

# Script1

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## 1 Set working space

### 1.1 Load libraries

```
rm(list=ls())
library(vegan)
library(ggplot2)
library(dplyr)
library(readxl)
library(xlsx)
library(Hmisc)
library(ggcorrplot)
library(ggthemes)
```

```
library(corrplot)
library(vegan)
library(cowplot)
library(car)
library("pgirmess")
library(microbiome)
library(adespatial)
```

## 1.2 Load morphological and molecular of relative abundances dataset

```
Morpho=data.frame(read_xlsx("data\\Table S2.xlsx",sheet="Morphological",na=""))
head(Morpho)[,1:5]
```

```
## Station Location pH Achnanthidium.minutisimum Achnanthidium.caledonicum
## 1 BAI01 BAI 5.45 1.4 0
## 2 BAI04 BAI 4.95 0.0 0
## 3 BAI06 BAI 5.05 0.0 0
## 4 BAI07 BAI 5.10 8.8 0
## 5 BPU04 BPU 6.22 2.8 0
## 6 BSC03 BSC 7.55 6.4 0
```

```
COI=data.frame(read_xlsx("data\\Table S2.xlsx",sheet="COI",na=""))
head(COI)[,1:5]
```

```
## Station Location pH BOGC_000020776 BOGC_000020780
## 1 BAI01 BAI 5.45 0 0
## 2 BAI04 BAI 4.95 0 0
## 3 BAI06 BAI 5.05 0 0
## 4 BAI07 BAI 5.10 0 0
## 5 BPU04 BPU 6.22 0 0
## 6 BSC03 BSC 7.55 0 0
```

```
X18s=data.frame(read_xlsx("data\\Table S2.xlsx",sheet="18S rRNA",na=""))
head(X18s)[,1:5]
```

```
## Station Location pH BOGS_000001575 BOGS_000014580
## 1 BAI01 BAI 5.45 0 0.000000
## 2 BAI04 BAI 4.95 0 0.000000
## 3 BAI06 BAI 5.05 0 0.000000
## 4 BAI07 BAI 5.10 0 0.000000
## 5 BPU04 BPU 6.22 0 0.000000
## 6 BSC03 BSC 7.55 0 1.886792
```

```
#assigned rownames
rownames(Morpho)=(Morpho$Station)
rownames(COI)=(COI$Station)
rownames(X18s)=(X18s$Station)
```

```
#Select the numeric data
dato_morpho<-Morpho[4:386]
dato_coi<-COI[4:280] #277 otus
dato_18s<-X18s[4:169] #166 otus
```

```
#square root transformation of relative abundance data
datos.trans_morpho <- sqrt(dato_morpho)
```

```
datos.trans_coi <- sqrt(dato_coi)
datos.trans_18s<- sqrt(dato_18s)
```

## 2 Procrustes analysis

### 2.1 Distances calculation

```
dist.BC_morpho <- vegdist(datos.trans_morpho)
dist.BC_coi <- vegdist(datos.trans_coi)
dist.BC_18s <- vegdist(datos.trans_18s)

#distances
out.mds_morpho <- cmdscale(dist.BC_morpho,eig=T,k=2)
out.mds_coi <- cmdscale(dist.BC_coi,eig=T,k=2)
out.mds_18s <- cmdscale(dist.BC_18s ,eig=T,k=2)
```

### 2.2 Procrustes between morphological to molecular identification

```
#Residual calculation
#COI
pro_morpho_coi<- procrustes(X = out.mds_morpho,
                           Y = out.mds_coi, symmetric = TRUE,choices = c(1,2))
Res.coi=data.frame(resid=residuals(pro_morpho_coi),
                   Station=names(residuals(pro_morpho_coi)),Method="COI")

#18S
pro_morpho_18s<- procrustes(X = out.mds_morpho,
                           Y = out.mds_18s, symmetric = TRUE)
Res.18s=data.frame(resid=residuals(pro_morpho_18s),
                   Station=names(residuals(pro_morpho_18s)),Method="18S")

#Concatenated residual of procrustes dataframe
Res.all<-rbind(Res.coi,Res.18s)

#a merge is performed with the data matrix to add the location, station and pH.
Res.all=merge(Res.all,Morpho[,1:3],by="Station",sort = FALSE)

#Set method as a factor
Res.all$Method=as.factor(Res.all$Method)

#reorder factors
Res.all$Method=factor(Res.all$Method,c("COI","18S"))
head(Res.all)
```

```
##   Station      resid Method Location   pH
## 1  BAI01 0.02872557   COI      BAI 5.45
## 2  BAI01 0.05415680   18S      BAI 5.45
## 3  BAI04 0.03054378   18S      BAI 4.95
## 4  BAI04 0.05504238   COI      BAI 4.95
## 5  BAI06 0.02213755   18S      BAI 5.05
## 6  BAI06 0.09839896   COI      BAI 5.05
```

## 2.3 Create the residual figure

```
#Display panel figure 3C
Residuals=Res.all%>%
  ggplot(aes(Method,resid))+
  geom_boxplot(outlier.shape = NA)+
  geom_jitter(width = 0.2,shape=21,size=2)+labs(x=" ",y="Residuals",title = "")+
  ylim(0,0.18)+
  theme_bw()+ theme(strip.text.x = element_text(size=12),
                    strip.background = element_blank(),
                    panel.grid.minor = element_blank(),
                    panel.grid.major = element_blank(),
                    axis.text.x = element_text(size=12,color="black",
                                                  angle=0,vjust = 0.5),
                    axis.text.y = element_text(size=8,color="black",
                                                  angle=0,vjust = 0.5),
                    axis.title.x = element_text(color="black", size=10,
                                                  margin = margin(t = 5)),
                    axis.title.y = element_text(color="black", size=10),
                    text = element_text(family="serif"))

#Normality test
tapply(Res.all$resid,Res.all$Method,shapiro.test)

## $COI
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.93771, p-value = 0.002481
##
##
## $`18S`
##
##  Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.94212, p-value = 0.004018

#Statistical test
wilcox.test(Res.all$resid~Res.all$Method)

##
##  Wilcoxon rank sum test with continuity correction
##
## data:  Res.all$resid by Res.all$Method
## W = 2345, p-value = 0.4486
## alternative hypothesis: true location shift is not equal to 0
```

## 2.4 Procrustes calculation

```
#Morphological v/s COI procrustes calculation
pro_morpho_coi<- procrustes(X = out.mds_morpho,
                           Y = out.mds_coi, symmetric = TRUE,choices = c(1,2))

#stadistical test
```

```

protest(X = out.mds_morpho,
        Y = out.mds_coi, scores = "sites", permutations = 999)

##
## Call:
## protest(X = out.mds_morpho, Y = out.mds_coi, scores = "sites",      permutations = 999)
##
## Procrustes Sum of Squares (m12 squared):      0.3974
## Correlation in a symmetric Procrustes rotation: 0.7763
## Significance: 0.001
##
## Permutation: free
## Number of permutations: 999
#Get data from pro_morfo_coi
tmp.pro<-rbind(pro_morpho_coi[["Yrot"]],pro_morpho_coi[["X"]])

pro.df<-data.frame(X=tmp.pro[,1],
                   Y=tmp.pro[,2],
                   Method=rep(c("Molecular","Morphological"),each=66),
                   Station=row.names(tmp.pro))

pro.df$Method<-as.factor(pro.df$Method)
pro.df$Method<-factor(pro.df$Method, c("Molecular","Morphological"))

pro.df<-merge(pro.df,COI[,1:3],by="Station",sort = FALSE)

#Display panel figure 3A
pro_morpho_coi=pro.df%>%
  ggplot(aes(X,Y,color=pH,shape=Method))+
  geom_line(aes(group=Station),color="black")+
  geom_point(size=4)+
  theme_bw()+labs(x="Dimension 1",y="Dimension 2",title = "Morphological vs. COI")+
  theme(strip.text.x = element_text(size=12),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        plot.title = element_text(size=12,hjust =0.5),
        axis.title.x =element_text(size=10,color="black"),
        axis.title.y= element_text(size=10,color="black"),
        axis.text.x = element_text(size=8,color="black",angle=0,vjust = 0.5),
        axis.text.y = element_text(size=8,color="black",angle=0,vjust = 0.5),
        text = element_text(family="serif"))+
  scale_color_gradientn(colours = c("#F27F0C","yellow","#7ed348","#26B170","#01377d"))+
  theme(legend.position = "none")

#Morphological v/s 18S procrustes calculation
pro_morpho_18s<- procrustes(X = out.mds_morpho, Y = out.mds_18s, symmetric = TRUE)
#Statistical test
protest(X = out.mds_morpho, Y = out.mds_18s, scores = "sites", permutations = 999)

##
## Call:
## protest(X = out.mds_morpho, Y = out.mds_18s, scores = "sites",      permutations = 999)
##
## Procrustes Sum of Squares (m12 squared):      0.317
## Correlation in a symmetric Procrustes rotation: 0.8264

