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# Supplementary Data 1. R code to analyse diversity-multifunctionality relationships in theoretical communities.
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#  
# Investigate the diversity-multifunctionality relationship with theoretical data. #  
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library(MASS)  
library(devEMF)
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setwd('fil-in-working-directory-path')
```

```
# richness_values  
no_communities <- 100  
richness_values <- rep(c(1:5),no_communities/5)  
average_correlation_value <- c(-0.07,0,0.25,0.5,0.75,1)  
regional_richness <- 20  
no_functions <- 15  
repetitions_artificial <- 100  
function_variability <- 1 # standard deviation EF values  
interaction_strength <- 0  
EF_mean <- 0
```

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filename <- paste("multifunctionality_artificial_data3_", ".emf", sep="")  
emf(file = filename,width=15,height=20)  
par(mfrow=c(3,2))
```

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feta_each <- c()  
for(a in 1:length(average_correlation_value)){  
  effect_size_all <- c()  
  lower_limit_all <- c()  
  upper_limit_all <- c()  
  p_value_all <- c()  
  feta_all <- c()  
  for(b in 1:repetitions_artificial){  
    # calculate ecosystem values  
    ecosystem_function <- matrix(ncol=no_functions,nrow=no_communities)  
    if(function_variability>0){  
      Sigma2=diag(no_functions)  
      Sigma2[which(diag(no_functions)==0)]<-average_correlation_value[a]  
      function_values_pool <- mvrnorm(n = regional_richness,  
mu=rep(EF_mean,no_functions),Sigma=Sigma2,tol=1e-6,empirical=FALSE,EISPACK=FALSE)  
      for(i in 1:no_communities){  
        function_values <-  
function_values_pool[sample(c(1:regional_richness),richness_values[i]),]  
        interaction_values <- interaction_strength*sample(abs(function_values),  
(richness_values[i]*richness_values[i]),replace=T)  
        if(richness_values[i]==1){interaction_values<-0}  
        if(richness_values[i]==1){  
          for(j in 1:no_functions){  
            ecosystem_function[i,j] <- function_values[j]  
          }  
        }else{  
          for(j in 1:no_functions){  
            ecosystem_function[i,j] <- mean(function_values[,j])+sum(interaction_values)  
          }  
        }  
      }  
    }else{  
      for(i in 1:no_communities){  
        for(j in 1:no_functions){  
          ecosystem_function[i,j] <- 0  
        }  
      }  
    }  
  }  
}
```

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    }
  }
}

ecosystem_function2 <- ecosystem_function
for(i in 1:no_communities){
  for(j in 1:no_functions){
    ecosystem_function2[i,j] <- (ecosystem_function[i,j] - min(ecosystem_function[,j]))
/
    (max(ecosystem_function[,j]) - min(ecosystem_function[,j]))
  }
}

ecosystem_function <- ecosystem_function2

teller <- var(rowSums(function_values_pool))
sd_values <- c()
for(i in 1:no_functions){
  sd_values[i] <- sd(function_values_pool[,i])
}
noemer <- (sum(sd_values))^2
feta_all[b] <- teller / noemer

# calculate multifunctionality
multifunctionality <- matrix(ncol=100,nrow=no_communities)
for(k in 1:100){
  for(i in 1:no_communities){
    performance <- c()
    for(j in 1:no_functions){
      threshold <- min(ecosystem_function[,j])+(k/100)*(max(ecosystem_function[,j])-
min(ecosystem_function[,j]))
      if(ecosystem_function[i,j]>threshold){performance[j]<-1}else{performance[j]<-0}
    }
    multifunctionality[i,k] <- sum(performance)
  }
}
multifunctionality <- data.frame(multifunctionality)
column_names <- c()
for(i in 1:100){
  column_names[i] <- paste("multifunctionality",i,sep="")
}
names(multifunctionality) <- column_names
row_names <- c()
for(i in 1:no_communities){
  row_names[i] <- paste("community_",i,sep="")
}
row.names(multifunctionality) <- row_names

multifunctionality_data <- cbind(richness_values,multifunctionality)

# test richness effect
p_value <- c()
effect_size <- c()
upper_limit <- c()
lower_limit <- c()
intercept <- c()
for(i in 1:99){
  j = i + 1
  multifunctionality_data2 <-
cbind(multifunctionality_data,"focal_multifunctionality"=multifunctionality_data[,j])
  m1 <- lm(focal_multifunctionality~richness_values,data=multifunctionality_data2)
  output <- summary(m1)
  lower_limit[i] <- confint(m1, 2, level = 0.95)[1]
  upper_limit[i] <- confint(m1, 2, level = 0.95)[2]
  p_value[i] <- output$coefficients[2,4]
}

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effect_size[i] <- output$coefficients[2,1]
intercept[i] <- output$coefficients[1,1]
}
effect_size_all <- cbind(effect_size_all,effect_size)
lower_limit_all <- cbind(lower_limit_all,lower_limit)
upper_limit_all <- cbind(upper_limit_all,upper_limit)
p_value_all <- cbind(p_value_all,p_value)
}
effect_size_mean <- c()
lower_limit_mean <- c()
upper_limit_mean <- c()
p_value_mean <- c()
for(i in 1:99){
  effect_size_mean[i] <- mean(effect_size_all[i,])
  lower_limit_mean[i] <- mean(lower_limit_all[i,])
  upper_limit_mean[i] <- mean(upper_limit_all[i,])
  p_value_mean[i] <- mean(p_value_all[i,])
}
print(a)
print(which(p_value_mean<0.05))
#effect_size_mean
print(which(effect_size_mean==max(effect_size_mean)))
print(max(effect_size_mean))
print(max(effect_size_mean)/(15/4))
print(which(effect_size_mean==min(effect_size_mean)))
print(min(effect_size_mean))
print(min(effect_size_mean)/(15/4))

feta_each[a] <- mean(feta_all)

plot(effect_size_mean~c(1:99),xlab="threshold value (%)",ylab="change in number of
functions per added species",
cex.axis=1.6,cex.lab=1.6,pch=19,xlim=c(1,100),ylim=c(-1.5,1.5))
lines(c(1:99), upper_limit_mean, col = 'grey')
lines(c(1:99), lower_limit_mean, col = 'grey')
polygon(c(c(1:99), rev(c(1:99))), c(upper_limit_mean, rev(lower_limit_mean)),col =
"grey", border = NA)
points(effect_size_mean~c(1:99),pch=19)
lines(x=c(-10,1000),y=c(0.0,0.0),col="black",lty=2)
}
dev.off()

feta_each

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