Seasonal and spatial variations of greenhouse gas (CO₂, CH₄ and N₂O) emissions from urban ponds in Brussels

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Highlights :

- First survey of GHG emissions from Brussels ponds
- Small ponds are more subject to edge effects and have higher pCO2
- Macrophytes enhance methane production in ponds
- City center ponds have higher N₂O emissions due to atmospheric deposition
- Brussels ponds emissions were equivalent to the carbon sink estimated for the urban green spaces.

Revised manuscript (clean version)

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2	and N ₂ O) emissions from urban ponds in Brussels
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10	Abstract:
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12	Freshwaters have been recognized as important sources of greenhouse gases
13	(GHG) to the atmosphere. However, urban ponds have received little attention even though
14	their number is increasing due to expanding urbanisation globally. Ponds are frequently
15	associated to urban green spaces that provide several ecosystemic services such as cooling
16	local climate, regulating the water cycle, and acting as small carbon sinks This study aims
17	to identify and understand the processes producing GHGs (CO2, CH4, and N2O) in the urban
18	ponds of the temperate European city of Brussels in Belgium. 22 relatively small ponds (0.1-
19	4.6 ha) surrounded by contrasted landscape (strictly urban, bordered by cropland or by
20	forest), were sampled during four seasons in 2021-2022. The mean ± standard deviation
21	was 3,667 \pm 2,904 ppm for the partial pressure of CO ₂ (pCO ₂), 2,833 \pm 4,178 nmol L ⁻¹ for
22	CH ₄ , and 273 \pm 662 % for N ₂ O saturation level (%N ₂ O). Relationships of GHGs with oxygen
23	and water temperature suggest that biological processes controlled pCO2, CH4
24	concentration and %N2O. However, pCO2 was also controlled by external inputs as indicated
25	by the higher values of pCO ₂ in the smaller ponds, more subject to external inputs than
26	larger ones. The opposite was observed for CH4 concentration that was higher in larger
27	ponds, closer to the forest in the city periphery, and with higher macrophyte cover. N_2O
28	concentrations, as well as dissolved inorganic nitrogen, were higher closer to the city center,
29	where atmospheric nitrogen deposition was potentially higher. The total GHG emissions

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from the Brussels ponds were estimated to 1kT CO₂-eq per year and were equivalent to the
 carbon sink of urban green spaces.

32 Keywords:

33 carbon dioxide, methane, nitrous oxide, Brussels, urban ponds, urban ecology,

34 *macrophytes*

35

36 1. Introduction

37

The three main greenhouse gases (GHGs) emitted to the atmosphere by human 38 39 activities are carbon dioxide (CO₂), methane (CH₄) with a global warming potential (GWP) 34 times greater than CO₂ on a 100-year time scale, and nitrous oxide (N₂O) with a GWP 40 41 298 times greater than CO₂ on a 100-year time scale (Myrhe et al., 2013). In inland waters, the production of CO₂ and CH₄ are mainly due to the degradation of organic matter (OM) 42 43 through mainly aerobic processes for CO₂ (Del Giorgio et al., 1999; Cole and Caraco, 2001) and by anaerobic archaeal methanogenesis for CH₄ (Conrad, 2020). The OM in inland 44 45 waters can be autochthonous or allochthonous. Autochthonous production of OM is from phytoplankton (McClure et al., 2020; Bartosiewicz et al., 2021) or aquatic macrophytes 46 (Grasset et al., 2019; Desrosiers et al., 2022). The relation between autochthonous biomass 47 (phytoplankton and macrophytes) and CH₄ emissions is well established (DelSontro et al., 48 2018; Borges et al., 2022; Bastviken et al., 2023). The impact of autochthonous biomass on 49 CO₂ depends on the phase of the bloom development, high concentrations of chlorophyll-a 50 (Chl-a) during the growth and peak phases coinciding with intense photosynthesis and CO₂ 51 consumption (Grasset et al., 2020; Borges et al. 2022;), and phases of senescence leading 52 to CO₂ production. Allochthonous OM in small water bodies (e.g., ponds) comes from fallen 53 leaves of the surrounding vegetation and from particulate and dissolved OM from surface 54 runoff or from soil-water and groundwater inputs (Gasith and Hosier, 1976; Weyhenmeyer 55 et al., 2015). The N₂O production is mainly due to microbial nitrification and denitrification 56 and is dependent on the availability of dissolved inorganic nitrogen (DIN) and O₂ (Codispoti 57 and Christensen, 1985; Mengis et al., 1997). Anthropogenic nitrogen inputs from leakage of 58 fertilizers from croplands or from atmospheric deposition may enhance the N₂O emissions 59 in freshwater wetlands, lakes and rivers (McCrackin & Elser, 2011; Bettez & Groffman, 2013; 60 Sønderup et al., 2016; D'Acunha et al., 2019; Decina et al., 2020; Bonetti et al., 2022). 61

The emissions from inland waters (rivers-streams, lakes and reservoirs) of GHGs were estimated at 14.3 Pg CO₂ yr⁻¹ for CO₂ (Drake et al., 2018), 6.03 Pg CO₂ equivalent (CO₂-q) yr⁻¹ for CH₄ (Rosentreter et al., 2021) and 0.05 Pg CO₂-eq yr⁻¹ for N₂O (Lauerwald et al., 2019). Yet, these estimates remain highly uncertain because of the low number of measurements that is insufficient to account for the strong spatial and temporal variability of the fluxes within a given system, as well as the high diversity of systems due to differences in size, climate, catchment morphology, land cover, and anthropogenic pressures.

Several studies have shown the importance of small water bodies (surface area < 0.1 69 ha) as hotspots for GHG emissions (Holgerson and Raymond, 2016; Grinham et al., 2018; 70 Rosentreter et al., 2021; Peacock et al., 2021). Although only representing 9% of the total 71 area of lentic waters, small water bodies account for 15% and 40% of diffusive emissions 72 from lakes of CO₂ and CH₄, respectively according to Holgerson and Raymond (2016). 73 Urban areas have a large number of small water bodies in the form of ponds mostly 74 associated to green spaces such as public parks, and their number is increasing due to rapid 75 urbanisation worldwide (Brans et al., 2018; Audet et al., 2020). Yet, only a very limited 76 number of studies have investigated GHG emissions from urban ponds (Singh et al., 2000; 77 Natchimuthu et al., 2014; van Bergen et al., 2019; Audet et al., 2020; Peacock et al., 2021). 78 Urban ponds are small and thus have high ratio of perimeter to surface area (Hanson et al., 79 2007), surrounded by impervious surfaces (Davidson et al., 2015; Peacock et al., 2021), 80 and a high stormwater runoff that combined result in high inputs of OM and DIN that should 81 sustain emissions of CO₂, CH₄, and N₂O to the atmosphere. Additionally, urban ponds are 82 usually very shallow so there is a strong influence on the water column of GHG production 83 in sediments. 84

Brussels is the most densely populated area in Belgium and the largest city of the 85 country and contains 158 ponds totalling a surface area of 101ha. The population is more 86 than 1 million inhabitants, equating to less than 1m² of water body per inhabitant, indicating 87 a high degree of anthropogenic pressure on these ecosystems. However, not all urban 88 ponds are subject to the same degree of anthropogenic pressure, as the periphery of the 89 city is bordered by cropland and forest. Yet, the ponds in Brussels are highly eutrophied with 90 in some cases high algal biomass but also macrophytes which compete with phytoplankton 91 for nutrients (Peretyatko et al., 2007; de Backer et al., 2010; Peretyatko et al., 2012; Descy 92 et al., 2016). Yet, the equilibrium between the prevalence of macrophytes and phytoplankton 93 entails a gradual shift rather than an abrupt process and is influenced by additional factors 94 other than nutrient levels (Van Nes et al., 2002; Davidson et al., 2023). It can be 95

hypothesized that water bodies with a dominance of phytoplankton and a dominance of
macrophytes should have different GHG dynamics, although this has seldom been tested
(Harpenslager et al., 2022 ; Baliña et al., 2023).

Here, we report CO₂, CH₄, and N₂O dissolved concentrations and complementary 99 variables (water temperature, chlorophyll-a (Chl-a), DIN, and O₂) sampled in 22 small urban 100 ponds (0.1-4.6 ha) of the city of Brussels, during the four seasons in 2021-2022. Potential 101 drivers of CO₂, CH₄, and N₂O dynamics are investigated by comparing pond size, presence 102 of macrophytes, relative distance to the city center, surrounding landscape (strictly urban, 103 bordered by cropland or by forest). We test if the most frequent accepted hypotheses of the 104 drivers of CO₂, CH₄ and N₂O seasonal and spatial variations in natural lakes and ponds also 105 apply to urban ponds The CO₂, CH₄, and N₂O diffusive emissions are computed and 106 compared to the official inventory of GHG emissions from the city of Brussels. 107

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109 **2. Material and methods**

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111 **2.1. Pond selection and sampling frequency**

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22 ponds were selected within the Brussels region (Fig. 1). All of the ponds are 113 artificial and were built during the past century, the majority with the primary aim of 114 landscaping and embellishing parks. Recreational fishing activities are allowed in four of the 115 sampled ponds (Table S1). Some of the sampled ponds are connected to a river network 116 (flow-through ponds) and others are fed directly by groundwater and small watercourses 117 (overflow ponds), where water level is kept constant by an overflow outlet (Table S1). The 118 sampled ponds are in the four main catchment areas of the city of Brussels (Molenbeek 119 (n=2), Neerpedebeek-Vogelzangbeek (n=4), Maelbeek (n=6), and Woluwe (n=10) 120 catchments) (Table S1). Four sampling campaigns were carried out on the 22 ponds, 121 corresponding to the four seasons: November 2021 (Fall), February 2022 (Winter), May 122 2022 (Spring) and August 2022 (Summer). 123

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125 **2.2. Meteorological data**

Air temperature, rainfall and wind speed were retrieved from <u>https://wow.meteo.be/fr</u> for the meteorological station of the Royal Meteorological Institute of St-Lambert (50.8408 °N, 4.4234 °E) in Brussels, located between 1 and 10 km from the sampled ponds. Wind speeds and air temperatures were averaged over 24 hours to obtain a daily average value. The rainfall was integrated on each day to obtain the daily rainfall.

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133 **2.3. Sampling in field**

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Water was sampled and collected from pontoons. Water pH, temperature, conductivity and oxygen saturation level ($\%O_2$) were measured by a VWR MU 6100 H probe. Water was collected in 2L polypropylene bottles for subsequent sampling for Chl-*a*, total suspended matter (TSM) and dissolved inorganic nutrients (ammonium (NH₄⁺), nitrite (NO₂⁻), nitrate (NO₃⁻), soluble reactive phosphorus (SRP)). Three 50mL falcon tubes were filled with unfiltered water to which was added 200µL of HNO₃ (65%) for total phosphorus (Ptot) analysis.

Macrophytes were defined as vascular aquatic plants, excluding filamentous algae, mosses and liverworts, (Bowden et al., 2017). The percentage of macrophyte cover was estimated visually during sampling. Dominant species of macrophytes (*Potamogeton pectinatus, Ceratophyllum demersum, Chara sp., Nitella sp., Lemna trisulca, Zannichellia palustris*) were identified in August 2022, when macrophyte cover was maximal. This list of species of macrophytes agreed with past studies in Brussels ponds (Peretyatko et al., 2009; de Backer et al., 2010).

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150 **2.4. Laboratory analysis**

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Water was filtered on Whatman filters 0.7 μ m GF/F glass microfibers with diameter of 47mm for TSM and Chl-*a* (stored frozen). Filtered water was stored in 50 ml plastic vials and stored frozen for dissolved nutrients analysis. Chl-*a* was measured on extracts with 90% acetone by fluorimetry (Kontron SFM 25 model) (Yentsch and Menzel, 1963). NH₄⁺ was determined with the nitroprusside-hypochlorite-phenol coloration method (Grasshoff and Johannsen, 1972). NO₂⁻ and NO₃⁻ were determined before and after reduction of NO₃⁻ to NO₂⁻ by a cadmium-copper column, with Griess' reagent in acidic medium coloration method
 (Grasshoff et al., 1983). SRP was determined with ammonium molybdate, ascorbic acid and
 potassium antimony tartrate coloration method (Koroleff, 1983).

Ptot was determined by inductively coupled plasma spectroscopy (ICP) on an ICP-OES Perkin Elmer Avio 200 model. The assay protocol was based on the US EPA (1994) method 200.7 for analysis of metals and trace elements in water by ICP with prior microwave acid digestion based on US EPA, (2007) method 3015A.

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166 **2.5. Measurement of GHG concentrations and computation of fluxes**

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CO₂ measurements were carried out on the field with a Li-Cor Li-840 CO₂/H₂O gas 168 analyser using the headspace technique with 4 polypropylene syringes of 60mL. The 169 technique consisted in equilibrating inside the syringe 30mL of sample water with 30mL of 170 atmospheric air by vigorous shaking during 5 minutes. The headspace of each syringe was 171 injected sequentially in the Li-840 and a fifth syringe was used to measure atmospheric CO₂. 172 The final pCO₂ value was computed taking into account the partitioning of CO₂ between 173 water and the headspace, as well as equilibrium with HCO₃⁻ (Dickson et al. 2007) using 174 175 water temperature measured in-stream and after equilibration, and total alkalinity (data not shown). Samples for total alkalinity were conditioned, stored and analysed as described by 176 Borges et al. (2019). The Li-Cor Li-840 was calibrated before and after each sampling period 177 with ultrapure N₂ and a suite of gas standards (Air Liquide Belgium) with CO₂ mixing ratios 178 179 of 388, 813, 3788 and 8300 ppm. The overall precision of pCO₂ measurements was ±2%.

Samples for dissolved CH₄ and N₂O were collected directly in surface waters with 60 180 ml plastic syringes and transferred with a plastic tube into two 60mL borosilicate serum 181 bottles and poisoned with 200µL of saturated HgCl₂ solution. The vials were sealed with a 182 butyl stopper and crimped with an aluminium cap. Measurements were carried out using the 183 headspace technique with a SRI 8610C gas chromatograph with a flame ionisation detector 184 (FID) for CH₄ and an electron capture detector (ECD) for N₂O, calibrated with CH₄:N₂O:N₂ 185 gas mixtures (Air Liquide Belgium) with mixing ratios of 1, 10, 30, 509, and 2010 ppm for 186 CH₄, and 0.2, 2.0 and 6.0 ppm for N₂O. The precision of measurement was ±11% for CH₄ 187 and ±6% for N₂O based on 176 replicates. Ebullitive CH₄ fluxes were not measured during 188 this study, but were measured with inverted funnels (e.g. Keller and Stallard, 1994) in four 189

ponds (Leybeek, Pêcheries Royales, Silex, Ten Reuken) for the period of 29/03/22 to
09/09/23 (unpublished data).

The CO₂ concentration is expressed in terms of partial pressure of CO₂ (pCO₂ in 192 ppm), the CH₄ dissolved concentration in nmol L⁻¹ corresponding to the usual convention in 193 topical literature. N₂O concentrations oscillated around atmospheric equilibrium, so the data 194 are presented as a percentage of the saturation level (%N₂O), where 100% corresponds to 195 atmospheric equilibrium. The atmospheric pCO₂ was measured on the field with the Li-Cor 196 Li-840. The equilibrium with atmosphere for N₂O was calculated from the average air mixing 197 ratios of N₂O provided by the Global Monitoring Division (GMD) of the National Oceanic and 198 Atmospheric Administration (NOAA) Earth System Research Laboratory (ESRL) (Dutton 199 and Hall, 2023). A constant atmospheric concentration of 1.9 ppm was assumed for CH₄. 200

The exchange of CO₂, CH₄, and N₂O between surface water and the atmosphere was calculated according to:

203 $F_G = k_g \Delta C$

F_G (mmol m⁻²d⁻¹) is the flux of a specific gas (G), k_g (cm h⁻¹) is the specific gas transfer velocity, and ΔC is the gas concentration gradient in water and the atmosphere (mmol L⁻¹).