```

```

## Significance: 0.001
##
## Permutation: free
## Number of permutations: 999

#Get data from pro_morfo_18s
tmp.pro_2<-rbind(pro_morfo_18s[["Yrot"]],pro_morfo_18s[["X"]])

pro.df_2<-data.frame(X=tmp.pro_2[,1],
                    Y=tmp.pro_2[,2],
                    Method=rep(c("Molecular","Morphological"),each=66),
                    Station=row.names(tmp.pro_2))

pro.df_2$Method<-as.factor(pro.df_2$Method)
pro.df_2$Method<-factor(pro.df_2$Method, c("Molecular","Morphological"))

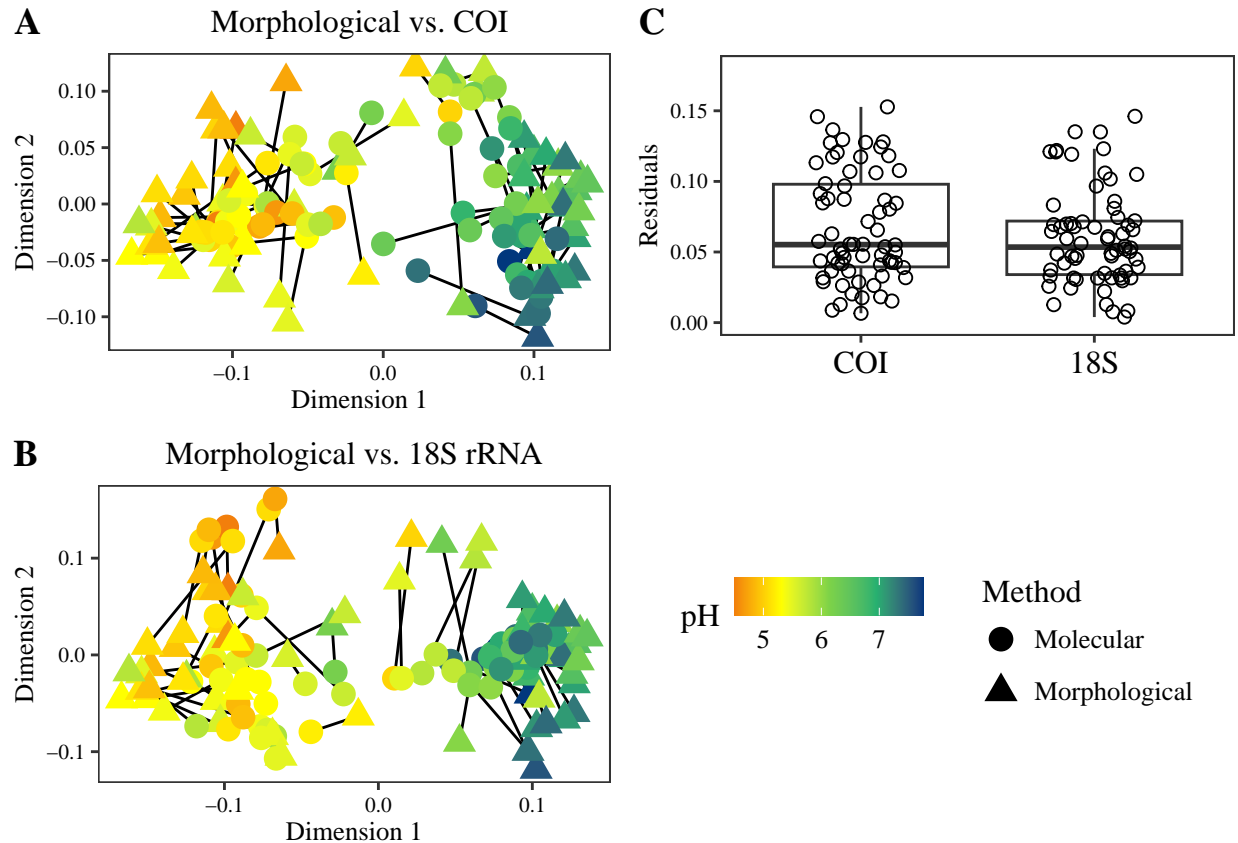
pro.df_2<-merge(pro.df_2,X18s[,1:3],by="Station",sort = FALSE)

#Display panel figure 3B
pro_morfo_18s=pro.df_2%>%
  ggplot(aes(X,Y,color=pH,shape=Method))+
  geom_line(aes(group=Station),color="black")+
  geom_point(size=4)+
  theme_bw()+labs(x="Dimension 1",y="Dimension 2",title = "Morphological vs. 18S rRNA")+
  theme(strip.text.x = element_text(size=12),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        plot.title = element_text(size=12,hjust = 0.5),
        axis.title.x =element_text(size=10,color="black"),
        axis.title.y= element_text(size=10,color="black"),
        axis.text.x = element_text(size=8,color="black",angle=0,vjust = 0.5),
        axis.text.y = element_text(size=8,color="black",angle=0,vjust = 0.5),
        text = element_text(family="serif"))+
  scale_color_gradientn(colours = c("#F27F0C","yellow","#7ed348","#26B170","#01377d"))+
  theme(legend.position = "bottom",legend.key.size = unit(0.5, 'cm'),
        legend.text = element_text(size=10),
        legend.title = element_text(size = 12))+
  guides(shape = guide_legend(title.position = "top",ncol=1))
legenda_18s_pro=get_plot_component(pro_morfo_18s, 'guide-box-bottom', return_all = TRUE)
pro_morfo_18s=pro_morfo_18s+theme(legend.position = "none")

```

## 2.5 Display Figure 3

Figure 3: Comparison of MDS through Procrustes analysis between morphological vs. COI (A) and morphological vs. 18S (B). Residual values from comparing morphological identification and the two molecular identifications (18S and COI) (C).



### 3 Lineal model

#### 3.1 Load metadata, alpha diversity calculation and Statistical analysis

```
#Load metadata
metadata=data.frame(read_xlsx("data\\Table S1.xlsx",sheet="Metadata",na=""))

#alpha diversity calculation
metadata$H_morpho=vegan::diversity(t(t(Morpho[,4:386])))
metadata$S_morpho=vegan::specnumber(t(t(Morpho[,4:386])))
metadata$J_morpho=evenness((t(Morpho[,4:386])), 'pielou')[,1]

metadata$H_COI=vegan::diversity(t(t(COI[,4:280])))
metadata$S_COI=vegan::specnumber(t(t(COI[,4:280])))
metadata$J_COI=evenness((t(COI[,4:280])), 'pielou')[,1]

metadata$H_18s=vegan::diversity(t(t(X18s[,4:169])))
metadata$S_18s=vegan::specnumber(t(t(X18s[,4:169])))
metadata$J_18s=evenness((t(X18s[,4:169])), 'pielou')[,1]
head(metadata)[,1:5]
```

```
## Station Location WTD GDD Rad
## 1 BAI01 BAI -1.0 1532 1835
## 2 BAI04 BAI -0.5 1410 1745
## 3 BAI06 BAI -13.0 1532 1835
```

```

## 4   BAI07      BAI   -1.5 1593 1742
## 5   BPU04      BPU   -1.0 2903 1812
## 6   BSC03      BSC   -2.0 2906 1946

morpho_index=metadata[,1:2]
morpho_index$Type<-"Morpho"
morpho_index$Richness=metadata[,25]
morpho_index$Shannon=metadata[,24]
morpho_index$Pielou=metadata[,26]

coi_index=metadata[,1:2]
coi_index$Type<-"COI"
coi_index$Richness=metadata[,28]
coi_index$Shannon=metadata[,27]
coi_index$Pielou=metadata[,29]

x18s_index=metadata[,1:2]
x18s_index$Type<-"18S"
x18s_index$Richness=metadata[,31]
x18s_index$Shannon=metadata[,30]
x18s_index$Pielou=metadata[,32]

analisis=rbind(morpho_index,coi_index,x18s_index)
head(analisis)[,1:5]

##   Station Location   Type Richness  Shannon
## 1   BAI01      BAI Morpho      68 3.458610
## 2   BAI04      BAI Morpho      40 2.797134
## 3   BAI06      BAI Morpho      43 3.062109
## 4   BAI07      BAI Morpho      61 3.528707
## 5   BPU04      BPU Morpho      61 3.346099
## 6   BSC03      BSC Morpho      35 2.823989

#Richness
#Normality test
tapply(analisis$Richness,analisis$Type,shapiro.test)

## $`18S`
##
##   Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.97336, p-value = 0.1664
##
##
## $COI
##
##   Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.9613, p-value = 0.03773
##
##
## $Morpho
##

```



```

## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.9878, p-value = 0.7653
#homoscedasticity test
bartlett.test(analysis$Richness~analysis$Type)

##
## Bartlett test of homogeneity of variances
##
## data:  analysis$Richness by analysis$Type
## Bartlett's K-squared = 46.003, df = 2, p-value = 1.024e-10
#kruskal Wallis test calculation
kruskal_S=kruskal.test(analysis$Richness~analysis$Type)
kruskal_S

##
## Kruskal-Wallis rank sum test
##
## data:  analysis$Richness by analysis$Type
## Kruskal-Wallis chi-squared = 131.8, df = 2, p-value < 2.2e-16
#Post test to evaluate where are the significant differences
kruskalmc(analysis$Richness~analysis$Type,alpha=0.001)

## Multiple comparison test after Kruskal-Wallis
## alpha: 0.001
## Comparisons
##      obs.dif critical.dif stat.signif
## 18S-COI      37.28030      35.78934      TRUE
## 18S-Morpho 112.35606      35.78934      TRUE
## COI-Morpho  75.07576      35.78934      TRUE
#Shannon index
#Normality test
tapply(analysis$Shannon,analysis$Type,shapiro.test)

## $`18S`
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.94625, p-value = 0.006386
##
##
## $COI
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.95526, p-value = 0.0182
##
##
## $Morpho
##

```

```

## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.94707, p-value = 0.007016
#homoscedasticity test
bartlett.test(analysis$Shannon~analysis$Type)

##
## Bartlett test of homogeneity of variances
##
## data:  analysis$Shannon by analysis$Type
## Bartlett's K-squared = 18.125, df = 2, p-value = 0.000116
#kruskal Wallis test calculation
kruskal_H=kruskal.test(analysis$Shannon~analysis$Type)
kruskal_H

##
## Kruskal-Wallis rank sum test
##
## data:  analysis$Shannon by analysis$Type
## Kruskal-Wallis chi-squared = 75.169, df = 2, p-value < 2.2e-16
#Post test to evaluate where are the significant differences
kruskalmc(analysis$Shannon~analysis$Type,alpha=0.01)

## Multiple comparison test after Kruskal-Wallis
## alpha: 0.01
## Comparisons
##      obs.dif critical.dif stat.signif
## 18S-COI    30.75758      29.27852      TRUE
## 18S-Morpho  85.37879      29.27852      TRUE
## COI-Morpho  54.62121      29.27852      TRUE
#Pielou evenness
#Normality test
tapply(analysis$Pielou,analysis$Type,shapiro.test)

## $`18S`
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.86626, p-value = 3.992e-06
##
##
## $COI
##
## Shapiro-Wilk normality test
##
## data:  X[[i]]
## W = 0.83768, p-value = 5.165e-07
##
##
## $Morpho
##

```

```
## Shapiro-Wilk normality test
##
## data: X[[i]]
## W = 0.89811, p-value = 5.31e-05

#homoscedasticity test
bartlett.test(analisis$Pielou~analisis$Type)

##
## Bartlett test of homogeneity of variances
##
## data: analisis$Pielou by analisis$Type
## Bartlett's K-squared = 34.65, df = 2, p-value = 2.991e-08

#kruskal Wallis test calculation
kruskal_J=kruskal.test(analisis$Pielou~analisis$Type)
kruskal_J

##
## Kruskal-Wallis rank sum test
##
## data: analisis$Pielou by analisis$Type
## Kruskal-Wallis chi-squared = 1.2121, df = 2, p-value = 0.5455
```

