “Episodic” group, there are two sub-groups: one closer to the opportunist group with fixed in a tube or mobile polychaetes and one group with mostly bivalves with higher adult survival and lower juvenile survival (“true” episodic).

The **Opportunist** group is in a central position with a lower adult survival rate than the episodic group, but with a higher juvenile survival rate (closer to the precocial group). This group is phylogenetically and functionally diverse (**Figure 5.8; Figure 5.9**). These taxa are mostly tubicolous (e.g. *Terebellides stroemii* and *Melinna palmata*) or living in galleries such as *Heteromastus filiformis* or *Alitta succinea*. In this group, there are also co-occurring decapods, closer to the Precocial group (e.g. *Upogebia pussila* and *Crangon crangon*) which release large planktotrophic larvae after a long-brooding incubation period (i.e. mixed planktotrophic).

The **Precocial** group mainly includes small species with short life spans and early sexual maturity (**Figure 5.8**). These species produce a small number of large benthic juvenile offspring after internal incubation, often in a continuous reproductive mode (**Figure 5.8**). These biological traits are strongly specific to amphipods (e.g. *Ampelisca* sp., *Iphinoe* sp., *Corophium* sp., *Caprella* sp., *Microdeutopus* sp. and *Orchomene* sp.) and cumaceans such as *Eudorella truncatula* and *Cumella* sp. (**Figure 5.9**). Precocial group is homogeneous with some modalities of traits exclusively associated with precocial species (e.g. internal mode).

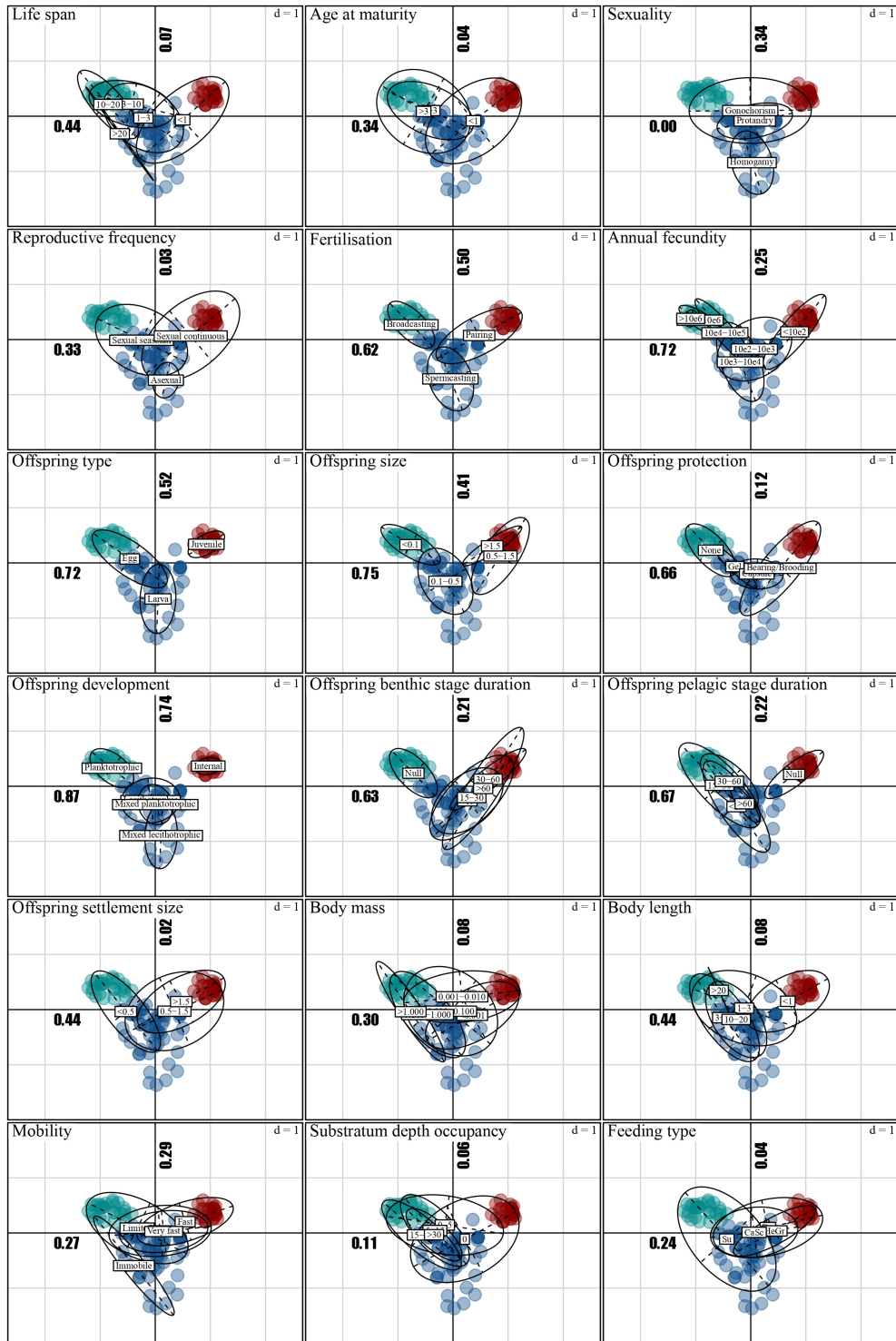


Figure 5.8: Fuzzy correspondence analysis applied on the response traits matrix. Distribution of trait modalities (ellipse) for each trait (windows). Taxa are represented by coloured dots; each colour corresponds to a life-history group with dark red for “P”, dark blue for “O” and dark cyan for “E”. The respective CR for each trait for the first and second axes are in bold (Table 5.3). The trait modalities are positioned at the centre of gravity of their respective taxa.

CHAPTER 5. TRAIT MODELLING TO PREDICT BENTHIC FUNCTIONS AND VULNERABILITIES OVER THE NORTHWESTERN SHELF OF THE BLACK SEA

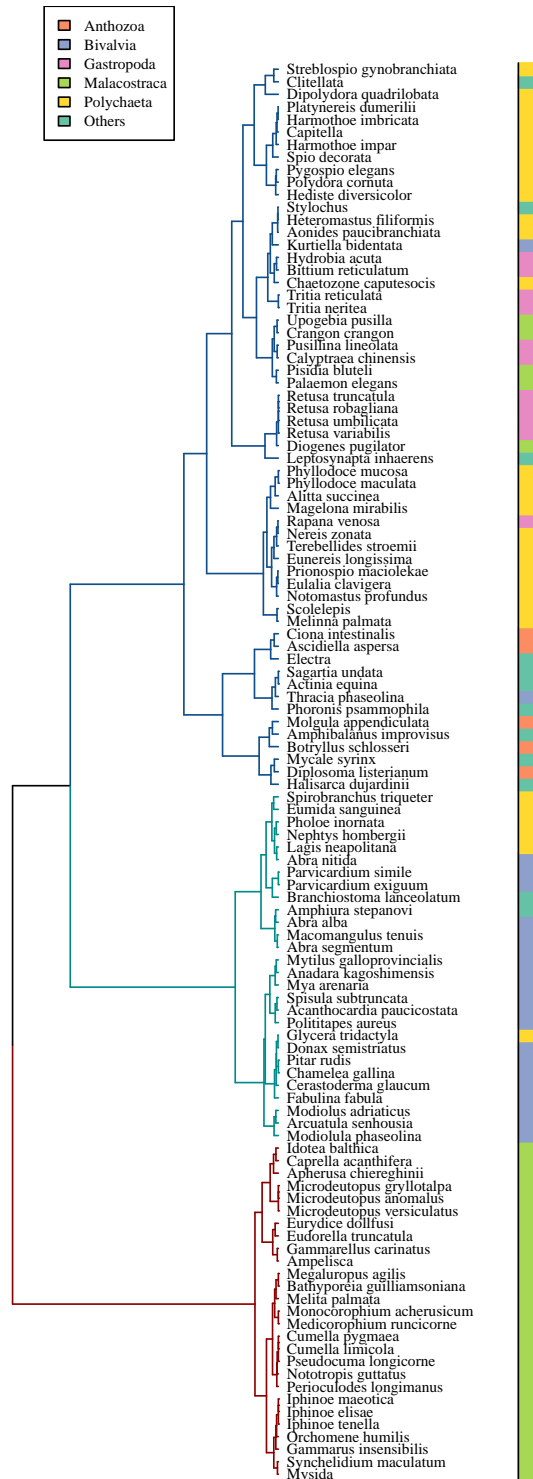


Figure 5.9: Main groups of life-history strategy derived from the response traits: “**Precocial**” (dark red), “**Opportunist**” (dark blue) and “**Episodic**” (dark cyan) species. On the right, the colour bar corresponds to the taxon class, with the legend in the top right corner (class level, others merge classes for which the number of species was less than five).

5.4.1.2 Ecological indicators

The three life history groups exhibit different combinations of effect traits (**Figure 5.10**), reflecting different adaptations (**Figure 5.11**). Most species occupy the upper 5 cm of sediment, with no ventilation or bioconstruction, and are primarily biodiffusors (**Figure 5.10a**). The opportunistic group includes the highest proportion of deep burrowers, greatest ventilation activity, complex endo-bioconstruction structures, and upward or downward conveyors (**Figure 5.10b**). Episodic species are mainly suspension feeders, often epifaunal, with limited mobility and no bioconstruction. Precocial species are mobile deposit feeders, mostly epifaunal or shallow burrowers, with biodiffusive or downward conveying activity (**Figure 5.10c**).

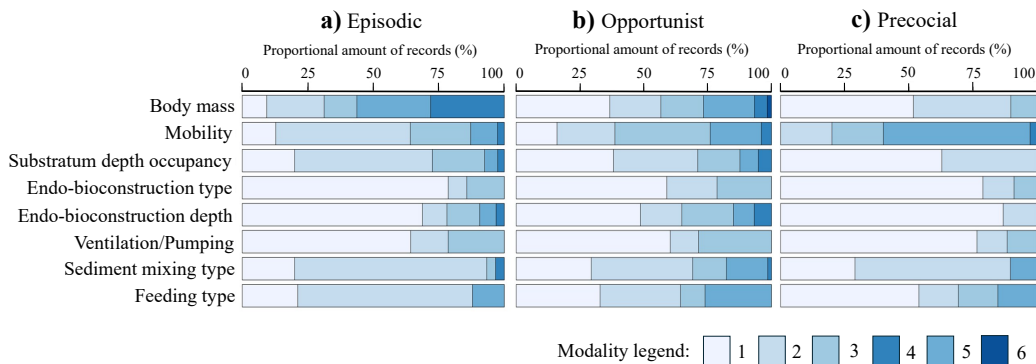


Figure 5.10: Relative amounts of records within effect traits per life history group: a) “E” - Episodic, b) “O” - Opportunist and c) “P” - Precocial. The modality codes correspond to those in (**Table 5.1**).

These differences in effect traits are directly reflected in the ecological indicators derived from those traits (**Figure 5.11a**). Higher biodeposition scores are mostly associated with episodic, followed by opportunist and precocial. On the other hand, precocial deposit feeding scores are greater than episodic scores (**Figure 5.11d**). The highest scores for biomixing and bioirrigation are associated with opportunist followed by episodic (**Figure 5.11b,c**).

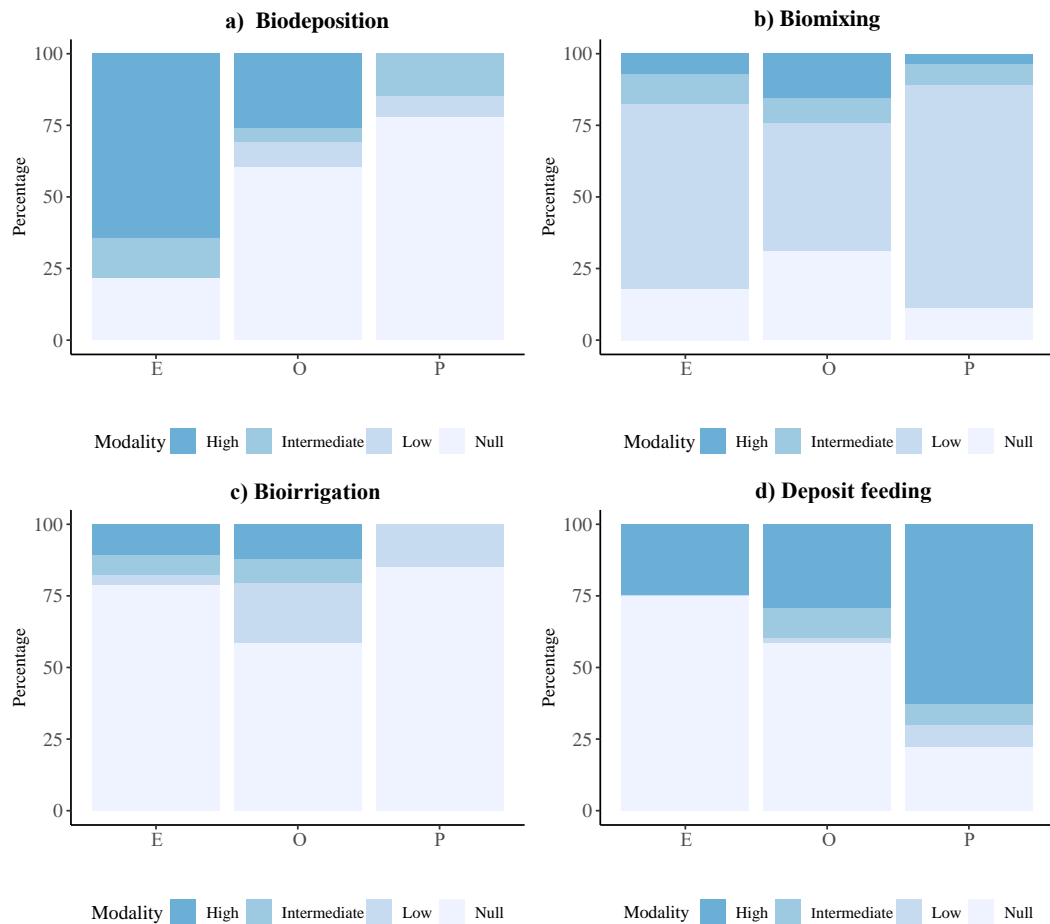


Figure 5.11: Contribution to the four ecological indicators: a) biodeposition, b) biomixing, c) bioirrigation and d) deposit feeding, defined in Table 3) at the species level per life history group (“E”- episodic, “O”- opportunist and “P”- precocial). These indicators are classified into four categories: null (0), low (0 to 0.1), intermediate (0.1 to 0.2) and high (> 0.2). The colour bar on the bottom represents, from light blue to dark blue, increasing ecological indicator scores.

5.4.2 Local trait-environment relationships

In the following sub-sections, we present the results of the RLQ analysis combined with fourth-corner on the traits and multi-traits indicators (in a similar way than Dray et al. (2014)). If the RLQ analysis is no significant, the fourth-corner analysis is not applied on the RLQ axis. Further details on the RLQ analysis, including its significance and the spatial distribution of sampling station scores on RLQ axes, are also available in **Appendix E**.

5.4.2.1 Life history strategy groups

From the preliminary tests, no significant correlation was found between abiotic predictors and life history groups (**Table E.1**). We only select the content of *POC* with the lowest *p*-value. The RLQ analysis is not significant (*p*-value Model 4 = 0.31 and *p*-value Model 2 = 0.001**). On the first axis of the RLQ analysis, the group Opportunist is associated either with (very) high or very low *POC* either at shallow coastal or very deep stations (**Figure 5.12**). Precocial and episodic strategies are associated with lower *POC* at intermediate depths or shallower stations in the southern part, along the Bulgarian coast (**Figure 5.12**).

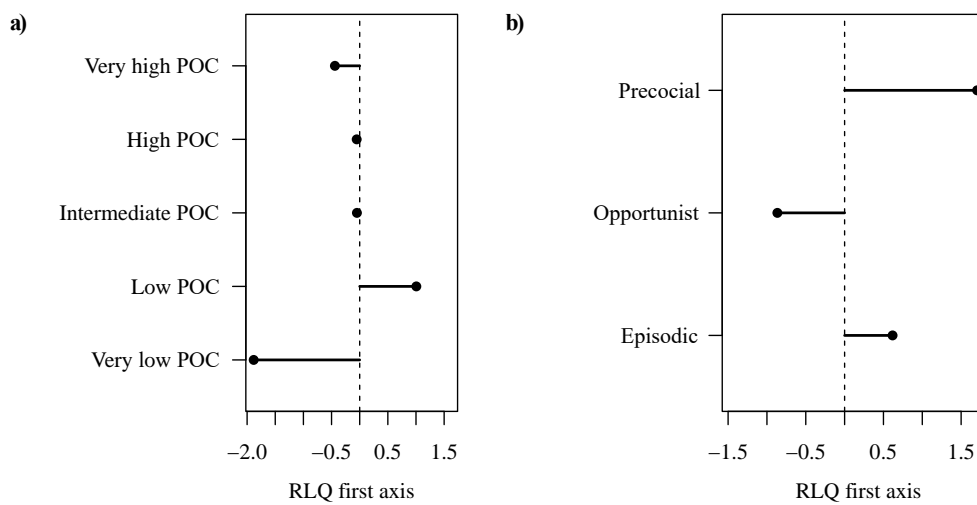


Figure 5.12: a) Abiotic descriptor scores and b) life history strategy scores along the first axis of the RLQ analysis applied on life-history strategies.

5.4.2.2 Response traits

A preliminary test is done to select the environmental descriptors that are significantly linked to the selected response traits. salinity and dissolved oxygen are significantly linked to the response traits (**Table E.3**). Then, we evaluate correlations between individual traits and the selected subset of abiotic variables. Only three traits (i.e. sexuality, offspring development and mobility) are significantly correlated with salinity and dissolved oxygen (**Table E.4**).

The RLQ analysis evidences a globally significant correlation between the selected traits and environmental variables (p -value Model 4 = 0.00052*** and p -value Model 2 = 0.00002***). The first two axes of the RLQ are significantly correlated with sexuality, mobility, and offspring development, as well as oxygen and salinity (**Table 5.4**). The first axis of the RLQ is associated with a depth-gradient: low oxygen and high salinity values (typical of shelf break conditions) are found in the negative part, while higher oxygen content and lower salinity are found in the positive part (**Figure 5.13a**). In contrast to shallower well-oxygenated waters, which include species from the three life history groups with multiple modalities of traits, deep hypoxic stations are inhabited only by opportunists (**Figures 5.13b** and **5.14**). The second axis of the RLQ differentiates the stations based on their depths and oxygen levels. In shallow and well-oxygenated waters (positive scores on both axes, **Figure 5.13a**), we find either mixed-planktotrophic development, very fast mobility or limited mobility (**Figure 5.13b**). On positive scores on the second axis, there are also very deep hypoxic stations with only some opportunists with slow mobility, homogamy and (mixed)- lecithotrophic development (**Figure 5.13b** and **Figure 5.14**). At intermediate depths and oxygen levels (positive and negative scores for the first and second axes, respectively, **Figure 5.13a**), we show either fast species with an internal mode of development (typical of precocial strategists, **Figure 5.14**) or species with no mobility such as episodic bivalves or fixed opportunist worms (**Figure 5.13** and **Figure 5.14**).

At deeper depths and lower oxygen contents, (negative scores on both axes, **Figure 5.13a**), we find mostly opportunistic species (**Figure 5.14**) with protandry, mixed-lecithotrophic development and no mobility (**Figure 5.13b**).