The k_g values were derived from the gas transfer velocity normalized to Schmidt number of 600 (k_{600}) according to :

208 $k_g = k_{600} \times (\text{Sc}/600)^{-0.5}$

209 where Sc is the Schmidt number of the given gas in freshwater at in-situ water 210 temperature computed according the algorithms given by Wanninkhof (1992).

k₆₀₀ was computed from wind speed using the parameterisation of Cole and Caraco
 (1998):

213 $k_{600} = 2.07 + 0.215 u^{1.7}$

214 where u is the wind speed (m s^{-1}).

The average daily u value on the day of sampling of each sampled pond was used. The u data were measured at the meteorological station of the Royal Meteorological Institute of St-Lambert (50.8408 °N, 4.4234 °E) in Brussels (located between 1 and 10 km from the sampled ponds) (retrieved from <u>wow.meteo.be</u>). The area-weighted average of F_G from the 22 sampled ponds (totaling a surface area of 34 ha) for each of the 4 seasons was averaged

to obtain an annual value that was extrapolated to the total surface area (101 ha) of the allof the 158 urban ponds of the city of Brussels.

The emissions of CH_4 and N_2O were expressed in CO_2 equivalents (CO_2 -eq) by multiplying the respective fluxes by global warming potential (GWP) values of 34 for CH_4 and 298 for N_2O corresponding to a time horizon of 100 years (Myrhe et al., 2013).

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226 **2.6. Statistical analysis**

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The statistical analyses were conducted using R version 4.1.1 (2022) (R Core Team, 2022) and graphs were produced with GraphPad Prism v9. Prior to analysis, data underwent logarithmic or square root transformations to ensure normality. Shapiro tests were performed to assess the normality of the distribution.

To investigate relationships between variables, quantile regressions were performed with the quantreg package v5.95 (Koenker, 2005). Quantile regressions were chosen over linear regressions due to their ability to provide a more comprehensive understanding of potential causal relationships in ecological processes (Cade and Noon, 2003). Significance levels for the quantile regression were determined using Wald statistical tests.

For comparisons of medians in the boxplots, permutational multivariate analysis of 237 variance (PERMANOVA) on each Bray-Curtis distance matrix with 999 permutations was 238 performed. A post-hoc test was performed to establish significant differences between 239 sample pairs. For multiple comparisons, *p*-values were adjusted using the Bonferroni 240 correction. Simultaneously, a test for homogeneity of multivariate dispersions (PERMDISP) 241 was conducted. This test allowed for the examination of dispersion effects independent of 242 location effects, as significant differences could be attributed to variations in within-group 243 dispersion rather than differences in medians values among the groups (Anderson, 2006). 244 PERMANOVA and PERDISP were performed using the vegan package v2.6.4 (Oksanen et 245 246 al., 2013).

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248 2.7. Data availability

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The time-sampled and georeferenced data-set presented in this paper is publicly available (Bauduin et al., 2024).

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253 **3. Results**

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3.1. Seasonal variations of meteorological conditions and GHGs

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In 2022, the average air temperature in Brussels was 12.2°C, above the average of 257 11.0°C for the 1991-2020 period. With the exception of April, September and December, air 258 temperature monthly averages were above the reference value for the 1991-2020 period. A 259 total of 701 mm of rain fell in Brussels (837 mm for the 1991-2020 period), during 149 days 260 (190 days for the 1991-2020 period). The year 2022 was the fourth driest and the first (ex-261 aequo with 2020) warmest year of the current reference period. During the sampling period 262 from 22/11/21 to 18/08/22, the air temperature and wind speed averaged 11.9±6.3°C and 263 0.6±0.6 m s⁻¹, respectively, and the total precipitation was 321mm in the city of Brussels 264 (Fig. 2A). Air temperature was highest in summer (22.2±1.6°C) and lowest in fall (6.1±0.7°C). 265 The highest wind speed was in spring (Fig. 2B) (1.7±0.5 m s⁻¹). In the 15 days prior to 266 sampling, it rained a total of 13mm in fall, 53mm in winter, 13mm in spring, and less than 1 267 mm in summer. 268

During the four sampling periods, pCO_2 averaged 3,667 ± 2,904 ppm, CH₄ 269 concentration averaged 2,833 \pm 4,178 nmol L⁻¹, and %N₂O averaged 273 \pm 662 % in the 22 270 sampled ponds in the city of Brussels. The maximum pCO₂ value was observed in fall 271 (15,029 ppm) and the minimum value in summer (121 ppm). Median pCO₂ was higher in fall 272 273 than during the other seasons and was lower in winter than spring (Fig. 3A). The maximum and minimum CH₄ concentration values were both observed in summer, 29,190 and 10 nmol 274 L⁻¹, respectively (Fig. 3B). The CH₄ concentrations showed seasonal variations, with 275 significantly lower median concentration in winter than other seasons, and higher 276 concentrations during spring and summer (Fig. 3B). The minimum and maximum values of 277 %N₂O were both observed during summer (0 and 10,354%) (Fig. 3C). The median 278 279 concentration of %N₂O was lower during summer than fall and winter. The %N₂O values showed a higher dispersion during summer than other seasons (Fig. 3C). Undersaturation 280 281 of CO₂ with respect to atmospheric equilibrium was observed only on 7 occurrences (out of 88 observations), during spring (2 times) and summer (5 times). For N₂O, undersaturation 282

with respect to atmospheric equilibrium was observed at each season, 3 times in fall, 5 times
in winter, 7 times in spring and 10 times during summer. CH₄ concentrations were always
above saturation.

The average, median, minimum and maximum values of ancillary variables (%O₂, 286 Chl-a, SRP, Ptot, NH4⁺, NO3⁻ and NO2⁻) are given in Table S2. The median %O2 values were 287 significantly lower in fall than in the other seasons (Table S2). The median Chl-a 288 concentration value was significantly lower in winter than in fall and the highest 289 concentrations were measured in summer, when values were the most variable. There were 290 no differences between seasons for SRP. Ptot had higher concentrations in spring and 291 summer than in fall and winter. DIN concentrations were significantly higher in fall and winter 292 than during spring and summer. NH₄⁺, NO₃⁻ and NO₂⁻ concentrations were significantly 293 higher in fall than during the other seasons. 294

295 pCO₂, CH₄ concentration and %N₂O were negatively related to %O₂ (Figs. 4A, 4G, 4M). CH₄ concentration was positively related to water temperature (Fig. 4H) and %N₂O was 296 297 negatively related to water temperature (Fig. 4N). There was no relation between pCO₂ and water temperature (Fig. 4B), nor Chl-a concentration (Fig. 4C). pCO₂ was positively related 298 299 to SRP (Fig. 4E). CH₄ was negatively related to Chl-a (Fig. 4I). There was a positive correlation between CH₄ and Ptot (Fig. 4L), that was also observed during each individual 300 season, except in winter (Figs. S2A, S2B, S2C, S2D). The negative relation observed 301 between CH₄ and DIN (Fig. 4J) could be spurious and be attributed to parallel seasonal 302 changes of both variables, as there were no correlations between CH₄ and DIN when 303 analysed independently for each season, except in spring (Figs. S2E, S2F, S2G, S2H). 304 %N₂O was positively correlated to DIN for the whole dataset (Fig. 4P) as well as for each 305 season analysed independently (Figs. S2I, S2J, S2K, S2L). %N₂O was statistically more 306 strongly correlated to NO₂⁻ (Fig. 5C) and NO₃⁻ (Fig. 5D) than to NH₄⁺ (Fig. 5B). %N₂O was 307 negatively related to O₂:DIN (Fig. 5E), O₂:NO₂⁻ (Fig. 5G) and O₂:NO₃⁻ (Fig. 5H) but unrelated 308 to O₂:NH₄⁺ (Fig. 5F). 309

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311 3.2. Spatial variations of GHGs

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As pond perimeter (r = 0.90, p<0.0001) and depth (r=0.45, p 0.014) were correlated to surface area ; hereafter, only relationships between GHGs and pond surface area are

presented (Figs. 6A, 6E, 6I). Both pCO₂ and %N₂O were negatively related to pond surface 315 area (Figs. 6A, 6I) and CH₄ was positively related to pond surface area (Fig. 6E). The ponds 316 closer to the city center had lower CH₄ concentrations (Fig. 6F) and higher %N₂O (Fig. 6J) 317 than the ponds closer to the periphery of the city. %N₂O values showed a higher dispersion 318 in small ponds than larger ponds and in ponds closer to the city center compared to the 319 320 periphery of the city. pCO₂ did not show a clear pattern with regards to the city center (Fig. 6B). DIN, NO_{3⁻} and NO_{2⁻} concentrations decreased with distance from the city center but not 321 NH₄⁺ (Fig. 7). Atmospheric nitrogen dioxide (NO₂) concentration decreased with distance 322 from the city center (Fig. 7E). %N₂O was positively correlated to the atmospheric NO₂ 323 concentration (Fig. 7F). Elevated CH₄ concentrations were observed in ponds with higher 324 macrophyte cover (Fig. 6G). Ponds with high macrophyte cover were located farther away 325 from the city center (Fig. S3A), had larger surface areas (Fig. S3B), and lower residence 326 times (Fig. S3C). Ponds with high macrophyte cover exhibited lower concentrations of DIN 327 and Chl-a than ponds with low macrophyte cover (Fig. S3D,E). Ponds with high and low 328 macrophyte cover had no difference in SRP and Ptot concentration (Figs. S3F,G). No 329 significant correlation was observed between Chl-a and distance from city center (Fig. S4). 330

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332 3.3. Fluxes of GHGs

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The mean and median concentrations of the three GHGs in the sampled ponds were consistently higher than their respective equilibrium concentrations (Fig. 3). Consequently, the sampled ponds were sources of GHGs to the atmosphere during all four season, although negative flux values were observed on some rare occasions (Fig. S5). Seasonal variations of the GHG fluxes mirrored those of the respective concentrations because of modest variations of wind speed and computed k_{600} , although higher fluxes were observed during spring characterized by the highest daily wind speeds (Fig. S5A).

In terms of GHG emissions expressed in CO₂ equivalents, CO₂ was the dominant GHG emitted, with a median annual emission to the atmosphere of 732 g CO₂ m⁻² yr⁻¹. Median annual emission of CH₄ and N₂O was 97 and 9 g CO₂-eq m⁻² yr⁻¹, respectively. CO₂ contributed to 88% of the total CO₂ equivalent emissions, CH₄ contributed to 11%, and N₂O contributed to 1%. The total annual diffusive emissions of CO₂, CH₄, and N₂O amounted to 0.8 kT of CO₂-eq for all of the ponds in the city of Brussels (n=158) (Fig. 8). 347

348 **4. Discussion**

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350 **4.1. Seasonal variations**

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Seasonal variations of dissolved CO₂ in lacustrine systems at mid and high latitudes 352 depend on the seasonal alternation between periods of net ecosystem heterotrophy and 353 periods when the aquatic ecosystem evolves to more autotrophic conditions. This seasonal 354 alternation depends on light conditions, with a higher aquatic primary production in spring 355 and summer. Light conditions co-vary with temperature, and warmer conditions coincide 356 seasonally with the sunnier conditions in summer. Allochthonous organic matter inputs from 357 the watershed can also vary seasonally and are usually stronger in fall and winter than spring 358 and summer. 359

In the sampled ponds of the city of Brussels, the overall negative relation between 360 pCO₂ and %O₂ (Figs. 4A) reflected an alternation between autotrophy-heterotrophy 361 362 (balance between photosynthesis and community respiration), as also observed in other studies in rivers and lakes (Borges et al., 2015, 2018). This confirmed by the observed 363 relationship between pCO₂ and SRP (Fig. 4F) that may reflect the uptake of nutrients by 364 autotrophic groups leading to the lowering of SRP and pCO₂, and, conversely, high SRP 365 and pCO₂ resulting from the degradation of OM in the water column and sediments (Allesson 366 et al., 2020). The pCO₂ values in the sampled ponds of the city of Brussels were highest in 367 368 fall and lowest in summer. In fall, the senescence of photosynthetic organisms, induced by a decrease in light intensity and water temperatures, will lead to a large amount of potentially 369 degradable autochthonous OM (Bartosiewicz et al., 2021), which will be respired and lead 370 to higher pCO₂ in fall than in winter and summer. The input in fall of allochthonous OM to 371 ponds is increased with leaf fall from surrounding trees (Sønderup et al., 2016) and with 372 runoff due to higher precipitation . In spring, the occurrence of aquatic primary production in 373 response to the increase of the photoperiod and water temperature, lead to a CO₂ 374 consumption and O₂ production. 375

In summer, the median (average) pCO_2 was not significantly different compared to other seasons, but there was a higher dispersion of the values that explained the lack of relation between pCO_2 and water temperature and Chl-*a* (Figs. 4B, 4C), unlike CH₄ and

%N₂O that correlated to water temperature (Figs. 4H, 4N). The high dispersion of pCO₂ 379 values in summer may be attributed to sampling different stages of phytoplankton 380 development in various ponds during the summer campaign (Fig. S4). Low pCO₂ values 381 were observed in ponds where phytoplankton was blooming and high pCO₂ values were 382 observed in ponds where phytoplankton was senescent. The relation of pCO₂ and Chl-a 383 384 was further complicated by the variable density of macrophyte cover among the different sampled ponds. High macrophyte cover corresponded to low Chl-a concentrations, while 385 low macrophyte cover corresponded to higher Chl-a levels (Fig. S3D). 386

Low %O₂ values in the water column indicate a generalized net heterotrophy 387 within the sampled ponds including the sediments and therefore also a high 388 methanogenesis, leading to an overall negative relation between CH₄ and %O₂ (Fig. 4G), 389 as also observed in other urban ponds (Audet et al., 2020; Peacock et al., 2021). Since 390 methanogenesis occurs in sediments rich in organic matter, seasonal variations of 391 392 methanogenesis are very strongly dependent on temperature variations (Zeikus and Winfrey, 1976; Yvon-Durocher et al., 2014; Chen et al., 2021). Consequently, there was also 393 394 a highly significant positive correlation between CH₄ and water temperature (Fig. 4H). There is a highly significant positive correlation between Ptot and CH₄, as in other urban ponds 395 396 (Peacock et al., 2019; Rabaey and Cotner, 2022), probably reflecting the delivery of fresh, labile organic matter to sediments enhancing methanogenesis (Grasset et al., 2021; Gruca-397 Rokosz and Cieśla, 2021 ; Nijman et al., 2022). Ptot was similar in ponds with high and low 398 macrophyte cover (Fig. S3G), indicating Ptot gives a measure of potentially degradable OM 399 from both phytoplankton and macrophytes. Consequently, Ptot was higher in spring and 400 summer (Table S2), when primary production from phytoplankton and macrophytes is 401 highest. 402

N₂O in lakes and ponds results from nitrification and denitrification that show seasonal 403 variations, mainly as a function of DIN availability and O₂ conditions (in particular in thermally 404 stratified systems), as well as water temperature (Myrstener et al., 2016; Palacin-Lizarbe et 405 al., 2018; Mander et al., 2021). Additionally, the yield of N₂O production from both nitrification 406 and denitrification also depends on O₂ levels and water temperature. In the sampled ponds, 407 408 %N₂O was higher in fall than summer, as observed by Mander et al. (2021) in artificial wetlands. During fall, high heterotrophic degradation of organic matter and ammonification, 409 combined with lower uptake of NH4⁺ by phytoplankton, resulted in high concentrations of 410 NH4⁺ (Table S2). This led to high N₂O production from nitrification due to higher NH4⁺ 411 availability, and the N₂O yield from nitrification was possibly enhanced by low O₂ conditions 412 13