### 3.2 To make a new data frame with pH mean, minimum and max for location and alpha index

```
ind_ph=aggregate(pH~Location,data=metadata,mean)
ind_ph$pH.sd=aggregate(pH~Location,data=metadata,sd)[,2]
ind_ph$S_morpho=aggregate(S_morpho~Location,data=metadata,mean)[,2]
ind_ph$S_morpho.sd=aggregate(S_morpho~Location,data=metadata,sd)[,2]
ind_ph$S_COI=aggregate(S_COI~Location,data=metadata,mean)[,2]
ind_ph$S_COI.sd=aggregate(S_COI~Location,data=metadata,sd)[,2]
ind_ph$S_18S=aggregate(S_18s~Location,data=metadata,mean)[,2]
ind_ph$S_18S.sd=aggregate(S_18s~Location,data=metadata,sd)[,2]

ind_ph$H_morpho=aggregate(H_morpho~Location,data=metadata,mean)[,2]
ind_ph$H_morpho.sd=aggregate(H_morpho~Location,data=metadata,sd)[,2]
ind_ph$H_COI=aggregate(H_COI~Location,data=metadata,mean)[,2]
ind_ph$H_COI.sd=aggregate(H_COI~Location,data=metadata,sd)[,2]
ind_ph$H_18S=aggregate(H_18s~Location,data=metadata,mean)[,2]
ind_ph$H_18S.sd=aggregate(H_18s~Location,data=metadata,sd)[,2]

ind_ph$J_morpho=aggregate(J_morpho~Location,data=metadata,mean)[,2]
ind_ph$J_morpho.sd=aggregate(J_morpho~Location,data=metadata,sd)[,2]
ind_ph$J_COI=aggregate(J_COI~Location,data=metadata,mean)[,2]
ind_ph$J_COI.sd=aggregate(J_COI~Location,data=metadata,sd)[,2]
ind_ph$J_18S=aggregate(J_18s~Location,data=metadata,mean)[,2]
ind_ph$J_18S.sd=aggregate(J_18s~Location,data=metadata,sd)[,2]
head(ind_ph)

## Location      pH      pH.sd S_morpho S_morpho.sd      S_COI S_COI.sd
## 1 BAI 5.137500 0.21746647 53.00000 13.638182 16.75000 5.852350
## 2 BPU 6.220000 NA 61.00000 NA 18.00000 NA
## 3 BSC 7.556667 0.20008332 36.66667 10.598742 15.66667 11.930353
```

```
## 4      BV 5.360000 0.05656854 33.00000      8.485281 16.00000 4.242641
## 5     BValp 5.416667 0.50649120 41.00000      3.464102 12.33333 5.131601
## 6      CBR 5.425000 0.54447222 59.50000      4.949747 27.50000 3.535534
##      S_18S S_18S.sd H_morpho H_morpho.sd      H_COI H_COI.sd      H_18S H_18S.sd
## 1 11.000000 4.690416 3.211640      0.3443353 2.165513 0.3376578 1.434446 0.8080834
## 2 6.000000      NA 3.346099      NA 2.435441      NA 1.438600      NA
## 3 14.666667 2.081666 2.781214      0.5113690 1.411986 1.3421487 2.035147 0.1318731
## 4 12.000000 4.242641 2.776004      0.3641266 2.411507 0.3189316 1.712559 0.3619781
## 5 9.666667 4.041452 2.961465      0.1834044 1.643052 0.6098899 1.662429 0.4842821
## 6 12.500000 3.535534 3.256676      0.1061851 2.499575 0.2564536 2.064514 0.4135034
##      J_morpho J_morpho.sd      J_COI J_COI.sd      J_18S J_18S.sd
## 1 0.8126118 0.041235912 0.7806790 0.03414478 0.5893503 0.23983976
## 2 0.8139630      NA 0.8426046      NA 0.8028979      NA
## 3 0.7738577 0.082931258 0.4807212 0.35589246 0.7597906 0.03626807
## 4 0.7960840 0.045155725 0.8739245 0.03064695 0.6949203 0.04524402
## 5 0.7975268 0.032274672 0.6539655 0.14926927 0.7424420 0.08198796
## 6 0.7972762 0.009740795 0.7542136 0.04810213 0.8199782 0.07121246
```

### 3.3 Models and figures

#### 3.3.1 Richness

```
#Morphological
lineal.poly=lm(S_morpho~poly(pH,2),data=ind_ph)
summary(lineal.poly)

##
## Call:
## lm(formula = S_morpho ~ poly(pH, 2), data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -19.475  -4.691  -2.249   5.421  15.993
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    42.163      1.667   25.298 <2e-16 ***
## poly(pH, 2)1    -2.615      8.498   -0.308  0.7611
## poly(pH, 2)2   -21.889      8.498   -2.576  0.0169 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.498 on 23 degrees of freedom
## Multiple R-squared:  0.2263, Adjusted R-squared:  0.1591
## F-statistic: 3.365 on 2 and 23 DF,  p-value: 0.05227

lineal.lm=lm(S_morpho~pH,data=ind_ph)
summary(lineal.lm)

##
## Call:
## lm(formula = S_morpho ~ pH, data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
```

```
## -21.493 -6.510 -1.288 7.270 18.946
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)  46.1071    14.3602   3.211  0.00374 **
## pH          -0.6517     2.3531  -0.277  0.78420
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.443 on 24 degrees of freedom
## Multiple R-squared:  0.003185, Adjusted R-squared:  -0.03835
## F-statistic: 0.07669 on 1 and 24 DF, p-value: 0.7842
```

```
AIC(lineal.poly,lineal.lm)
```

```
##              df      AIC
## lineal.poly  4 189.8707
## lineal.lm    3 194.4601
```

```
#Display panel figure 4A
```

```
S_morpho=ind_ph%>%
  ggplot(aes(pH,S_morpho))+
  geom_point() +
  geom_errorbar(aes(ymin = S_morpho-S_morpho.sd,ymax = S_morpho+S_morpho.sd )) +
  geom_errorbarh(aes(xmin = pH-pH.sd,xmax = pH+pH.sd))+
  theme_bw()+
  theme(strip.text.x = element_text(size=10),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.x =element_blank(),axis.title.x= element_blank(),
        axis.text.y = element_text(size=10, color="black"),
        axis.title = element_text(color="black",size=12, face="bold"),
        text = element_text(family="serif"))+
  labs(y="Richness",x="pH", title = "      Morphological") +xlim(4,8)+ylim(0,70)+
  geom_smooth(se=FALSE,method="lm", formula=y~poly(x,2))+
  annotate("text",x=5.8,y=11,
         label=bquote("y"~.(round(summary(lineal.poly)[[4]][3,1],3))~
                       "x"~2~"+".(round(summary(lineal.poly)[[4]][2,1],3))~
                       "x"~"+".(round(summary(lineal.poly)[[4]][1,1],3))),
         cex = 2.5,col="black")+
  annotate("text",x=5.8,y=5,
         label=bquote("R"~2~"="~.(round(summary(lineal.poly)[[8]],3))),
         cex = 2.5,col="black")
```

```
#COI
```

```
lineal.poly=lm(S_COI~poly(pH,2),data=ind_ph)
summary(lineal.poly)
```

```
##
## Call:
## lm(formula = S_COI ~ poly(pH, 2), data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.6806  -4.2949   0.0523   2.8785  12.6813
##
## Coefficients:
```

```
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)    20.469      1.200  17.063 1.49e-14 ***
## poly(pH, 2)1     9.995      6.117   1.634  0.1159
## poly(pH, 2)2   -21.579      6.117  -3.528  0.0018 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.117 on 23 degrees of freedom
## Multiple R-squared:  0.3966, Adjusted R-squared:  0.3441
## F-statistic: 7.557 on 2 and 23 DF,  p-value: 0.003002
```

```
lineal.lm=lm(S_COI~pH,data=ind_ph)
summary(lineal.lm)
```

```
##
## Call:
## lm(formula = S_COI ~ pH, data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -11.203  -5.627  -1.670   4.369  14.941
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)    5.397      11.304   0.477   0.637
## pH             2.490       1.852   1.344   0.191
##
## Residual standard error: 7.434 on 24 degrees of freedom
## Multiple R-squared:  0.07004,    Adjusted R-squared:  0.0313
## F-statistic: 1.808 on 1 and 24 DF,  p-value: 0.1914
```

```
AIC(lineal.poly,lineal.lm)
```

```
##           df          AIC
## lineal.poly  4 172.7735
## lineal.lm    3 182.0179
```

```
#Display panel figure 4B
```

```
S_COI=ind_ph%>%
  ggplot(aes(pH,S_COI))+
  geom_point() +
  geom_errorbar(aes(ymin = S_COI-S_COI.sd,ymax = S_COI+S_COI.sd )) +
  geom_errorbarh(aes(xmin = pH-pH.sd,xmax = pH+pH.sd))+
  theme_bw()+
  theme(strip.text.x = element_text(size=10),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.x =element_blank(),axis.title.x= element_blank(),
        axis.text.y =element_blank(),axis.title.y= element_blank(),
        text = element_text(family="serif"))+
  labs(y="",x="pH", title = "          COI") +xlim(4,8)+ylim(0,70)+
  geom_smooth(se=FALSE,method="lm", formula=y~poly(x,2))+
  annotate("text",x=5.6,y=60,
        label=bquote("y=~.(round(summary(lineal.poly)[[4]][3,1],3))~
                      "x"^2~"+"~.(round(summary(lineal.poly)[[4]][2,1],3))~
                      "x"~"+"~.(round(summary(lineal.poly)[[4]][1,1],3))),
        cex = 2.5,col="black")+
  