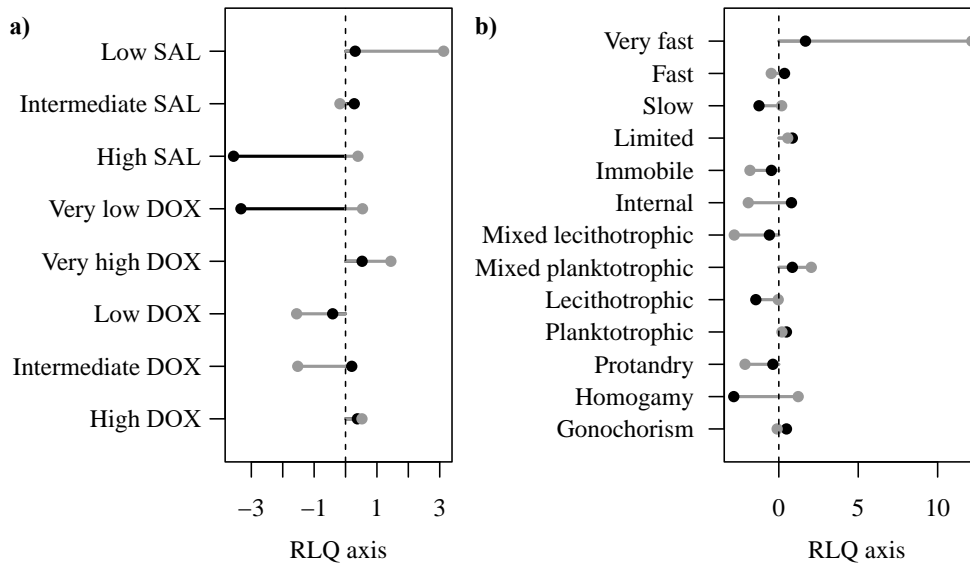


Figure 5.13: a) Abiotic descriptor scores and b) modalities scores on the first axis of the RLQ (black) and the second axis of the RLQ applied on response traits (grey).

Table 5.4: Combinations of fourth corner and RLQ results. A) Fourth-corner tests between the overall RLQ and the first two RLQ axes for the environmental gradient and traits. B) Fourth-corner tests between the overall RLQ and the first two RLQ axes for trait syndromes and environmental variables. Significant associations are marked by asterisks (p -value < 0.05).

A) Abiotic	Coinertia	Axis 1	Coinertia	Axis 2	Coinertia	Axis 1:2
DOX	0.102	0.00004***	0.031	0.133	0.133	0.00012***
SAL	0.145	0.00004***	0.010	0.396	0.155	0.00020***
B) Trait	Coinertia	Axis 1	Coinertia	Axis 2	Coinertia	Axis 1:2
Sexuality	0.130	0.007**	0.003	0.613	0.134	0.009**
Offspring development	0.068	0.048*	0.015	0.613	0.083	0.038*
Mobility	0.083	0.032*	0.011	0.613	0.093	0.035*

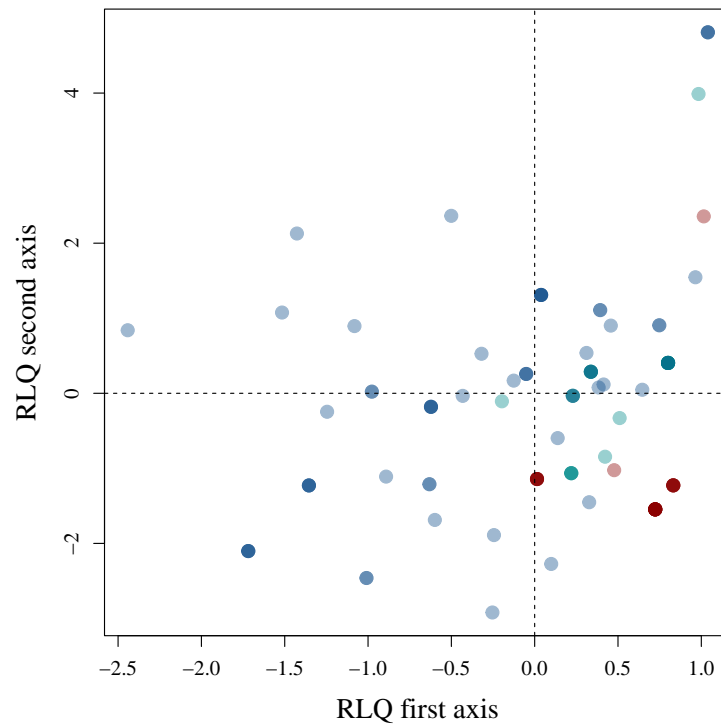


Figure 5.14: Species scores on the two first axes of the RLQ applied on the response traits. Each dot corresponds to a species. Each colour corresponds to a life history group, with dark red for Precocial, dark blue for Opportunist, and dark cyan for Episodic. The intensity of colours varies depending on the number of dots superimposed. A darker intensity of colour indicates that two or more taxa share the same scores on both axes of the RLQ.

5.4.2.3 Effect traits

From the preliminary tests, we only select salinity and substratum type as the two most likely correlated environmental predictors with effect traits (**Table E.6**). We keep the two most likely traits correlated with salinity and substratum (**Table E.7**): substratum depth occupancy (SD) and ventilation/pumping ability (VP).

The global RLQ analysis is not significant (p -value Model 4 = 0.14 and p -value Model 2 = 0.00002***). The first axis of the RLQ analysis separates mixed-coarse substrata at intermediate to high salinity with epibenthos and no ventilation/pumping (negative values, **Figure 5.15**) from the finer substrata, at low to intermediate salinity with endobenthic species with ventilation/pumping activities (positive values, **Figure 5.15**). The second axis separates low-salinity sites with higher ventilation pumping from intermediate to high salinity sites with lower ventilation/pumping ability (**Figure 5.15**).

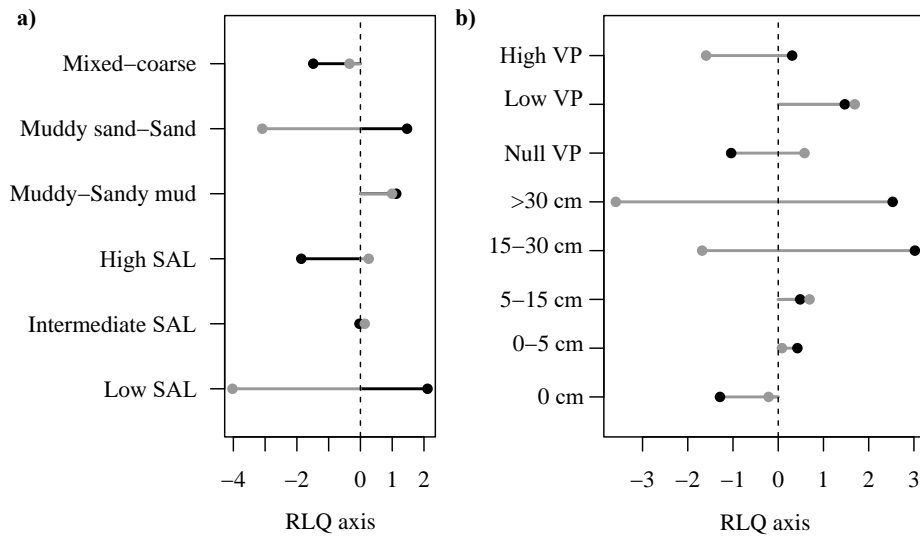


Figure 5.15: a) Abiotic descriptor scores and b) modalities scores on the first axis of the RLQ (black) and the second axis of the RLQ applied on effect traits (grey).

5.4.2.4 Ecological indicators

Here, we only selected salinity as the most likely correlated variables with ecological processes indicators (**Table E.9**). We do not find any significant relationships between salinity and the four indicators of ecological processes (**Table E.10**). Bioirrigation is at the limit of significance (i.e. p -value = 0.07). The RLQ analysis is not significant (p -value Model 4 = 0.064 and p -value Model 2 = 0.002**). On the first axis of the RLQ analysis, higher intensity of bioirrigation and deposit-feeding potential are associated with low salinity (**Figure 5.16**).

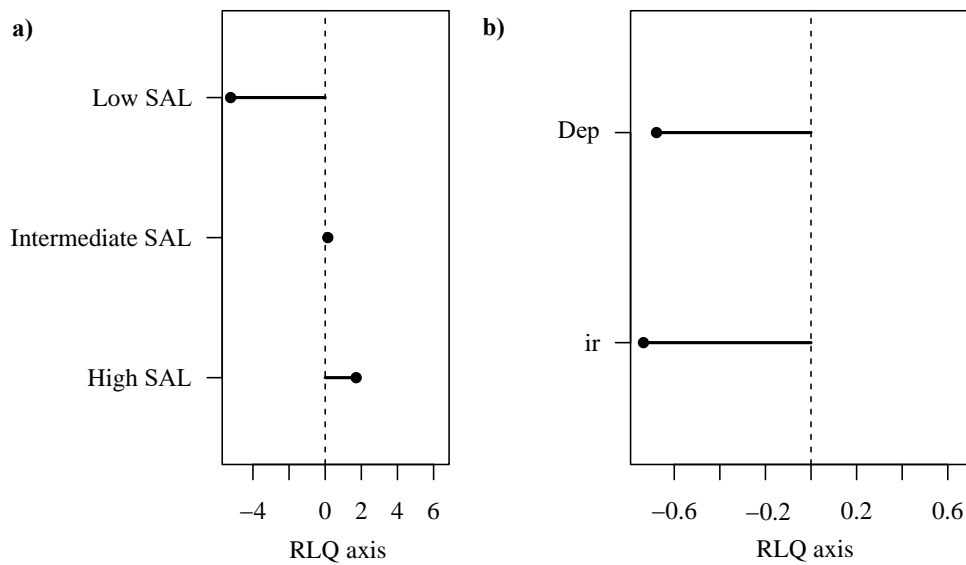


Figure 5.16: a) Abiotic descriptor scores and b) ecological indicator scores on the first axis of the RLQ applied on ecological indicators.

5.4.3 Mapping of traits

5.4.3.1 Spatial distribution of life history strategies

The distribution of episodic species is higher in the southern part and near the mouth of the Dnieper River in the upper north-eastern sector (**Figure 5.17a**). At the edge of the shelf, the benthic communities are dominated by opportunistic species (**Figure 5.17b**). The proportion of opportunists is also slightly greater in coastal areas close to the Danube mouth and in the northeastern part of the shelf (**Figure 5.17b**). The proportion of precocial species is higher at intermediate depths, particularly in the southern part of the shelf (**Figure 5.17c**).

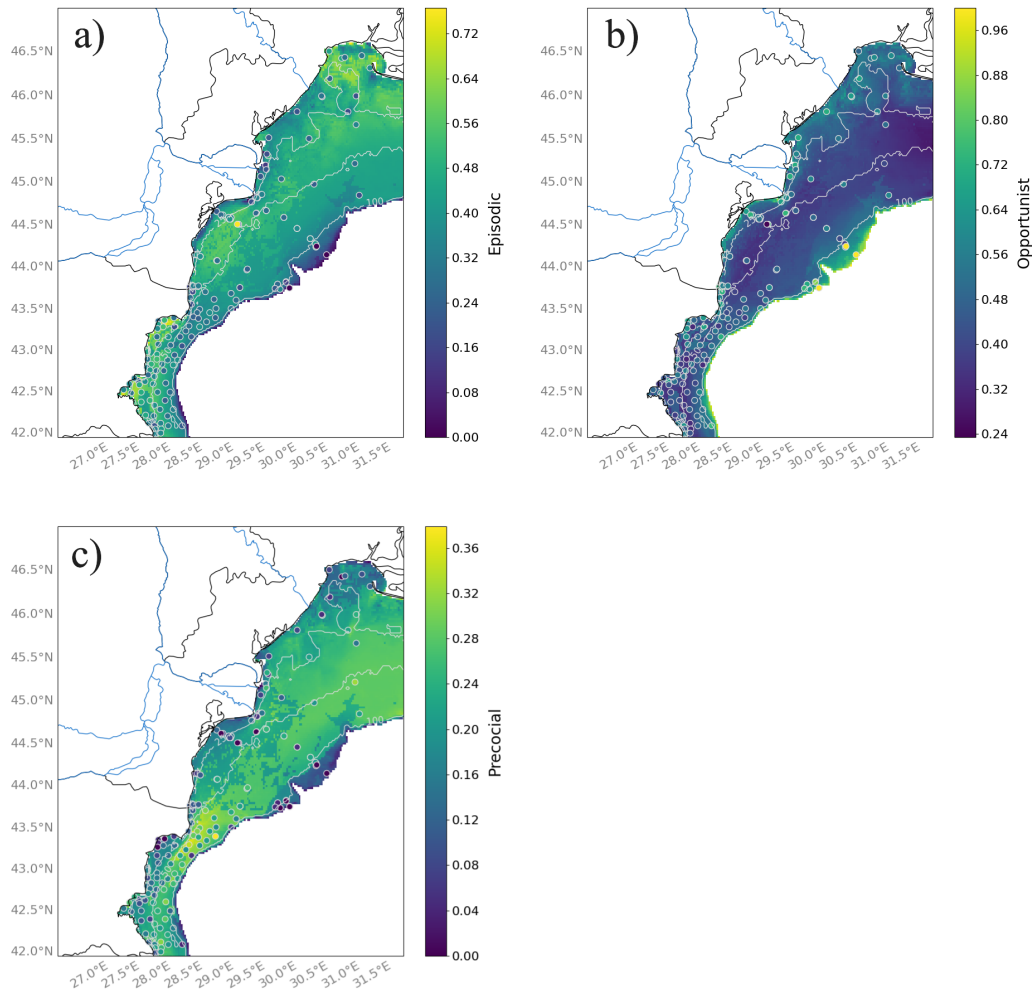


Figure 5.17: Spatial distribution of the three life history strategies across the northwestern shelf of the Black Sea: a) episodic, b) opportunist and c) precocial species. Dots represent sampling stations coloured according to the measured proportion of life history strategies within a community. The colour bar on the left ranges from dark purple (low values) to yellow (high values). Blue lines indicate rivers, black and white lines show bathymetry, and areas deeper than 200 m are coloured in white. Colour bar scales differ between maps and are not directly comparable.

5.4.3.2 Spatial distribution of response traits

The planktotrophic and mixed planktotrophic modes decrease along a depth gradient (**Figure 5.18a,c**). In contrast, the proportions of lecithotrophic and mixed lecithotrophic modes are higher at the edge of the shelf (**Figure 5.18b,d**). The spatial distribution of internal development (**Figure 5.18e**) follows the same pattern as that of the precocial group (**Figure 5.17c**), as this mode of development is found almost exclusively in this group.

CHAPTER 5. TRAIT MODELLING TO PREDICT BENTHIC FUNCTIONS AND VULNERABILITIES OVER THE NORTHWESTERN SHELF OF THE BLACK SEA

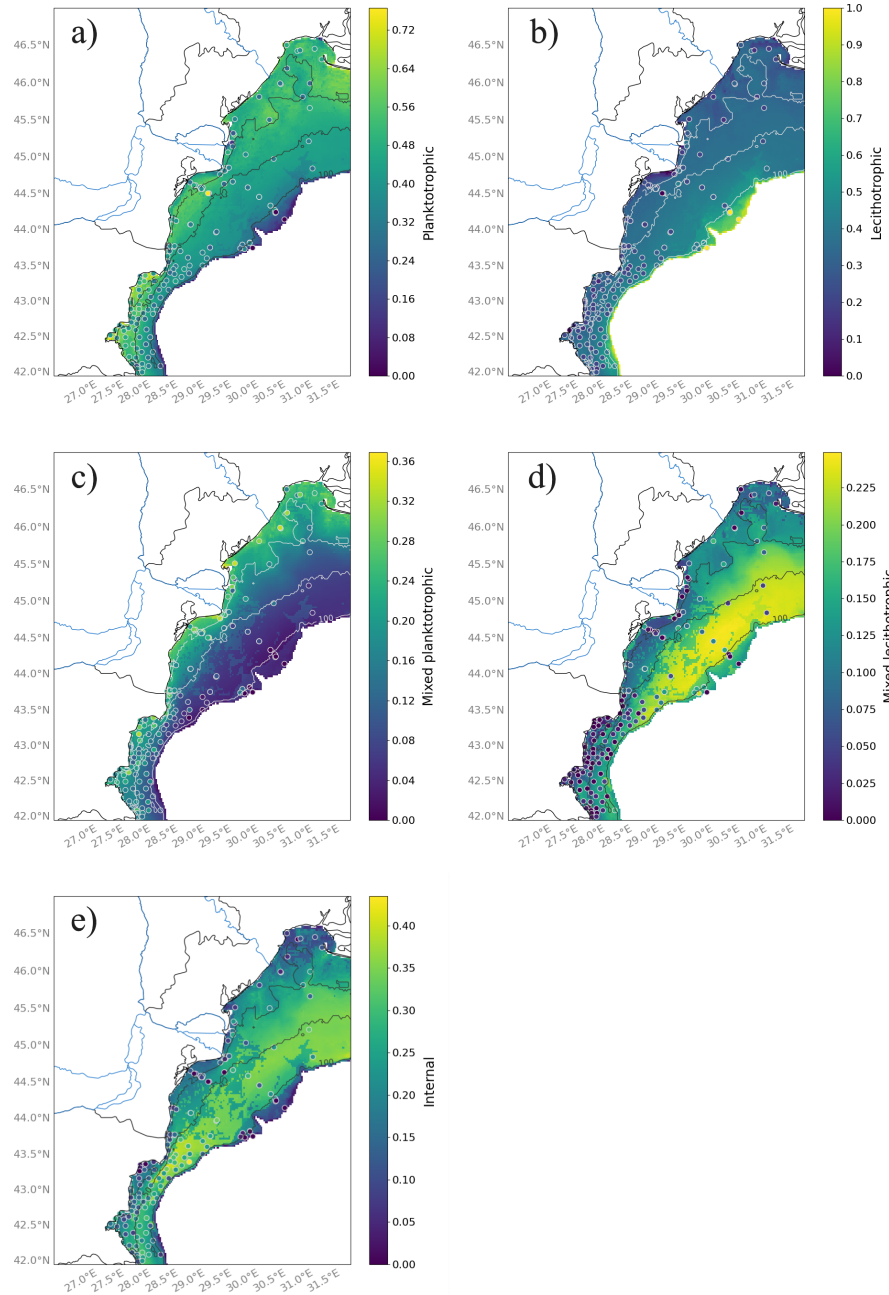


Figure 5.18: Spatial distribution of the five modalities of trait offspring development across the northwestern shelf of the Black Sea: a) planktotrophic, b) lecithotrophic, c) mixed-planktotrophic, d) mixed-lecithotrophic and e) internal. Dots represent sampling stations coloured according to the measured proportion of modalities of offspring development within a community. The colour bar on the left ranges from dark purple (low values) to yellow (high values). Blue lines indicate rivers, black and white lines show bathymetry, and areas deeper than 200 m are coloured in white. Colour bar scales differ between maps and are not directly comparable.

5.4.3.3 Spatial distribution of effect traits

The epifaunal distribution (i.e. 0 cm) is slightly higher at deeper water depths and lower in the area close to the mouth of the Danube Delta (**Figure 5.19a**). Other modalities (i.e. 0 – 5 cm and 5 – 15 cm) have less clear spatial patterns, but the proportion of burrowing species (from 0 to 15 cm) is, in general, higher at shallower coastal areas (**Figure 5.19b,c**). Very deep burrowers (i.e. > 15 cm) are mainly distributed near the coast close to the Danube Delta (**Figure 5.19d**).