(Rosamond et al., 2012; Soued et al., 2016). In spring and summer, the NH4⁺ was 413 assimilated by phytoplankton and there was substrate limitation for N₂O production by 414 nitrification in the water column. The combination of high and low nitrification in fall and 415 416 spring, respectively explained the observed overall relationship between %N₂O and NH₄⁺ (Fig. 5B). NO_{2⁻} and NO_{3⁻} are, respectively, intermediate and final product of the nitrification, 417 418 which explained the observed overall relationship between %N₂O and NO₂⁻ and NO₃⁻ (Figs. 5C, 5D) that was statistically more significant than the relation between %N₂O and NH₄⁺. 419 Low O₂ levels will favour N₂O production over NO₂⁻ production during nitrification (Goreau 420 et al., 1980; Ni et al., 2011), leading to a negative relation between %N₂O and O₂:DIN, 421 O₂:NO₂⁻ and O₂:NO₃⁻ (Figs. 5E, 5G, 5H). While the correlations between N₂O and NH₄⁺, 422 NO_{3⁻} and NO_{2⁻} (Figs. 5B, 5C, 5D) in the sampled ponds of the city of Brussels indicate the 423 importance of nitrification in driving the variability of N₂O, denitrification should also have 424 played a role in N₂O dynamics. An indication of this is given by the overall negative 425 relationship between %N₂O and water temperature (Fig. 4N), as the last step of 426 denitrification is inhibited at low temperatures (Liao et al., 2018; Velthuis and Veraart, 2022), 427 428 leading to an accumulation of N₂O in the water column during fall and winter. A pattern of higher N₂O levels during wintertime was also observed in boreal lakes and attributed to the 429 effect of temperature on the N₂O yield from denitrification (Kortelainen et al., 2020). 430

431

432 **4.2. Spatial variations in GHGs**

433

The levels of CO₂ and CH₄ in lakes vary as a function of lake size (Lapierre and del 434 Giorgio, 2012; Kankaala et al., 2013; Raymond et al., 2013; Holgerson and Raymond, 2016; 435 Casas-Ruiz et al., 2021; Borges et al., 2022; Chiriboga et al. 2024) as well as land cover 436 (Maberly et al., 2013). Combined lake size and land cover determine relative levels of 437 allochthonous organic matter inputs to aquatic systems and the general balance of 438 autotrophy and heterotrophy largely determining CO₂ levels (del Giorgio and Peters, 1994), 439 as well as benthic methanogenesis. The size of the water body will also largely determine 440 the relative importance of soil-water and ground-water inputs of CO₂ and CH₄ 441 (Weyhenmeyer et al., 2015; Olid et al. 2022). The size of the water body will also determine 442 to a larger extent its depth, with smaller systems being in general shallower (Wetzel, 2001). 443 Depth will in turn determine the degree of coupling between sedimentary processes and 444 surface waters, as well as the possibly development of macrophytes. Yet, given the low 445

number of studies, it has not been investigated if these general patterns established from studies in natural lakes and ponds can be applied to urban ponds. In urban ponds the catchment cover is obviously very different from more natural systems, and there is a very strong atmospheric nitrogen deposition. Inter-lake variations of N₂O have been less investigated than CO₂ and CH₄, although DIN levels and depth are expected to exert a strong influence on N₂O lacustrine levels (Lauerwald et al., 2019; Borges et al., 2022).

The surface area of the sampled ponds was overall low and the range of variation was small (0.1-4.6 ha), so that other properties than pond size might have driven intersystem differences. The analysis of spatial patterns of GHGs in the sampled ponds was complicated by the fact that the smaller systems were located towards the center of city, and the larger ones at the periphery in contact with either cropland to the West and the Sonian forest to the East. Furthermore, the larger systems were characterized by a more important macrophyte cover.

The pCO₂ values in the sampled ponds did not show a significant relation to the 459 distance from the city center or macrophyte cover (Figs. 6B, 6C). However, the pCO₂ values 460 were significantly higher in the smaller systems than the larger ones (Fig. 6A). This agrees 461 with several studies at local, regional or global scales showing a negative relation between 462 pCO₂ and lake surface area (Lapierre and Del Giorgio, 2012; Kankaala et al., 2013; 463 Raymond et al., 2013; Holgerson and Raymond, 2016; Casas-Ruiz et al., 2021). The causes 464 of such relations are related to the overall influence of lake size on inputs from the landscape 465 and lake carbon cycling and have been discussed in detail in several other studies (del 466 Giorgio and Peters, 1994; Sand-Jensen and Staehr, 2007; Borges et al., 2022). 467

The CH₄ concentration in the sampled ponds was significantly higher in the larger 468 systems than the smaller ones (Fig. 6E). Several studies reported a negative correlation 469 between CH₄ concentration and lake size (surface and/or depth) across spatial scales (local 470 to global) (Borges et al., 2011; Kankaala et al., 2013; Holgerson and Raymond, 2016; Borges 471 et al., 2022; Chiriboga et al. 2024), yet one study reported no relationship (Peacock et al., 472 2019) and another a positive correlation (Rabaey et al., 2022). The positive relationship 473 between CH₄ concentration and pond size in the ponds of the city of Brussels was probably 474 indirect and reflected another driver. The relationship of CH₄ concentrations in relation to the 475 distance from the city center and to macrophyte cover was statistically more significant than 476 to surface area (Figs. 6E, 6F, 6G). We hypothesize that macrophyte cover was the 477 478 predominant driver (given the stronger statistical significance) that was itself a function of distance from the city center. The statistically weak negative relationship between Chl-*a* and
CH₄ (Fig. 4J) could reflect the fact that ponds with high macrophyte cover tended to exhibit
lower Chl-*a* concentrations (Fig. S3D) but had significantly higher CH₄ concentrations (Fig.
6G).

Submerged macrophytes have a complex impact on CH₄ dynamic in lakes and 483 ponds, as synthesized by Bastviken et al. (2023), with both positive and negative effects on 484 the CH₄ emissions. In the urban ponds of the city of Brussels, the CH₄ concentration was 485 significantly higher in ponds with a high macrophyte cover (Fig. 6G), located in the periphery 486 of the city and closer to the Sonian Forest (Figs. 6F, 6H). In addition to the potential positive 487 impact of macrophytes on CH₄, the Sonian forest likely transfers allochthonous carbon as 488 plant litter promoting CH₄ production compared to other ponds. Litter from trees can induce 489 sediment anoxia (Mehring et al., 2014) enhancing CH₄ production. Yet, this potential 490 additional input of allochthonous organic carbon from the Sonian Forest did not seem to 491 492 affect the pCO₂ in ponds for which the effect of size seemed more important.

The %N₂O levels were significantly higher but also more variable in the sampled 493 urban ponds closer to the city center than at the periphery of the city (Figs. 6J, 6L). %N₂O 494 was strongly correlated to DIN, NO₂⁻ and NO₃⁻ concentrations (Fig. 5) that were higher close 495 to the city center (Fig. 7). The higher DIN, NO₂⁻ and NO₃⁻ concentrations in the city center 496 could have resulted from higher N deposition as indicated by the pattern of atmospheric NO₂ 497 (Fig. 7E). No correlation was found with macrophyte cover (Fig. 6K), although the presence 498 of macrophytes strongly influences nitrogen cycling in sediments (Barko et al., 1991; 499 Choudhury et al., 2018; Dan et al., 2021; Ni et al., 2022) and could in theory potentially have 500 affected N₂O levels. 501

502

4.3. Comparison with other urban and natural ponds

504

There are only a few equivalent studies of GHG emissions in urban ponds for comparison (Table S3). We restricted the comparison to dissolved GHG concentrations rather than fluxes across the air-water interface. Fluxes measured with floating chambers and computed fluxes from k can provide diverging results, so are not necessarily comparable (Duchemin et al., 1999; Guérin et al., 2007; Klaus and Vachon, 2020; Erkkilä et

al., 2018; Perolo et al., 2021). This is possibly related to measurement biases using floating 510 chambers (Belanger and Korzun, 1991; Lorke et al., 2015; Vingiani et al., 2021), although 511 the computation of k can be made with different parameterizations derived from several 512 methods (including floating chambers) and that provide different values particularly at high 513 wind speeds. It is unclear which wind parametrization is the most appropriate (Klaus and 514 515 Vachon, 2020), in particular with regards to effects of size and fetch limitation (Wanninkhof, 1992) which might become very important in small water bodies such as ponds (Holgerson 516 et al., 2017; Jansen et al., 2020). We computed the fluxes from dissolved concentrations of 517 CO₂, CH₄ and N₂O and the gas transfer velocity computed from the parameterization as a 518 function of wind speed of Cole and Caraco (1998). The parametrization of Cole and Caraco 519 (1998) was built from a large compilation of deliberate tracer based estimates of k₆₀₀ in 11 520 lakes with a wide range of maximum depth (1-109m) and surface area (0.2-487 km²). The 521 parametrization of Cole and Caraco (1998) parameterization as function of wind speed 522 intrinsically integrates other sources of turbulence (night-time convection due to cooling) and 523 fetch limitation, because it is built on compilation of repeated tracer measurements in several 524 lakes with a wide range of size and depth. Yet, the parameterization of Cole and Caraco 525 (1998) is based on k_{600} data acquired in water bodies with a surface area larger (>4 ha) than 526 the ponds we sampled (0.1-4.6 ha), which admittedly introduces uncertainty in the 527 computation of k_{600} . The computed k_{600} from the sampled ponds with the parametrization of 528 Cole and Caraco (1998) ranged from 2.1 to 2.6 cm h⁻¹ in summer and spring, respectively. 529 The overall annual average k_{600} value of 2.3 cm h⁻¹ was within the range of k_{600} values (0.9-530 3.0 cm h⁻¹) reported by Holgerson et al. (2017) in four ponds located in a dense forest, with 531 surface area ~0.02 ha. These four ponds located in a dense forest were smaller than the 532 ponds we sampled (0.1-4.6 ha) that were located in recreational urban parks, hence, less 533 534 sheltered from wind that in a dense forest. The values of computed k_{600} from the sampled ponds with the parameterization of Cole and Caraco (1998) are also comparable to the 535 536 values of k_{600} used by Holgerson and Raymond (2016), in a global analysis of CO₂ and CH₄ emissions from ponds, of 1.5 cm h⁻¹ for ponds with a surface area <0.1 ha and of 2.0 cm h⁻¹ 537 ¹ for ponds with a surface are of 0.1-1.0 ha. 538

The average of CH₄ dissolved concentrations (annual or summer average) in the sampled ponds of Brussels was comparable to the values reported in other urban ponds in the cities of Silkeborg (Denmark, Audet et al., 2020), Uppsala (Sweden, Peacock et al., 2019; Ray et al., 2023), Linkoping (Sweden, Natchimuthu et al., 2014), Minnesota (United States, Rabaey & Cotner, 2022; Ray et al., 2023) and in Salaspils (Latvia, Ray et al., 2023). The average dissolved CO₂ concentrations in the sampled ponds was also equivalent to the values in the ponds of Silkeborg (average of 4 seasons) (Audet et al., 2020) and possibly lower in summer than the ponds in Uppsala (Peacock et al., 2019).

The extreme variability of CH₄ and CO₂ concentrations seasonally and among ponds 547 indicated by the large values of the standard deviations was observed in all systems 548 (Brussels, Silkeborg, Uppsala, Minnesota and Salaspils) and probably obscured differences 549 among different studies that could have emerged from differences in inputs of allochthonous 550 carbon and nitrogen, despite the differences in the size of the cities, as Brussels (> 1 million 551 inhabitants) is larger than Silkeborg (<50,000 inhabitants) and Uppsala (140,000 552 inhabitants). The N₂O concentration in the sampled ponds of Brussels was higher than 553 values in the ponds of Silkeborg (average of 4 seasons) (Audet et al., 2020), which might 554 be a consequence of a larger anthropogenic input of nitrogen from atmospheric deposition 555 due to the larger size of the city of Brussels compared to Silkeborg. This was consistent with 556 a higher value of the annual average NH₄⁺ concentration in the ponds of Brussels (27±63) 557 µmol L⁻¹) compared to the ponds of Silkeborg (9±14 µmol L⁻¹, Audet et al., 2020). A stronger 558 eutrophication (nitrogen levels) sustaining a higher primary production, might also explain 559 potentially lower CO₂ concentrations in Brussels in summer than in Silkeborg and Uppsala 560 (Table S3). 561

Concentrations of CO₂ and CH₄ in the sampled ponds of Brussels were >3 times 562 higher than those in natural (non-urban) ponds of similar size reported by Holgerson and 563 Raymond (2016) (Fig. 9). This confirms the emerging idea that artificial inland water bodies 564 (agricultural reservoirs, urban ponds, storm-water retention basins, ...) have higher emission 565 rates per m² than natural systems (Martinez-Cruz et al., 2017; Grinham et al., 2018; Herrero 566 Ortega et al., 2019; Gorsky et al., 2019; Ollivier et al., 2019; Peacock et al., 2019, 2021; 567 Webb et al., 2019). This seems to result from more important anthropogenic allochthonous 568 inputs of carbon and nitrogen in artificial systems but might also reflect other differences 569 compared to natural systems such as in hydrology (Clifford and Heffernan, 2018). 570

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572 **4.4. Contribution of urban ponds to the GHG budget of the city of Brussels**

573

574 The CO₂, CH₄ and N₂O air-water fluxes were converted into CO₂ equivalents and the 575 average scaled to total surface area of ponds in the city of Brussels (101 ha). The diffusive

emissions were dominated by CO₂ (88%), followed by CH₄ (11%) and N₂O (1%). The GHG 576 emissions officially accounted from the city of Brussels are overwhelmingly dominated by 577 CO₂ (99.1%) with a very small contribution from N₂O (0.6%) and CH₄ (0.3%). The relative 578 additional contribution of emissions of GHG from ponds to the total urban emissions is 579 different for each of the three GHGs individually (Table S4). The diffusive emissions from 580 581 ponds correspond to a fraction of the other urban emissions of 1.2% for CH₄, 0.05% for N₂O and 0.03% for CO₂. The total annual GHG diffusive emissions from ponds of 0.8 kTCO₂eq 582 is several orders of magnitude lower and very marginal compared to the three major 583 anthropogenic emissions in the city of Brussels such as energy (1804 kTCO₂eq yr⁻¹), 584 transport (807 kTCO₂eq yr⁻¹), and industries (277 kTCO₂eq yr⁻¹) (Fig. 8). Ebullitive fluxes of 585 CH₄ were measured in four ponds (Leybeek, Pêcheries Royales, Silex, Ten Reuken) during 586 2021, 2022, and 2023 and ranged from 0.0 to 59.2 mmol CH₄ m⁻² d⁻¹ (unpublished data). 587 Based on an exponential relationship between ebullitive CH₄ and temperature in the 4 588 sampled ponds (n=39, r²=0.54, not shown), we scaled the ebullitive CH₄ fluxes to all of 589 ponds in the city of Brussels that summed correspond to an additional GHG emission of 0.2 590 kTCO₂-eq yr⁻¹. When additionally accounting for CH₄ ebullition, the total CH₄ emissions from 591 the urban ponds (0.2 kTCO₂-eq yr⁻¹) correspond to 3.7% of the total urban emissions of 592 CH₄.Yet, the total CH₄ emissions (including CH₄ ebullition) from the urban ponds only 593 represented 0.01% of the total GHG emissions of the city of Brussels (2885 kTCO₂-eq yr⁻¹). 594 This was consistent with the contribution of CH₄ emissions from streams and ponds of 595 0.004% to the total GHG emissions from the city of Berlin reported by Herrero Ortega et al 596 (2019). The total annual GHG emissions from urban ponds, (1.0 kTCO₂-eq yr⁻¹ including 597 CH₄ ebullition) is of the same order of magnitude as smaller GHG emissions such as waste 598 (5.3 kTCO₂eq yr⁻¹) and agriculture (1.3 kTCO₂eq yr⁻¹). The total annual GHG emissions from 599 600 urban ponds was nearly identical than the estimated sink of GHGs from urban green spaces (1.2 kTCO₂eq yr⁻¹). Urban green spaces (parks and gardens) provide several ecosystem 601 602 services reviewed by Pataki et al. (2011) including reducing or offsetting GHG emissions. Here, we showed that the emissions of GHGs from ponds almost fully balance the reported 603 GHG sink reported for green spaces in the city of Brussels. As such, GHG emissions from 604 ponds should be accounted among the negative consequences or tradeoffs of implementing 605 green infrastructure, the so-called "ecosystem disservices" according to Pataki et al. (2011). 606