```

```

    annotate("text",x=5.6,y=55,
            label=bquote("R"2~"="~.(round(summary(lineal.poly)[[8]],3))~"*"),
            cex = 2.5,col="black")

#18S rRNA
lineal.poly=lm(S_18S~poly(pH,2),data=ind_ph)
summary(lineal.poly)

##
## Call:
## lm(formula = S_18S ~ poly(pH, 2), data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8.6486 -1.6226 -0.0661  0.6997  7.5010
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   12.9353     0.6258  20.671 2.35e-16 ***
## poly(pH, 2)1    7.4697     3.1909   2.341  0.0283 *
## poly(pH, 2)2   -6.7718     3.1909  -2.122  0.0448 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.191 on 23 degrees of freedom
## Multiple R-squared:  0.3027, Adjusted R-squared:  0.2421
## F-statistic: 4.992 on 2 and 23 DF,  p-value: 0.01583

lineal.lm=lm(S_18S~pH,data=ind_ph)
summary(lineal.lm)

##
## Call:
## lm(formula = S_18S ~ pH, data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.2485 -1.9045 -0.7976  1.0671  8.6077
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    1.6713     5.1944   0.322  0.7504
## pH              1.8613     0.8512   2.187  0.0387 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.416 on 24 degrees of freedom
## Multiple R-squared:  0.1661, Adjusted R-squared:  0.1314
## F-statistic: 4.782 on 1 and 24 DF,  p-value: 0.03874

AIC(lineal.poly,lineal.lm)

##           df      AIC
## lineal.poly  4 138.9323
## lineal.lm    3 141.5820

```

```

#Display panel figure 4C
S_18s=ind_ph%>%
  ggplot(aes(pH,S_18S))+
  geom_point() +
  geom_errorbar(aes(ymin = S_18S-S_18S.sd,ymax = S_18S+S_18S.sd )) +
  geom_errorbarh(aes(xmin = pH-pH.sd,xmax = pH+pH.sd))+
  theme_bw()+
  theme(strip.text.x = element_text(size=10),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.x =element_blank(),axis.title.x= element_blank(),
        axis.text.y =element_blank(),axis.title.y= element_blank(),
        text = element_text(family="serif"))+
  labs(y="",x="pH", title = "      18S rRNA") +xlim(4,8)+ylim(0,70)+
  geom_smooth(se=FALSE,method="lm", formula=y~poly(x,2))+
  annotate("text",x=5.5,y=60,
        label=bquote("y="~.(round(summary(lineal.poly)[[4]][3,1],3))~
          "x"^2~"+"~.(round(summary(lineal.poly)[[4]][2,1],3))~
          "x"~"+"~.(round(summary(lineal.poly)[[4]][1,1],3))),
        cex = 2.5,col="black")+
  annotate("text",x=5.5,y=55,
        label=bquote("R"^2~"="~.(round(summary(lineal.poly)[[8]],3))~"*"),
        cex =2.5,col="black")

```

### 3.3.2 Shannon index

```

#Morphological
lineal.poly=lm(H_morpho~poly(pH,2),data=ind_ph)
summary(lineal.poly)

##
## Call:
## lm(formula = H_morpho ~ poly(pH, 2), data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.32312 -0.18990  0.01622  0.19568  0.52735
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   2.76003    0.07813  35.328  <2e-16 ***
## poly(pH, 2)1  -0.38742    0.39836  -0.973   0.341
## poly(pH, 2)2  -0.76594    0.39836  -1.923   0.067 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3984 on 23 degrees of freedom
## Multiple R-squared:  0.168, Adjusted R-squared:  0.0956
## F-statistic: 2.321 on 2 and 23 DF, p-value: 0.1207

lineal.lm=lm(H_morpho~pH,data=ind_ph)
summary(lineal.lm)

##
## Call:

```



```

## lm(formula = H_morpho ~ pH, data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.39375 -0.13488  0.07533  0.25067  0.60231
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  3.34424    0.63890   5.234 2.3e-05 ***
## pH          -0.09654    0.10469  -0.922  0.366
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4201 on 24 degrees of freedom
## Multiple R-squared:  0.03422,    Adjusted R-squared:  -0.006025
## F-statistic: 0.8503 on 1 and 24 DF,  p-value: 0.3657
AIC(lineal.poly,lineal.lm)

##              df      AIC
## lineal.poly   4 30.73655
## lineal.lm     3 32.61192
#Display panel figure 4D
H_morpho=ind_ph%>%
  ggplot(aes(pH,H_morpho))+
  geom_point() +
  geom_errorbar(aes(ymin = H_morpho-H_morpho.sd,ymax = H_morpho+H_morpho.sd )) +
  geom_errorbarh(aes(xmin = pH-pH.sd,xmax = pH+pH.sd))+
  theme_bw()+
  theme(strip.text.x = element_text(size=10),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.x =element_blank(),axis.title.x= element_blank(),
        axis.text.y = element_text(size=10, color="black"),
        axis.title = element_text(color="black",size=12, face="bold"),
        text = element_text(family="serif"))+
  labs(y="Shannon",x="pH", title = "") +xlim(4,8)+ylim(0,4)+
  geom_smooth(se=FALSE,method="lm", formula=y~poly(x,2))+
  annotate("text",x=5.6,y=1.0,
         label=bquote("y=~.(round(summary(lineal.poly)[[4]][3,1],3))~
                        "x"^2~"+"~.(round(summary(lineal.poly)[[4]][2,1],3))~
                        "x"~"+"~.(round(summary(lineal.poly)[[4]][1,1],3))),
         cex =2.5,col="black")+
  annotate("text",x=5.6,y=0.7,
         label=bquote("R"^2~"="~.(round(summary(lineal.poly)[[8]],3))),
         cex =2.5,col="black")

#COI
lineal.poly=lm(H_COI~poly(pH,2),data=ind_ph)
summary(lineal.poly)

##
## Call:
## lm(formula = H_COI ~ poly(pH, 2), data = ind_ph)
##
## Residuals:

```

```
##      Min      1Q   Median      3Q      Max
## -0.80267 -0.24028  0.04111  0.25047  0.54609
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   2.24431    0.07482  29.996 < 2e-16 ***
## poly(pH, 2)1  0.41430    0.38151   1.086   0.289
## poly(pH, 2)2 -2.03663    0.38151  -5.338 2.02e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3815 on 23 degrees of freedom
## Multiple R-squared:  0.5634, Adjusted R-squared:  0.5254
## F-statistic: 14.84 on 2 and 23 DF,  p-value: 7.264e-05

lineal.lm=lm(H_COI~pH,data=ind_ph)
summary(lineal.lm)
```

```
##
## Call:
## lm(formula = H_COI ~ pH, data = ind_ph)
##
## Residuals:
##      Min      1Q   Median      3Q      Max
## -1.1302 -0.4353  0.1408  0.4006  0.8761
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   1.6196    0.8498   1.906  0.0687 .
## pH            0.1032    0.1393   0.741  0.4657
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5588 on 24 degrees of freedom
## Multiple R-squared:  0.02239, Adjusted R-squared: -0.01835
## F-statistic: 0.5496 on 1 and 24 DF,  p-value: 0.4657
```

```
AIC(lineal.poly,lineal.lm)
```

```
##              df      AIC
## lineal.poly   4 28.48856
## lineal.lm     3 47.44596
```

```
#Display panel figure 4E
```

```
H_COI=ind_ph%>%
  ggplot(aes(pH,H_COI))+
  geom_point() +
  geom_errorbar(aes(ymin = H_COI-H_COI.sd,ymax = H_COI+H_COI.sd )) +
  geom_errorbarh(aes(xmin = pH-pH.sd,xmax = pH+pH.sd))+
  theme_bw()+
  theme(strip.text.x = element_text(size=10),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.x=element_blank(),axis.title.x= element_blank(),
        axis.text.y=element_blank(),axis.title.y= element_blank(),
        text = element_text(family="serif"))+
  labs(y="",x="pH", title = "") +xlim(4,8)+ylim(0,4)+
```

```
geom_smooth(se=FALSE,method="lm", formula=y~poly(x,2))+
annotate("text",x=5.5,y=3.9,
        label=bquote("y="~.(round(summary(lineal.poly)[[4]][3,1],3))~
          "x"^2~"+"~.(round(summary(lineal.poly)[[4]][2,1],3))~
          "x"~"+"~.(round(summary(lineal.poly)[[4]][1,1],3))),
        cex =2.5,col="black")+
annotate("text",x=5.5,y=3.6,
        label=bquote("R"^2~"="~.(round(summary(lineal.poly)[[8]],3))~"***"),
        cex =2.5,col="black")
```

*#18S rRNA*

```
lineal.poly=lm(H_18S~poly(pH,2),data=ind_ph)
summary(lineal.poly)
```

```
##
## Call:
## lm(formula = H_18S ~ poly(pH, 2), data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.74800 -0.18582  0.00243  0.22456  0.81629
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    1.9150     0.0714  26.821 < 2e-16 ***
## poly(pH, 2)1    1.0767     0.3641   2.957  0.00706 **
## poly(pH, 2)2   -1.0954     0.3641  -3.009  0.00626 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3641 on 23 degrees of freedom
## Multiple R-squared:  0.4363, Adjusted R-squared:  0.3872
## F-statistic: 8.899 on 2 and 23 DF,  p-value: 0.001372
lineal.lm=lm(H_18S~pH,data=ind_ph)
summary(lineal.lm)
```

```
##
## Call:
## lm(formula = H_18S ~ pH, data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.77370 -0.25286 -0.07173  0.26940  0.99530
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.2914     0.6398   0.455  0.6529
## pH             0.2683     0.1048   2.559  0.0172 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4207 on 24 degrees of freedom
## Multiple R-squared:  0.2144, Adjusted R-squared:  0.1816
## F-statistic: 6.549 on 1 and 24 DF,  p-value: 0.01722
```

```
AIC(lineal.poly,lineal.lm)
```

```
##           df      AIC
## lineal.poly  4 26.05427
## lineal.lm    3 32.68302
```

```
#Display panel figure 4F
```

```
H_18s=ind_ph%>%
  ggplot(aes(pH,H_18S))+
  geom_point() +
  geom_errorbar(aes(ymin = H_18S-H_18S.sd,ymax = H_18S+H_18S.sd )) +
  geom_errorbarh(aes(xmin = pH-pH.sd,xmax = pH+pH.sd))+
  theme_bw()+
  theme(strip.text.x = element_text(size=10),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        #axis.text.x = element_text(size=10, color="black",hjust=1),
        #axis.title.x = element_text(color="black",size=12,face="bold"),
        axis.text.x =element_blank(),axis.title.x= element_blank(),
        axis.text.y =element_blank(),axis.title.y= element_blank(),
        text = element_text(family="serif"))+
  labs(y="",x="pH", title = "") +xlim(4,8)+ylim(0,4)+
  geom_smooth(se=FALSE,method="lm", formula=y~poly(x,2))+
  annotate("text",x=5.5,y=3.8,
         label=bquote("y"~.(round(summary(lineal.poly)[[4]][3,1],3))~
                       "x"^2~"+"~.(round(summary(lineal.poly)[[4]][2,1],3))~
                       "x"~"+"~.(round(summary(lineal.poly)[[4]][1,1],3))),
         cex =2.5,col="black")+
  annotate("text",x=5.5,y=3.5,
         label=bquote("R"^2~"="~.(round(summary(lineal.poly)[[8]],3))~"*"),
         cex =2.5,col="black")
```