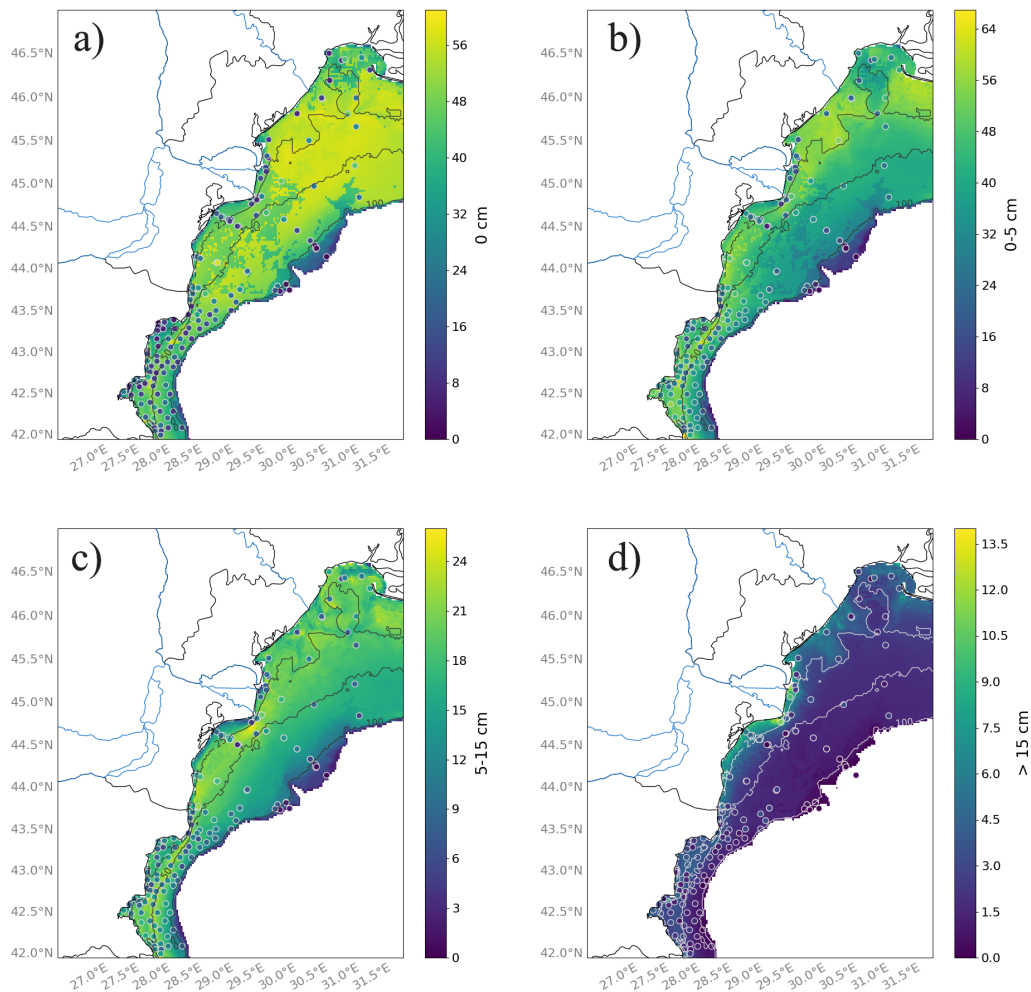


Figure 5.19: Spatial distribution of the modalities of trait substratum depth occupancy (in cm) across the northwestern shelf of the Black Sea: a) 0 cm, b) 0 – 5 cm, c) 5 – 15 cm and d) deeper than 15 cm in the sediment. Dots represent sampling stations coloured according to the measured intensity of modalities of substratum depth occupancy within a community. The colour bar on the left ranges from dark purple (low values) to yellow (high values). Blue lines indicate rivers, black and white lines show bathymetry, and areas deeper than 200 m are coloured in white. Colour bar scales differ between maps and are not directly comparable.

5.4.3.4 Spatial distribution of the ecological indicators

The four community ecological indicators exhibit similar spatial pattern across the shelf, with higher intensities occurring in shallower oxic waters and decreasing to zero at the edge of the shelf (**Figure 5.20a-d**). This finding aligns with the RLQ analysis, as higher intensities of bioirrigation and deposit feeding were linked to very shallow low-salinity sites. Higher bioirrigation potential is observed within a narrower coastal band (**Figure 5.20a**) compared to biomixing potential, with a more scattered distribution towards deeper depth (**Figure 5.20b**). We observe hotspots of biodeposition near the Dnieper River Delta in the northeastern region or in the coastal southern area of the shelf (**Figure 5.20c**). Deposit feeding potential is generally high across the whole shelf except in very deep area (**Figure 5.20d**).

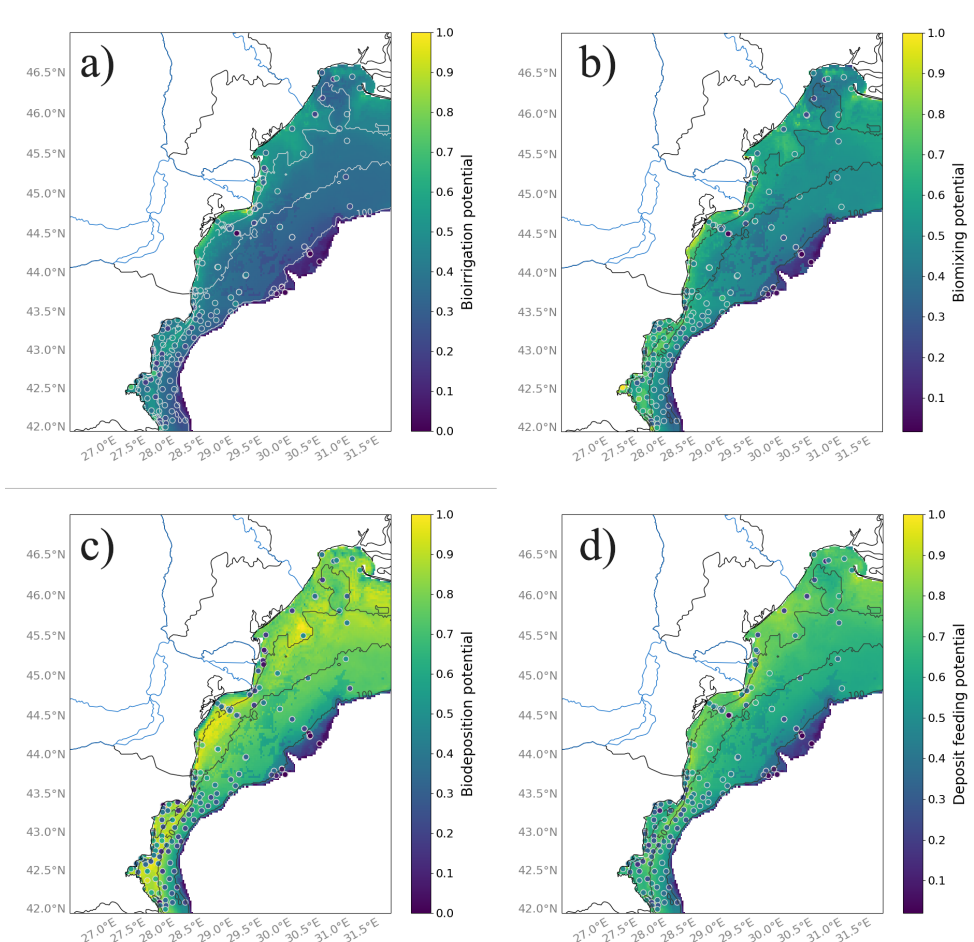


Figure 5.20: Spatial distribution of the ecological process indicators across the northwestern shelf of the Black Sea: a) bioirrigation, b) biomixing, c) biodeposition and d) deposit feeding potentials. Dots represent sampling stations coloured according to the measured intensity of ecological processes within a community. The colour bar on the left ranges from dark purple (low values) to yellow (high values). Blue lines indicate rivers, black and white lines show bathymetry, and areas deeper than 200 m are coloured in white. Colour bar scales differ between maps and are not directly comparable.

5.5 Discussion

One of the key challenges in biodiversity conservation is to upscale local sub-meter ecological processes to a broader scale, which is relevant for ecosystem-based management (Bianchi et al., 2021; Snelgrove et al., 2014). The mapping of ocean biodiversity lags far behind that of terrestrial ecosystems; one of the reasons is the lack of ocean biological and environmental information. For the past decade, quality-controlled physical and biogeochemical data for all European regional seas and the open ocean have been freely available through CMEMS. This opens doors to the development of biogeographic models providing the necessary scientific framework for mapping marine species, traits and functions at ecosystem scales. These maps offer a solid support for the implementation of marine policy to preserve biodiversity (e.g. design of MPAs) and to allocate blue growth activities (e.g. marine spatial planning as requested by the *Intergovernmental Science Policy Platform on Biodiversity and Ecosystem Services (IPBES)*).

This paper proposes a methodology to scale up local species data to obtain maps of ecological groups and functions at shelf scale. The methodology combines species-level trait assessments with the derivation of community-level ecological functions from these traits using empirical formulations. The link between the community traits and functions with environmental conditions was assessed using multivariate statistical approaches and machine learning. The mapping of the traits and functions is then performed using the environmental variables that are significantly correlated with the traits and that are predicted by a numerical ocean model. The methodology was developed for the northwestern shelf of the Black Sea, which constitutes an ideal test area because it hosts significant gradients of environmental conditions that are expected to affect trait composition and biodiversity. We focused on effect traits, which reflect the benthos' influence on benthic-pelagic coupling and on response traits that characterise species sensitivity to environmental disturbance.

The distribution of the data covers the northwestern shelf area from the river side and coastal areas to offshore up to the shelf break. This allows us to investigate the trait-environment link across a large environmental gradient dominated by oxygen (well oxygenated, hypoxic and anoxic waters), salinity (fresh to marine waters), depth (from 10 up to 150 m), and substratum (muddy, sandy and coarser grained with shells). We show that variability of environmental conditions, such as the gradient of bathymetry, bottom dissolved oxygen, sedimentary organic matter, and type of substratum, explain the spatial distribution of traits. In particular, the gradient of bathymetry, bottom dissolved oxygen, sedimentary organic matter and type of substratum are the environmental variables that explain the most of the variability in traits. The level of bottom oxygen is a dominant factor that explains the diversity of traits and the distribution of ecological groups: at very low oxygen levels, we find low functional biodiversity with only some opportunistic species, while in well-oxygenated waters, higher trait divergence is observed with a marked diversity of offspring and adult traits.

In previous studies, it was found that trait composition changes dramatically as oxygen concentration decreases (Kendzierska & Janas, 2024; Villnäs et al., 2012), with only asexually reproducing polychaetes found in oxygen-deficient zones (Bon et al., 2021; Pacheco et al., 2011).

Three of the four life history strategies proposed in Kindsvater et al. (2016) are present in the Black Sea macrozoobenthos: opportunist, precocial and episodic. Half of the taxa are grouped as opportunists, which is consistent with historical disturbances, including widespread bottom hypoxia and eutrophication (Gomoiu, 1992; Mee et al., 2005; Friedrich et al., 2014). The opportunistic species are highly functionally heterogeneous, and their distribution is not habitat specific. They are mostly found in areas characterised by a high seasonality of oxygen, such as the northeastern part of the shelf, which has a high sedimentary organic matter content and seasonal hypoxia, and the shelf break where upwellings of anoxic waters occur from the deep basin. In these two distinct seascapes, opportunist species are separated according to their offspring development mode and mobility. In shallow, oxygen-rich areas, mixed-planktotrophic development is prevalent, consistently with the reproductive strategies of deep-burrower opportunist decapods. In contrast, mixed-lecithotrophic and lecithotrophic modes are associated with deeper, hypoxic environments and taxa such as opportunistic tubicolous worms (e.g. capitellids and spionids), which usually persist in disturbed environments (Beauchard et al., 2022). We also show that homogamy is restricted to hypoxic zones, as this reproductive mode may confer advantages under sparse population densities, as documented in Holbrook and Grassle (1984). The small fraction of the episodic group, which is only present in well-oxygenated areas, and the absence of survivors, indicate that the benthic communities are still in a phase of ecological recovery (Chevalier et al., 2024; Stevens et al., 2019). Planktotrophic development, which is typical of the group episodic, is most prevalent at shallow and intermediate depths. In shallower areas, episodic species are mostly shallow-burrowing clams with limited mobility in opposition to immobile epifaunal episodic species, such as mussels found at intermediate levels of oxygen and depth. There are also mobile species, such as epifaunal carnivorous worms or small precocial amphipods, documented to be hosted by biogenic reefs of mussels (Begun et al., 2010; Teacă et al., 2006; Surugiu, 2011).

The spatial distributions of ecological processes follow a clear environmental gradient with diverse functionally seascapes along a depth-gradient. Shallow coastal waters, with finer-grained substrata, support deep-burrowing taxa with high ventilation activity, leading to higher scores for both biomixing and bioirrigation. These habitats are also characterised by higher species densities and trait diversity, amplifying the contribution of macrobenthos to those ecological processes. At intermediate depths, epifaunal species with no ventilation or pumping abilities are associated with mixed and coarse substrata. Coarser sediments have already been shown to generally support more epifaunal species, whereas fine sandy sediments are more favourable for burrowers (Bolam et al., 2017; Breine et al., 2018; Bremner et al., 2006a).

Despite their smaller contributions to biomixing or bioirrigation, large-bodied suspension feeders (e.g. bivalves, sponges, and ascidians) can contribute significantly to ecosystem functioning through biodeposition (Graf & Rosenberg, 1997; Norkko et al., 2013). At greater depth, hypoxic shelf-edge zones are dominated by impoverished benthic communities with negligible impacts on ecosystem functions. Our findings support previous works evidencing that bioturbation declines along increasing hypoxic gradients (Villnäs et al., 2012; Levin et al., 2009; Van Colen et al., 2012).

Our study has several limitations inherent to the sampling methods and available data. Firstly, the contribution of deep-dwelling taxa is likely underestimated by the Van Veen grab sampling method (Beukema, 1974). The resolution of the ocean model that delivers the environmental condition is at the km scale (i.e. 2.5 km), while that of the community is at the metre scale. This mismatch of spatial scales can explain the limited number of significant trait-environment relationships found in this study. *Artificial Neural Networks (ANNs)* can help to deal with non-linearities in trait-environment relationships and give sufficiently good results in mapping the modalities of the traits or indicators. However, our model has limitations in predicting extreme values that are less often sampled. Finally, many trait-based indicators are empirical, and their validation is limited due to the lack of in situ data or laboratory experiments (Martini et al., 2021). Several mathematical formulations for the same indicator have been proposed (Beauchard et al., 2023; Queirós et al., 2013; Renz et al., 2018; Wrede et al., 2018). In this study, we focus on the effect traits that impact benthic-pelagic coupling. Combinations of these traits are used via empirical formulations to assess ecological functions such as biomixing, bioirrigation and biodeposition. The aim is to use the variability of these ecological indicators to account the effect of benthic life in ocean biogeochemical models. However, these indicators are not necessarily directly linked with biogeochemical model parameters (Gogina et al., 2017). For instance, an information on the variability of biomixing cannot be used straightforwardly in the parametrisation of the sediment mixing layer depth. This highlights the potential benefit of mapping the modalities of traits as a better proxy. By mapping trait modalities, such as the sediment living depth, we are closer to the property measured at the species level and avoid the use of empirical equations to compute imperfect indicators, which are still poorly validated. The obtained trait maps can be used to incorporate the variability of benthos characteristics in the parametrisation of ocean models and, for instance, in the formulation of the resuspension and deposition processes and the exchanges of solutes and solids. This approach will constitute a significant step in the development of coupled benthic-pelagic biogeochemical models that will allow the modellers to refine the assessment of the role of continental shelves and macrobenthos in the mediation of biogeochemical cycles.

5.6 Conclusion

The developed trait distribution models offer the possibility of estimating how the spatial distribution of traits or indicators might change in response to climate change, enabling us to predict potential changes in the ecosystem functions delivered by the macrobenthos under different shared socioeconomic pathways. The mapping of traits can support spatial planning and ecosystem-based management by identifying areas where benthic communities deliver key ecological functions or are particularly vulnerable to disturbance (Cogan et al., 2009; Bolam et al., 2023; Miatta et al., 2021). In the Black Sea, beam trawling has been shown to impact benthic biodiversity negatively, including that of mussel biogenic reefs (Nenciu et al., 2023), highlighting the need to protect those habitats that provide shelter for a large diversity of free-living and sessile epifauna. In addition, immobile suspension-feeder episodic species, such as mussels, may be especially important for maintaining benthic-pelagic fluxes through biodeposition. They should be prioritised for protection because they provide key ecosystem functions and are the most vulnerable to disturbances. Moreover, shallow coastal finer-grained areas with high biomixing and bioirrigation potentials could play a role in enhancing blue carbon sequestration. Given the current gaps in spatial and temporal data for benthic ecosystems, particularly for the Black Sea, there is an urgent need for validated trait-based indicators and high-resolution ecological maps to support targeted conservation strategies and sustainable management of benthic ecosystems.

GENERAL DISCUSSION

6.1 Main results of this PhD work

Benthic biodiversity plays a crucial role in ecosystem functions and biogeochemical cycles in coastal ecosystems, but the lack of detailed spatial data on benthic functions and vulnerabilities hinders effective conservation and management efforts. This PhD study focuses on mapping benthic functions and vulnerabilities across the *North-Western Shelf (NWS)* of the Black Sea. We used a **functional approach** to biodiversity meaning that the biological communities are represented by their characteristics with an impact on ecosystem (i.e. effect traits as proxy of **functions**) or reacting in a certain way to environmental pressures (i.e. response traits as proxy of **vulnerabilities**). For this purpose, we compile a comprehensive macrozoobenthic dataset for the **NWS of the Black Sea**, collected between 1995 and 2017. The data set includes the occurrences of 215 macrozoobenthic taxa and 27 biological traits documented for 127 taxa, along with abiotic data on physical and chemical variables. The dataset aims to address the functional knowledge gap in the Black Sea, providing insights into ecosystem functions, biodiversity conservation, and management.

We propose a multidisciplinary approach to link our set of biological traits to abiotic conditions (e.g. bottom oxygen and temperature) at the shelf scale. Specifically, we combine field sampling, multiple traits based indicators as proxy of functions and vulnerabilities of the macrobenthos, abiotic data derived from a coupled biogeochemical-physic model and *TDMs* to predict the spatial distribution of traits from their local distribution at the community scale and maps of bottom abiotic conditions. The ultimate objective of this PhD work is to produce maps of the biological traits of the benthos to inform *Ecosystem-Based Management (EBM)* to protect biodiversity and to understand ecosystem functioning.

We applied and developed this methodology for the *NWS* of the Black Sea, an area strongly affected by eutrophication and its subsequent bottom hypoxia in the past (e.g. Gomoiu (1992); Zaitsev and Mamaev (1997)). At the end of this thesis, we can return to the research questions defined in the introduction (**Chapter 1**):

“How to upscale local observations of benthic biodiversity to the mapping of benthic functions and vulnerabilities at the shelf scale?”

1) What are the relationships between macrobenthos’ species and abiotic conditions at the local scale? (Chapter 3)

1.a) What are the changes in environmental conditions 1990s and 2010s?

We evidence a clear reduction in the duration and spatial extent of bottom hypoxia in the north-eastern part of the shelf (**Figure 3.5**), in relation to a decrease in riverine nutrient loads (**Figure 3.4**). However, bottom hypoxia persists in the north-eastern sector where an anticyclonic gyre carried the outputs of the Danube and this accumulation of organic matter leads to hypoxia in this area during the summer period (**Figure 6.1**).

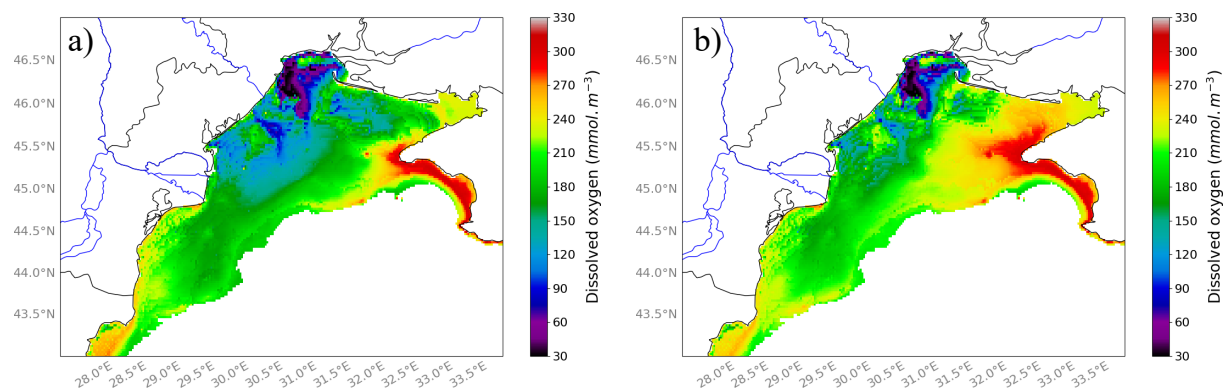


Figure 6.1: Spatial distribution of the bottom oxygen concentration in summer (from July to September) over the northwestern of the Black Sea. a) 1990-2000 and b) 2008-2018 climatology. Colour bar at the right from very low values in black to high values in red. Main rivers are in blue.