607

608 5. Conclusions

Our study showed that small urban ponds in the public parks of the city of Brussels 609 emit CO₂, CH₄ and N₂O to the atmosphere. The GHG concentrations and emissions in the 610 sampled ponds showed geographical variations that were partly related to differences in 611 morphology (size) but also the position with regards to the city center, and the surrounding 612 landscape for those in the urban periphery (cropland versus forest), as well as the presence 613 614 of macrophytes. Smaller ponds received larger amounts of allochthonous carbon, either directly as CO₂ or as OM resulting in larger CO₂ production from respiration. Larger ponds 615 emitted more CH₄ contrary to the pattern most commonly described in the literature. This 616 effect was due to the presence of macrophytes and the proximity of the forest in larger ponds 617 at the periphery of the city compared to the smaller ponds located closer to the city center. 618 N₂O levels were higher closer to the city center that we hypothesize resulting from a stronger 619 atmospheric nitrogen deposition based on the patterns of atmospheric NO₂. 620

Biological processes were the main drivers of GHG production, as showed from 621 622 relationships of the three GHGs with water temperature, oxygen saturation and nutrients. CH₄ concentrations were positively related to temperature on seasonal time scales due to 623 624 the dependence of methanogenesis on temperature, as usually observed in aquatic systems (Yvon-Durocher et al., 2014), including urban ponds (Bartosiewicz et al., 2016). On the 625 626 contrary the N₂O levels were negatively related to temperature possibly reflecting incomplete denitrification at low temperatures, although DIN levels also increased in winter, 627 so it is difficult to separate the impact of both potential drivers. 628

The GHG emissions from the studied ponds were lower than the other GHG sources 629 630 from the city such as road transport and building heating. However, urban green spaces are described as important carbon sinks by Pataki et al. (2011), but this sink was offset by 631 632 emissions from the ponds within these same green spaces in the city of Brussels. CH₄ emissions from inland waters were incorporated in the latest IPCC assessment report 633 (IPCC, 2019) and should now be included in current national inventories. We show that the 634 total CH₄ emissions from the urban ponds (including diffusive and ebullitive components) 635 correspond to a larger fraction of the total respective urban emissions (3.73%) than CO₂ 636 (0.05%) and N₂O (0.03%). The emissions of GHGs from the urban ponds were equivalent 637 638 to the reported GHG sink reported for green spaces in the city of Brussels.

639 Audet et al. (2020) proposed to consider the GHG emissions from ponds in the overall 640 GHG assessment of cities to provide a more comprehensive quantification of a city's carbon 641 footprint and fully quantify the ecosystem services provided by urban green spaces. Yet, the

total CO₂, N₂O, and CH₄ emissions (including CH₄ ebullition) from the urban ponds only represented 0.04% of the total GHG emissions of the city of Brussels (2885 kTCO₂-eq yr⁻¹).

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Figure 1 : Map of the metropolitan area of the region of Brussels delineated by the black line and surrounding region of Flanders in Belgium showing land cover and sampling locations in ponds (red diamonds). The star corresponds to the center of the city (50.8504°N, 4.3487°E). The acronyms correspond to the official names of the ponds of Bruxelles Environnement, the regional water management agency (Baud 1 : Etang Parc Roi Baudouin Phase I ; Baud2 : Etang Parc Roi Baudouin Phase II ; Btsf : Etang de Boitsfort ; Era 1 : Etang Erasme amont ; Era 2 : Etang Erasme aval ; GrMe : Grand étang Mellaerts ; Jos1 : Parc Josaphat - Etang aux canards ; Jos2 : Parc Josaphat – Etang aux pigeons ; Jos3 : Parc Josaphat – Etang de la laiterie ; Leop : Etang parc Leopold ; Leyb : Etang du Leybeek ; Pech : Etang Pêcheries Royales ; Ped1 : Etang de la Pède Grand Etang ; Ped2 : Etang de la Pède ; RC2 : Grand étang des Clabots (Etang 2) ; RC3 : Etang du moulin (Etang 3) ; Slx : Etang rond – Parc de Woluwe ; XL1 : Etang d'Ixelles Nord ; XL2 : Etang d'Ixelles Sud). The shape of each pond and connection to the hydrological network is shown in Figure S1.



Figure 2 : Daily mean air temperature (light grey line), daily precipitation (black line) (A) and daily wind speed (dark grey, right) (B). The empty red squares on (A) indicate the in-situ water temperature during sampling and the empty black diamonds on (B) indicate the average wind during each sampling period. Meteorological data were obtained from crowdsourcing platform of the Royal Belgian Meteorological Institute (wow.meteo.be) at the Brussels–Woluwe Saint Lambert station (50.8408 °N, 4.4234 °E).



Figure 3 : Seasonal variations of partial pressure of CO_2 (p CO_2 in ppm) (A); dissolved CH_4 concentration (CH_4 in nmol L⁻¹) (B) and N₂O saturation level (N_2O in N_2) (C) in 22 ponds in the city of Brussels during 4 seasons. Dashed lines indicate the equilibrium with atmosphere (493 ppm for p CO_2 (mean of measurements) and 100% for N_2O). Box plots show median (horizontal line), mean (cross), and 25-75% percentiles (box limits). Whiskers extend from min to max. The common letters above the box plots indicate groups that are not significantly different (Permanova). Statistical results of Permanova and Betadisper are summarized in Table S5.



Figure 4 : Partial pressure of CO₂ (pCO₂, ppm), dissolved CH₄ concentration (CH₄, nmol L⁻¹), and N₂O saturation level (%N₂O, %), versus oxygen saturation level (%O₂, %), water temperature (°C), concentration of chlorophyll-*a* (Chl-*a*, in µg L⁻¹), concentration of dissolved inorganic nitrogen (DIN= NH₄⁺ + NO₂⁻ + NO₃⁻, in µmol L⁻¹), concentration of soluble reactive phosphorus (SRP, in µmol L⁻¹) and total phosphorus concentration (P tot, mg L⁻¹) in 22 ponds in the city of Brussels during 4 seasons. Median quantile regression shown as solid lines, 25th-75th quantile regression as dashed lines. Significance indicated by ns (non-significant) * (p<0.05), ** (p<0.01), ****(p<0.0001). Statistical results of quantile regressions are summarized in Table S7.



Figure 5 : N₂O saturation level (%N₂O, %) versus dissolved inorganic nitrogen (DIN= NH₄⁺ + NO₂⁻ + NO₃⁻, in µmol L⁻¹) (A), ammonium concentration (NH₄⁺, µmol L⁻¹) (B), nitrite concentration (NO₂⁻, µmol L⁻¹) (C), nitrate concentration (NO₃⁻, µmol L⁻¹) (D), ratio of dissolved inorganic nitrogen (DIN= NH₄⁺ + NO₂⁻ + NO₃⁻, in µmol L⁻¹) to dissolved oxygen (O₂, in µmol L⁻¹) (E), ratio of NH₄⁺ to O₂ (µmol:µmol) (F), ratio of NO₂⁻ to O₂ (µmol:µmol) (G), ratio of NO₃⁻ to O₂ (µmol:µmol) (H) in 22 sampled ponds in the city of Brussels during 4 seasons. Median quantile regression shown as solid lines, 25th-75th quantile regression as dashed lines. Significance indicated by * (p<0.05), ****(p<0.0001). Statistical results of quantile regressions are summarized in Table S6.



Figure 6 : Median quantile regressions of mean partial pressure of CO_2 (p CO_2 , ppm), dissolved CH_4 concentration (CH_4 , nmol L^{-1}), and N_2O saturation level ($\%N_2O$, %) as function of surface area (A, E, I) and distance from city center (B,F, J) and boxplots of mean GHG with low (<40%, n=14) and high (>40%, n=8) macrophyte cover (MC) (C,G, K). Median quantile regression shown as solid lines, 25th-75th quantile regression as dashed lines. Significance indicated by ns (non-significant) * (p<0.05), ** (p<0.01). Box plots show median (horizontal line), mean (cross), and 25-75% percentiles (box limits). Whiskers extend from min to max. The common letters above the box plots indicate groups that are not significantly different (Permanova). Graphs D, H, L represent respectively mean p CO_2 (ppm), CH_4 , (nmol L^{-1}) and $\%N_2O$ (%) for each pond during the four spatial surveys plotted against coordinates (°E,°N). City perimeter shown in light grey, city center (50.8504°N, 4.3487°E) marked with star. Red circles evidence zones with highest values of CH_4 concentrations and $\%N_2O$. Statistical results of Permanova and Betadisper are summarized in Table S6. Statistical results of quantile regressions are summarized in Table S7.



Figure 7 : Mean dissolved inorganic nitrogen (DIN= $NH_4^+ + NO_2^- + NO_3^-$, in µmol L⁻¹) (A), ammonium (NH_4^+ , in µmol L⁻¹) (B), nitrite (NO_2^- , in µmol L⁻¹) (C), nitrate (NO_3^- , in µmol L⁻¹) (D), and atmospheric NO_2 (µg m⁻³) (E) concentrations as function of the distance from city center (50.8504°N, 4.3487°E) in 22 ponds in the city of Brussels during 4 seasons. N₂O saturation level (N_2O , N_2O) as function of atmospheric NO_2 concentration (µg m⁻³) (F). Median quantile regression shown as solid lines, 25th-75th quantile regression as dashed lines. Significance (ns non-significant, * p<0.05, ** p<0.01) indicated top right. Statistical results of quantile regressions are summarized in Table S6. The atmospheric nitrogen dioxide (NO_2) concentration (µg m⁻³) was extracted from the Curieuzenair initiative, which analyzed 2483 air samples in September 2021 covering the whole of the city of Brussels with a homogeneous distribution (<u>https://curieuzenair.brussels/en/the-results/</u>). Each data point corresponds to the average value of the six atmospheric NO_2 stations closest to each sampled pond, a distance between 50 and 450m.



Figure 8 : Official inventory of annual emissions of greenhouse gases (kT CO₂ equivalents per year) for the city of Brussels in 2020 (<u>https://environnement.brussels</u>) Energy (residential/tertiary buildings), Transport (road), Industry (industrial processes and product use), Waste (mainly wastewater treatment), Agriculture (few soils), Forest (mainly Sonian forest), greenspace (private and public gardens, grassland) and diffusive emissions estimated from Brussels ponds (Ponds) in this study (2021-2022). The specific budget for each GHG is given in Table S4.



Figure 9: dissolved CO₂ concentration (CO₂, μ mol L⁻¹) (A) and dissolved CH₄ concentration (CH₄, nmol L⁻¹) (B) in 22 sampled ponds in Brussels during 4 seasons (n=88) and natural ponds of similar size reported by Holgerson and Raymond (2016) (n=81 for CO₂, n=53 for CH₄). Box plots show median (horizontal line), mean (cross), and 25-75% percentiles (box limits). Whiskers extend from min to max. Both comparisons are significant based on Permanova test. Statistical results of Permanova and Betadisper are summarized in Table S5.

Seasonal and spatial variations of greenhouse gas (CO₂, CH₄ and N₂O) emissions from urban ponds in Brussels

(Supplementary Tables and figures)

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Figure S1 : Map of the city of Brussels and the sampled ponds. The acronyms correspond to the official names of the ponds of Bruxelles Environnement, the regional water management agency (Baud 1 : Etang Parc Roi Baudouin Phase I ; Baud2 : Etang Parc Roi Baudouin Phase II ; Btsf : Etang de Boitsfort ; Era 1 : Etang Erasme amont ; Era 2 : Etang Erasme aval ; GrMe : Grand étang Mellaerts ; Jos1 : Parc Josaphat - Etang aux canards ; Jos2 : Parc Josaphat – Etang aux pigeons ; Jos3 : Parc Josaphat – Etang de la laiterie ; Leop : Etang parc Leopold ; Leyb : Etang du Leybeek ; Pech : Etang Pêcheries Royales ; Ped1 : Etang de la Pède Grand Etang ; Ped2 : Etang du Silex ; Solv : Grand étang des Clabots (Etang 2) ; RC3 : Etang Parc Ten Reuken ; WolR : Etang rond – Parc de Woluwe ; XL1 : Etang d'Ixelles Nord ; XL2 : Etang d'Ixelles Sud). The stream names are italicized. The city center (50.8504°N, 4.3487°E) is marked with star.



Figure S2: Dissolved CH4 concentration (CH₄, nmol L⁻¹) versus the total phosphorus concentration (P tot, mg L⁻¹) (A,B,C,D) and versus dissolved inorganic nitrogen (DIN= NH₄⁺ + NO₂⁻ + NO₃⁻, in µmol L⁻¹) (E,F,G,H), N₂O saturation level (%N₂O, %) versus dissolved inorganic nitrogen (DIN= NH₄⁺ + NO₂⁻ + NO₃⁻, in µmol L⁻¹) (I,J,K,L) in 22 sampled ponds in the city of Brussels during 4 seasons. Median quantile regression shown as solid lines, 25th-75th quantile regression as dashed lines. Significance indicated by ns (non-significant) * (p<0.05), ** (p<0.01), ***(p<0.001), ****(p<0.001). Statistical results of quantile regressions are summarized in Table S6.



Figure S3: Distance from city center (km) (A), pond surface (ha) (B), pond water residence time (days) (C), mean chlorophyll-a (μ g L⁻¹) (D), dissolved inorganic nitrogen (DIN= NH₄⁺ + NO₂⁻ + NO₃⁻, in μ mol L⁻¹) (E), soluble reactive phosphorus (SRP, in μ mol L⁻¹) (F), total phosphorus (Ptot, in mg L⁻¹) (G), in 22 ponds with macrophyte cover below and above 40% during four seasons in the city of Brussels. Leopold Pond is indicated by an empty blue square and was excluded from the analysis due to extreme values for some variables. Box plots show median (horizontal line), mean (cross), and 25-75% percentiles (box limits). Whiskers extend from min to max. The common letters above the box plots indicate groups that are not significantly different (Permanova). Statistical results of Permanova and Betadisper are summarized in Table S6.



Figure S4: Mean chlorophyll-*a* concentration (Mean Chl-*a*, in μ g L⁻¹) in 22 sampled ponds during the four seasons in the city of Brussels as a function of the distance from city center (50.8504°N, 4.3487°E) (A) and on a map showing the city perimeter (line) and center (star, 50.8504°N, 4.3487°E) (B).



Figure S5 : Daily wind speed (dark grey, m s⁻¹) obtained from crowdsourcing platform of the Royal Belgian Meteorological Institute (<u>wow.meteo.be</u>) at the Brussels – Woluwe Saint Lambert station (50.8408 °N, 4.4234 °E) (A). squares indicate the mean wind for all samplings during a campaign. Seasonal comparisons of diffusive fluxes of CO₂ (B), CH₄ (C) and N₂O (D) (mmol m⁻² d⁻¹) in 22 ponds in the city of Brussels during 4 seasons. Box plots show median (horizontal line), mean (cross), and 25-75% percentiles (box limits). Whiskers extend from min to max. The common letters above the box plots indicate groups that are not significantly different (Permanova). Statistical results of Permanova and Betadisper are summarized in Table S4.



Figure S6: Chlorophyll-a concentration in (Chl-*a*, in μ g L⁻¹) in the Leybeek (Leyb) (A) and Pêcheries (Pech) (B) ponds from June 2021 to December 2022. The solid squares indicate the median values of the 22 ponds sampled at four different seasons.