### 3.3.3 Pielou evenness

```
#Morphological
```

```
lineal.poly=lm(J_morpho~poly(pH,2),data=ind_ph)
summary(lineal.poly)
```

```
##
## Call:
## lm(formula = J_morpho ~ poly(pH, 2), data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.28127 -0.03504  0.01220  0.05196  0.12054
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.74030    0.01658  44.649  <2e-16 ***
## poly(pH, 2)1 -0.10425    0.08454  -1.233    0.230
## poly(pH, 2)2 -0.10164    0.08454  -1.202    0.241
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.08454 on 23 degrees of freedom
```

```
## Multiple R-squared:  0.1142, Adjusted R-squared:  0.03721
## F-statistic: 1.483 on 2 and 23 DF,  p-value: 0.2479

lineal.lm=lm(J_morpho~pH,data=ind_ph)
summary(lineal.lm)

##
## Call:
## lm(formula = J_morpho ~ pH, data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.29065 -0.02830  0.02678  0.04397  0.11547
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.89751    0.12975   6.917 3.74e-07 ***
## pH          -0.02598    0.02126  -1.222   0.234
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.08532 on 24 degrees of freedom
## Multiple R-squared:  0.05856,    Adjusted R-squared:  0.01934
## F-statistic: 1.493 on 1 and 24 DF,  p-value: 0.2336

AIC(lineal.poly,lineal.lm)

##              df      AIC
## lineal.poly   4 -49.8686
## lineal.lm     3 -50.2839

#Display panel figure 4G
J_morpho=ind_ph%>%
  ggplot(aes(pH,J_morpho))+
  geom_point() +
  geom_errorbar(aes(ymin = J_morpho-J_morpho.sd,ymax = J_morpho+J_morpho.sd )) +
  geom_errorbarh(aes(xmin = pH-pH.sd,xmax = pH+pH.sd))+
  theme_bw()+
  theme(strip.text.x = element_text(size=10),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.x = element_text(size=10, color="black",hjust=1),
        axis.title.x = element_text(color="black",size=12,face="bold"),
        axis.title = element_text(color="black",size=12, face="bold"),
        text = element_text(family="serif"))+
  labs(y="Pielou evenness",x="pH", title = "") +xlim(4,8)+ylim(0,1)+
  geom_smooth(se=FALSE,method="lm", formula=y~poly(x,2))+
  annotate("text",x=5.6,y=0.32,
        label=bquote("y=~.(round(summary(lineal.poly)[[4]][3,1],3))~
                      "x"^2~"+~.(round(summary(lineal.poly)[[4]][2,1],3))~
                      "x"~"+~.(round(summary(lineal.poly)[[4]][1,1],3))),
        cex =2.5,col="black")+
  annotate("text",x=5.6,y=0.25,
        label=bquote("R"^2~"=~.(round(summary(lineal.poly)[[8]],3))),
        cex =2.5,col="black")
```

```

#COI
lineal.poly=lm(J_COI~poly(pH,2),data=ind_ph)
summary(lineal.poly)

##
## Call:
## lm(formula = J_COI ~ poly(pH, 2), data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.127062 -0.057129  0.004795  0.046659  0.169382
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.75152    0.01561  48.155 < 2e-16 ***
## poly(pH, 2)1 -0.01611    0.07958  -0.202   0.841
## poly(pH, 2)2 -0.43867    0.07958  -5.513 1.32e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.07958 on 23 degrees of freedom
## Multiple R-squared:  0.5695, Adjusted R-squared:  0.5321
## F-statistic: 15.21 on 2 and 23 DF,  p-value: 6.171e-05

lineal.lm=lm(J_COI~pH,data=ind_ph)
summary(lineal.lm)

##
## Call:
## lm(formula = J_COI ~ pH, data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.26476 -0.07588  0.02782  0.09231  0.13270
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.775819    0.180484   4.299 0.000247 ***
## pH            -0.004015    0.029575  -0.136 0.893148
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1187 on 24 degrees of freedom
## Multiple R-squared:  0.0007673, Adjusted R-squared:  -0.04087
## F-statistic: 0.01843 on 1 and 24 DF,  p-value: 0.8931

AIC(lineal.poly,lineal.lm)

##              df      AIC
## lineal.poly   4 -53.01649
## lineal.lm     3 -33.12181

#Display panel figure 4H
J_COI=ind_ph%>%
  ggplot(aes(pH,J_COI))+
  geom_point() +

```

```

geom_errorbar(aes(ymin = J_COI-J_COI.sd,ymax = J_COI+J_COI.sd )) +
geom_errorbarh(aes(xmin = pH-pH.sd,xmax = pH+pH.sd))+
theme_bw()+
theme(strip.text.x = element_text(size=10),strip.background = element_blank(),
      panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
      axis.text.x = element_text(size=10, color="black",hjust=1),
      axis.title.x = element_text(color="black",size=12,face="bold"),
      axis.text.y =element_blank(),axis.title.y= element_blank(),
      axis.title = element_text(color="black",size=12, face="bold"),
      text = element_text(family="serif"))+
labs(y="",x="pH", title = "") +xlim(4,8)+ylim(0,1)+
geom_smooth(se=FALSE,method="lm", formula=y~poly(x,2))+
annotate("text",x=5.8,y=0.32,
        label=bquote("y="~.(round(summary(lineal.poly)[[4]][3,1],3))~
          "x"^2~"+"~.(round(summary(lineal.poly)[[4]][2,1],3))~
          "x"~"+"~.(round(summary(lineal.poly)[[4]][1,1],3))),
        cex = 2.5,col="black")+
annotate("text",x=5.8,y=0.25,
        label=bquote("R"^2~"="~.(round(summary(lineal.poly)[[8]],3))~"***"),
        cex =2.5,col="black")

```

#### #18S rRNA

```

lineal.poly=lm(J_18S~poly(pH,2),data=ind_ph)
summary(lineal.poly)

```

```

##
## Call:
## lm(formula = J_18S ~ poly(pH, 2), data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.190008 -0.034590 -0.002687  0.057936  0.143541
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.75325    0.01674  45.007 < 2e-16 ***
## poly(pH, 2)1   0.28137    0.08534   3.297 0.003152 **
## poly(pH, 2)2  -0.32960    0.08534  -3.862 0.000791 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.08534 on 23 degrees of freedom
## Multiple R-squared:  0.5286, Adjusted R-squared:  0.4876
## F-statistic: 12.89 on 2 and 23 DF,  p-value: 0.0001755

lineal.lm=lm(J_18S~pH,data=ind_ph)
summary(lineal.lm)

```

```

##
## Call:
## lm(formula = J_18S ~ pH, data = ind_ph)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.270101 -0.086528  0.009426  0.061715  0.197407

```

```
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.32896    0.16311   2.017  0.0550 .
## pH           0.07011    0.02673   2.623  0.0149 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1073 on 24 degrees of freedom
## Multiple R-squared:  0.2228, Adjusted R-squared:  0.1904
## F-statistic: 6.881 on 1 and 24 DF,  p-value: 0.0149

AIC(lineal.poly,lineal.lm)

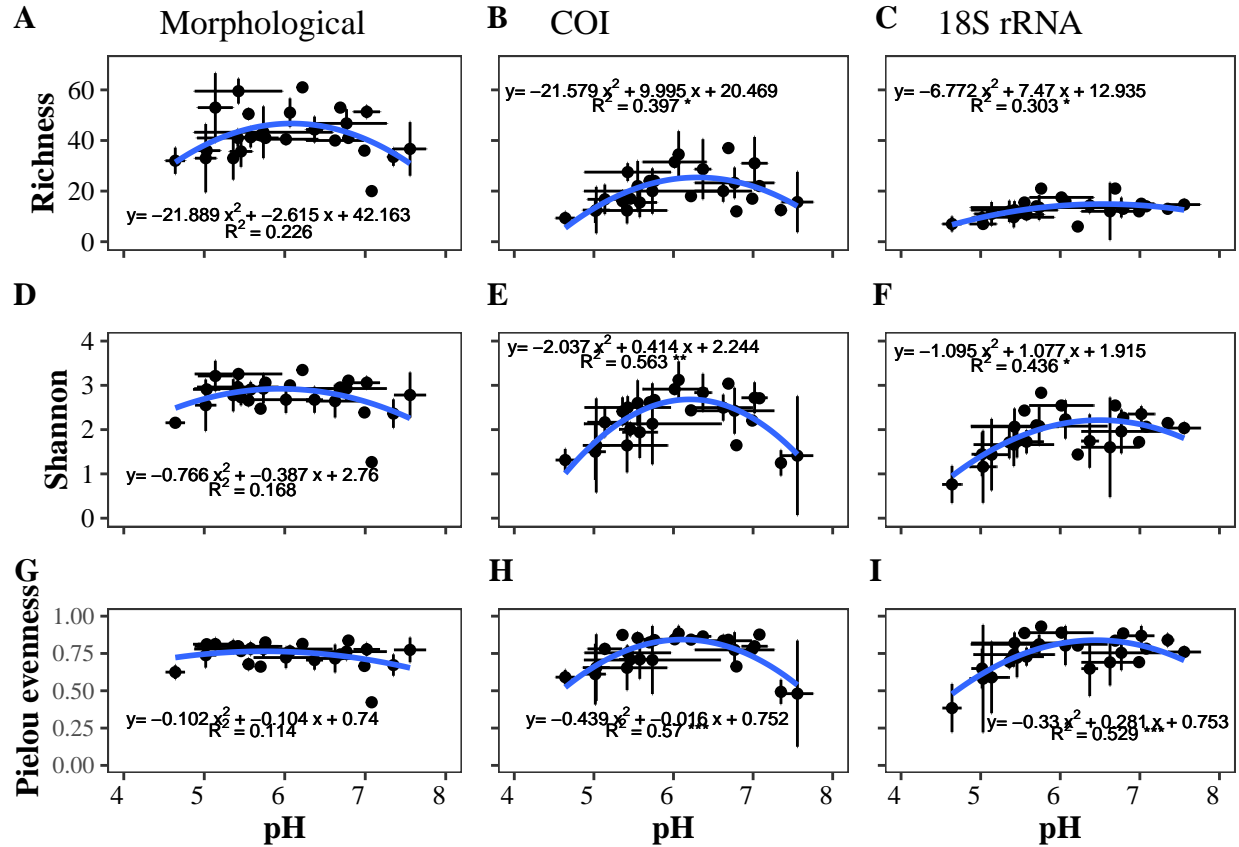
##              df      AIC
## lineal.poly   4 -49.38161
## lineal.lm      3 -38.38386

#Display panel figure 4I
J_18s=ind_ph%>%
  ggplot(aes(pH,J_18S))+
  geom_point() +
  geom_errorbar(aes(ymin = J_18S-J_18S.sd,ymax = J_18S+J_18S.sd )) +
  geom_errorbarh(aes(xmin = pH-pH.sd,xmax = pH+pH.sd))+
  theme_bw()+
  theme(strip.text.x = element_text(size=10),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.text.x = element_text(size=10, color="black",hjust=1),
        axis.title.x = element_text(color="black",size=12,face="bold"),
        axis.text.y =element_blank(),axis.title.y= element_blank(),
        axis.title = element_text(color="black",size=12, face="bold"),
        text = element_text(family="serif"))+
  labs(y="",x="pH", title = "") +xlim(4,8)+ylim(0,1)+
  geom_smooth(se=FALSE,method="lm", formula=y~poly(x,2))+
  annotate("text",x=6.6,y=0.30,
        label=bquote("y=~.(round(summary(lineal.poly)[[4]][3,1],3))~
          "x"^2~"+"~.(round(summary(lineal.poly)[[4]][2,1],3))~
          "x"~"+"~.(round(summary(lineal.poly)[[4]][1,1],3))),
        cex = 2.5,col="black")+
  annotate("text",x=6.6,y=0.23,
        label=bquote("R"^2~"="~.(round(summary(lineal.poly)[[8]],3))~"***"),
        cex =2.5,col="black")
```