1.b) Are the species compositional changes associated with these changes, particularly regarding organic pollution ?

Through an adapted multivariate analysis (i.e. detrending of physics using *VPA*; **Figure 3.3**), we are able to identify benthic community changes solely due to organic pollution variations between 1995 and 2017. Our results show a clear sign of recovery with an increase in taxonomic richness and proportion of species vulnerable to organic enrichment (e.g. long-live bivalves or some crustaceans). These changes were directly related to a decrease in eutrophication on the *NWS* and improved water quality. However, these positive signs of recovery should be taken care with caution due to the proliferation of opportunistic invasive species and the persistence of bottom hypoxia. These results need to be taken with caution because of sampling artefacts and taxonomic uncertainties.

In addition, some long-lived species typical of the area have not fully recovered, suggesting incomplete ecosystem restoration or unmet environmental conditions.

2. What are the biological traits of the macrobenthos' species ? (Chapter 4)

2.a) How to define the vulnerabilities of the macrobenthos ?

Our set of response traits allows for the representation of **life history trade-offs** (here the *POSE* concept proposed in Kindsvater et al. (2016)) through specific combinations of trait values. We evidenced three major life-history strategy groups from the clustering of the scores of the *FCA* applied on the response traits matrix. The three life-history strategy groups are: "**Episodic**" with high mortality of juveniles and low mortality of adults, "**Precocial**" with low mortality of juveniles and high mortality of adults and "**Opportunist**" with high mortality of juveniles and adults (**Figures 5.8 and 5.7**). The first group "Episodic" mostly gathers long-lived larger bivalves which need longer time and stability to achieve reproductive success, making them key conservative targets for marine management. The relative low proportion of "Episodic" species in our dataset supports the hypothesis that the macrobenthic communities are currently in a recovery state (Chevalier et al., 2024). The second group "Precocial" is the most homogeneous and represents the small short-lived amphipods and isopods with internal mode development (meaning vivipary). The third group "Opportunist" is the most heterogeneous and can in fact sub-divided in three sub-categories: a first group with species characterised by mixed planktotrophy, typical of co-occurring decapods such as *Upogebia pusilla* and *Crangon crangon*. They are mostly observed in sandier substratum in shallow well-oxygenated waters (**Figure 5.18c**). The second sub-groups gather carnivorous worms closer to long-lived "Episodic" species and tunicates, sponges co-occurring on mixed-coarse substratum with *Mytilus* and *Modiolula* at intermediate levels of depth and oxygen. Then the last sub-group merge species with homogamy, (mixed)-lecithotrophy, mostly infaunal species in muddier substratum with limited mobility such as *Prionospio maciolekae*, *Heteromastus filiformis*. They can be found at any depth but only these "true" opportunists are observed at the hypoxic sites at the edge of the shelf (**Figure 5.17b**).

2.b) How to define the functions of the macrobenthos ?

We combine multiple effect traits (e.g. substratum depth occupancy, type of burrows) to build ecological indicators as proxy of key **ecological processes** (e.g. biomixing, bioirrigation, biodeposition and deposit feeding). We evidenced that most species in the Black Sea have low biomixing and bioirrigation potentials (**Figure 5.11**). Several studies observed a similar pattern, with a few species contributing strongly to bioturbation or bioirrigation in other shelf areas (Gogina et al., 2017, 2020; Weinert et al., 2022; Wrede et al., 2017). Indeed, only a few species have a high impact on ecological processes with either species in high abundance or species in lowest abundance but with particular combination of traits. For instance, it is the case for *Upogebia pusilla*, with small numbers of individuals, but high biomixing and bioirrigation scores at the species level.

Accordingly, D'Andrea and DeWitt (2009) and Pillay et al. (2011), have shown that bioturbating thalassinid shrimps are important ecosystem engineers that exert major influences over ecosystem processes and community structure.

3) What are the relationships between (multi)-traits and abiotic conditions at the local scale ? (Chapter 5)

In this work, we evaluated the connection between environmental stressors and the trait composition of macrozoobenthic communities on the northwestern shelf with a combination of **RLQ and 4th corner analyses** (as proposed in Dray et al. (2014)). Our results showed that the distribution of biological traits is not randomly distributed across the shelf with the concentration of oxygen and the type of substratum being the main drivers of spatial variability in traits. A graphical summary of the traits influenced by abiotic drivers and associated life-history strategies is provided in **Figure 6.2**. Oxygen levels are a critical driver of macrobenthic community traits. Well-oxygenated areas support diverse communities with high functional trait diversity. They exhibit high biomixing and bioirrigation on muddier and sandier substrata, as well as high biodeposition on coarser substrata associated with mussel reef communities. These communities contribute significantly to ecosystem functions, such as sediment oxygenation, nutrient turnover, and organic matter recycling. In opposition, hypoxic (low oxygen) zones, typically found at the edge of the shelf, are inhabited by functionally impoverished macrozoobenthic communities. These communities are dominated by opportunistic species with a combination of a few typical opportunistic traits. Their contribution to ecological processes is negligible. Overall, ecological processes intensities, such as biomixing, bioirrigation, biodeposition and deposit feeding, decrease towards the edge of the shelf where the concentration of oxygen drop near zero. On the northwestern shelf of the Black Sea, (Lenstra et al., 2019) have evidenced that high rates of bio-irrigation (from *in-situ* measurements) drive the release of iron and manganese from the sediment to the overlying water column in the shallow coastal area, supporting our results.

The type of substrate also plays a crucial role in determining the depth at which organisms reside and their ventilation activities (**Figure 5.15**). Indeed, finer-grained sediments favour burrowing and sediment-mixing species, while coarser substrata are more suitable for epifaunal suspension feeders that contribute to biodeposition. Sandier substratum are typically found in shallower, well-oxygenated areas and they support high biomixing and bioirrigation activities. This type of substrate is favourable for deep-burrowing species that contribute to sediment mixing and nutrient recycling. Coarser sediments are often associated with mussel reef communities and are found at intermediate depths (between 40 and 60 metres). These substrata support epifaunal species with no mobility and suspension-feeding traits (e.g. bivalves, tunicates, sponges). Biodeposition is a key ecological function in these areas, as large-bodied suspension feeders (e.g. mussels) contribute to benthic-pelagic coupling by filtering particles and depositing carbon- and nitrogen-enriched materials

A major feature is the dominant implication of offspring traits (**Table 5.3**). This contrasts with the study of Paganelli et al. (2012), who found poor correlations between traits related to reproductive strategy and patterns of development, and *FCA* axes. In this study, planktotrophic mode is observed in shallow well-oxygenated waters or intermediate depth (**Figure 5.13**). That was previously observed in Shi et al. (2023) with planktotrophic mode at coastal high nutrients areas. (Mixed)-lecithotrophic modes are associated with deep low-oxygen level waters (**Figure 5.13**). Previous studies evidenced negative relationship between lecithotrophic larval development and chlorophyll-a content as these taxa can survive in areas of low food availability for their planktonic stage (Sutton et al., 2021).

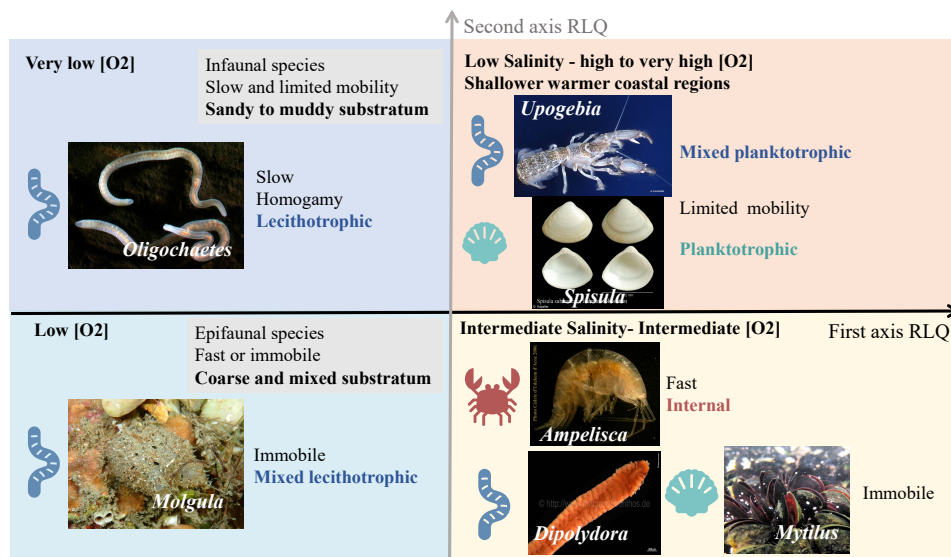


Figure 6.2: Graphical summary of the RLQ analysis applied on response traits: first axis in black and second axis in grey. The three life history strategies are in colours (dark blue for Opportunistic, dark cyan for Episodic and dark red for Precocial). Sources for pictures: Oligochaete by Stephen Moore taken from Landcare Research Group; *Spisula subtruncata* by Kapeller, Rudolf taken from WoRMS, *Upogebia pusilla* by J. Lecompte, *Molgula appendiculata* by Frédéric André taken from Wikimedia, *Ampelisca* spp. by Cédric d’Udeken d’Acoz taken from WoRMS, *Dipolydora quadrilobata* by Hermann Marko taken from WoRMS and *Mytilus galloprovincialis* by Pillon Roberto taken from WoRMS.

4) How to upscale this local information to map the distribution of these traits at the shelf case? (Chapter 5)

Biological traits are upscaled at the community level by crossing species observations and their traits to obtain the community weighted mean matrix (i.e. mean composition of traits at one sampling site). We developed statistical modelling of the trait-environment relationships using *TDMs*. We used **neural networks** to link traits to environmental drivers, provided by a biogeochemical model, at high temporal and spatial resolution, run in an operational mode by *CMEMS*. We chose to develop *ANNs* to deal with the non-linearities of traits-abiotic relationships.

Then, we combined our developed *TDMs* with spatial maps of bottom abiotic conditions to map continuous distribution of the community trait values from punctual observations. In this PhD work, we combined *ANNs* with multivariate statistic (RLQ combined with fourth-corner) to help understanding the underlying relationships between traits and abiotic predictors. A few studies used *ANNs* to assess the spatial distribution of fish (Brosse, Guegan, Tourenq, & Lek, 1999; Valavanis et al., 2008) and macroinvertebrates in freshwater (Goethals, Dedecker, Gabriels, Lek, & De Pauw, 2007; Gabriels, Goethals, Dedecker, Lek, & De Pauw, 2007) and marine systems (Lee et al., 2013; Bandelj, Curiel, Lek, Rismondo, & Solidoro, 2009).

6.2 Methodological advances and strengths

6.2.1 The case of the Black Sea

The northwestern shelf of the Black Sea was well-studied, in terms of macrozoobenthic communities (e.g. Băcescu et al. (1971); Begun et al. (2010, 2022); Gomoiu and Tiganus (1977); Teacă et al. (2019b, 2020); Todorova and Konsulova (2000); Todorova et al. (2020, 2021)), but to the best of my knowledge, before this work, no published trait database was accessible in open access for the Black Sea. One study about functional trait composition was found on the coast of Turkey (Van & Gümüş, 2021). The delivery of a complete set of biological traits for the Black Sea can help for future studies on the health of benthic communities in this region. In previous studies, RLQ and fourth-corner analyses have successfully identified key abiotic factors that influence the trait composition of macrobenthos assemblages in marine ecosystems (e.g. Piló et al. (2016); Hu et al. (2019); Morais et al. (2019); J.-Y. Dong et al. (2021); Vesal et al. (2021); Beauchard et al. (2022); Ribeiro et al. (2016)). To my knowledge, RLQ analysis and/or fourth-corner method have never been applied to the *NWS* of the Black Sea due to the lack of trait information for this area. Thus, in this PhD work, we provided the first study to apply these multivariate statistic approaches on the macrobenthic communities found over the *NWS* of the Black Sea.

6.2.2 General advances

Benthic research still lags behind phytoplankton (e.g. Litchman et al. (2013) and freshwater studies (e.g. Verberk et al. (2008) in considering multiple traits simultaneously, particularly the concept of trade-offs (Martini et al., 2021). Globally, marine community ecology based on multiple biological traits is still poorly studied (Beauchard et al., 2017). Previous studies have successfully assessed life-history strategies based on species traits for macrobenthos (Beauchard et al., 2022; Quell et al., 2021; Villnäs et al., 2018); coral reefs (Darling, Alvarez-Filip, Oliver, McClanahan, & Côté, 2012); copepods (Benedetti, Vogt, Righetti, Guilhaumon, & Ayata, 2018); freshwater invertebrates (C. R. Townsend & Hildrew, 1994; Usseglio-Polatera et al., 2000); phytoplankton (Edwards, Litchman, & Klausmeier, 2013); and marine fishes (Beukhof et al., 2019).

In this PhD work, response traits, traits that respond in a specific way to environmental factors, are combined (e.g. age at maturity, longevity, offspring size) to derive specific life-history strategies. Mapping of these strategies can be useful for ecosystem health assessment and may be helpful for conservation policy. The development of predictive modelling tools for trait distribution represents a significant breakthrough in the field of marine functional biogeography, historically based on the concept of species (Violle et al., 2014). The application of *Species Distribution Models (SDMs)* in the marine environment is recent (Robinson et al., 2011) and has been essentially focused on the distribution of commercial fish species (Cheung et al., 2010) given the need for ecosystem-based fisheries management. In comparison, very few studies have applied *SDMs* to the simulation of patterns of marine benthos (Degraer et al., 2008). This lack of distribution models is even more critical when it comes to the biological traits of the species, but it is worth noting that it is different for the mapping of bioturbation potential with maps over various coastal ecosystems using trait distribution models (e.g. regression model, machine learning) or spatial interpolation (e.g. Weinert et al. (2022); Gogina et al. (2017, 2022)).

6.3 Ways to improve this work

Although the methodology and analyses were conducted on the northwestern shelf of the Black Sea, it is important to note that the use of a trait-based approach can facilitate the transferability of the results obtained in this thesis to other marine ecosystems with different taxonomic compositions (Poff, 1997; Bremner et al., 2003). At the end of this thesis, we suggest several directions that could be pursued to further improve the study of the relationships between the traits of macrobenthos and environmental conditions on the shelf scale. First, further studies can expand the trait data set by adding new physiological or functional traits. For example, the presence of haemocyanin, which increases affinity for dissolved O_2 , could be used as a marker of adaptation to hypoxia (Pihl et al., 1991). The abiotic data set can also be extended by including additional higher-resolution abiotic predictors. In particular, the best approach will be to use *in-situ* measurements of bottom abiotic conditions to have a better correspondence between observed traits distributions and local abiotic drivers. One other way to improve this work can be achieved through an **ensemble modelling approach**, that will combine multiple types of *TDMs* to build an "average" predicted distribution. Ensemble modelling has been suggested as one way to overcome the difficulties in choosing an appropriate method and maximising the benefit from each approach (Grenouillet et al., 2011; Araújo & New, 2007). Therefore, ensemble models can improve the accuracy of the predictions and enable more robust forecasts (Araújo & New, 2007; Elith et al., 2006). An important perspective for future work concerns the mapping of uncertainties in predicted spatial distributions. Providing spatially explicit uncertainty maps would allow users to assess the reliability of model outputs and facilitate their use in management and decision-making contexts (Elith et al., 2002).

For this purpose, Gaussian Process Regression represents a particularly promising approach, as it directly yields uncertainty estimates associated with model predictions without requiring computationally expensive bootstrapping procedures (Golding & Purse, 2016). Finally, **functional biodiversity** indices can be computed from our set of biological traits. Functional diversity refers to the number, type, and distribution of functions performed by organisms within an ecosystem (Tilman, 2001). Functional diversity can be assessed through different diversity indices, such as the three indices proposed by Villéger et al. (2008), which quantify each facet of functional diversity for a community with species distributed in a multidimensional functional space. These three indices are: **functional richness** (i.e. volume of the functional space occupied by the community), **functional evenness** (i.e. regularity of the distribution of abundance in this volume), and **functional divergence** (i.e. divergence in the distribution of abundance in this volume). The three diversity indices are independent of each other and have the potential to highlight the role of biodiversity on ecosystem functioning Villéger et al. (2008). Patterns and changes in functional diversity can inform about spatial and temporal variation in trait diversity, about the processes that drive assembly, and whether assemblages are likely to contain redundant species (Petchey & Gaston, 2006). Future studies should compare multiple periods (e.g. mid-nineties and the current period 2008-2017) using functional biodiversity indices to assess long-term changes in macrobenthic functioning.

6.4 Perspectives of this work

Building on this PhD work, I outline several paths for future development in the following section. The next step of this work will be to derive ecosystem functions and services from the distribution of ecological processes estimated in this thesis. Although the importance of biodiversity to ecosystem functioning is widely acknowledged (Cardinale et al., 2012; Worm et al., 2006), there are still few studies exploring the mechanistic links between biodiversity and ecosystem functioning (*Biodiversity - Ecosystem Functions (BEF)*) at spatial scales relevant for environmental management (Harrison et al., 2014). The climate and biodiversity crisis add urgency to develop new tools able to predict how changes in biodiversity will impact ecosystem functions and deliver benefits that support life and economies (Bulling et al., 2010; Snelgrove et al., 2014). To fill this gap, we propose to extend this PhD work by incorporating benthic traits variability into an ocean model to better understand the links between biodiversity, biophysical relations and ecosystem functions and services at scales (**Figure 6.3**).

- What are the relationships between benthic functional biodiversity, environmental conditions, and ecosystem functions at the shelf scale (Perspective 1, P1) ?
- How can we predict the impact of global changes and human pressures on benthic biodiversity and ecosystem functions (Perspective 2, P2) ?
- How can ecosystem functions be translated into ecosystem services and economic value (Perspective 3, P3) ?

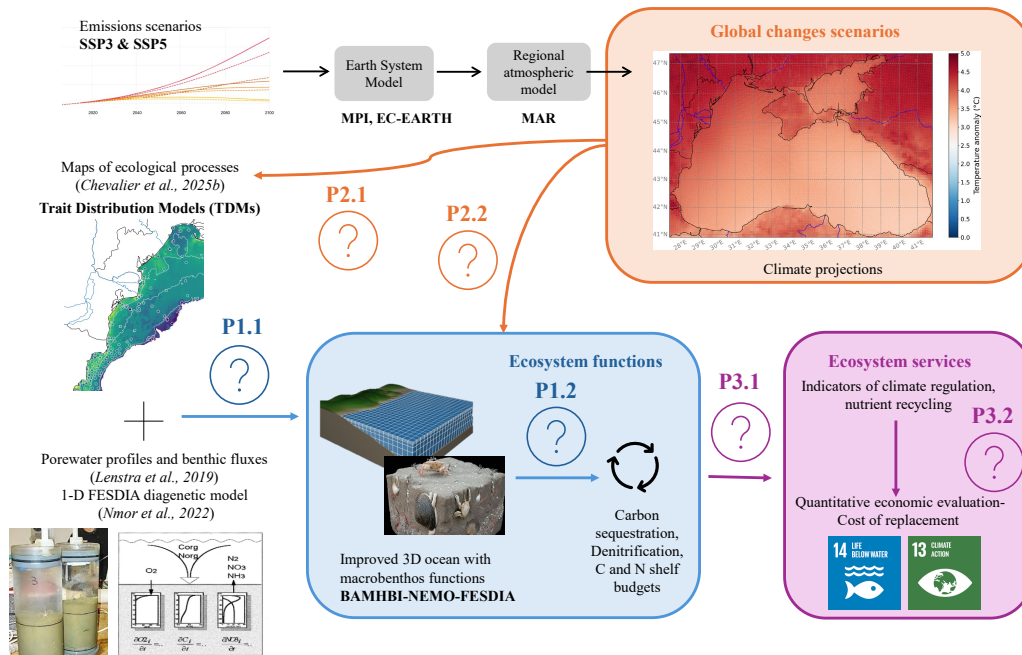


Figure 6.3: Flowchart of the different perspective of this PhD work. Map of ecological process (here biomixing potential) from Chevalier et al. (2025b) and SSP from PwC New Zealand. Map of temperature (°C) at the horizon 2100 from Modèle Atmosphérique Regional (MAR) was given by courtesy from J.-F. Grailet.