Table S1: Name, acronym, catchment, geographical coordinates, surface area, depth, perimeter, residence time, distance from the city center (50.8504°N, 4.3487°E), summer macrophyte cover, over-flow or flow-through regime, fishing activity, and storm-water regulation in 22 ponds in the city of Brussels. Surface area and perimeter data were calculated from satellite images in Google Earth and the other data were obtained from data resources from Brussels Environment, the regional water management agency.

puod	9	0	0	es	es	0	0	0	0	0	0	0	es	es	0	0	0	0	0	9	9	0
	-	2	2	7	7	~	~	~	~	~	2	~	7	7	~	-	~	~	~	~	~	~
bnoq gniAsi٦	No	No	No	No	No	No	No	No	No	No	No	Yes	No	No	No	Yes	No	No	No	No	Yes	Yes
əmigəЯ	Overflow	Overflow	Flow-through	Overflow	Overflow	Overflow	Overflow	Overflow	Overflow	Overflow	Overflow	Flow-through	Overflow	Overflow	Flow-through	Flow-through	Flow-through	Flow-through	Flow-through	Overflow	Flow-through	Flow-through
cover (%) Summer macrophyte	25	60	60	25	2	50	0	15	15	50	0	15	30	20	25	40	06	80	80	15	20	15
Distance from city כפחtפר (m)	3.9	4.0	8.0	7.2	6.8	6.5	3.1	3.0	3.0	2.5	7.6	6.1	6.0	5.8	8.2	7.5	8.0	7.8	7.4	6.3	3.5	3.2
Residence time (days)	30.0	100.0	7.0	45.0	30.0	78.0	115.0	66.4	35.9	150.0	300.0	7.8	25.0	100.0	13.0	11.0	0.3	12.0	12.0	10.7	200.0	250.0
Perimeter (m)	211	671	904	316	283	939	438	206	481	515	511	558	1084	713	664	989	464	331	1148	600	774	589
ha) Surface	0.3	0.5	2.5	0.4	0.3	3.6	0.4	0.2	0.1	0.7	0.7	1.4	4.6	1.3	2.1	3.4	1.0	0.4	3.2	1.8	2.1	1.2
Average depth Average depth	0.8	0.8	0.3	0.7	0.7	1.2	0.5	0.6	0.7	1.0	0.6	1.3	1.3	0.7	2.0	1.1	1.1	1.4	1.4	1.4	1.8	1.2
(∃°) Longitude	4.3276	4.3185	4.4171	4.2666	4.2722	4.4347	4.3903	4.3878	4.3830	4.3802	4.4210	4.4119	4.2726	4.2766	4.4477	4.4463	4.4151	4.4123	4.4278	4.4317	4.3743	4.3716
(N∘) ∋buřiřaJ	50.8833	50.8813	50.7930	50.8118	50.8123	50.8274	50.8606	50.8611	50.8629	50.8390	50.8000	50.8127	50.8262	50.8254	50.8105	50.8118	50.7917	50.7925	50.8066	50.8300	50.8229	50.8257
tnemtotsD	Molenbeek	Molenbeek	Woluwe	Neerpedebeek	Neerpedebeek	Woluwe	Maelbeek	Maelbeek	Maelbeek	Maelbeek	Woluwe	Woluwe	Neerpedebeek	Neerpedebeek	Woluwe	Woluwe	Woluwe	Woluwe	Woluwe	Woluwe	Maelbeek	Maelbeek
noitsiverddA	Baud1	Baud2	Btsf	Era1	Era2	GrMe	Jos1	Jos2	Jos3	Leop	Leyb	Pech	Ped1	Ped2	RC2	RC3	SIX	Solv	Trk	WolR	XL1	XL2
Pond name (in French, according to Bruxelles Environnement)	Etang Parc Roi Baudouin Phase I (Sacré Cœur)	Etang Parc Roi Baudouin Phase II - Sud	Etang de Boitsfort ou du Moulin	Etang Erasme Amont	Etang Erasme Aval	Grand étang Mellaerts	Parc Josaphat - Etang aux canards	Parc Josaphat - Etang aux pigeons	Parc Josaphat - Etang de la laiterie	Etang Parc Leopold	Etang Parc du Leybeek	Etang Pêcheries Royales - Parc de la Héronnière	Etang de la Pède - Grand Etang	Etang de la Pede	Grand étang des Clabots (Etang 2) - Rouge-Cloître	Etang du Moulin (Etang 3) - Rouge-Cloître	Etang du Silex	Grand étang - Parc Tournay-Solvay	Etang Parc Ten Reuken	Etang rond - Parc de Woluwe	Etang d'Ixelles Nord	Etang d'Ixelles Sud

Table S2: Mean, standard deviation (SD), minimal value (min), maximal value (max), median and interquartile range (IQR) and number of samples (n) of Water temperature (in °C), pH, conductivity (in μ S cm⁻¹), Oxygen saturation (in %), Ammonium, Nitrite, Nitrate, Soluble reactive phosphorus (SRP) concentrations (in μ mol L⁻¹), total phosphorus concentration (in mg L⁻¹), Chlorophyll-a concentration (in μ g L⁻¹), total suspended matter concentration (in mg L⁻¹), partial pressure of CO₂ (pCO₂ in ppm); dissolved CH₄ concentration (CH₄ in nmol L⁻¹) and N₂O saturation level (in %) measured at 22 ponds in the city of Brussels during 4 seasons. Statistical results of Permanova and Betadisper for comparisons between seasons are summarized in Table S4.

November 2021					
Parameter	Units	mean ± SD	(min;max)	median (IQR)	n
Water temperature	°C	6.1 ± 0.7	(4.5:7.1)	6.2 (5.5 : 6.6)	22
nH		7.51 ± 0.42	$(7.12 \cdot 9.23)$	7 30 (7 10 : 7 74)	22
Conductivity	us om-1	050 ± 056	(7.12, 0.23)	705 (675 - 019)	22
Conductivity		030 ± 230	(0.440)	795 (075, 916)	22
Oxygen saturation	^{%0}	52 ± 29	(9;110)	51 (34.6; 70.9)	22
Ammonium	µmol L ⁻	57.7 ± 90.4	(2.1;447.0)	29.8 (20.4 ; 62.6)	22
Nitrite	µmol L ⁻¹	2.7 ± 3.6	(0.6; 18.3)	1.7 (1.2 ; 2.9)	22
Nitrate	umol L ⁻¹	77.2 ± 115.4	(9.3:432.8)	34.6 (21.1 : 47.4)	22
SRP	umol L ⁻¹	18+12	(0.5:4.8)	12(10.28)	22
Total phoophorup	mg 1 -1	0.1 ± 0.2	(0.1 : 0.9)	0.1 (0.1 : 0.2)	22
Oblassical	IIIg L	0.1 ± 0.2	(0.1, 0.0)	0.1 (0.1, 0.3)	22
Chiorophyli-a	µg L	9.5 ± 10.0	(0.5;33.6)	4.5 (2.1; 15.2)	22
I otal suspended matter	mg L ⁻ '	8.3 ± 14.2	(1.2;72.4)	4.6 (3.9 ; 7.6)	22
pCO ₂	ppm	5453 ± 3058	(1655 ; 15029)	4650 (3568 ; 7094)	22
CH ₄	nmol L ⁻¹	2260 ± 3523	(201; 15132)	1073 (773 ; 1714)	22
N ₂ O saturation	%	360 ± 677	(84:3324)	161 (123 : 247)	22
February 2022	1				
Water temperature	°C	60 + 12	(5.6.11.9)	6 5 (6 2 : 7 1)	21
	C	0.9 ± 1.3	(3.0, 11.0)	0.3(0.2, 7.1)	21
рн	- 1	8.11 ± 0.47	(7.05;8.75)	8.13 (7.93; 8.49)	21
Conductivity	µS cm ⁻ '	796 ± 265	(520 ; 1364)	685 (585 ; 942)	21
Oxygen saturation	%	96 ± 23	(48; 139)	102 (80 ; 111)	21
Ammonium	umol L ⁻¹	12.3 ± 27.9	(0.7 : 127.6)	2.6 (1.9 : 7.5)	21
Nitrite	umol L ⁻¹	1.6 ± 1.5	(0.1:5.0)	1.1 (0.5 : 3.1)	21
Nitrate	umol L ⁻¹	110 3 + 1/6 6	(4.6 : 619.3)	65 2 (30 8 : 129 5)	21
SDD	umol L -1	22+20	(0.2 · 2 0)	0.6 (0.4 + 2.0)	21
JRF Tatal shaashas		2.2 1 2.9	(0.2, 0.9)	0.0(0.4, 2.0)	21
i otai pnospnorus	mg L'	0.3 ± 0.2	(0.1; 0.7)	0.2 (0.2; 0.4)	21
Chlorophyll-a	µg L-'	19.7 ± 14.3	(1.8;52.8)	16.2 (7.2 ; 30.7)	21
Total suspended matter	mg L ⁻¹	9.5 ± 4.5	(3.7; 20.3)	9.2 (5.7 ; 11.1)	21
pCO ₂	ppm	2064 ± 1295	(610; 5283)	1822 (1121 ; 2524)	21
CH4	nmol 1 ⁻¹	860 + 841	$(105 \cdot 3552)$	697 (416 · 861)	21
N=O saturation	0/	182 ± 05	(77 : 425)	150 (117 : 231)	20
	70	102 1 95	(11,425)	150 (117 , 251)	20
May 2022		10000	((
Water temperature	°C	16.2 ± 2.0	(13.3 ; 20.2)	16.1 (14.6 ; 17.4)	22
pH		8.06 ± 0.50	(7.37; 9.12)	7.91 (7.72 ; 8.44)	22
Conductivity	µS cm ⁻¹	776 ± 272	(402 ; 1424)	709 (604 ; 835)	22
Oxygen saturation	%	90 ± 41	(17:164)	87 (61 : 104)	22
Ammonium	umol L ⁻¹	10 9 + 25 0	$(0.5 \cdot 117.6)$	29(18.75)	22
Nitrito	umol L ⁻¹	12+10	(0.1 : 6.6)	0.4(0.1:1.0)	22
Nillite		1.3 ± 1.0	(0.1, 0.0)	0.4 (0.1, 1.8)	22
Nitrate	µmol L 1	45.5 ± 64.9	(0.9;218.0)	10.7 (2.9 ; 69.5)	22
SRP	µmol L ⁻	1.6 ± 1.5	(0.2 ; 5.6)	1.0 (0.8 ; 1.2)	22
Total phosphorus	mg L ⁻¹	1.4 ± 0.5	(1.0; 3.3)	1.3 (1.1 ; 1.6)	22
Chlorophyll-a	ua L ⁻¹	11.9 ± 8.8	(1.7:29.0)	8.5 (5.2 : 19.2)	22
Total suspended matter	ma L ⁻¹	107+74	$(3.2 \cdot 38.5)$	83(73:121)	22
		3802 + 2003	(100 · 12320)	3787 (1218 : 5805)	22
	ppin pmol L ⁻¹	2456 ± 2201	(190, 12329)	2190 (1546 - 4700)	22
	nmoi L	3450 ± 230 I	(000,9507)	3160 (1546 ; 4790)	22
N ₂ O saturation	%	289 ± 432	(79; 1963)	125 (96 ; 250)	22
August 2022					
Water temperature	°C	22.2 ± 1.6	(19.4 ; 25.2)	22.1 (20.8 ; 23.3)	22
nH .		7 99 + 0 56	(7.08 9.10)	7 90 (7 65 8 25)	22
Conductivity	uS cm ⁻¹	667 + 270	(435 · 1557)	537 (499 · 792)	22
Oxygen saturation	%	87 + 33	(27 · 150)	93 (71 · 101)	22
	70	07 1 33	(27, 159)	95 (71, 101)	22
Ammonium		23.1 ± 10.1	(0.1;330.6)	3.0 (0.7; 15.0)	22
Nitrite	µmol L	0.4 ± 0.9	(0.0;4.2)	0.1 (0.0 ; 0.4)	22
Nitrate	µmol L ⁻	42.1 ± 112.4	(0.2;467.7)	7.8 (1.7 ; 11.8)	22
SRP	µmol L ⁻¹	2.6 ± 2.6	(0.1;8.4)	1.4 (0.4 ; 4.2)	22
Total phosphorus	mg L ⁻¹	2.0 ± 1.4	(0.2 ;5.3)	1.8 (1.2 ; 2.8)	22
Chlorophyll-a	ug 1 ⁻¹	105 9 + 149 9	$(0.1 \cdot 488.4)$	11 4 (2 8 201 9)	22
Total suspended matter	ma I -1	33 2 + 37 4	(3.2 · 110.2)	12 1 (6 1 : 59 3)	22
	ing L	2405 + 2002	(101 + 0070)	0707 (400 - 5040)	22
pCO2	ppm	3100 ± 2902	(121;8970)	2131 (409; 5218)	22
CH4	nmol L '	4691 ± 6713	(10;29190)	2790 (883 ; 4956)	22
N ₂ O saturation	%	656 ± 2182	(0;10354)	113 (77; 159)	22
All seasons					
Water temperature	°C	12.8 ± 6.9	(4.5; 25.2)	12.6 (6.4 ; 19.5)	87
Ha		7.91 ± 0.53	(7.05 ; 9.12)	7.87 (7.55 : 8.20)	87
Conductivity	uS cm ⁻¹	775 + 271	(402 · 1557)	699 (580 · 903)	87
	μο στη %	81 + 36	(9 • 164)	83 (54 + 102)	07
	70	01 1 30	(9, 104)	03 (34, 103) E 6 (24, 24.0)	0/
Ammonium	µmoi L"	20.7 ± 02.9	(0.1;447.0)	5.0 (2.1; 24.0)	87
Nitrite	µmol L ⁻	1.5 ± 2.3	(0.0; 18.3)	0.8 (0.2 ; 2.1)	87
Nitrate	µmol L ⁻¹	68.4 ± 114.6	(0.2;619.3)	24.8 (7.0 ; 66.7)	87
SRP	µmol L ⁻¹	2.0 ± 2.2	(0.1;8.9)	1.1 (0.6 ; 2.9)	87
Total phosphorus	ma L ⁻¹	1.0 ± 1.1	(0.1:5.3)	0.6 (0.15 ; 1.4)	87
Chlorophyll-a	ug L ⁻¹	36.6 + 84.6	(0.1 : 488.4)	10.0 (3.6 · 24.9)	87
Total suspended metter	mg 1-1	15 5 ± 22 0	$(1.2 \cdot 110.2)$	80(52.14 0)	07
		13.3 ± 22.0	(1.2, (10.2)	0.0 (0.2 , 11.0)	0/
pcO ₂	ppm	3007 ± 2904	(121;15029)	3016 (1394 ; 5343)	87
CH ₄	nmol L ⁻	2833 ± 4178	(10 ; 29190)	1276 (705 ; 3604)	87
N ₂ O saturation	%	273 ± 662	(0; 10354)	138 (96 ; 232)	86

Table S3: Dissolved concentrations of CO_2 , CH_4 and N_2O in urban ponds reported in literature and the present study in 22 urban ponds in the city of Brussels during 4 seasons.