### 3.4 Display Figure 4

Figure 4: Correlation between pH and alpha diversity variables (Richness, Shannon index, and Pielou) through morphological (A, D, G), COI (B, E, H), and 18S rRNA (C, F, I) datasets ordered by increasing pH. The dot is the average pH vs the average of the corresponding alpha diversity variable of each locality. Lines are the standard deviations of pH on the horizontal axis and the corresponding alpha on the vertical axis. The blue line is an adjusted polynomial model of order two.





## 4 Correlation and log2 fold changes

### 4.1 Log2 fold changes

```
meta.diff.coi<-metadata[,1:2]
meta.diff.18s<-metadata[,1:2]

#Richnnes log2 fold changes
meta.diff.coi$L2FC<-log2(metadata$S_COI/metadata$S_morpho)
meta.diff.coi$Method<-"COI"
meta.diff.18s$L2FC<-log2(metadata$S_18s/metadata$S_morpho)
meta.diff.18s$Method<-"18S"

meta.diff.rich<-rbind(meta.diff.coi,meta.diff.18s)
head(meta.diff.rich)
```

##	Station	Location	L2FC	Method
## 1	BAI01	BAI	-1.502500	COI
## 2	BAI04	BAI	-1.415037	COI
## 3	BAI06	BAI	-2.104337	COI
## 4	BAI07	BAI	-1.760812	COI
## 5	BPU04	BPU	-1.760812	COI
## 6	BSC03	BSC	-1.544321	COI

```
#Display panel figure 2B
plot.L2FC.rich<-meta.diff.rich%>%
```

```

ggplot(aes(L2FC,Method,fill=Method))+
geom_boxplot(outlier.shape = NA,alpha=0.4)+
geom_jitter(aes(L2FC,Method,fill=Method),position=position_jitterdodge(0.4),
            shape=21,size=1)+
scale_fill_manual(values = c("#6600CC","#FF9900"))+geom_vline(xintercept = 0,
                                                            linetype="dashed")+

labs(x="",y="Richness")+xlim(-3.5,2)+
theme_bw()+theme(strip.text.x = element_text(size=10),
                 strip.background = element_blank(),
                 panel.grid.minor = element_blank(),
                 panel.grid.major = element_blank(),
                 axis.text.x =element_blank(),
                 axis.title.x= element_blank(),
                 axis.text.y = element_text(size=8,color="black",
                                             angle=0,vjust = 0.5),
                 axis.title.y = element_text(color="black", size=10),
                 text = element_text(family="serif"))+
theme(legend.position = "right",legend.key.size = unit(0.5, 'cm'),
      legend.text = element_text(size=8))
legenda_rich=get_legend(plot.L2FC.rich)
plot.L2FC.rich=plot.L2FC.rich+theme(legend.position = "none")

#Normality test
shapiro.test(meta.diff.rich$L2FC[meta.diff.rich$Method=="COI"])

##
##  Shapiro-Wilk normality test
##
## data:  meta.diff.rich$L2FC[meta.diff.rich$Method == "COI"]
## W = 0.99045, p-value = 0.8952

shapiro.test(meta.diff.rich$L2FC[meta.diff.rich$Method=="18S"])

##
##  Shapiro-Wilk normality test
##
## data:  meta.diff.rich$L2FC[meta.diff.rich$Method == "18S"]
## W = 0.96967, p-value = 0.1057

#t-student
t.test(meta.diff.rich$L2FC[meta.diff.rich$Method=="COI"],mu=0)

##
##  One Sample t-test
##
## data:  meta.diff.rich$L2FC[meta.diff.rich$Method == "COI"]
## t = -16.661, df = 65, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
##  -1.352556 -1.063000
## sample estimates:
## mean of x
## -1.207778

t.test(meta.diff.rich$L2FC[meta.diff.rich$Method=="18S"],mu=0)

```

```
##
## One Sample t-test
##
## data: meta.diff.rich$L2FC[meta.diff.rich$Method == "18S"]
## t = -25.065, df = 65, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -1.958868 -1.669743
## sample estimates:
## mean of x
## -1.814306
```

*#Shannon index log2 fold changes*

```
meta.diff.sha.coi<-metadata[,1:2]
meta.diff.sha.18s<-metadata[,1:2]
```

```
meta.diff.sha.coi$L2FC<-log2(metadata$H_COI/metadata$H_morpho)
meta.diff.sha.coi$Method<-"COI"
meta.diff.sha.18s$L2FC<-log2(metadata$H_18s/metadata$H_morpho)
meta.diff.sha.18s$Method<-"18S"
```

```
meta.diff.shannon<-rbind(meta.diff.sha.coi,meta.diff.sha.18s)
head(meta.diff.shannon)
```

```
## Station Location      L2FC Method
## 1 BAI01      BAI -0.4055937 COI
## 2 BAI04      BAI -0.3950673 COI
## 3 BAI06      BAI -0.7746882 COI
## 4 BAI07      BAI -0.7254976 COI
## 5 BPU04      BPU -0.4582974 COI
## 6 BSC03      BSC -1.3970570 COI
```

*#Display panel figure 2D*

```
plot.L2FC.shannon<-meta.diff.shannon%>%
  ggplot(aes(L2FC,Method,fill=Method))+
  geom_boxplot(outlier.shape = NA,alpha=0.4)+
  geom_jitter(aes(L2FC,Method,fill=Method),position=position_jitterdodge(0.4),
              shape=21,size=1)+
  scale_fill_manual(values = c("#6600CC","#FF9900"))+geom_vline(xintercept = 0,
                                                                linetype="dashed")+
  labs(x="",y="Shannon")+xlim(-3.5,2)+
  theme_bw()+theme(strip.text.x = element_text(size=10),
                   strip.background = element_blank(),
                   panel.grid.minor = element_blank(),
                   panel.grid.major = element_blank(),
                   axis.text.x =element_blank(),
                   axis.title.x= element_blank(),
                   axis.text.y = element_text(size=8,color="black",
                                                angle=0,vjust = 0.5),
                   axis.title.y = element_text(color="black", size=10),
                   text = element_text(family="serif"))+
  theme(legend.position = "right",legend.key.size = unit(0.5, 'cm'),
        legend.text = element_text(size=8))
legenda_shannon=get_legend(plot.L2FC.shannon)
plot.L2FC.shannon=plot.L2FC.shannon+theme(legend.position = "none")
```

```

#Normality test
shapiro.test(meta.diff.shannon$L2FC[meta.diff.shannon$Method=="COI"])

##
## Shapiro-Wilk normality test
##
## data: meta.diff.shannon$L2FC[meta.diff.shannon$Method == "COI"]
## W = 0.91502, p-value = 0.0002481
shapiro.test(meta.diff.shannon$L2FC[meta.diff.shannon$Method=="18S"])

##
## Shapiro-Wilk normality test
##
## data: meta.diff.shannon$L2FC[meta.diff.shannon$Method == "18S"]
## W = 0.89594, p-value = 4.397e-05
#Statistical test
wilcox.test(meta.diff.shannon$L2FC[meta.diff.shannon$Method=="COI"],mu=0)

##
## Wilcoxon signed rank test with continuity correction
##
## data: meta.diff.shannon$L2FC[meta.diff.shannon$Method == "COI"]
## V = 242, p-value = 3.529e-08
## alternative hypothesis: true location is not equal to 0
wilcox.test(meta.diff.shannon$L2FC[meta.diff.shannon$Method=="18S"],mu=0)

##
## Wilcoxon signed rank test with continuity correction
##
## data: meta.diff.shannon$L2FC[meta.diff.shannon$Method == "18S"]
## V = 44, p-value = 1.221e-11
## alternative hypothesis: true location is not equal to 0
#Pielou evenness log2 fold changes
meta.diff.pie.coi<-metadata[,1:2]
meta.diff.pie.18s<-metadata[,1:2]

meta.diff.pie.coi$L2FC<-log2(metadata$J_COI/metadata$J_morpho)
meta.diff.pie.coi$Method<-"COI"
meta.diff.pie.18s$L2FC<-log2(metadata$J_18s/metadata$J_morpho)
meta.diff.pie.18s$Method<-"18S"

meta.diff.pielou<-rbind(meta.diff.pie.coi,meta.diff.pie.18s)
head(meta.diff.pielou)

## Station Location L2FC Method
## 1 BAI01 BAI 0.003337448 COI
## 2 BAI04 BAI 0.050860836 COI
## 3 BAI06 BAI -0.066749608 COI
## 4 BAI07 BAI -0.217307570 COI
## 5 BPU04 BPU 0.049892634 COI
## 6 BSC03 BSC -0.880257895 COI