6.4.1 Scaling-up Biodiversity-Ecosystem Functioning (BEF) relationships

Current *BEF* studies are often limited to highly controlled, small-scale laboratory experiments, making it challenging to extrapolate the findings to natural, heterogeneous ecosystems (Snelgrove et al., 2014; Thrush et al., 2017). **Scaling up *BEF*** relationships to spatial and temporal scales relevant to *EBM* is a major challenge that could improve management and conservation efforts (Snelgrove et al., 2014). In marine systems, strong non-linearity and context dependency limit the transferability of results from local experiments to larger scales or long-term predictions (Raffaelli et al., 2003). Consequently, the role of benthic biodiversity in regulating ecosystem functioning on continental shelves remains poorly explored. To address this gap, we propose a trait-based approach of biodiversity to bridge scales between individual species characteristics (e.g. burrowing depth, mobility) to ecosystem functions at the shelf scale. Scaling approaches in ecology assume that biological traits of species are the primary factors through which organisms impact ecosystem processes (McGill et al., 2006). These traits influence ecological processes such as sediment mixing (biomixing) or ventilation of burrows (bioirrigation), and are precursors to ecosystem functions such as carbon sequestration, nutrient recycling or habitat creation (**Figure 1.2**). Trait-based ecology is now at the forefront of ecological research and offers a promising method for upscaling *BEF* relationships in complex marine environments (De Bello et al., 2010).

6.4.2 Linking biodiversity, environment and ecosystem functions at the shelf scale (P1)

6.4.2.1 Integrating benthic life in ocean numerical model (P1.1)

Building on this PhD work, we propose to improve a current ocean numeric model to mechanistically link biodiversity, environmental conditions, and global biogeochemistry and to quantify the essential role of the coastal ocean in global carbon and nutrient cycling (**Perspective P1.1** in **Figure 6.3**). Although the role of benthic fauna is widely recognised in benthic processes, most state-of-the-art ocean models still ignore life at the seafloor and its variability (Snelgrove et al., 2014; Bianchi et al., 2021). This shortcoming seriously limits our ability to accurately evaluate the importance of macrobenthos in delivering key ecosystem functions, such as carbon sequestration and nutrient recycling. We propose to address this deficiency by revising the current formulation of sediment processes in a numerical ocean model to explicitly account for ecological processes delivered by the macrobenthos (e.g. biomixing, bioirrigation). This new formulation may improve the representation of the benthic-pelagic coupling in large scale ocean model, correct shelf budgets considering the influence of the sea floor biota. This aspect is almost entirely absent from current state-of-the-art modelling frameworks and will constitute a significant breakthrough for the development of coupled benthic-pelagic biogeochemical models that would allow refining the assessment of the role of the world-wide continental shelf in the mediation of biogeochemical cycles.

The formulation of sedimentary processes in *BAMHBI* is currently oversimplified, ignoring the biology and vertical variability of sedimentary processes. The transport of solutes and solids through the sediment largely affects diagenesis and is dominated by the bioturbation activities of benthic organisms (Boudreau, 1997). However, so far, the parameterisation of biomixing and bioirrigation transports is still elementary not only in *BAMHBI* but in most of the benthic parts of ocean 3D models (Snelgrove et al., 2014). Currently, the parameterisation of biomixing and bioirrigation mechanisms in diagenetic models does not vary in space using constant mixing coefficients (D_b), mixed layer (L) and bioirrigation velocity. However, previous studies have shown that the diffusion coefficient can vary by up to three orders of magnitude (0.01–10 cm²/day) between contrasted seafloor settings (W. Zhang et al., 2019), highlighting the importance of changing the formulation of benthic processes in the model. In this PhD work, we showed a strong gradient in the intensity of biomixing and bioirrigation, with decreasing intensities towards the edge of the northwestern shelf of the Black Sea (**Figure 5.20**). Our maps of ecological processes (biomixing and bioirrigation) will be used to integrate the variability of benthic life into an existing diagenetic model (**FESDIA**; Nmor et al. (2022)). The improved model will be calibrated and validated with *in-situ* pore water concentrations and benthic-pelagic fluxes from a previous field campaign (Lenstra et al., 2019), covering a gradient from coastal stations with high benthic-pelagic fluxes to the edge of the shelf governed by anaerobic processes.

The model **OMEXdia** is a 1-D model that solves diagenetic processes that includes oxic and anoxic mineralisation in the marine sediment (Soetaert et al., 1996). The model has six state variables: oxygen, nitrate, ammonium, oxygen demand unit (ODU; meaning the lump sum of reduced substances generated in anoxic mineralisation) and two classes of organic carbon with fast and slow degradability. Advection, molecular diffusion (for solutes) and bioturbation (for solids) cause transport of modelled substances between layers of the model. The FESDIA model (Iron and Sulphur diagenesis) was developed from OMEXdia to explicitly model the oxic and anoxic biogeochemical processes occurring in the sediment (Nmor et al., 2022). In FESDIA, biomixing is modelled as a diffusive process, while bioirrigation is described as an enhancement of sediment-water exchange over molecular diffusion. The choice of FESDIA is motivated by the importance of suboxic and anoxic remineralisation processes in the Black Sea. For example, previous study evidenced that manganese (*Mn*) reduction appears to be a major component of sediment organic carbon remineralisation in oxic Black Sea sediments (Thamdrup et al., 2000). FESDIA will be coupled with the existing biogeochemical model of the water column (i.e. *BAMHBI*).

6.4.2.2 Ecosystem Functions (P1.2)

This new model formulation will also help to better quantify key ecosystem functions such as nutrient recycling, denitrification, and carbon sequestration (**Perspective P1.2** in **Figure 6.3**). For example, a previous study estimated that 324×10^3 tons of nitrogen per year are lost through denitrification based on in situ data analysis and model results (Grégoire & Friedrich, 2004). Therefore, shelf sediments are an efficient sink for riverine nitrogen, trapping about 50% of the annual river discharge in total inorganic nitrogen. However, this evaluation may underestimate the true potential of denitrification in shelf sediments as the influence of macrobenthos was not considered. Quantifying denitrification is crucial for effectively managing eutrophication in coastal ecosystems and mitigating its consequences, such as bottom hypoxia and the formation of harmful algal blooms.

6.4.3 Global change impacts on marine benthic biodiversity and ecosystem functions (P2)

Environmental stressors such as climate change, deoxygenation, and eutrophication are expected to significantly alter benthic ecosystems and biodiversity, with cascading effects on biogeochemical cycles and ecosystem functioning at the shelf scale, in ways that we currently ignore and lack comprehensive approaches to deal with this issue (Díaz et al., 2006). Additionally, in the coastal zone, anthropogenic stressors such as eutrophication, pollution, and habitat destruction worsen the situation, making it difficult to understand and predict how living communities will respond to disturbances due to non-linear combination of climatic and non-climatic stressors. The importance of biodiversity changes on a large scale is less clear, as lessons from local experiments and theory do not seem to extend to long-term, large scale management decisions due to the strong nonlinearity of ocean processes (Raffaelli et al., 2003).

Understanding how benthic biodiversity and ecosystem functioning respond to the combined effects of climatic and non-climatic stressors is, therefore, a major scientific challenge.

6.4.3.1 Global change impacts on marine benthic biodiversity (P2.1)

The diversity of bottom environmental conditions on the *NWS* (e.g. oxic/hypoxic, eutrophicated/oligotrophic), benthic habitats (e.g. biogenic reefs of mussels) as well as the synergetic actions of stressors (e.g. eutrophication, deoxygenation, climate change) offer unique conditions for connecting the benthic biota traits, the environment, and stressors. The Black Sea is a hotspot of global changes due to its landlocked nature and strong coupling with the atmosphere (see 1.5.4). The *Modelling for Aquatic SysTems (MAST)* team predicted an increase of 0.43 ± 0.02 °C per decade in sea surface temperature between the period 2015-2100 under the worst-case scenario. This sharp increase will likely affect biodiversity, leading to unprecedented changes in the marine ecosystem. A strong decrease in the volume of oxygenated waters is also predicted under both SSP1-2.6 (-10%) and SSP5-8.5 (-57%).

In this PhD work, we showed that biomixing and bioirrigation are inhibited below a certain threshold of dissolved oxygen concentrations, raising concerns about the ability of biodiversity to support key ecosystem functions. In the next steps of this work, we will combine scenarios of environmental changes with our developed *TDMs*, in **Chapter 5**, to predict changes in the spatial distribution of traits of the macrobenthos (**Perspective P2.1** in **Figure 6.3**). We will benefit from climate projections (for the period 1950-2100) produced in the framework of the NECCTON project. In short, they used two global earth system models, MPI (Giorgetta et al., 2013) and EC-Earth (Döscher et al., 2022), under the two *Shared Socioeconomic Pathway (SSP)* SSP1 and SSP5 (respectively the most optimistic and the most pessimistic scenarios), to force the regional climate model *Modèle Atmosphérique Régional (MAR)*, developed at the University of Liège (Grailet et al., 2025). *MAR* provides the atmospheric forcing for the coupled biogeochemical-physics model *BAMHBI-NEMO*.

6.4.3.2 Global change impacts on ecosystem functions (P2.2)

The potential modifications in macrobenthic trait distribution and bottom environmental conditions are expected to impact benthic-pelagic coupling. These changes will be evaluated using the revised diagenetic model that integrates information on biota characteristics. The effect of climate change on key ecosystem functions, mediated through biodiversity changes, may be assessed (**Perspective P2.2** in **Figure 6.3**). The outputs from the the improved model will be used to estimate changes in multiple ecosystem functions such as carbon sequestration and denitrification.

6.4.4 Evaluating how do biodiversity underpin ecosystem services (P3.1)

Biodiversity provides a wide range of **ecosystem services** essential to human well-being, with a first global estimate valuing these services at \$33 trillion (Costanza et al., 1997). Despite the vital importance of ecosystem services, their value is often underestimated in policy and management decisions because we still struggle to quantitatively evaluate them. Most publications about ecosystem services are conceptual and barely include empirical measurements of ecosystem services (Carpenter et al., 2009). This lack of knowledge is even more pronounced for marine ecosystems, with most studies on ecosystem services focusing on the terrestrial realm (Liquete et al., 2013). This lack of quantitative information hampers informed decision-making and sustainable management of coastal ecosystems (Costanza et al., 2017). Based on the model incorporating benthic functional biodiversity, it will be possible to estimate ecosystem services driven by benthic biodiversity, suitable for integration into established economic valuation frameworks and ecosystem-based management (**Perspective P3.1** in **Figure 6.3**).

Assessing and quantifying every aspect of ecosystem services is a challenging task (Kremen & Ostfeld, 2005) and the use of indicators (e.g. proxies for complex phenomena) can facilitate this process (Hattam et al., 2015). Quantitative indicators can be used to connect ecosystem functions estimated in **Perspective P1.2** with key ecosystem services. For example, indicators may include carbon burial (tons of $C\ m^{-2}\ year^{-1}$) and depth of carbon in sediment for carbon sequestration through bioturbation; as well as burial of nitrogen (tons of $C\ m^{-2}\ year^{-1}$) and denitrification (tons of $N\ m^{-2}\ year^{-1}$) for nitrogen removal.

6.4.4.1 Economic valuation of biodiversity-driven ecosystem services (P3.2)

The economic evaluation of ecosystem services based on ecosystem functions is the ultimate step in *EBM* work. Ecosystem services can be valued economically using various methods that aim to quantify their contributions to human welfare, both in market (e.g. carbon and nutrient credits) and non-market terms such as willingness-to-pay surveys, cost savings or replacement costs (Costanza et al., 1997) (**Perspective P3.2** in **Figure 6.3**). For regulating services (i.e. storing and recycling of nutrients, climate mitigation), replacement or avoided damage costs have been recommended as appropriate valuation techniques (Schenau et al., 2022).

6.5 General conclusions and recommendations for management

This PhD thesis advances our understanding of the functional biodiversity of the northwestern shelf of the Black Sea, a region where trait-based ecology was not studied before this work. We evidenced a clear relationship between the macrozoobenthic species and abiotic conditions related to organic pollution (**Chapter 3**). In particular, the reduction of river inputs and the associated decrease in bottom hypoxia led to an improvement in the ecological status of macrobenthic communities, as well as an increase in taxonomic richness. Building on the analysis of species and abiotic conditions, we built a comprehensive set of response and effect traits for macrobenthic species (**Chapter 4**), enabling the transition from species-environment relationships to trait-environment relationships (**Chapter 5**). Knowing the local relationships between the abiotic data and the traits (meaning at one punctual sampling sites), we were able to predict trait distributions in areas lacking field observations. By developing *TDMs* based on neural networks, we produced continuous, full-coverage maps of benthic traits across the study area (**Chapter 5**). Two complementary types of trait-based maps were generated: maps of life-history strategies, highlighting spatial patterns in sensitivity to environmental pressures, and maps of ecological processes (e.g. biomixing, biodeposition), illustrating spatial variability in functional potential.

These outputs represent a significant step forward in the spatial assessment of benthic habitats in the Black Sea and provide a robust foundation for ecosystem-based management. In contrast to conventional approaches that rely primarily on physical seabed maps, which do not capture biological diversity or ecosystem functioning (Galparsoro et al., 2014; Gerovasileiou et al., 2019), trait-based maps offer ecologically meaningful indicators that are directly relevant to management objectives. They are particularly valuable in the context of policy frameworks such as the *MSFD*, which aims to achieve *Good Environmental Status (GES)* across European seas.

Firstly, maps of benthic functions can support decision-making for a wide range of stakeholders, including environmental agencies, fisheries, aquaculture, and educators. These maps enable more targeted and efficient spatial management by identifying areas of high functional importance and areas of high vulnerability to disturbance (Bolam et al., 2023). For example, maps can inform licensing decisions by avoiding activities such as dredging and bottom trawling in regions that host functionally important and sensitive communities to ensure sustainable management. In this PhD thesis, maps help to identify hotspots of functions (e.g. areas of higher biomixing or biodeposition) and highlight areas with more vulnerable species that need longer time with no disturbances to achieve reproductive success (such as "Episodic" long-lived bivalves). Conservation of "Episodic" species should be a key priority as these species need a longer time with no perturbation to achieve reproductive success. These species should be properly monitored as they are indicators of the recovery of the Black Sea. "Episodic" species are also associated with higher biodeposition potential thus secure the delivery of key ecosystem functions and services.

The restoration of bivalves populations has been proposed as a mitigation mechanism of climate change as shellfish reefs contain significant pools of carbon, facilitate atmospheric CO_2 drawdown via filtration and biodeposition (Chowdhury et al., 2019; Fodrie et al., 2017). These habitats of key ecosystem engineers should be protected from human impacts through environmental legislation and management. Protection measures should also include restriction or regulation of activities that degrade sediment (dredging, bottom trawling) in the areas with the more vulnerable species, and we need to prioritise monitoring programs in vulnerable zones to detect functional loss early.

Furthermore, our trait maps can be superimposed with spatial data on anthropogenic pressures (bottom trawling, dredging, pollution) and climate change stressors (e.g. hypoxia) to identify zones of highest risk. This integrative approach supports the assessment of cumulative impacts and facilitates the implementation of holistic, ecosystem-based management strategies (Halpern et al., 2009, 2015; Goodsir et al., 2015). Benthic maps can also help for the establishment of *MPAs* to safeguard benthic habitats and enhance ecosystem resilience. Evidence suggest that protected areas can improve the resilience of species to extreme events, thereby supporting adaption to climate change (Micheli et al., 2012). In addition, maps are also powerful communication tools that enhance public understanding of marine ecosystems and are the missing link between decision-making and science (Wright, 2016).

Beyond management applications, this PhD work also open new perspectives for improving benthic-pelagic representation in ocean models. By integrating spatial variability in macrobenthic traits into the benthic compartment, ocean models can better represent ecosystem functions such as carbon burial and nutrient cycling (Snelgrove et al., 2018). This is particularly important for recognising the role of benthic life in climate regulation and move forward the inclusion of benthic ecosystems in national and international climate adaptation and mitigation strategies (Solan et al., 2020). In addition, our Trait Distribution Models in **Chapter 5** can be used to predict future distributions of benthic functions under global change scenarios and to inform policy-makers on the expected changes to come. Such predictive capacity in the long-term (several decades) is essential for supporting adaptive management and ensuring the sustainability of marine ecosystem services in a changing world.

Overall, this PhD work highlights the potential of a trait-based modelling approach to bridge the gap between local benthic biodiversity observations and maps relevant for large-scale and management-relevant applications. The delivery of maps detailing benthic functions and vulnerabilities represents a major step toward preserving the *GES* of the Black Sea, informing the design of *MPAs*, and supporting recommendations based on ecological considerations for biodiversity conservation.

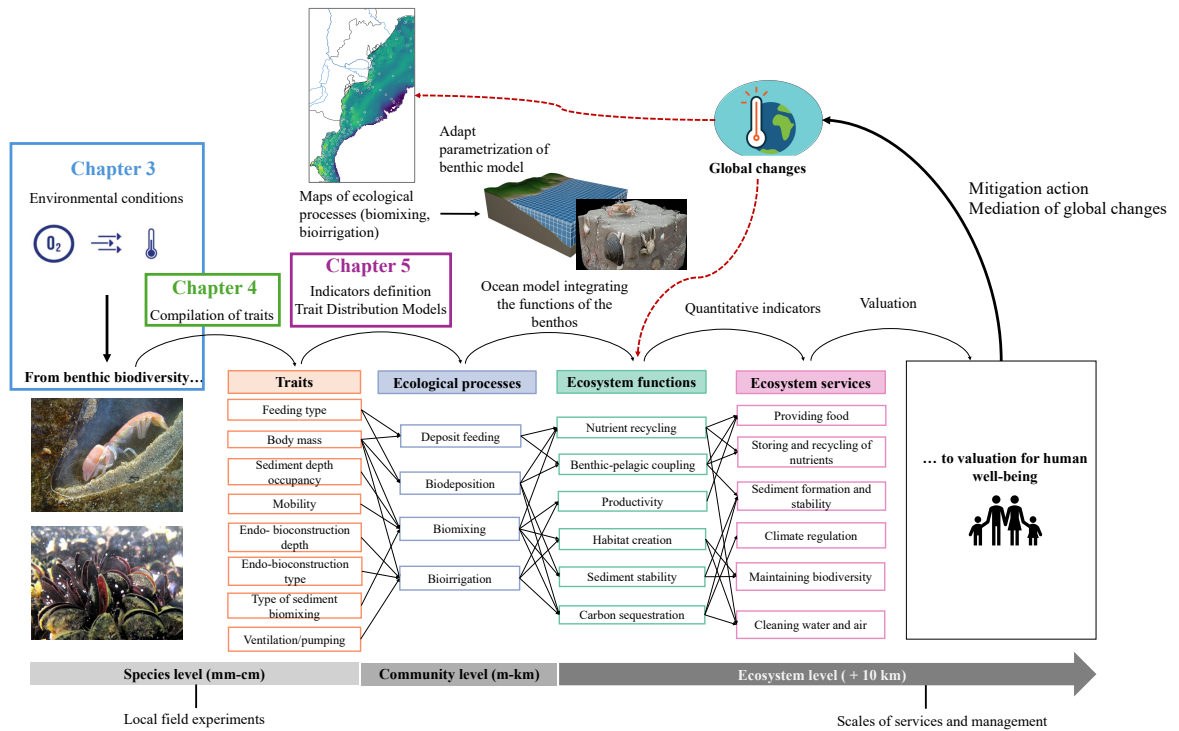


Figure 6.4: Framework describing the links between macrobenthic fauna, biological traits, ecological processes and ecosystem functions and services. The contributions of each chapter are highlighted with a colour box: **Chapter 3** linked benthic biodiversity and environmental conditions, **Chapter 4** linked benthic species to traits and **Chapter 5** linked traits to the maps of ecological processes and vulnerabilities at the shelf scale, which can directly support ecosystem-based management. The maps of ecological processes can be incorporated into ocean models to derive ecosystem functions and services and to help decision-making to mitigate global changes. The red dashed lines correspond to the impact of global changes on benthic traits and ecosystem functions. These impacts can be mitigated through adequate mitigation plans. Adapted from Snelgrove et al. (2014); Lam-Gordillo et al. (2020); Cardinale et al. (2012). First species picture: Monterey Bay Aquarium and second species picture: Pillon Roberto from Worms. Stressors picture: istock.