Desien	-	Coverence	CO ₂	CH4	N ₂ O	Deference	
Region	п	Coverage	(µmol L ⁻¹)	(nmol L ⁻¹)	(nmol L ⁻¹)	Reiefelice	
Brussel (Belgium)	22	4 seasons	189 ± 155	2,745 ± 4,120	39 ± 95	This study	
Brussel (Belgium)	22	Summer	123 ± 112	4,691 ± 6,713	8 ± 158	This study	
Linkoping (Sweden)	1	4 seasons	-	1,733	-	Natchimuthu et al. (2014)	
Minnesota (United Sates)	7	Summer	6 ± 180	2,530 ± 32,360		Ray et al. (2023)	
Minnesota (United States)	18	Summer	170 ± 180	15,794 ± 23,107	18 ± 12	Rabaey and Cotner (2022)	
Salaspils (Lativa)	1	Summer	133 ± 17	4,914 ± 3,779		Ray et al. (2023)	
Silkeborg (Denmark)	37	4 seasons	162 ± 184	3,667 ± 16,500	28 ± 64	Audet et al. (2020)	
Silkeborg (Denmark)	37	Summer	202 ± 302	6,000 ± 32,200	-	Audet et al. (2020)	
Uppsala (Sweden)	40	Summer	157 ± 153	4,528 ± 5,712	-	Peacock et al. (2019)	
Uppsala (Sweden)	2	Summer	66 ± 74	804 ± 2,378		Ray et al. (2023)	

Table S4 : Reported annual emissions of CO_2 , CH_4 , and N_2O in CO_2 equivalents from the Brussels region for the year 2020 (<u>https://environnement.brussels</u>) and the CO_2 , CH_4 and N_2O emissions from the urban ponds (n=158, total area = 101ha), and their relative contribution. The CH₄ emissions have a diffusive component (Diff.) computed from the diffusive flux estimates in the 22 ponds during the 4 seasons, and an ebullitive component (Ebull.) computed from water temperature in the 22 ponds during the 4 seasons based on a relation between CH₄ ebullition and water temperature collected in 4 ponds (Leybeek, Pêcheries, Silex and Ten Reuken) from 2021-2023 (unpublished data).

		CO ₂	CH₄	N ₂ O
			kT CO₂ eq y⁻¹	
	Energy	1796.18	6.15	1.63
	Transport	799.01	0.73	6.87
	Industries	275.45	0.04	2.23
Officially reported	Waste	0.00	0.29	4.98
	Agriculture	0.04	0.53	0.71
	Forest	-8.68	0.00	0.50
	Greenspace	-1.19	0.00	0.00
This study	Ponds	0.74	0.09 (Diff.) 0.21 (Ebull.)	0.01
			Relative contribution (%)	
Ponds (expressed a	s fraction of the total)	0.03	1.12 (Diff.) 2.61 (Ebull.)	0.05

Table S5: Outcomes of the Betadisper and Permanova statistical tests for pairs of seasons on the medians of Water temperature (Temperature, in °C), Oxygen saturation ($\%O_2$, in %), Soluble reactive phosphorus (SRP, in µmol L⁻¹), total phosphorus concentration (Ptot, in mg L⁻¹), ammonium concentration (NH₄⁺, in µmol L⁻¹); nitrite concentration (NO₂⁻, in µmol L⁻¹), nitrate concentration (NO₃⁻, in µmol L⁻¹), dissolved inorganic nitrogen (DIN= NH₄⁺ + NO₂⁻ + NO₃⁻, in µmol L⁻¹), Chlorophyll-a concentration (Chl-a, in µg L⁻¹), total suspended matter concentration (TSM, in mg L⁻¹), dissolved CH₄ concentration (CH₄, in nmol L⁻¹), partial pressure of CO₂ (pCO₂, in ppm); N₂O saturation level ($\%N_2O$, in %), diffusive fluxes of CO₂ (FCO₂), CH₄ (FCH₄) and N₂O (FN₂O) (mmol m⁻² d⁻¹). Transformations have been applied to the data to ensure a normal distribution, and are indicated in the Transformation column.

			p-values of the Permanova test								
Figure or table	Variable	Transformation	Fall vs Winter	Fall vs Spring	Fall vs Summer	Winter vs Spring	Winter vs Summer	Spring vs Summer			
	Temperature (°C)	logarithmic	0.005 **	0.001 ***	0.001 ***	0.001 ***	0.001 ***	0.001 ***			
	%O ₂ (%)	logarithmic	0.001 ***	0.004 **	0.003 **	0.241	0.170	0.925			
	SRP (µmol L ⁻¹)	square root	0.847	0.486	0.610	0.771	0.615	0.342			
	Ptot (mg L ⁻¹)	square root	0.452	0.001 ***	0.001 ***	0.001 ***	0.001 ***	0.172			
Tab 62	NH4 ⁺ (µmol L ⁻¹)	square root	0.002 **	0.001 ***	0.004 **	0.804	0.729	0.582			
Tab. 32	NO2 ⁻ (µmol L ⁻¹)	square root	0.041 *	0.006 **	0.001 ***	0.218	0.001 ***	0.04 *			
	NO3 ⁻ (µmol L ⁻¹)	square root	0.354	0.119	0.022 *	0.015 *	0.004 **	0.419			
	DIN (µmol L ⁻¹)	square root	0.532	0.003 **	0.008 **	0.037 *	0.038 *	0.864			
	Chl-a (µg L ⁻¹)	square root	0.007 **	0.076	0.088	0.100	0.928	0.413			
	TSM (mg L ⁻¹)	square root	0.039*	0.044*	0.003 **	0.333	0.106	0.237			
	CH ₄ (nmol L ⁻¹)	logarithmic	0.008 **	0.007 **	0.092	0.001 ***	0.014 *	0.368			
Fig. 3	%N2O (%)	logarithmic	0.262	0.477	0.024*	0.764	0.047*	0.422			
	pCO ₂ (ppm)	square root	0.001 ***	0.031 *	0.002 **	0.039 *	0.373	0.337			
	FCO ₂ (mmol m ⁻² d ⁻¹)	logarithmic (value+median)	0.001 ***	0.238	0.002 **	0.018*	0.147	0.041*			
Fig. S5	FCH ₄ (mmol m ⁻² d ⁻¹)	logarithmic	0.019*	0.001***	0.092	0.001***	0.002**	0.124			
	FN ₂ O (mmol m ⁻² d ⁻¹)	logarithmic (value+median)	0.262	0.954	0.028*	0.279	0.047*	0.881			
					p-values of th	e Betadisper tes	st				
					P						
Figure or table	Variable	Transformation	Fall vs Winter	Fall vs Spring	Fall vs Summer	Winter vs Spring	Winter vs Summer	Spring vs Summer			
Figure or table	Variable Temperature (°C)	Transformation	Fall vs Winter 0.689	Fall vs Spring 0.515	Fall vs Summer 0.093	Winter vs Spring 0.987	Winter vs Summer 0.121	Spring vs Summer 0.093			
Figure or table	Variable Temperature (°C) %O ₂ (%)	Transformation logarithmic logarithmic	Fall vs Winter 0.689 0.005 **	Fall vs Spring 0.515 0.381	Fall vs Summer 0.093 0.127	Winter vs Spring 0.987 0.068	Winter vs Summer 0.121 0.134	Spring vs Summer 0.093 0.503			
Figure or table	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹)	Transformation logarithmic logarithmic square root	Fall vs Winter 0.689 0.005 ** 0.102	Fall vs Spring 0.515 0.381 0.858	Fall vs Summer 0.093 0.127 0.002 **	Winter vs Spring 0.987 0.068 0.105	Winter vs Summer 0.121 0.134 0.565	Spring vs Summer 0.093 0.503 0.071			
Figure or table	Variable Temperature (°C) %O2 (%) SRP (μmol L ⁻¹) Ptot (mg L ⁻¹)	Transformation logarithmic logarithmic square root square root	Fall vs Winter 0.689 0.005 ** 0.102 0.908	Fall vs Spring 0.515 0.381 0.858 0.879	Fall vs Summer 0.093 0.127 0.002 ** 0.001 ***	Winter vs Spring 0.987 0.068 0.105 0.779	Winter vs Summer 0.121 0.134 0.565 0.103	Spring vs Summer 0.093 0.503 0.071 0.002 **			
Figure or table	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹)	Transformation logarithmic logarithmic square root square root square root	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143	Fall vs Spring 0.515 0.381 0.858 0.879 0.138	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629	Winter vs Spring 0.987 0.068 0.105 0.779 0.942	Winter vs Summer 0.121 0.134 0.565 0.103 0.439	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397			
Figure or table	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2 ⁻ (µmol L ⁻¹)	Transformation logarithmic logarithmic square root	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397 0.057			
Figure or table	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2 ⁻ (µmol L ⁻¹) NO3 ⁻ (µmol L ⁻¹)	Transformation logarithmic logarithmic square root	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397 0.057 0.596			
Figure or table	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2 ⁻ (µmol L ⁻¹) NO3 ⁻ (µmol L ⁻¹) DIN (µmol L ⁻¹)	Transformation logarithmic logarithmic square root	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412 0.407	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665 0.722	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852 0.912	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614 0.624	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364 0.523	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397 0.057 0.596 0.829			
Figure or table	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2 ⁻ (µmol L ⁻¹) NO3 ⁻ (µmol L ⁻¹) DIN (µmol L ⁻¹) Chl-a (µg L ⁻¹)	Transformation logarithmic logarithmic square root	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412 0.407 0.721	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665 0.722 0.340	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852 0.912 0.001 ***	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614 0.624 0.194	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364 0.523 0.002 **	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397 0.057 0.596 0.829 0.002 **			
Figure or table	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2 ⁻ (µmol L ⁻¹) DIN (µmol L ⁻¹) DIN (µmol L ⁻¹) Chl-a (µg L ⁻¹) TSM (mg L ⁻¹)	Transformation logarithmic logarithmic square root	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412 0.407 0.721 0.933	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665 0.722 0.340 0.814	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852 0.912 0.001 *** 0.127	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614 0.624 0.194 0.799	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364 0.523 0.002 ** 0.006 **	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397 0.057 0.596 0.829 0.002 ** 0.002 **			
Figure or table	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2 ⁻ (µmol L ⁻¹) DIN (µmol L ⁻¹) DIN (µmol L ⁻¹) Chl-a (µg L ⁻¹) TSM (mg L ⁻¹) CH4 (nmol L ⁻¹)	Transformation logarithmic logarithmic square root square root	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412 0.407 0.721 0.933 0.934	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665 0.722 0.340 0.814 0.842	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852 0.912 0.001 *** 0.159 0.066	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614 0.624 0.194 0.799 0.9919	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364 0.523 0.002 ** 0.006 ** 0.083	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397 0.057 0.596 0.829 0.002 ** 0.002 ** 0.002 ** 0.004 ** 0.045			
Figure or table Tab. S2 Fig. 3	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2 ⁻ (µmol L ⁻¹) DIN (µmol L ⁻¹) DIN (µmol L ⁻¹) ChI-a (µg L ⁻¹) TSM (mg L ⁻¹) CH4 (nmol L ⁻¹) %N2O (%)	Transformation logarithmic logarithmic square root logarithmic logarithmic logarithmic	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412 0.407 0.721 0.933 0.934 0.445	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665 0.722 0.340 0.814 0.842 0.861	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852 0.912 0.001 *** 0.159 0.066 0.022*	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614 0.624 0.194 0.799 0.919 0.326	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364 0.523 0.002 ** 0.083 0.026*	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397 0.057 0.596 0.829 0.002 ** 0.002 ** 0.002 ** 0.004 ** 0.045 0.031*			
Figure or table Tab. S2 Fig. 3	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2* (µmol L ⁻¹) NO3* (µmol L ⁻¹) DIN (µmol L ⁻¹) ChI-a (µg L ⁻¹) TSM (mg L ⁻¹) CH4 (nmol L ⁻¹) %N2O (%) pCO2 (ppm)	Transformation logarithmic logarithmic square root logarithmic logarithmic logarithmic square root	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412 0.407 0.721 0.933 0.934 0.445 0.119	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665 0.722 0.340 0.814 0.842 0.861 0.197	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852 0.912 0.001 *** 0.159 0.066 0.022* 0.122	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614 0.624 0.194 0.799 0.919 0.326 0.049 **	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364 0.523 0.002 ** 0.083 0.026* 0.059	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.057 0.596 0.829 0.002 ** 0.002 ** 0.002 ** 0.002 ** 0.004 ** 0.045 0.031* 0.417			
Figure or table Tab. S2 Fig. 3	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2 ⁻ (µmol L ⁻¹) NO3 ⁻ (µmol L ⁻¹) DIN (µmol L ⁻¹) ChI-a (µg L ⁻¹) TSM (mg L ⁻¹) CH4 (nmol L ⁻¹) %N2O (%) pCO2 (ppm) FCO2 (mmol m ² d ⁻¹)	Transformation logarithmic logarithmic square root logarithmic logarithmic logarithmic logarithmic logarithmic (value+median)	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412 0.407 0.721 0.933 0.934 0.445 0.119 0.903	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665 0.722 0.340 0.814 0.842 0.861 0.197 0.064	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852 0.912 0.001 *** 0.159 0.066 0.022* 0.122 0.031*	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614 0.624 0.194 0.799 0.919 0.326 0.049 ** 0.019*	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364 0.523 0.002 ** 0.006 ** 0.083 0.026* 0.059 0.017*	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.057 0.596 0.829 0.002 ** 0.004 ** 0.045 0.031* 0.417			
Figure or table Tab. S2 Fig. 3 Fig. S5	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NH4* (µmol L ⁻¹) NO2 ⁻ (µmol L ⁻¹) NO3 ⁻ (µmol L ⁻¹) DIN (µmol L ⁻¹) ChI-a (µg L ⁻¹) TSM (mg L ⁻¹) CH4 (nmol L ⁻¹) %N2O (%) pCO2 (ppm) FCO2 (mmol m ⁻² d ⁻¹)	Transformation logarithmic logarithmic square root logarithmic logarithmic logarithmic (value+median) logarithmic	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412 0.407 0.721 0.933 0.934 0.445 0.119 0.903 0.928	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665 0.722 0.340 0.814 0.842 0.861 0.197 0.064	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852 0.912 0.001 *** 0.159 0.066 0.022* 0.122 0.031* 0.084	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614 0.624 0.194 0.799 0.919 0.326 0.049 ** 0.566	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364 0.523 0.002 ** 0.083 0.026* 0.059 0.017* 0.075	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397 0.557 0.596 0.829 0.002 ** 0.004 ** 0.045 0.031* 0.417 0.969 0.095			
Figure or table Tab. S2 Fig. 3 Fig. S5	Variable Temperature (°C) %O2 (%) SRP (µmol L ⁻¹) Ptot (mg L ⁻¹) NO2 ⁻ (µmol L ⁻¹) NO3 ⁻ (µmol L ⁻¹) DIN (µmol L ⁻¹) DIN (µmol L ⁻¹) ChI-a (µg L ⁻¹) TSM (mg L ⁻¹) CH4 (nmol L ⁻¹) %N2O (%) pCO2 (ppm) FCH2 (mmol m ⁻² d ⁻¹) FCH4 (mmol m ⁻² d ⁻¹)	Transformation logarithmic logarithmic square root logarithmic logarithmic logarithmic (value+median) logarithmic (value+median) logarithmic	Fall vs Winter 0.689 0.005 ** 0.102 0.908 0.143 0.801 0.412 0.407 0.721 0.933 0.934 0.445 0.119 0.903 0.928 0.273	Fall vs Spring 0.515 0.381 0.858 0.879 0.138 0.492 0.665 0.722 0.340 0.814 0.842 0.861 0.197 0.064 0.508	Fall vs Summer 0.093 0.127 0.002 ** 0.001 *** 0.629 0.392 0.852 0.912 0.001 *** 0.159 0.066 0.022* 0.122 0.031* 0.084 0.046*	Winter vs Spring 0.987 0.068 0.105 0.779 0.942 0.584 0.614 0.624 0.194 0.799 0.919 0.326 0.049 ** 0.566 0.174	Winter vs Summer 0.121 0.134 0.565 0.103 0.439 0.078 0.364 0.523 0.002 ** 0.083 0.026* 0.059 0.017* 0.075	Spring vs Summer 0.093 0.503 0.071 0.002 ** 0.397 0.576 0.596 0.829 0.002 ** 0.004 ** 0.045 0.031* 0.417 0.969 0.095			

Table S6: Outcomes of the Betadisper and Permanova statistical tests for variables shown in box plots from the listed Figures. Comparisons marked with a square (**•**) are not shown in boxplots but were conducted using quantile regressions. These additional comparisons were performed to assess a statistical value of the dispersion, which is not provided in quantile regressions. Transformations have been applied to the data to ensure a normal distribution, and are indicated in the Transformation column.