```

```

#Display panel figure 2F
plot.L2FC.pielou<-meta.diff.pielou%>%
  ggplot(aes(L2FC,Method, fill=Method))+
  geom_boxplot(outlier.shape = NA,alpha=0.4)+
  geom_jitter(aes(L2FC,Method,fill=Method),position=position_jitterdodge(0.4),
              shape=21,size=1)+
  scale_fill_manual(values = c("#6600CC","#FF9900"))+geom_vline(xintercept = 0,
                                                                linetype="dashed")+
  labs(x=expression("Log" [2] * " fold change"),y="Pielou evenness")+xlim(-3.5,2)+
  theme_bw()+theme(strip.text.x = element_text(size=10),
                  strip.background = element_blank(),
                  panel.grid.minor = element_blank(),
                  panel.grid.major = element_blank(),
                  axis.text.x = element_text(size=8,color="black",
                                              angle=0,vjust = 0.5),
                  axis.text.y = element_text(size=8,color="black",
                                              angle=0,vjust = 0.5),
                  axis.title.x = element_text(color="black", size=10,
                                              margin = margin(t = 0,l=0)),
                  axis.title.y = element_text(color="black", size=10),
                  text = element_text(family="serif"))+
  guides(fill=guide_legend(ncol=2))+
  theme(legend.position="none",legend.key.size = unit(0.4, 'cm'),
        legend.text = element_text(size=4),legend.title = element_text(size=6))
legenda_pielou=get_legend(plot.L2FC.pielou)

```

#### #Normality test

```
shapiro.test(meta.diff.pielou$L2FC[meta.diff.pielou$Method=="COI"])
```

```

##
##  Shapiro-Wilk normality test
##
## data:  meta.diff.pielou$L2FC[meta.diff.pielou$Method == "COI"]
## W = 0.84996, p-value = 1.21e-06

```

```
shapiro.test(meta.diff.pielou$L2FC[meta.diff.pielou$Method=="18S"])
```

```

##
##  Shapiro-Wilk normality test
##
## data:  meta.diff.pielou$L2FC[meta.diff.pielou$Method == "18S"]
## W = 0.85743, p-value = 2.072e-06

```

#### #Statistical test

```
wilcox.test(meta.diff.pielou$L2FC[meta.diff.pielou$Method=="COI"],mu=0)
```

```

##
##  Wilcoxon signed rank test with continuity correction
##
## data:  meta.diff.pielou$L2FC[meta.diff.pielou$Method == "COI"]
## V = 1052, p-value = 0.7349
## alternative hypothesis: true location is not equal to 0
wilcox.test(meta.diff.pielou$L2FC[meta.diff.pielou$Method=="18S"],mu=0)

```

```
##
```

```
## Wilcoxon signed rank test with continuity correction
##
## data: meta.diff.pielou$L2FC[meta.diff.pielou$Method == "18S"]
## V = 1029, p-value = 0.6273
## alternative hypothesis: true location is not equal to 0
```

## 4.2 Correlation analysis

### *#Richness*

```
corr_rich<-rcorr(as.matrix(metadata[,c(25,28,31)]),type="spearman")
corr_rich
```

```
##           S_morpho S_COI S_18s
## S_morpho      1.00  0.58  0.32
## S_COI         0.58  1.00  0.56
## S_18s         0.32  0.56  1.00
##
## n= 66
##
##
## P
##           S_morpho S_COI S_18s
## S_morpho           0.0000 0.0083
## S_COI       0.0000           0.0000
## S_18s       0.0083  0.0000
```

### *#significance*

```
round(as.dist(corr_rich$P),4)
```

```
##           S_morpho S_COI
## S_COI      0.0000
## S_18s      0.0083 0.0000
```

### *#Display panel figure 2A*

```
richness_corr=ggcorrplot(corr_rich$r,type = "upper",
                          outline.col = "white",lab=TRUE,insig = "blank",lab_size = 3)+
  theme_bw()+
  labs(x="",y="")+
  theme_tufte(base_size = 7)+
  theme(strip.text.x = element_text(size=12),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.title.x =element_blank(),axis.title.y= element_blank(),
        axis.text.x = element_text(size=8,color="black",angle=0,vjust = 0.5),
        axis.text.y = element_text(size=8,color="black",angle=0,vjust = 0.5),
        text = element_text(family="serif"))+
  theme(legend.position = "none",legend.key.size = unit(0.5, 'cm'),
        legend.text = element_text(size=8))
```

### *#Shannon index*

```
corr_sha<-rcorr(as.matrix(metadata[,c(24,27,30)]),type="spearman")
corr_sha
```

```
##           H_morpho H_COI H_18s
## H_morpho      1.00  0.20  0.26
## H_COI         0.20  1.00  0.49
## H_18s         0.26  0.49  1.00
```

```
##
## n= 66
##
##
## P
##           H_morpho H_COI  H_18s
## H_morpho           0.1119 0.0375
## H_COI      0.1119           0.0000
## H_18s      0.0375    0.0000

#significance
round(as.dist(corr_sha$P),4)

##           H_morpho  H_COI
## H_COI      0.1119
## H_18s      0.0375 0.0000

#Display panel figure 2C
shannon_corr=ggcorrplot(corr_sha$r,type = "upper",
                        outline.col = "white",lab=TRUE,insig = "blank",lab_size = 3)+
  labs(x="",y="")+
  theme_bw()+
  theme_tufte(base_size = 7)+
  theme(strip.text.x = element_text(size=12),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.title.x =element_blank(),axis.title.y= element_blank(),
        axis.text.x = element_text(size=8,color="black",angle=0,vjust = 0.5),
        axis.text.y = element_text(size=8,color="black",angle=0,vjust = 0.5),
        text = element_text(family="serif"))+
  theme(legend.position = "none",legend.key.size = unit(0.5, 'cm'),
        legend.text = element_text(size=8))

#Pielou evenness
corr_pie<-rcorr(as.matrix(metadata[,c(26,29,32)]),type="spearman")
corr_pie

##           J_morpho J_COI J_18s
## J_morpho      1.00 -0.04  0.17
## J_COI        -0.04  1.00  0.23
## J_18s         0.17  0.23  1.00
##
## n= 66
##
##
## P
##           J_morpho J_COI  J_18s
## J_morpho           0.7376 0.1771
## J_COI      0.7376           0.0673
## J_18s      0.1771    0.0673

#Significance
round(as.dist(corr_pie$P),4)

##           J_morpho  J_COI
## J_COI      0.7376
## J_18s      0.1771 0.0673
```

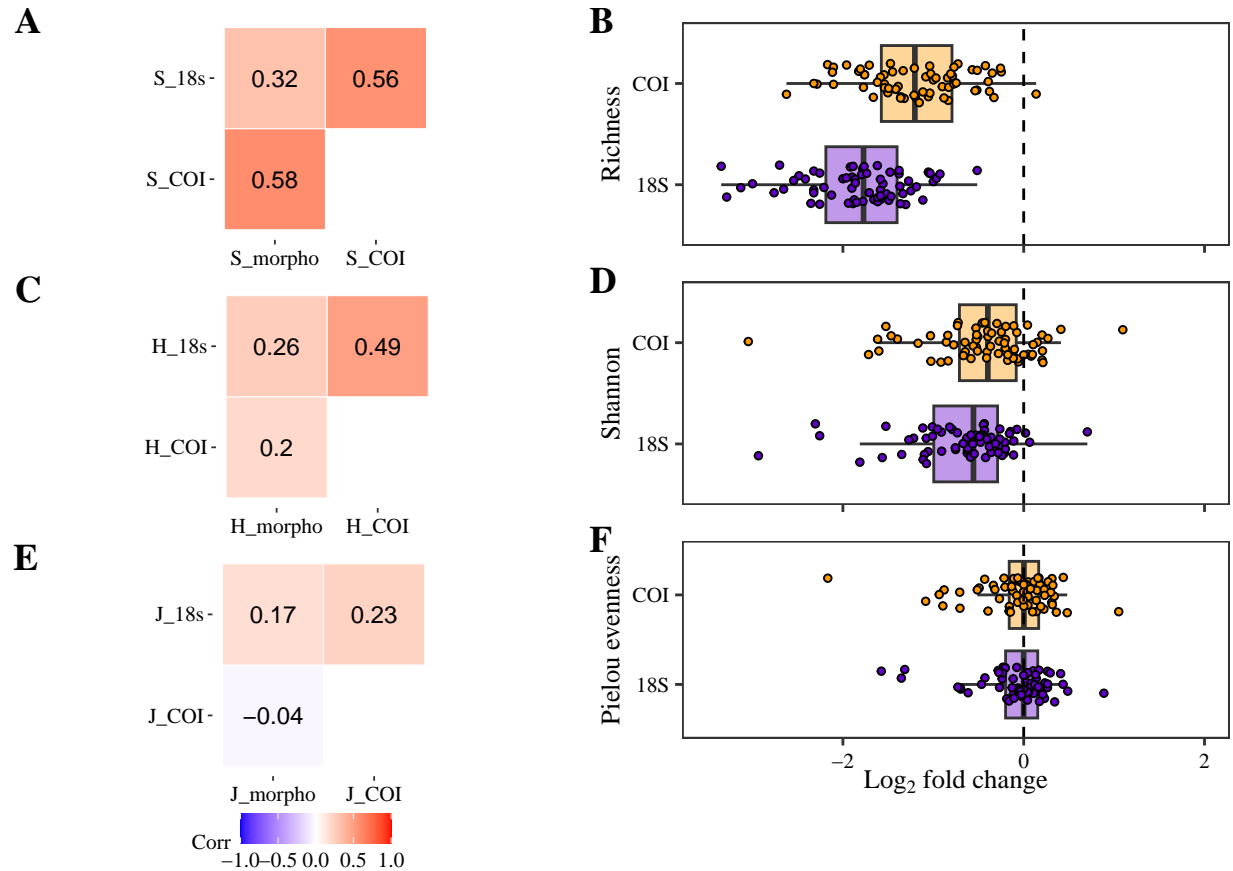
```

#Display panel figure 2E
pielou_corr=ggcorrplot(corr_pie$r,type = "upper",
                      outline.col = "white",lab=TRUE,insig = "blank",lab_size=3)+
  labs(x="",y="")+
  theme_bw()+
  theme_tufte(base_size = 7)+
  theme(strip.text.x = element_text(size=12),strip.background = element_blank(),
        panel.grid.minor = element_blank(),panel.grid.major = element_blank(),
        axis.title.x =element_blank(),axis.title.y= element_blank(),
        axis.text.x = element_text(size=8,color="black",angle=0,vjust = 0.5),
        axis.text.y = element_text(size=8,color="black",angle=0,vjust = 0.5),
        text = element_text(family="serif"))+
  theme(legend.position = "bottom",legend.key.size = unit(0.4, 'cm'),
        legend.text = element_text(size=8),
        legend.title = element_text(size = 8))
legenda_pielou_corr=get_plot_component(pielou_corr, 'guide-box-bottom',
                                     return_all = TRUE)
pielou_corr=pielou_corr+theme(legend.position = "none")

```

### 4.3 Display Figure 2

Figure 2: Correlation between molecular and morphological identification in richness (A), Shannon index (C), and Pielou evenness (E). Comparison of alpha diversity indices in the three approaches to diatom communities of high mountain mires in richness (B), Shannon index (D), and Pielou evenness (F), molecular with morphological identification.