LIST OF PUBLICATIONS

The name of the PhD student is underlined in the citations below.

A.1 Published publications

- Chevalier, S., Beauchard, B., Teacă, A., Soetaert, K., & Grégoire, M. (2024). Partial recovery of macrozoobenthos on the northwestern shelf of the Black Sea. *Marine Pollution Bulletin*, 207, 116857. <https://doi.org/10.1016/j.marpolbul.2024.116857>.
- Chevalier, S., Beauchard, O., Teacă, A., Begun, T., Todorova, V., Vandenbulcke, L., K. Soetaert., & Grégoire, M. (2025a). A macrozoobenthic data set of the Black Sea northwestern shelf. *Scientific data*, 12(1), 957. <https://doi.org/10.1038/s41597-025-05311-2>. Dataset associated on Figshare (<https://doi.org/10.6084/m9.figshare.27888843.v1>).
- Chevalier, S., Beauchard, O., Vandenbulcke, L., Teacă, A., Begun, T., Todorova, V., K. Soetaert., & Grégoire, M. Trait modelling to predict benthic functions and vulnerabilities across Black Sea seascapes. *Scientific Reports*, 15(1), 39076. <https://doi.org/10.1038/s41598-025-24508-4>.
- Grégoire, M., Vandenbulcke, L., Chevalier, S., Choblet, M., Drozd, I., Graillet, J.-F.,...Soetaert, K. (2026). The BiogeochemicAl model for hypoxic and benthic influenced areas: BAMHBI v1.0. *Geoscientific Model Development*, 19(5), 2137–2175. <https://doi.org/10.5194/gmd-19-2137-2026>.

A.2 Ongoing publication

- Chevalier, S., Drozd, I., Grailet, J.-F., Vandenbulcke, L., Soetaert, K., & Grégoire, M. (2026). Predicting functional biodiversity changes under various hypoxia scenarios over the northwestern shelf of the Black Sea, to be submitted in *Environmental Research Letters* in May 2026.

A.3 Oral communication

- Chevalier, S., Beauchard, O., Teacă, A., Barth, A., Soetaert, K., & Grégoire, M. Mapping lifehistory strategies of macrobenthic communities. 2nd International Joint Conference MARBLUE 2024, Constanta, Romania, 23-25 October 2024.
- Chevalier, S., Beauchard, O., Teacă, A., Begun, T., Vandenbulcke, L., Soetaert, K., & Grégoire, M. From benthic functional biodiversity to the mapping of ecosystem functions: a case study over the Black Sea northwestern shelf. EGU General Assembly 2025, Vienna, Austria, 27 Apr–2 May 2025, EGU25-12826, <https://doi.org/10.5194/egusphere-egu25-12826>, 2025.
- Chevalier, S., Beauchard, O., Teacă, A., Begun, T., Vandenbulcke, L., Soetaert, K., & Grégoire, M. Linking Deoxygenation to Macrobenthic Fauna and Ecosystem Functioning in the Black Sea. Ocean Sciences Meeting 2026, Glasgow, United Kingdom, 22-27 February 2026.

CHAPTER 3 : SUPPLEMENTARY MATERIALS

In this Appendix, we provided the main supplementary materials associated with **Chapter 3**. The complete supplementary information is available on the online version of the paper published in Marine Pollution Bulletin.

The spatial distribution of the taxonomic richness for the two periods is shown in **Figure B.1** and a summary of key statistics compared both periods is provided in **Table B.1**. In 1995, the taxonomic richness is higher in the southwestern part of the shelf compared to the northeastern part and a north-south gradient with respect to the Danube's mouth is evidenced. During 2008-2017, this north-south gradient is not apparent, and the distribution of the taxonomic richness is more homogeneous. The lowest specific richness is observed at the deepest offshore stations from the 100 m isobath where the upwellings of euxinic waters from the deep-sea basin occur and in the very shallow coastal sites located north of the Danube's mouth.

Table B.1: Summary of taxonomic richness per period and sampling station: for each period: total number of sites, total number of taxa and mean taxonomic richness (\pm standard deviation (std)) per sampling site.

Period	Total number of sites	Total number of taxa	Mean (\pm std)	Max	Min
1995	27	62	9 +/- 5	24	2
2008-2017	111	152	22 +/- 10	45	1

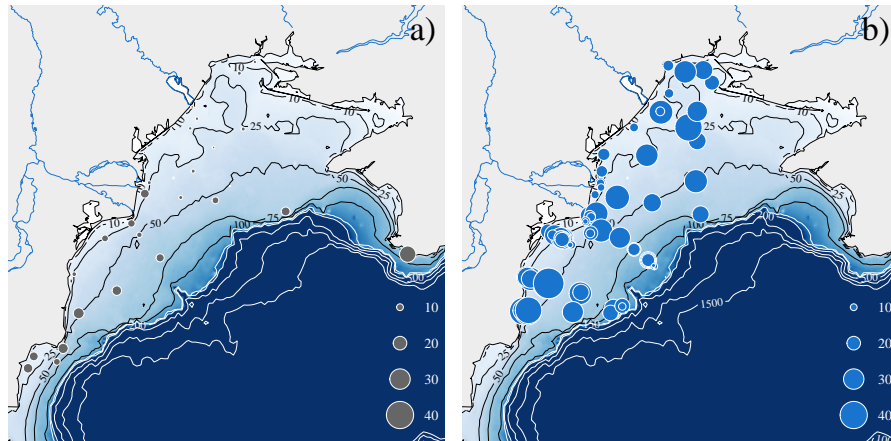


Figure B.1: Spatial distribution of taxonomic richness per sampling station over the Black Sea northwestern shelf. a) Grey dots, in 1995 and b) blue dots, in 2008-2017. Legend, dot size proportional to the taxonomic richness per sampling site.

Figure B.2 gives the *BIC* obtained for regime shift models from 1 to 5 breakpoints for the different biogeochemical properties of the river inputs (i.e. nitrates, phosphates, ammonium, particulate organic carbon and nitrogen, dissolved carbon and nitrogen quantities discharging in the Black Sea (mmol/month)). A summary with the optimal number of breakpoint(s) and the corresponding date for each variable is provided in **Table B.2**.

Table B.2: Summary of the regime shifts analysis of the river data. For each environmental variable (expressed in mmol/month, the optimal number of breakpoints according to the *BIC* and the corresponding date.

Biogeochemical variables	Optimal number of breakpoint(s)	Date
Nitrates	1	March 1996
Phosphates	2	April 1994 and June 1999
Ammonium	1	March 1996
Particulate Organic Carbon	1	March 1996
Particulate Organic Nitrogen	1	March 1996
Dissolved Organic Carbon	1	April 1995
Dissolved Organic Nitrogen	1	April 1995

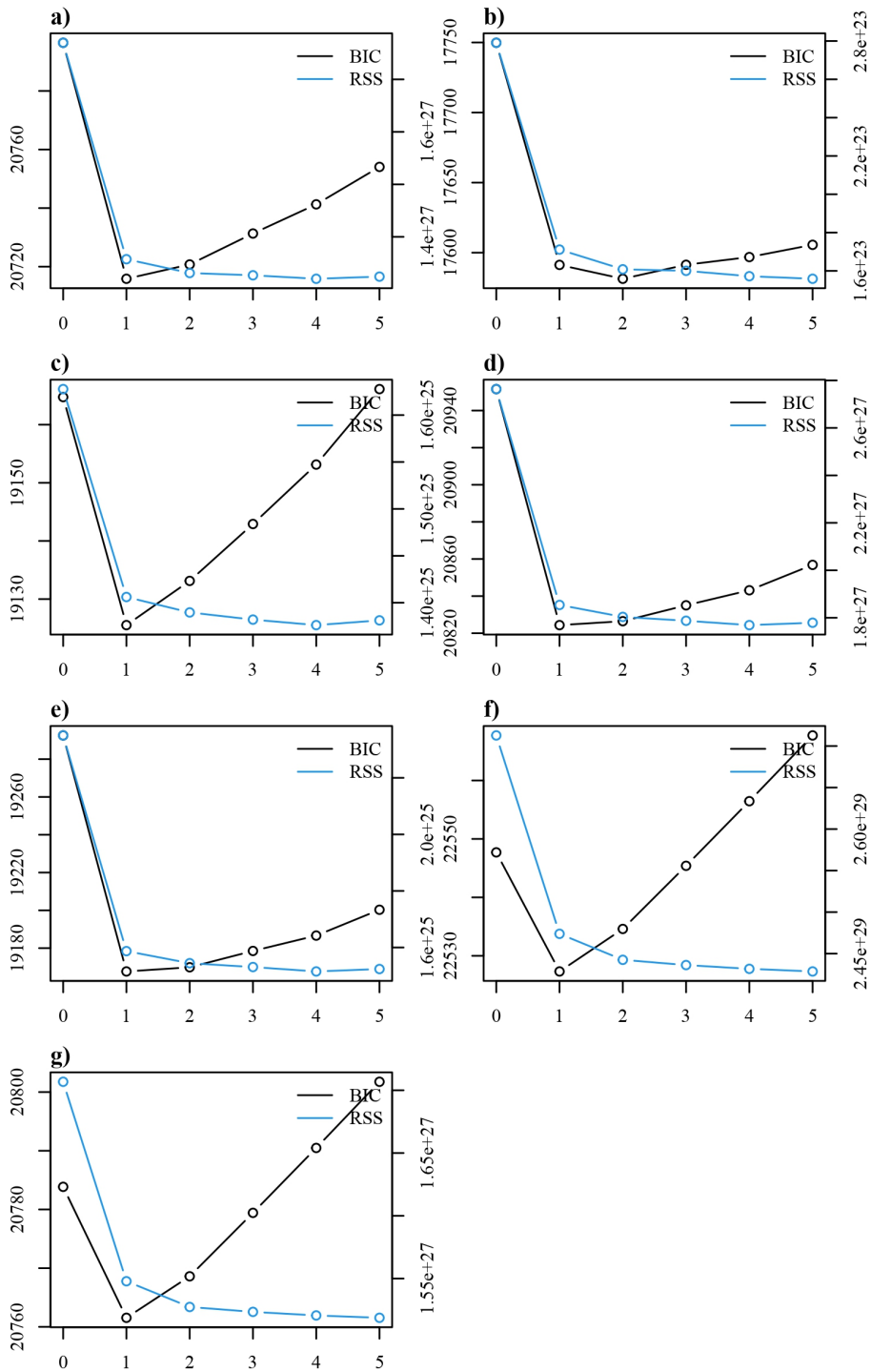


Figure B.2: BIC (right scale) and residual sum of squares (RSS, left scale) of regime shifts models from 1 to 5 breakpoints. From top-right to bottom-left: a) nitrates, b) phosphates, c) ammonium, d) particulate organic carbon, e) particulate organic nitrogen, f) dissolved organic carbon and g) dissolved organic nitrogen.

Figure B.3 compares the proportion of the *AMBI* ecological groups between 1995 and 2008-2017. From 1995 to 2008-2017, there is an increase in the proportion of the sensitive taxa (+ 7%) and indifferent taxa (+ 13%) while there is a slight decrease in the proportion of opportunistic taxa (-5%) and of taxa tolerant to an enrichment in organic matter (- 15%).

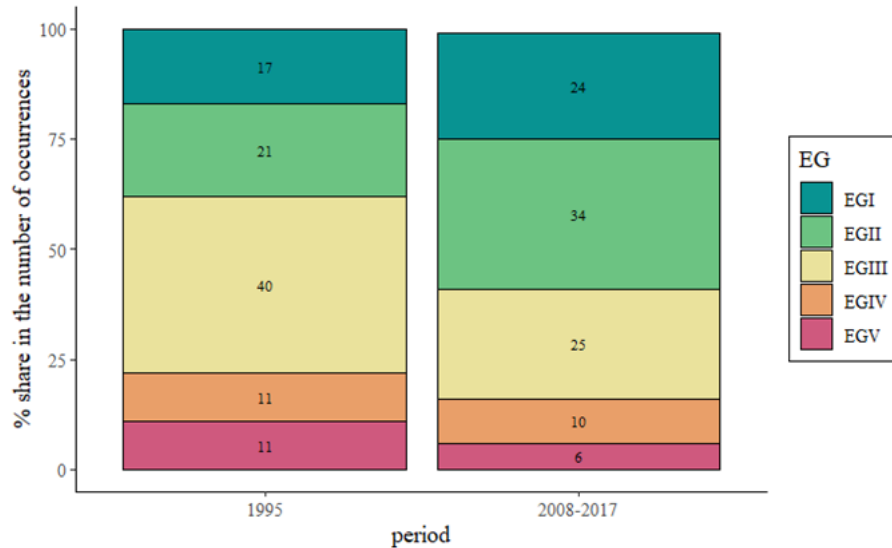


Figure B.3: Comparison of the percentage share in the number of occurrences per saprobic group (EG) between 1995 and 2008-2017. EG I (in green), EG II (in light green), EG III (in yellow), EG IV (in orange) and EG V (in red).

Table B.3 gives the conversion into qualitative modalities for each abiotic descriptor. Substrate type and depth are *in-situ* variables while the other are derived from the hydrodynamical-biogeochemical coupling of *NEMO 4.2* with *BAMHBI*.

Table B.3: Detailed information on abiotic descriptors; each qualitative descriptor is divided into modalities with corresponding range value.

Descriptor	Modality for qualitative descriptor	Range of values
Physical variables		
Depth	Very shallow	< 20
	Shallow	[20; 40[
	Intermediate	[40; 60[
	Deep	[60; 80[
	Very deep	≥ 80
Substrate type	Sand and muddy sand	-
	Mud and sandy mud	-
	Mixed sediments	-
	Very Cold	< 8

	Cold Intermediate Warm Very warm	[8; 10[[10; 12[[12; 14[≥ 14
Thermal seasonality	Very low Low Intermediate High Very high	< 10 [10; 20[[20; 30[[30; 40[≥ 40
Salinity (SAL)	Low Intermediate High	< 17 [17; 19[≥ 19
Bottom shear stress (SHEAR)	Very low Low Intermediate High Very high	< 0.01 [0.01; 0.02[[0.02; 0.03[[0.03; 0.04[≥ 0.04
Chemical variables		
Photosynthetically active radiation (PAR)	Very low Low Intermediate High Very high	< 0.005 [0.005; 0.015[[0.015; 0.025[[0.025; 0.035[≥ 0.035
Particulate organic carbon content (POC)	Very low Low Intermediate High Very high	< 5 [5; 10[[10; 15[[15; 20[≥ 20
Fast (or labile) sedimentary stock of organic content (fCSED)	Very low Low Intermediate High Very high	< 50 [50; 100[[100; 150[[150; 200[≥ 200
Slow (or semi-labile) sedimentary stock of organic content (sCSED)	Very low Low Intermediate High Very high	< 2500 [2500; 5000[[5000; 7500[[7500; 10000[≥ 10000

<p>Dissolved oxygen concentration (DOX)</p>	<p>Very low Low Intermediate High Very high</p>	<p>< 125 [125; 175[[175; 225[[225; 275[≥ 275</p>
<p>DOX seasonality</p>	<p>Very low Low Intermediate High Very high</p>	<p>< 10 [10; 30[[30; 50[[50; 70[≥ 70</p>



CHAPTER 4 : SUPPLEMENTARY MATERIALS

This appendix provides some guidelines to handle the provided trait data in R (<https://www.r-project.org/>) and synthesise new traits based on the provided ones. The code and data used are freely accessible on the repository *Figshare*. This code is related to the publication published in Scientific Data (Chevalier et al., 2025a) corresponding to **Chapter 4** of this PhD work.

C.1 Handling the data

Load the provided commands and the package `ade4` (Chessel et al., 2004) for multivariate ordination.

```
Source (".../ bind_traits.R")
source (".../ syn_trait.R")
require(ade4)
```

Load the trait data and labels.

```
data <- read.csv("traits.csv", h = T)
lab <- read.csv("labels.csv", h = T)
head(data)
```

The trait data frame is vertically partitioned, each trait being a block of trait modality variables. The label data frame **lab** lists the successive traits coded in the column **Trait.code** and named in **Trait**. Within each trait, the column **Modality** provides the modality labels. The column **Column.name** contains the column labels of the trait data frame, hereafter called **tab** (see below).

```
head(lab, 20)
```

Column.name	Trait.code	Modality.code	Trait	Modality	
1	T1.M1	1	1	Life span	<1
2	T1.M2	1	2	Life span	1-3
3	T1.M3	1	3	Life span	3-10
4	T1.M4	1	4	Life span	10-20
5	T1.M5	1	5	Life span	>20
6	T2.M1	2	1	Age at maturity	<1
7	T2.M2	2	2	Age at maturity	1-3
8	T2.M3	2	3	Age at maturity	>3
9	T3.M1	3	1	Sexuality	Gonochorism
10	T3.M2	3	2	Sexuality	Homogamy
11	T3.M3	3	3	Sexuality	Protandry
12	T4.M1	4	1	Reproductive frequency	Sexual seasonal
13	T4.M2	4	2	Reproductive frequency	Sexual continuous
14	T4.M3	4	3	Reproductive frequency	Asexual
15	T5.M1	5	1	Fertilisation	Broadcasting
16	T5.M2	5	2	Fertilisation	Spermcasting
17	T5.M3	5	3	Fertilisation	Pairing
18	T6.M1	6	1	Annual fecundity	<10e2
19	T6.M2	6	2	Annual fecundity	10e2-10e3
20	T6.M3	6	3	Annual fecundity	10e3-10e4
...					