Low (<40%) vs high (>40%) macrophyte cover Fig. 6 pCO2 (ppm) square root 0.641 0.136 Fig. 6 CH4 (nmol L') logarithmic 0.589 0.005** %N-O (%) logarithmic 0.524 0.361 Part (ng L') logarithmic 0.294 0.003** Part (ng L') none 0.117 0.127 SRP (unol L') none 0.663 0.048* Distance (km) logarithmic 0.633 0.048* Surface (ha) none 0.669 0.041* Distance (km) logarithmic 0.142 0.028* Surface (ha) none 0.669 0.041* DIN (umol L') square root 0.1472 0.028* Fig. 9 pCO2 (umol L') logarithmic 0.1472 0.028* Fig. 9 pCO2 (umol L') logarithmic 0.472 0.028* OLN (umol L') square root 0.142 0.041* DIN (umol L') square root 0.142 0.041*		Variable	Transformation	Betadisper	Permanova						
pCO2 (ppm) square root 0.641 0.136 Fig. 6 CH4 (nmol L ¹) logarithmic 0.989 0.005** 5%A/O (%) logarithmic 0.524 0.361 Ch1-a (µg L ¹) logarithmic 0.294 0.003** Pe, (mg L ³) none 0.833 0.975 SRP (µmol L ³) none 0.263 0.144 Distance (Ime (days) none 0.263 0.144 Distance (Ime) logarithmic 0.633 0.048* Surface (ha) none 0.669 0.041* DIN (µmol L ¹) square root 0.1472 0.028* 0.425 0.014* 0.125 0.001** 0.42* 0.003** 0.42* 0.003** 0.42* 0.003** 0.42* 0.003** 0.42* 0.003** 0.42* 0.003**		Low (<40%) vs high (>40%) macrophyte cover									
Fig. 6 CH4 (nmol L ⁻¹) logarithmic 0.989 0.005** %N2 (%) logarithmic 0.524 0.361 PA (mg L ⁻¹) logarithmic 0.294 0.003** PA (mg L ⁻¹) none 0.833 0.975 SRP (µmol L ⁻¹) none 0.117 0.127 SRP (µmol L ⁻¹) none 0.669 0.044* Distance (km) logarithmic 0.63 0.048* Surface (ha) none 0.669 0.044* DIN (µmol L ⁻¹) square root 0.147 0.028* PpC02 (µmol L ⁻¹) logarithmic 0.125 0.014* PpC02 (µmol L ⁻¹) logarithmic 0.125 0.014* Pig. 9 %N20 (%) logarithmic 0.02* 0.00*** Minol L ⁻¹) square root 0.132 0.041* None NN4* (µmol L ⁻¹) square root 0.142 0.00*** None NN4* (µmol L ⁻¹) square root 0.132 0.00*** None NN4* (µmol L ⁻¹)		pCO ₂ (ppm)	square root	0.641	0.136						
%N2O (%) logarithmic 0.524 0.381 PA: (mg L ⁻¹) logarithmic 0.294 0.003** PA: (mg L ⁻¹) none 0.833 0.975 SRP (um L ⁻¹) none 0.833 0.975 SRP (um L ⁻¹) none 0.833 0.975 SRP (um L ⁻¹) none 0.63 0.44* Distance (km) logarithmic 0.63 0.04* Surface (ha) none 0.669 0.04* DIN (umol L ⁻¹) square root 0.1472 0.02* Natural (Holgerson and Raymond, 2016) vs Brussels poids 0.01** Fig. 9 pCO ₂ (umol L ⁻¹) logarithmic 0.12 0.001** PCO ₂ (umol L ⁻¹) logarithmic 0.02* 0.003** MoCh ² (umol L ⁻¹) logarithmic 0.12 0.004** OUT Center vs Periphery (a) MoL ² (umol L ⁻¹) square root 0.132 0.003** None %No ² (umol L ⁻¹) square root 0.104<	Fig. 6	CH ₄ (nmol L ⁻¹)	logarithmic	0.989	0.005**						
Chl-a (µg L ⁻¹) logarithmic 0.294 0.003** P _m (mg L ⁻¹) none 0.833 0.975 SRP (µm0 L ⁻¹) none 0.117 0.127 Residence time (days) none 0.263 0.144 Distance (km) logarithmic 0.63 0.048* Surface (ha) none 0.669 0.041* DIN (µmol L ⁻¹) square root 0.1472 0.028* Natural (Holgerson and Raymond, 2016) vs Brussels ponds (This Study) Tecture vs Periphery (ar) 0.014** CH ₄ (nmol L ⁻¹) logarithmic 0.02* 0.001*** SKh2O (%) logarithmic 0.02* 0.003** DIN (µmol L ⁻¹) square root 0.132 0.041* NA* (µmol L ⁻¹) square root 0.132 0.041* NO ₅ (µmol L ⁻¹) square root 0.125 0.009** NO ₅ (µmol L ⁻¹) square root 0.125 0.009** NO ₅ (µmol L ⁻¹) square root 0.125 0.009** NO ₅ (µmol L ⁻¹		%N2O (%)	logarithmic	0.524	0.361						
Pack (mg L ⁻¹) none 0.833 0.975 SRP (µmol L ⁻¹) none 0.117 0.127 Residence time (days) none 0.263 0.144 Distance (km) logarithmic 0.63 0.048* Surface (ha) none 0.669 0.041* DIN (µmol L ⁻¹) square root 0.127 0.028* Natural (Holgeron and Raymond, 2016) vs Brussels ponds (This Study) Tetre vs Pariphery (a) PCO ₂ (µmol L ⁻¹) logarithmic 0.973 0.001** Fig. 3 pCO ₂ (µmol L ⁻¹) logarithmic 0.02* 0.003** PCO ₂ (µmol L ⁻¹) logarithmic 0.02* 0.003** One Square root 0.125 0.009** NN ₂ (µmol L ⁻¹) square root 0.125 0.009** NO ₂ (µmol L ⁻¹) square root 0.125 0.009** NO ₂ (µmol L ⁻¹) square root 0.125 0.009** NO ₂ (µmol L ⁻¹) square root 0.073		Chl-a (µg L ⁻¹)	logarithmic	0.294	0.003**						
SRP (µmol L ⁻¹) none 0.117 0.127 Residence time (days) none 0.263 0.144 Distance (km) logarithmic 0.63 0.048* Surface (na) none 0.669 0.041* DIN (µmol L ⁻¹) square root 0.1472 0.028* Natural (Holgerson and Raymond, 2016) vs Brussels ponds (This Study) Center vs Poriphery (a) Genter vs Poriphery (a) Surface (na) 0.003** OL92 (µmol L ⁻¹) logarithmic 0.02* 0.003** Genter vs Poriphery (a) Surface (rs) 0.003** OL92 (µmol L ⁻¹) logarithmic 0.02* 0.003** OL92 (µmol L ⁻¹) square root 0.132 0.041* None Surface (rs) is square root OL92 None (¹) square root OL92 None (¹) Square root OL92		P _{tot} (mg L ⁻¹)	none	0.833	0.975						
Fig. S3 Residence time (days) none 0.263 0.144 Distance (km) logarithmic 0.633 0.048* Surface (ha) none 0.669 0.041* DIN (umol L*) square root 0.1472 0.028* DIN (umol L*) logarithmic 0.125 0.014* DS CH4 (nmol L*) logarithmic 0.973 0.001*** Fig. 9 PCO ₂ (umol L*) logarithmic 0.02* 0.003** Fig. 9 PKN2O (%) logarithmic 0.02* 0.004** None NN4 (umol L*) square root 0.132 0.004** NN4 (umol L*) square root 0.144 0.265 0.009** NOs (umol L*) square root 0.132 0.009** NOs (umol L*) square root 0.144 0.007** PCO ₂ (ppm) square root 0.174 0.009** NOs (umol L*) square root 0.073 0.328 OH (umol L*) square root 0.03* 0.012* <		SRP (µmol L ⁻¹)	none	0.117	0.127						
Distance (km) logarithmic 0.63 0.048* Surface (ha) none 0.669 0.041* DIN (µmol L ⁻¹) square root 0.1472 0.208* Natural (Holgerson and Raymond, 2016) vs Brussels ponts This Study Terminal (Holgerson and Raymond, 2016) vs Brussels ponts Din (Jmol L ⁻¹) logarithmic 0.125 0.014* Fig. 9 pCO ₂ (µmol L ⁻¹) logarithmic 0.02* 0.003** Din (Jmol L ⁻¹) logarithmic 0.02* 0.003** Din (Jmol L ⁻¹) square root 0.481 0.265 None DIN (Jmol L ⁻¹) square root 0.441 0.009** NO ₂ (Jmol L ⁻¹) square root 0.104 0.009** NO ₂ (Jmol L ⁻¹) square root 0.104 0.009** NO ₂ (Jmol L ⁻¹) square root 0.014 0.009** NO ₂ (Jmol L ⁻¹) square root 0.03* 0.012* pCO ₂ (ppm) square root 0.03* 0.012* DIN (µmol L ⁻¹) square root <td>Fig. S3</td> <td>Residence time (days)</td> <td>none</td> <td>0.263</td> <td>0.144</td>	Fig. S3	Residence time (days)	none	0.263	0.144						
Surface (ha) none 0.669 0.041* DIN (µmol L ¹) square root 0.1472 0.028* Natural (Holgerson and Raymond, 2016) vs Brussels ponds (This Study) Fig. 9 PCO ₂ (µmol L ⁻¹) Iogarithmic 0.125 0.001** Fig. 9 %N2O (%) Iogarithmic 0.02* 0.003** MN2O (%) Iogarithmic 0.02* 0.009** NOne Small surfaces (<1ha) vs High Surfaces (>1ha) 0.009** MN2O (%) Iogarithmic 0.03* 0.009** MN2O (%) <th co<="" colspan="2" td=""><td></td><td>Distance (km)</td><td>logarithmic</td><td>0.63</td><td>0.048*</td></th>	<td></td> <td>Distance (km)</td> <td>logarithmic</td> <td>0.63</td> <td>0.048*</td>			Distance (km)	logarithmic	0.63	0.048*				
DlN (µmol L ⁻¹) square root 0.1472 0.028* Natural (Holgerson and Raymond, 2016) vs Brussels pows (This Study) Fig. 9 pCO ₂ (µmol L ⁻¹) logarithmic 0.125 0.014* Center vs Periphery (a) Center vs Periphery (a) ØN ₂ O (%) logarithmic 0.02* 0.003** ONO ² 0.004* None ⁹ N ₂ O (%) logarithmic 0.02* 0.004* None ⁹ N ₂ O (%) logarithmic 0.02* 0.004* None ⁹ N ₂ O (%) logarithmic 0.481 0.265 NOne' (µmol L ⁻¹) square root 0.114 0.009** NOs' (µmol L ⁻¹) square root 0.104 0.007** Small surfaces (<1ha) vs High Surfaces (>1ha) (#) PCO ₂ (ppm) square root 0.03* 0.012* ONO ² (µmol L ⁻¹) square root 0.03* 0.012* ONO ² (µmol L ⁻¹) square root 0.03 0.037* </td <td></td> <td>Surface (ha)</td> <td>none</td> <td>0.669</td> <td>0.041*</td>		Surface (ha)	none	0.669	0.041*						
Natural (Holgerson and Raymond, 2016) vs Brussels ponds (This Study) Fig. 9 pCO ₂ (µmol L ⁻¹) logarithmic 0.125 0.014* CH4 (nmol L ⁻¹) logarithmic 0.973 0.001*** Center vs Periphery (m) Center vs Periphery (m) Mono C%) 0.003** Mono L ⁻¹) square root 0.009** None Mono L ⁻¹) square root 0.003** Mono L ⁻¹) square root 0.009** Mono L ⁻¹) square root 0.009** Small surfaces (<1ha) vs High Surfaces (>1ha) vs High Surfaces (>1ha) 0.012* Mono C%) Oli (µmol L ⁻¹) square root 0.03* 0.043* Mono C%)		DIN (µmol L ⁻¹)	square root	0.1472	0.028*						
Fig. 9 pCO ₂ (µmol L ⁻¹) logarithmic 0.125 0.014* CH ₄ (nmol L ⁻¹) logarithmic 0.973 0.001*** Center vs Periphery (a) MA ₂ O (%) logarithmic 0.02* 0.003** DIN (µmol L ⁻¹) square root 0.132 0.041* NH4* (µmol L ⁻¹) square root 0.481 0.265 NO ₂ (µmol L ⁻¹) square root 0.104 0.007** NO ₅ (µmol L ⁻¹) square root 0.104 0.007** NO ₅ (µmol L ⁻¹) square root 0.104 0.007** pCO ₂ (ppm) square root 0.104 0.007** pCO ₂ (ppm) square root 0.073 0.328 CH ₄ (nmol L ⁻¹) logarithmic 0.573 0.009** MNeP MN ₂ O (%) logarithmic 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.103 0.03* 0.012* MNeP MN ₂ O (%) logarithmic 0.03* 0.012* NOne NN ₂ O (%) logarithmi			Natural (Holgerson and Raymond, 201	6) vs Brussels ponds (This Study)							
Fig. 3 CH4 (nmol L ⁻¹) logarithmic 0.973 0.001*** Center vs Periphery (m) %N2O (%) logarithmic 0.02* 0.003** DIN (µmol L ⁻¹) square root 0.132 0.041* NH4* (µmol L ⁻¹) square root 0.481 0.265 NO2 (µmol L ⁻¹) square root 0.125 0.009** NO3 (µmol L ⁻¹) square root 0.104 0.007** pCO2 (ppm) square root 0.073 0.328 CH4 (nmol L ⁻¹) logarithmic 0.03* 0.009** pCO2 (ppm) square root 0.073 0.328 CH4 (nmol L ⁻¹) logarithmic 0.03* 0.012* pCO2 (ppm) square root 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.0953 0.043* NOne NN4* (µmol L ⁻¹) square root 0.103 0.037* NOne NN4* (µmol L ⁻¹) square root 0.103 0.037* NOne NN4* (µmol L ⁻¹) square root 0.107	Fig. 0	pCO ₂ (µmol L ⁻¹)	logarithmic	0.125	0.014*						
Center vs Periphery (a) %N20 (%) logarithmic 0.02* 0.003** DIN (µmol L ⁻¹) square root 0.132 0.041* None 0.481 0.265 0.009** NO2: (µmol L ⁻¹) square root 0.104 0.007** NO2: (µmol L ⁻¹) square root 0.104 0.007** NO2: (µmol L ⁻¹) square root 0.104 0.007** pCO2 (ppm) square root 0.073 0.328 CH4, nmol L ⁻¹) logarithmic 0.573 0.009** DIN (µmol L ⁻¹) logarithmic 0.03* 0.012* MO2: (µmol L ⁻¹) square root 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.03* 0.043* MO2: (µmol L ⁻¹) square root 0.11 0.567 None NO2: (µmol L ⁻¹) square root 0.103 0.037* None NO2: (µmol L ⁻¹) square root 0.107 0.047* None NO2: (µmol L ⁻¹) square root 0.107 0.047*	Fig. 9	CH ₄ (nmol L ⁻¹)	logarithmic	0.973	0.001***						
%N20 (%) logarithmic 0.02* 0.003** DIN (µmol L ⁻¹) square root 0.132 0.041* NNne NH4* (µmol L ⁻¹) square root 0.481 0.265 NO ₂ (µmol L ⁻¹) square root 0.104 0.007** NO ₂ (µmol L ⁻¹) square root 0.104 0.007** PCO ₂ (ppm) square root 0.073 0.328 CH ₄ (nmol L ⁻¹) logarithmic 0.573 0.009** DIN (µmol L ⁻¹) logarithmic 0.03* 0.012* PCO ₂ (ppm) square root 0.03* 0.012* DIN (µmol L ⁻¹) logarithmic 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.103 0.037* None MA ₄ * (µmol L ⁻¹) square root 0.11 0.567 None NO ₂ (µmol L ⁻¹) square root 0.103 0.037* None NO ₂ (µmol L ⁻¹) square root 0.107 0.047* None NO ₂ (µmol L ⁻¹) square root 0.107 0.047*			Center vs Per	iphery (=)							
DIN (µmol L ⁻¹) square root 0.132 0.041* N0ae NH4* (µmol L ⁻¹) square root 0.481 0.265 N0ae N0ae (µmol L ⁻¹) square root 0.125 0.009** N0ae (µmol L ⁻¹) square root 0.104 0.007** pC0a (µmol L ⁻¹) square root 0.073 0.328 cH4 (nmol L ⁻¹) logarithmic 0.573 0.009** CH4 (nmol L ⁻¹) logarithmic 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.03* 0.04* None NNae (Nae (1) square root 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.11 0.567 None NOae (µmol L ⁻¹) square root 0.103 0.037* None NOae (µmol L ⁻¹) square root 0.107 0.047* None NOae (µmol L ⁻¹) square root 0.107 0.047* NOae (µmol L ⁻¹) square root 0.107 0.047* NOae (µmol L ⁻¹) square root </td <td></td> <td>%N₂O (%)</td> <td>logarithmic</td> <td>0.02*</td> <td>0.003**</td>		%N ₂ O (%)	logarithmic	0.02*	0.003**						
None NH4* (µmol L ⁻¹) square root 0.481 0.265 NO2: (µmol L ⁻¹) square root 0.125 0.009** NO3: (µmol L ⁻¹) square root 0.104 0.007** pCO2 (ppm) square root 0.073 0.328 CH4 (nmol L ⁻¹) logarithmic 0.573 0.009** Small surfaces (<1ha) vs High Surfaces (>1ha) (m) %N2O (%) logarithmic 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.0953 0.043* Nore MA* (µmol L ⁻¹) square root 0.103 0.037* Nore NO2: (µmol L ⁻¹) square root 0.103 0.037* Nore NO2: (µmol L ⁻¹) square root 0.103 0.037* Nore NO2: (µmol L ⁻¹) square root 0.107 0.047* NO2: (µmol L ⁻¹) square root 0.107 0.047* NO3: (µmol L ⁻¹) square root 0.107 0.047* NO2: (ppm) square root 0.128 0.		DIN (µmol L ⁻¹)	square root	0.132	0.041*						
None NO2 [·] (µmol L ⁻¹) square root 0.125 0.009** NO3 [·] (µmol L ⁻¹) square root 0.104 0.007** pCO2 (ppm) square root 0.073 0.328 CH4 (nmol L ⁻¹) logarithmic 0.573 0.009** Small surfaces (<1ha) vs High Surfaces (>1ha) (m) Small surfaces (<1ha) vs High Surfaces (>1ha) (m Small surfaces (<1ha) vs High Surfaces (>1ha) (m O D O O		NH4 ⁺ (μmol L ⁻¹)	square root	0.481	0.265						
NO3 [•] (µmol L ⁻¹) square root 0.104 0.007** pCO2 (ppm) square root 0.073 0.328 CH4 (nmol L ⁻¹) logarithmic 0.573 0.009** Small surfaces (<1ha) vs High Surfaces (>1ha) (=) DIN (µmol L ⁻¹) logarithmic 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.103 0.043* NOa ^c (µmol L ⁻¹) square root 0.103 0.037* NOa ^c (µmol L ⁻¹) square root 0.103 0.037* NOa ^c (µmol L ⁻¹) square root 0.103 0.037* NOa ^c (µmol L ⁻¹) square root 0.107 0.047* NOa ^c (µmol L ⁻¹) square root 0.107 0.047* NOa ^c (µmol L ⁻¹) square root 0.107 0.047* PCO2 (ppm) square root 0.128 0.018* CH4 (nmol L ⁻¹) logarithmic 0.873 0.011*	None	NO2 ⁻ (µmol L ⁻¹)	square root	0.125	0.009**						
pCO2 (ppm) square root 0.073 0.328 CH4 (nmol L ⁻¹) logarithmic 0.573 0.009** Small surfaces (<1ha) vs High Surfaces (>1ha) (=) MA2O (%) logarithmic 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.0953 0.043* NOa 0.012* 0.012* 0.012* NNee NH4* (µmol L ⁻¹) square root 0.11 0.567 NO2 (µmol L ⁻¹) square root 0.103 0.037* NO3 (µmol L ⁻¹) square root 0.107 0.047* NO3 (µmol L ⁻¹) square root 0.107 0.047* NO2 (µmol L ⁻¹) square root 0.107 0.047* PCO2 (ppm) square root 0.128 0.018* CH4 (nmol L ⁻¹) logarithmic 0.873 0.011*		NO3 ⁻ (μmol L ⁻¹)	square root	0.104	0.007**						
CH4 (nmol L ⁻¹) logarithmic 0.573 0.009** Small surfaces (<1ha) vs High Surfaces (>1ha) (=)		pCO ₂ (ppm)	square root	0.073	0.328						
Small surfaces (<1ha) vs High Surfaces (>1ha) (m) %N2O (%) logarithmic 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.0953 0.043* NOne NH4* (µmol L ⁻¹) square root 0.11 0.567 NO.2° (µmol L ⁻¹) square root 0.103 0.037* NO.3° (µmol L ⁻¹) square root 0.107 0.047* PCO2 (ppm) square root 0.128 0.018* CH4 (nmol L ⁻¹) logarithmic 0.873 0.011*		CH ₄ (nmol L ⁻¹)	logarithmic	0.573	0.009**						
%N2O (%) logarithmic 0.03* 0.012* DIN (µmol L ⁻¹) square root 0.0953 0.043* NH4* (µmol L ⁻¹) square root 0.11 0.567 NO2 ⁻ (µmol L ⁻¹) square root 0.103 0.037* NO3 ⁻ (µmol L ⁻¹) square root 0.107 0.047* PCO2 (ppm) square root 0.128 0.018* CH4 (nmol L ⁻¹) logarithmic 0.873 0.011*			Small surfaces (<1ha) vs H	igh Surfaces (>1ha) (∎)							
DIN (µmol L ⁻¹) square root 0.0953 0.043* NH4* (µmol L ⁻¹) square root 0.11 0.567 NO2 ⁻ (µmol L ⁻¹) square root 0.103 0.037* NO3 ⁻ (µmol L ⁻¹) square root 0.107 0.047* pCO2 (ppm) square root 0.128 0.018* CH4 (nmol L ⁻¹) logarithmic 0.873 0.011*		%N ₂ O (%)	logarithmic	0.03*	0.012*						
Nne NH4* (µmol L-1) square root 0.11 0.567 NO2: (µmol L-1) square root 0.103 0.037* NO3: (µmol L-1) square root 0.107 0.047* pCO2 (ppm) square root 0.128 0.018* CH4 (nmol L-1) logarithmic 0.873 0.011*		DIN (µmol L ⁻¹)	square root	0.0953	0.043*						
None NO2 ⁻ (µmol L ⁻¹) square root 0.103 0.037* NO3 ⁻ (µmol L ⁻¹) square root 0.107 0.047* pCO2 (ppm) square root 0.128 0.018* CH4 (nmol L ⁻¹) logarithmic 0.873 0.011*		NH4 ⁺ (µmol L ⁻¹)	square root	0.11	0.567						
NO3 ⁻ (µmol L ⁻¹) square root 0.107 0.047* pCO2 (ppm) square root 0.128 0.018* CH4 (nmol L ⁻¹) logarithmic 0.873 0.011*	None	NO2 ⁻ (µmol L ⁻¹)	square root	0.103	0.037*						
pCO2 (ppm) square root 0.128 0.018* CH4 (nmol L ⁻¹) logarithmic 0.873 0.011*		NO3 ⁻ (µmol L ⁻¹)	square root	0.107	0.047*						
CH ₄ (nmol L ⁻¹) logarithmic 0.873 0.011*		pCO ₂ (ppm)	square root	0.128	0.018*						
		CH ₄ (nmol L ⁻¹)	logarithmic	0.873	0.011*						