Extract the trait data. **tab** is the taxa × trait modalities data frame linked to **lab** by its column names sorted in the same order.

```
data.frame(colnames(data))
tab <- data[7:124]
rownames(tab) <- data$Taxon
```

Selection of two sets of response traits. Firstly, identify them.

```
unique(lab[c("Trait.code", "Trait")])
```

Trait.code	Trait	
1	1	Life span
6	2	Age at maturity
9	3	Sexuality
12	4	Reproductive frequency
15	5	Fertilisation
18	6	Annual fecundity
24	7	Offspring type
27	8	Offspring size
31	9	Offspring protection
35	10	Offspring development
40	11	Offspring benthic stage duration
45	12	Offspring pelagic stage duration
50	13	Offspring settlement size

```

54      14      Body mass
60      15      Body length
65      16      Mobility
70      17      Substratum depth occupancy
75      18      Epi-bioconstruction type
81      19      Epi-bioconstruction extension
85      20      Epi-bioconstruction size
92      21      Endo-bioconstruction type
95      22      Endo-bioconstruction depth
100     23      Endo-bioconstruction width
104     24      Ventilation/Pumping
107     25      Sediment mixing type
112     26      Biostabilisation
115     27      Feeding type

```

```

tab1 <- tab[lab$Trait.code %in% 15:16]
tab2 <- tab[lab$Trait.code %in% 1:13]
lab1 <- lab[lab$Trait.code %in% 15:16,]
lab2 <- lab[lab$Trait.code %in% 1:13,]

```

Bind **tab1** and **tab2** in a single data frame with an adapted label data frame by means of the command **bind.traits**.

```

w <- bind.traits(list(tab1, tab2), list(lab1, lab2))
tab.resp <- w$tab
lab.resp <- w$lab
head(tab.resp[1:10])

```

	T1.M1	T1.M2	T1.M3	T1.M4	T1.M5	T2.M1	T2.M2	T2.M3	T2.M4
Abra alba	0	1	0	0	0	0	1	0	0
Abra nitida	0	1	0	0	0	0	1	0	0
Abra segmentum	0	1	0	0	0	0	1	0	0
Acanthocardia paucicostata	0	0	1	0	0	0	1	1	0
Actinia equina	0	0	1	0	0	1	0	0	0
Alitta succinea	0	0	0	1	0	0	0	1	1

```


```

	T2.M5			
Abra alba	0			
Abra nitida	0			
Abra segmentum	0			
Acanthocardia paucicostata	0			
Actinia equina	0			
Alitta succinea	0			
...				

```

head(lab.resp, 50)

```

	Column.name	Trait.code	Modality.code	Trait	Modality
1	T1.M1	1	1	Body length	<1
2	T1.M2	1	2	Body length	1-3

3	T1.M3	1	3	Body length	3-10
4	T1.M4	1	4	Body length	10-20
5	T1.M5	1	5	Body length	>20
6	T2.M1	2	1	Mobility	Immobile
7	T2.M2	2	2	Mobility	Limited
8	T2.M3	2	3	Mobility	Slow
9	T2.M4	2	4	Mobility	Fast
10	T2.M5	2	5	Mobility	Very fast
11	T3.M1	3	1	Life span	<1
12	T3.M2	3	2	Life span	1-3
13	T3.M3	3	3	Life span	3-10
14	T3.M4	3	4	Life span	10-20
15	T3.M5	3	5	Life span	>20
16	T4.M1	4	1	Age at maturity	<1
17	T4.M2	4	2	Age at maturity	1-3
18	T4.M3	4	3	Age at maturity	>3
...					

C.2 Multivariate ordination of taxa and traits

Prior to ordination, the data are fuzzy coded according to (Chevenet et al., 1994).

```
w <- prep.fuzzy.var(tab.resp, table(lab.resp$Trait.code))
```

Fuzzy Correspondence Analysis.

```
fca <- dudi.fca(w, scan = F)
```

Check the dimensionality of the data with the eigenvalues (**Figure C.1**).

```
barplot(fca$eig)
```

Represent the taxa (black dots) and traits (one per graphical window) along the two first axes.

```
par(mfrow = c(5, 3))
for(i in 1:15){
  s.label(fca$li, clab = 0)
  s.distri(fca$li, tab.resp[lab.resp$Trait.code == i],
    lab = lab.resp$Modality[lab.resp$Trait.code == i],
    cpoint = 2, clab = 1.5, cstar = 0,
    sub = unique(lab.resp$Trait)[i],
    possub = "topleft", csub = 2, add.p = T)
}
```

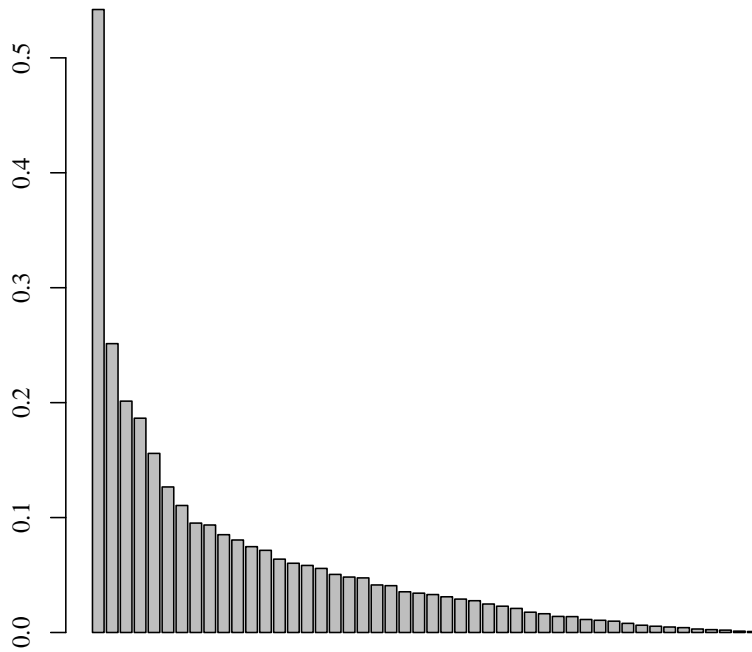


Figure C.1: Barplot of the eigenvalues of the fuzzy correspondence analysis applied on the traits table

The pattern clearly highlights the three life strategies identified (Beauchard et al., 2022): on the right side, stress-resistant taxa (Precocial-strategists); on the top, disturbance-resilient taxa (Opportunist-strategists); on the left, long-lived and slow growing species typical of low hydrodynamics (Episodic-strategists) (**Figure C.2**).

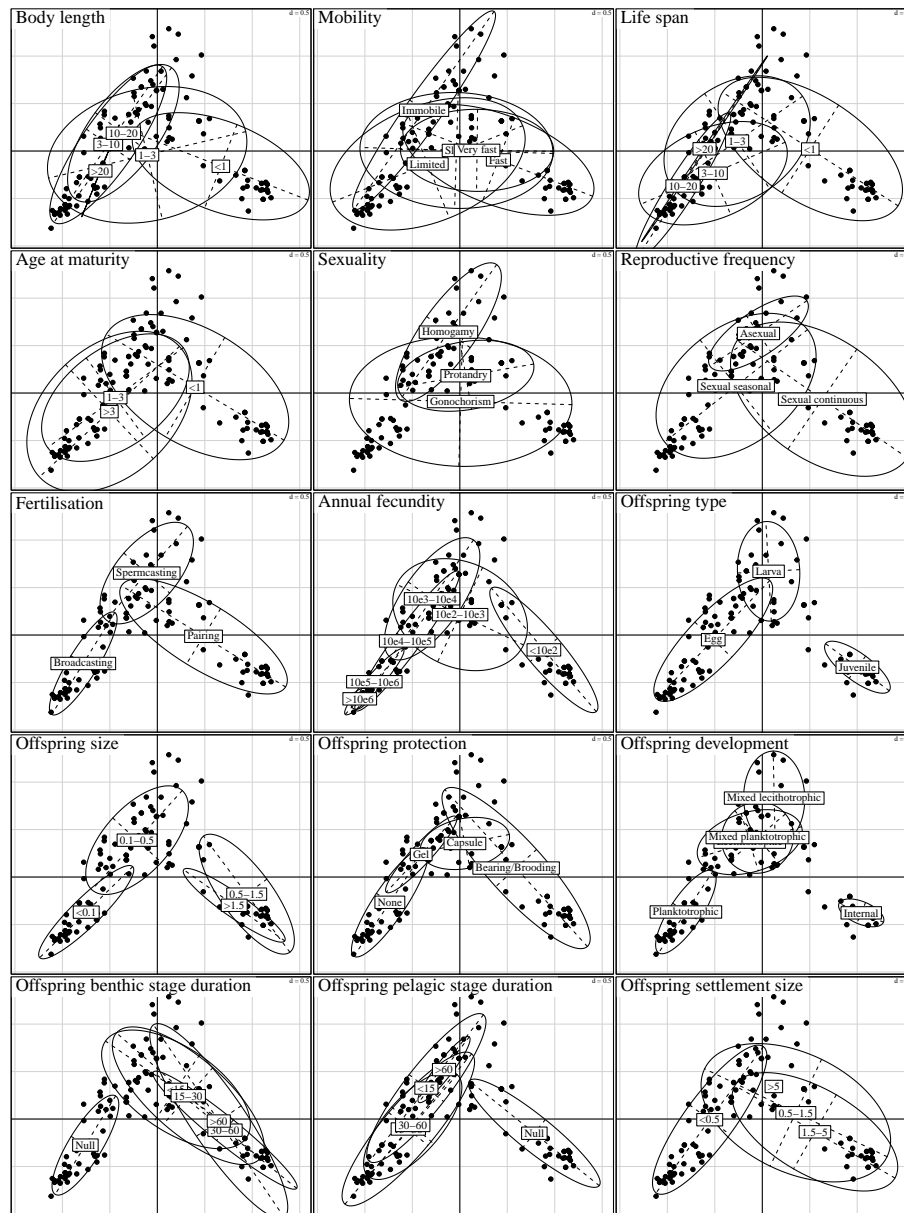


Figure C.2: Distribution of trait modalities (ellipse) for each trait (windows) on the first two axes of the FCA. Taxa are represented by black dots. The trait modalities are positioned at the centre of gravity of their respective taxa. Size of the grid in the top-right corner.

C.3 Synthesising a combined trait indicator

The data set is composed of traits commonly described in natural history studies. However, some processes, especially ecosystem functions, require the combination of several traits so that they be relevantly expressed. The command `syn.trait` enables the combination of several traits with the desired formula. It was designed in a flexible way to specifically handle traits as blocks of trait modality variables for which transformations such as fuzzy coding and scaling can be needed to adjust trait weights in the calculation. **Bioturbation**, that includes sediment bioturbation types

and bioirrigation (Kristensen et al., 2012), is a relevant example. The calculation proposed here is illustrated in (Beauchard et al., 2023) for several effect traits. **Biomixing**: There are four different types of sediment biomixing (Kristensen et al., 2012; François et al., 1997): biodiffusion, upward conveying, downward conveying and regeneration. They are the four last modalities of the trait 25 (the first one being “None”, i.e. no affinity for none of them). Following a formula that make consensus nowadays (Queirós et al., 2013), here is the calculation of biodiffusive mixing potential at the species level.

Trait 25:

```
lab[lab$Trait == "Sediment mixing type",]
```

	Column.name	Trait.code	Modality.code	Trait	Modality
107	T25.M1	25	1	Sediment mixing type	None
108	T25.M2	25	2	Sediment mixing type	Diffusion
109	T25.M3	25	3	Sediment mixing type	Upward conveying
110	T25.M4	25	4	Sediment mixing type	Downward conveying
111	T25.M5	25	5	Sediment mixing type	Regeneration

We are interested in the modality “Diffusion”, but also in the modality “None” as some epibenthic taxa are intermittently buried and weakly diffusive (e.g. mysids). Therefore, they are scored 1/1/0/0/0; through fuzzy coding, their diffusive ability is lower (0.5/0.5/0.0/0.0/0.0) than those also biodiffusers but permanently and more continuously buried (0.0/1.0/0.0/0.0/0.0). Hence, a new trait is specifically created:

```
w1 <- tab[lab$Trait.code == 25][1:2]
w2 <- lab[lab$Trait.code == 25,][1:2,]
w2$Trait <- rep("Biodiffusion", nrow(w2))
w <- bind.traits(list(tab, w1), list(lab, w2))
lab <- w$lab
tab <- w$tab
tail(lab, 4)
...
```

	Column.name	Trait.code	Modality.code	Trait	Modality
117	T27.M3	27	3	Feeding type	HeGr
118	T27.M4	27	4	Feeding type	CaSc
119	T28.M1	28	1	Biodiffusion	None
120	T28.M2	28	2	Biodiffusion	Diffusion

Then, identify additional traits contributing to biodiffusive mixing rate (amount of sediment displaced in all directions per time unit).

```
unique(lab[c("Trait.code", "Trait")])
```

	Trait.code	Trait
1	1	Life span
6	2	Age at maturity
9	3	Sexuality
12	4	Reproductive frequency
15	5	Fertilisation
18	6	Annual fecundity
24	7	Offspring type
27	8	Offspring size
31	9	Offspring protection
35	10	Offspring development
40	11	Offspring benthic stage duration
45	12	Offspring pelagic stage duration
50	13	Offspring settlement size
54	14	Body mass
60	15	Body length
65	16	Mobility
70	17	Substratum depth occupancy
75	18	Epi-bioconstruction type
81	19	Epi-bioconstruction extension
85	20	Epi-bioconstruction size
92	21	Endo-bioconstruction type
95	22	Endo-bioconstruction depth
100	23	Endo-bioconstruction width
104	24	Ventilation/Pumping
107	25	Sediment mixing type
112	26	Biostabilisation
115	27	Feeding type
119	28	Biodiffusion

The formula firstly considers the affinity for biodiffusion which is trait 28 (**which.traits = c(28,...)**); the modalities of this trait are multiplied by the scores 0/1 so that only effective biodiffusive mixing be non-null (**list.sco = list(c(0,1),...**). Following this scoring procedure, the next considered traits are body mass (14), mobility (16) and substratum depth occupancy (17). These traits are multiplied by the performance score profiles 1/2/3/4/5/6 (from < 0.001 to > 10.000 g AFDM), 0/1/2/3/4 (from immobile to very fast) and 1/1/1/1/1 (from 0 to > 30 cm deep), respectively.

Note that the score 0 for the modality “Immobile” in mobility induces a null performance. Prior to multiplication by performance scores, the command **fuz** indicates whether the raw scores (0 or 1) are processed by fuzzy coding so that the values for a species within a trait sum to 1. Then, the sum of the values following multiplication by performance scores returns a mean trait value (i.e. non-uniformly weighted mean). In the case of the last trait (vertical distribution), **fuz** is attributed **F** since we want to express the thickness of the sediment matrix in which the organism prospers. **trait.scale = T** enables the rescaling of each trait between 0 and 1 before the calculation (the traits get the same importance), specified by formula. Optionally, the final synthetic trait can be rescaled between 0 and 1 with **index.scale = T**.

```
dif <- syn.trait(tab = tab, lab = lab,
  which.traits = c(28, 14, 16, 17),
  list.sco = list(c(0, 1),
    c(1:6),
    c(0:4),
    c(rep(1, 5))),
  fuz = c(T, T, T, F),
  fun = c("sum", "sum", "sum", "sum"),
  formula = "x1 * x2 * x3 * x4",
  sco.scale = T,
  trait.scale = T,
  index.scale = T)

sort(dif, decreasing = T)[1:30]
```

Upogebia pusilla	Alitta succinea	Glycera tridactyla
1.0000000	0.8333333	0.8333333
Hediste diversicolor	Crangon crangon	Rapana venosa
0.8333333	0.5833333	0.5000000
Branchiostoma lanceolatum	Diogenes pugilator	Scolelepis
0.4687500	0.4375000	0.4166667
Eurydice dollfusi	Tritia neritea	Tritia reticulata
0.3750000	0.3333333	0.3333333
Nephtys hombergii	Perinereis cultrifera	Phyllodoce maculata
0.3125000	0.3125000	0.3125000
Phyllodoce mucosa	Corophium volutator	Donax semistriatus
0.3125000	0.2500000	0.2500000
Anadara kagoshimensis	Leptosynapta inhaerens	Synchelidium maculatum
0.2083333	0.1875000	0.1875000
Iphinoe elisae	Iphinoe maeotica	Iphinoe tenella
0.1666667	0.1666667	0.1666667
Retusa robagliana	Retusa truncatula	Retusa umbilicata
0.1666667	0.1666667	0.1666667
Retusa variabilis	Cerastoderma glaucum	Eunereis longissima
0.1666667	0.1562500	0.1562500
...		

The resulting synthetic trait (**dif**), when multiplied by organism densities (individual or biomass density) and summed across species, gives the sediment biodiffusive potential of the species community. The same can be done for the three other biomixing types (with an adapted scoring to the biomixing mechanism); the sum of the four can, if carefully scaled between 0 and 1 (**index.norm = T**), gives the total sediment biomixing potential.

Bioirrigation: We use here ability to ventilate (24), body mass (14), burrow type (21) and burrow depth (22); optionally, body length and burrow width could also be considered as all these traits are directly involved in bioirrigation (Kristensen & Kostka, 2005).

```

irr <- syn.trait(tab = tab, lab = lab,
               which.traits = c(24, 14, 21, 22),
               list.sco = list(c(1:3),
                              c(1:6),
                              c(1:3),
                              c(1:5)),
               fuz = c(T, T, T, T),
               fun = "mean",
               formula = "x1 * x2 * x3 * x4",
               sco.scale = T,
               trait.scale = T,
               index.scale = T)

sort(irr, decreasing = T)[1:30]

```

Upogebia pusilla	Alitta succinea	Glycera tridactyla
1.00000000	0.90000000	0.90000000
Hediste diversicolor	Nereis zonata	Mya arenaria
0.80000000	0.75000000	0.58333333
Leptosynapta inhaerens	Nephtys hombergii	Malacoceros fuliginosus
0.45000000	0.45000000	0.33333333
Notomastus profundus	Perinereis cultrifera	Thracia phaseolina
0.30000000	0.30000000	0.30000000
Amphiura stepanovi	Corophium volutator	Lagis neapolitana
0.25000000	0.25000000	0.25000000
Magelona mirabilis	Terebellides stroemii	Scolecopsis
0.20000000	0.20000000	0.19444444
Prionospio maciolekae	Heteromastus filiformis	Polydora cornuta
0.15000000	0.13333333	0.12500000
Pygospio elegans	Arcuatula senhousia	Melinna palmata
0.12500000	0.11111111	0.11111111
Ampelisca	Dipolydora quadrilobata	Cerastoderma glaucum
0.11111111	0.10000000	0.08333333
Capitella	Spio decorata	Streblospio gynobranchiata
0.08333333	0.08333333	0.08333333
...		



CHAPTER 5 : SUPPLEMENTARY MATERIALS FOR COMPUTING SYNTHETIC TRAIT FROM EFFECT TRAITS

In this Appendix, an example is provided to compute a multi-trait based indicator (proxy) of ecological process at the species level. For more details, please refer to Beauchard et al. (2023).

Biodeposition potential of the species *Abra alba*: It is the product of its score of biomass and feeding type for the modality “suspension feeding”:

- Body mass is between 0.001 – 0.01 (g AFDW); *A. alba* shares affinity with both deposit-feeding and suspension-feeding mode, after fuzzy coding, a value of 0.5 was given to these two modalities.
- A score of 2 was assigned to the modality “0.001 – 0.01” and a score of 1 for the suspension—feeding mode (the other type of feeding was score to 0)
- Modalities scores were normalised from 0 to 1. Each trait was given the same importance (even if the number of modalities was different between traits or if there was a mix between quantitative and qualitative traits).
- Then, we multiplied modalities values * normalised scores for each trait.
- Biodeposition = Body Mass * Suspension Feeding = $0.33 * 0.5 = 0.17$ (after rescaling = $0.17 / \max(de) = 0.2$)

NB: The highest score of biodeposition will be given to a large-bodied species with a strict suspension feeding mode (i.e. both scores = 1). Here, the highest biomass for the suspension feeding mode is “1 – 10 g” (the highest score is then equal to 0.83, after rescaling 1).

Table D.1: Example of the potential of biodeposition for the species *A. alba* with the respective modality of the trait body mass in bold. Body mass is expressed in g (AFDW). More details can be found in Beauchard et al. (2023).

Trait	Body Mass					
Modalities	> 0.001	0.001 – 0.01	0.01 – 0.1	0.1 – 1	1 – 10	> 10
Modalities values	0	1	0	0	0	0
Modalities scores	1	2	3	4	5	6
Normalized scores	0.16	0.33	0.5	0.67	0.83	1

Table D.2: Example of the potential of biodeposition for the species *A. alba* with the respective modality of the trait feeding type in bold. Feeding type is divided into four modalities: deposit feeder (De), suspension feeder (Su), Herbivory-Grazer (HeGr) and Carnivory or Scavenger (CaSc). More details can be found in Beauchard et al. (2023).

Trait	Feeding type			
Modalities	<i>De</i>	<i>Su</i>	<i>HeGr</i>	<i>CaSc</i>
Modalities values	0.5	0.5	0	0
Modalities scores	0	1	0	0
Normalized scores	0	1	0	0

CHAPTER 5 : SUPPLEMENTARY MATERIALS FOR RLQ ANALYSIS

In this Appendix, more details of the RLQ analyses done in **Chapter 5** are provided. For each RLQ analyses, preliminary tests was done to select to best set of abiotic and traits data. The spatial distribution of sampling station scores on RLQ axes is also presented in this Appendix.

E.1 Life-history strategy groups based on response traits

Table E.1: Results of tests for multivariate statistics. Tests between individual environmental conditions and life history strategy groups. The selected variables are in bold. The simulated statistics were generated by 999 random permutations.

Abiotic	Coinertia	<i>p</i> -value
DOX	0.03	0.57
SHEAR	0.02	0.46
botfluxPOC	0.02	0.39
SAL	0.03	0.43
TEMP	0.01	0.85
POC	0.04	0.29
PAR	0.01	0.76
FSED	0.01	0.78
SSED	0.02	0.47
Depth	0.03	0.48
Substratum	0.01	0.51
TEMPCV	0.02	0.60
DOXCV	0.01	0.57

Table E.2: Models 2 and 4 observations and their associated p -values. 49 999 random permutations.

Test	Obs	Std. Obs	p -value
Model 2	0.023	15.859	0.001 ***
Model 4	0.023	0.313	0.310

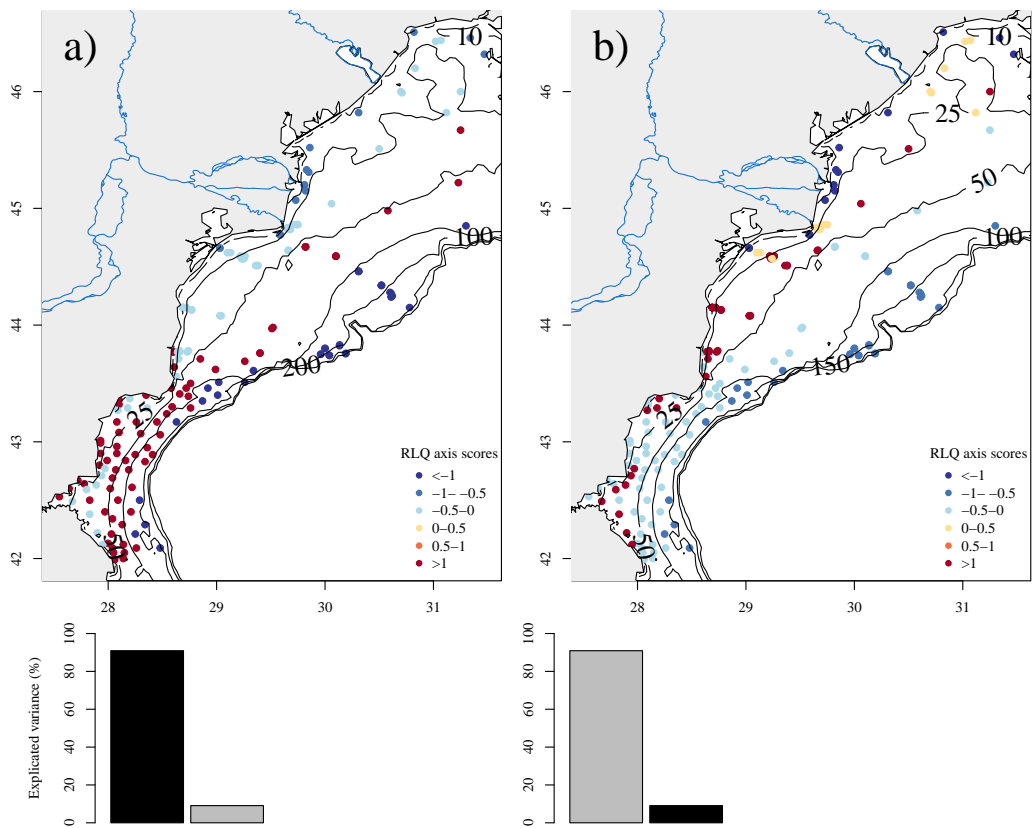


Figure E.1: Sampling station scores on the first axis of the RLQ analysis (from dark blue negative scores to red positive scores on the first axis), explained variance (%) by the first axis (91%) and second axis (9%) of the RLQ analysis.

E.2 Response traits

Table E.3: Results of tests for multivariate statistics. Tests between individual environmental conditions and response traits. The selected variables for the RLQ analysis are in bold (p -value < 0.10). The simulated statistics were generated by 999 random permutations.

Abiotic	Coinertia	p -value
DOX	0.93	0.08
SHEAR	0.45	0.41
botfluxPOC	0.50	0.37
SAL	0.77	0.05
TEMP	0.56	0.26
POC	0.82	0.20
PAR	0.45	0.11
FSED	0.54	0.18
SSED	0.72	0.13
Depth	0.93	0.15
Substratum	0.39	0.42
TEMPCV	0.69	0.33
DOXCV	0.38	0.48

Table E.4: Results of tests for multivariate statistics. Tests between response traits and previously selected abiotic descriptors. The selected variables for the RLQ analysis are marked by an asterisk (p -value < 0.05). The simulated statistics were generated by 999 random permutations.

Traits	Coinertia	p -value
Life span	0.096	0.263
Age at maturity	0.023	0.856
Sexuality	0.258	0.002**
Reproductive frequency	0.104	0.091
Fertilisation	0.027	0.859
Annual fecundity	0.048	0.963
Offspring type	0.037	0.679
Offspring size	0.055	0.642
Offspring protection	0.129	0.087
Offspring development	0.155	0.041*
Offspring benthic stage duration	0.109	0.236
Offspring pelagic stage duration	0.105	0.272
Offspring settlement size	0.096	0.105
Body mass	0.065	0.706
Body length	0.118	0.143
Mobility	0.177	0.024*
Substratum depth occupancy	0.062	0.582
Feeding type	0.036	0.839

Table E.5: Models 2 and 4 observations and their associated p -values. 49 999 random permutations.

Test	Obs	Std. Obs	p -value
Model 2	0.057	40.715	0.00002***
Model 4	0.057	4.797	0.00052***

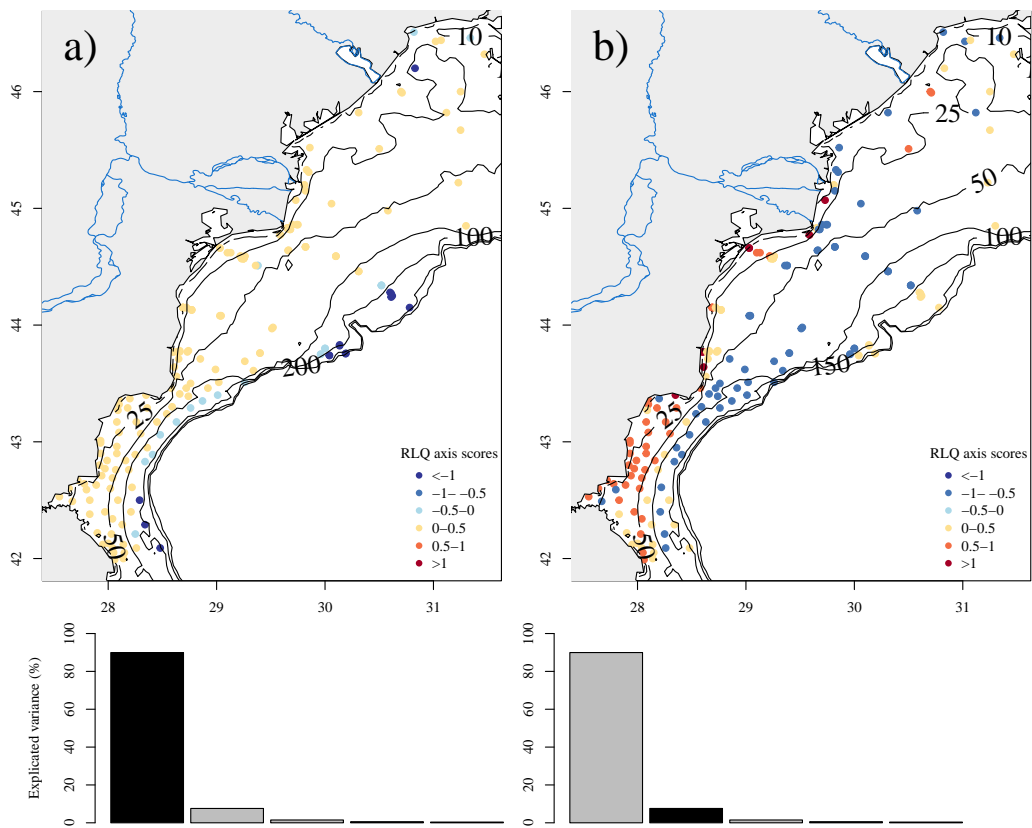


Figure E.2: a) Mapping of the scores of the sampling sites on the first axis of the RLQ analysis and b) second axis. Bottom line. The eigenvalues of the first and second axes are expressed as percentages, with 90% for the first axis and 8% for the second axis.

E.3 Effect traits

Table E.6: Results of tests for multivariate statistics. Tests between individual environmental conditions and traits. The selected variables for the RLQ analysis are in bold (i.e. p -value < 0.10). The simulated statistics were generated by 999 random permutations.

Abiotic	Coinertia	p -value
DOX	0.128	0.458
SHEAR	0.175	0.213
botfluxPOC	0.209	0.09
SAL	0.082	0.017*
TEMP	0.231	0.220
POC	0.230	0.290
PAR	0.106	0.441
FSED	0.125	0.416
SSED	0.142	0.578
Depth	0.321	0.151
Substratum	0.235	0.067
TEMPCV	0.235	0.382
DOXCV	0.087	0.184

Table E.7: Results of tests for the multivariate statistic. Tests between individual traits and previously selected abiotic descriptors. The selected variables for the RLQ analysis are in bold (i.e. p -value < 0.1). The simulated statistics were generated by 999 random permutations.

Traits	Coinertia	p -value
BM	0.031	0.277
MB	0.018	0.582
SD	0.046	0.057
BT	0.018	0.266
maxBD	0.030	0.262
VP	0.045	0.021*
AF	0.038	0.107
FT	0.009	0.818

Table E.8: Models 2 and 4 observations and their associated p -values. 49 999 random permutations.

Test	Obs	Std. Obs	p -value
Model 2	0.014	29.405	0.00002***
Model 4	0.014	0.903	0.14

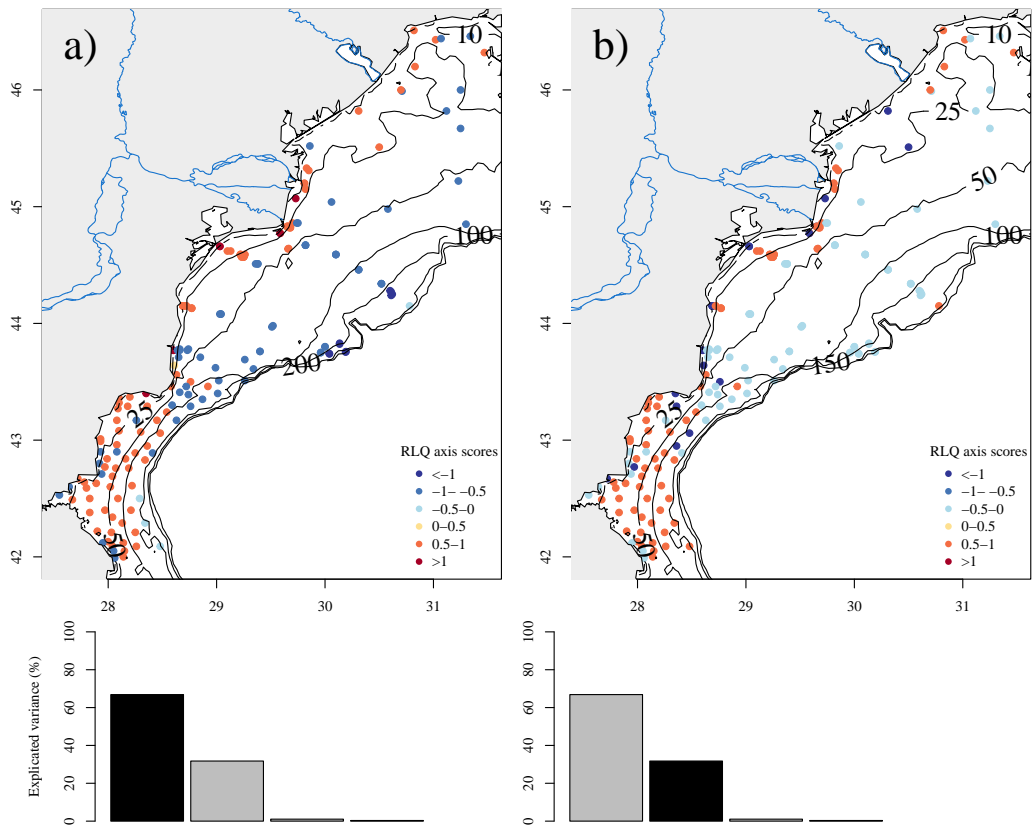


Figure E.3: a) Mapping of the scores of the sampling sites on the first axis of the RLQ analysis and b) second axis. Bottom line. The eigenvalues of the first and second axes are expressed as percentages, with 67% for the first axis and 32% for the second axis.

E.4 Ecological processes based on effect traits

Table E.9: Results of tests for multivariate statistics. Tests between individual environmental conditions and ecological process indicators. The selected variables are in bold (i.e. p -value < 0.1). The simulated statistics were generated by 999 random permutations.

Abiotic	Coinertia	p -value
DOX	0.01	0.49
SHEAR	0.01	0.51
botfluxPOC	0.08	0.37
SAL	0.01	0.10
TEMP	0.03	0.24
POC	0.01	0.76
PAR	0.01	0.40
FSED	0.01	0.69
SSED	0.00	0.99
Depth	0.04	0.24
Substratum	0.02	0.44
TEMPCV	0.02	0.51
DOXCV	0.01	0.59

Table E.10: Results of tests for multivariate statistics. Tests between ecological process indicators and previously selected abiotic descriptors. The selected variables are in bold (i.e. p -value < 0.15). The simulated statistics were generated by 999 random permutations.

Indicators	Coinertia	p -value
de	0.0005	0.69
mi	0.002	0.17
ir	0.003	0.07
Dep	0.003	0.13

Table E.11: Models 2 and 4 observations and their associated p -values. 49 999 random permutations.

Test	Obs	Std. Obs	p -value
Model 2	0.006	6.245	0.002**
Model 4	0.006	1.717	0.064

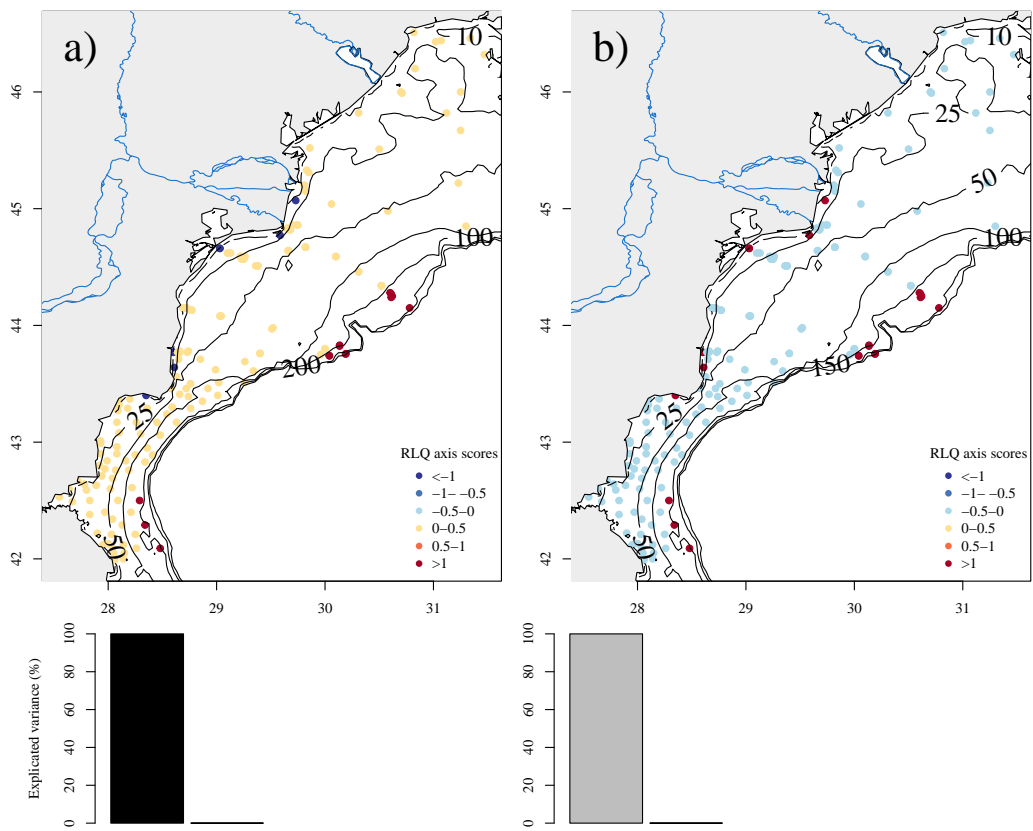


Figure E.4: a) Sampling station scores on the first axis of the RLQ analysis (from dark blue negative scores to red positive scores on the first axis), explained variance (%) by the first axis (99.9%) and second axis (0.01%) of the RLQ analysis.



CHAPTER 5 : SUPPLEMENTARY MATERIALS FOR NEURAL NETWORKS

In this Appendix, error metrics are provided for the neural network models developed in **Chapter 5**: R^2 (proportion of variance in the data explained by the model), RMSE (Root Mean Square Error), Bias and Standardized RMSE.

Table F.1: List of error metrics associated with the neural network for the three main **life-history strategies** (episodic, opportunist and precocial).

Metric	Episodic	Opportunistic	Precocial
Mean prediction	0.29853	0.55651	0.14306
R^2	0.13716	0.30975	0.32609
RMSE	0.07203	0.08224	0.05100
Bias	0.00025	-0.00302	0.00087
Standardized RMSE	0.64882	0.58801	0.59547

Table F.2: List of error metrics associated with the neural network for the **offspring development** trait subdivided into five modalities ((mixed)- planktotrophic, (mixed)- lecithotrophic and internal).

Metric	Planktotrophic	Lecithotrophic	Mixed plankto.	Mixed lecitho.	Internal
Mean prediction	0.35007	0.27913	0.17365	0.04286	0.15173
R^2	0.15743	0.33470	0.49348	0.17911	0.41007
RMSE	0.06240	0.05997	0.04451	0.03590	0.05033
Bias	0.00094	-0.00189	-0.00120	-0.00204	0.00164
Standardized RMSE	0.58863	0.41539	0.52135	0.66515	0.55364

Table F.3: List of error metrics associated with the neural network for the **substratum depth occupancy** trait subdivided into four modalities (0 cm, 0-5 cm, 5-15 cm and > 15 cm).

Metric	0 cm	0–5 cm	5–15 cm	> 15 cm
Mean prediction	0.31143	0.46245	0.18070	0.03675
R^2	0.08718	0.11605	0.03681	0.40448
RMSE	0.07358	0.05852	0.03582	0.01581
Bias	-0.00112	0.00062	-0.00117	-0.00016
Standardized RMSE	0.68204	0.68371	0.70669	0.47247

Table F.4: List of error metrics associated with the neural network for the four multi-trait based indicators of **ecological processes** (biodeposition, deposit feeding, bioirrigation and biomixing).

Metric	Biodeposition	Deposit feeding	Bioirrigation	Biomixing
Mean prediction	0.37895	0.41026	0.37170	0.46845
R^2	0.23267	0.31405	0.37091	0.34081
RMSE	0.13199	0.10099	0.10095	0.11376
Bias	0.00365	0.00244	0.00203	0.00300
Standardized RMSE	0.66419	0.61412	0.56598	0.60330

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