Figure	Database	Variable 1 (response)	Variable 2 (explicative)	Equation of quantile regression	p-value (Wald test)	Statistical significance	
			%O ₂ (%)	$\sqrt{pCO_2} = 198.06 - 76.83 \times log(\%O_2)$	0.0000	****	
			Temperature (°C)	$\sqrt{pCO_2} = 66.2 - 9.10 \times \log(T)$	0.5400	ns	
		nCO (nnm)	Chl-a (µg L-1)	$\sqrt{pCO_2} = 64.14 - 2.18 \times \sqrt{Chla}$	0.1095	ns	
		pCO ₂ (ppm)	DIN (µmol L ⁻¹)	$\sqrt{pCO_2} = 45.42 + 1.40 \times \sqrt{DIN}$	0.1624	ns	
			SRP (µmol L ⁻¹)	$\sqrt{pCO_2} = 24.37 + 22.60 \times \sqrt{SRP}$	0.0004	***	
			Ptot (mg L ⁻¹)	$\sqrt{pCO_2} = 47.86 + 12.21 \times \sqrt{P_{tot}}$	0.4448	ns	
			%O ₂ (%)	$log(CH_4) = 4.60 - 0.80 \times log(\%O_2)$	0.0124	*	
			Temperature (°C)	$log(CH_4) = 2.22 + 0.91 \times log(T)$	0.0000	****	
Fig. 4		$C \vdash (nmol \downarrow -1)$	Chl-a (µg L ⁻¹)	$log(CH_4) = 3.26 - 0.04 \times \sqrt{Chla}$	0.0296	*	
			DIN (µmol L ⁻¹)	$log(CH_4) = 3.52 - 0.04 \times \sqrt{DIN}$	0.0069	**	
			SRP (µmol L ⁻¹)	$log(CH_4) = 3.06 + 0.02 \times \sqrt{SRP}$	0.8903	ns	
			Ptot (mg L ⁻¹)	$log(CH_4) = 2.67 + 0.63 \times \sqrt{P_{tot}}$	0.0000	****	
	All		%O ₂ (%)	$log(\%N_2O) = 2.78 - 0.33 \times log(\%O_2)$	0.0488	*	
			Temperature (°C)	$log(\%N_20) = 2.36 - 0.21 \times log(T)$	0.0356	*	
			Chl-a (µg L-1)	$log(\%N_2O) = 2.13 + 0.01 \times \sqrt{Chla}$	0.6087	ns	
			DIN (µmol L ⁻¹)	$log(\%N_2O) = 1.88 + 0.04 \times \sqrt{DIN}$	0.0000	****	
			SRP (µmol L ⁻¹)	$log(\%N_2O) = 1.96 + 0.15 \times \sqrt{SRP}$	0.0956	ns	
			P _{tot} (mg L ⁻¹)	$log(\%N_2O) = 2.21 - 0.07 \times \sqrt{P_{tot}}$	0.2153	ns	
	_	%N ₂ O (%)	NH4 ⁺ (µmol L ⁻¹)	$log(\%N_2O) = 2.06 + 0.03 \times \sqrt{(NH_4^+)}$	0.0436	*	
			NO3 ⁻ (µmol L ⁻¹)	$log(\%N_2O) = 1.91 + 0.04 \times \sqrt{NO_3}$	0.0000	****	
			NO2 ⁻ (µmol L ⁻¹)	$log(\%N_2O) = 1.84 + 0.36 \times \sqrt{NO_2^-}$	0.0000	****	
Fig. 5			O2:DIN (mol:mol)	$log(%N_2O) = 2.41 - 0.35 \times log(O_2:DIN)$	0.0001	****	
			O ₂ : NH ₄ ⁺ (mol:mol)	$log(%N_2O) = 2.27 - 0.07 \times log(O_2:NH_4^+)$	0.1007	ns	
			O ₂ : NO ₃ ⁻ (mol:mol)	$log(\%N_2O) = 2.46 - 0.31 \times log(O_2:NO_3^-)$	0.0001	****	
			O2: NO2 ⁻ (mol:mol)	$log(\%N_20) = 2.94 - 0.32 \times log(O_2:NO_2^-)$	0.0001	****	
			Surface (ha)	$\sqrt{pCO_2} = 68.64 - 6.74 \times Surface$	0.0020	**	
		pCO ₂ (ppm)	Distance (km)	$\sqrt{pCO_2} = 57.64 - 0.23 \times Distance$	0.9203	ns	
=	Average for		Surface (ha)	$log(CH_4) = 3.18 + 0.13 \times Surface$	0.0168	*	
Fig. 6	each pond	CH4 (nmoi L ⁻)	Distance (km)	$log(CH_4) = 2.92 + 0.08 \times Distance$	0.0036	**	
			Surface (ha)	$log(%N_2O) = 2.36 - 0.12 \times Surface$	0.0479	*	
		%N ₂ O (%)	Distance (km)	$log(%N_2O) = 2.73 - 0.09 \times Distance$	0.0081	**	
	Fall			$log(CH_4) = 1.99 + 1.35 \times \sqrt{P_{tot}}$	0.0273	*	
	Winter	_		$log(CH_4) = 2.87 - 0.09 \times \sqrt{P_{tot}}$	0.8880	ns	
	Spring	_	P _{tot} (mg L ⁻¹)	$log(CH_4) = 2.29 + 0.46 \times \sqrt{P_{tot}}$	0.0123	*	
	Summer	-		$log(CH_4) = 2.55 + 0.38 \times \sqrt{P_{tot}}$	0.0038	**	
	Fall	- CH ₄ (nmol L ⁻¹)		$log(CH_4) = 3.07 + -0.01 \times \sqrt{DIN}$	0.8728	ns	
F 1. 00	Winter	_		$log(CH_4) = 2.92 - 0.01 \times \sqrt{DIN}$	0.6101	ns	
Fig. S2	Spring	_		$log(CH_4) = 3.86 - 0.07 \times \sqrt{DIN}$	0.0080	**	
	Summer	_		$log(CH_4) = 3.43 + 0.01 \times \sqrt{DIN}$	0.9405	ns	
	Fall		DIN (µmol L ⁻)	$log(\%N_2O) = 1.78 + 0.05 \times \sqrt{DIN}$	0.0005	***	
	Winter	-		$log(\%N_2O) = 1.90 + 0.03 \times \sqrt{DIN}$	0.0000	****	
	Spring	– %N ₂ O (%)		$log((N_2 O)) = 1.86 + 0.06 \times \sqrt{DIN}$	0.0071	**	
	Summer	-		$log(\%N_2O) = 1.77 + 0.08 \times \sqrt{DIN}$	0.0317	*	
		DIN (µmol L ⁻¹)		$\sqrt{DIN} = 13.25 - 0.98 \times Distance$	0.0294	*	
		NH4 ⁺ (µmol L ⁻¹)		$\sqrt{(NH_4^+)} = 2.27 + 0.33 \times Distance$	0.3210	ns	
 –		NO3 ⁻ (µmol L ⁻¹)	Distance (km)	$\sqrt{NO_3^{-}} = 15.22 - 1.45 \times Distance$	0.0015	**	
Fig. 7	Average for	NO2 ⁻ (µmol L ⁻¹)		$\sqrt{NO_2} = 2.01 - 0.16 \times Distance$	0.0311	*	
	each pond	atm NO2 ⁻ (µg m ⁻³)		$atm NO_2^- = 37.56 - 3.35 \times Distance$	0.0003	***	
			%N ₂ O (%)	atm NO ₂ - (µg m-3)	$\log(\%N_20) = 1.66 + 0.04 \times atm NO_2^-$	0.0043	**
Fig. S4	_	Chl-a (µg L ⁻¹)	Distance (km)	$\sqrt{\text{Chla}} = 1.61 - 0.05 \times \text{Distance}$	0.3410	ns	

Table S7: Equation of quantiles regressions in listed figures and p-value (Wald test).