## 5 dbRDA

```
Metadata=data.frame(read_xlsx("data\\Table S1.xlsx",sheet="Metadata",na=""))

x<-Metadata[3:23]
head(x)
```

	WTD	GDD	Rad	Prec	pH	Vascular_plants	Sphagnum	Brown_mosses											
##																			
## 1	-1.0	1532	1835	1487.9	5.45		58.5	58	2.5										
## 2	-0.5	1410	1745	1504.0	4.95		63.5	0	75.0										
## 3	-13.0	1532	1835	1482.7	5.05		62.5	0	8.0										
## 4	-1.5	1593	1742	1462.3	5.10		55.5	2	16.0										
## 5	-1.0	2903	1812	1037.7	6.22		123.5	21	3.0										
## 6	-2.0	2906	1946	947.4	7.55		99.5	0	8.5										
##																			
##																			
## 1																			
## 2																			
## 3																			
## 4																			
## 5																			
## 6																			
##																			
##																			
## 1																			
## 2																			
## 3																			
## 4																			
## 5																			
## 6																			
##																			
##																			
## 1																			
## 2																			
## 3																			
## 4																			
## 5																			
## 6																			

```
#data transformation
datos.trans_morpho <- sqrt(dato_morpho)
datos.trans_coi <- sqrt(dato_coi)
datos.trans_18s<- sqrt(dato_18s)
```

### 5.1 dbRDA Morphology

```
analisis_morpho <- capscale(datos.trans_morpho ~ WTD+GDD+Rad+
                           Sphagnum+Brown_mosses+Acrocarp_mosses+Liverworts+
                           Bryophytes+Vascular_plants+
                           pH+Prec+K+Mg+Mn+Na+P+S+Si+Al+Ca+Fe, Metadata,
                           dist="bray")

anova (analisis_morpho) # 0.001 *** - it is significant
```

```
## Permutation test for capscale under reduced model
## Permutation: free
## Number of permutations: 999
##
## Model: capscale(formula = datos.trans_morpho ~ WTD + GDD + Rad + Sphagnum + Brown_mosses + Acrocarp_mosses + Liverworts + Bryophytes + Vascular_plants + pH + Prec + K + Mg + Mn + Na + P + S + Si + Al + Ca + Fe, Metadata, dist = "bray")
##           Df SumOfSqs      F Pr(>F)
## Model    20   10.454 2.1575 0.001 ***
## Residual 45    10.902
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
adjR2.tbrda <- RsquareAdj (analysis_morpho)$adj.r.squared
```

```
#Forward selection
```

```
sel.fs <- forward.sel (Y = datos.trans_morpho , X=x, adjR2thresh = adjR2.tbrda)
```

```
## Testing variable 1
## Testing variable 2
## Testing variable 3
## Testing variable 4
## Testing variable 5
## Procedure stopped (alpha criteria): pvalue for variable 5 is 0.055000 (> 0.050000)
```

```
sel.fs
```

```
##   variables order      R2      R2Cum AdjR2Cum      F pvalue
## 1      pH      5 0.15039348 0.1503935 0.1371184 11.328989 0.001
## 2      GDD      2 0.02687798 0.1772715 0.1511531 2.058167 0.004
## 3      Mg     16 0.02897985 0.2062513 0.1678441 2.263627 0.001
## 4     Prec      4 0.01903143 0.2252827 0.1744816 1.498504 0.032
```

```
analysis_morpho_fs <- capscale(datos.trans_morpho ~ GDD+pH+Prec+Mg, Metadata,
                               dist="bray")
summary(analysis_morpho_fs)
```

```
##
```

```
## Call:
```

```
## capscale(formula = datos.trans_morpho ~ GDD + pH + Prec + Mg,      data = Metadata, distance = "bray"
```

```
##
```

```
## Partitioning of squared Bray distance:
```

```
##           Inertia Proportion
## Total      21.357      1.0000
## Constrained    5.735      0.2685
## Unconstrained 15.622      0.7315
```

```
##
```

```
## Eigenvalues, and their contribution to the squared Bray distance
```

```
##
```

```
## Importance of components:
```

```
##           CAP1      CAP2      CAP3      CAP4      MDS1      MDS2      MDS3
## Eigenvalue      4.3332 0.61005 0.41251 0.37904 2.03714 1.16655 1.0316
## Proportion Explained 0.2029 0.02856 0.01932 0.01775 0.09539 0.05462 0.0483
## Cumulative Proportion 0.2029 0.23146 0.25078 0.26852 0.36391 0.41853 0.4668
##           MDS4      MDS5      MDS6      MDS7      MDS8      MDS9      MDS10
## Eigenvalue      0.80461 0.79114 0.68426 0.58200 0.51403 0.50802 0.47569
## Proportion Explained 0.03767 0.03704 0.03204 0.02725 0.02407 0.02379 0.02227
## Cumulative Proportion 0.50451 0.54155 0.57359 0.60084 0.62491 0.64870 0.67097
##           MDS11      MDS12      MDS13      MDS14      MDS15      MDS16      MDS17
## Eigenvalue      0.42582 0.41709 0.38130 0.36409 0.34544 0.32508 0.3053
## Proportion Explained 0.01994 0.01953 0.01785 0.01705 0.01617 0.01522 0.0143
## Cumulative Proportion 0.69091 0.71044 0.72829 0.74534 0.76152 0.77674 0.7910
##           MDS18      MDS19      MDS20      MDS21      MDS22      MDS23      MDS24
## Eigenvalue      0.29707 0.28655 0.26682 0.25278 0.22812 0.21450 0.208557
## Proportion Explained 0.01391 0.01342 0.01249 0.01184 0.01068 0.01004 0.009765
## Cumulative Proportion 0.80494 0.81836 0.83085 0.84269 0.85337 0.86341 0.873180
##           MDS25      MDS26      MDS27      MDS28      MDS29      MDS30
```

```

## Eigenvalue      0.204296 0.19307 0.177620 0.168212 0.163369 0.153479
## Proportion Explained 0.009566 0.00904 0.008317 0.007876 0.007649 0.007186
## Cumulative Proportion 0.882746 0.89179 0.900103 0.907979 0.915629 0.922815
##               MDS31   MDS32   MDS33   MDS34   MDS35   MDS36
## Eigenvalue      0.14139 0.136333 0.128620 0.12642 0.11063 0.106262
## Proportion Explained 0.00662 0.006384 0.006022 0.00592 0.00518 0.004976
## Cumulative Proportion 0.92944 0.935819 0.941841 0.94776 0.95294 0.957916
##               MDS37   MDS38   MDS39   MDS40   MDS41   MDS42
## Eigenvalue      0.105807 0.095932 0.085993 0.084972 0.08244 0.072014
## Proportion Explained 0.004954 0.004492 0.004026 0.003979 0.00386 0.003372
## Cumulative Proportion 0.962871 0.967363 0.971389 0.975368 0.97923 0.982600
##               MDS43   MDS44   MDS45   MDS46   MDS47   MDS48
## Eigenvalue      0.060781 0.053766 0.047686 0.044456 0.039331 0.032095
## Proportion Explained 0.002846 0.002517 0.002233 0.002082 0.001842 0.001503
## Cumulative Proportion 0.985446 0.987963 0.990196 0.992277 0.994119 0.995622
##               MDS49   MDS50   MDS51   MDS52   MDS53
## Eigenvalue      0.027498 0.0171959 0.0160493 0.0147608 0.0106180
## Proportion Explained 0.001288 0.0008052 0.0007515 0.0006911 0.0004972
## Cumulative Proportion 0.996909 0.9977146 0.9984660 0.9991572 0.9996544
##               MDS54   MDS55   MDS56
## Eigenvalue      0.0044508 0.0026375 2.935e-04
## Proportion Explained 0.0002084 0.0001235 1.374e-05
## Cumulative Proportion 0.9998628 0.9999863 1.000e+00
##
## Accumulated constrained eigenvalues
## Importance of components:
##               CAP1   CAP2   CAP3   CAP4
## Eigenvalue      4.3332 0.6100 0.41251 0.37904
## Proportion Explained 0.7556 0.1064 0.07193 0.06609
## Cumulative Proportion 0.7556 0.8620 0.93391 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 6.084609
##
##
## Species scores
##
##               CAP1   CAP2   CAP3
## Achnanthidium.minutisimum      0.9252808 -5.790e-02 -3.124e-01
## Achnanthidium.caledonicum     -0.0090569 -2.228e-02 -4.820e-02
## Achnanthidium.rivulare        0.0475790 5.322e-02 4.478e-02
## Achnanthidium.lineare        0.0751045 -5.117e-02 -4.356e-02
## Achnanthidium.delmontii     -0.0080806 1.942e-01 3.291e-02
## Achnanthidium.macrocephalum  -0.0066334 -3.104e-03 -1.996e-02
## Achnanthidium.sp7           -0.0425992 2.309e-03 5.209e-03
## Achnanthidium.sp12          0.0177455 -1.702e-02 1.684e-02
## Achnanthidium.affine         -0.0054446 7.743e-03 6.799e-04
## Achnanthidium.trinode        0.0092824 7.384e-04 6.984e-03
## Achnanthes.lapidosa.var..lapidosa -0.0070239 6.306e-02 2.086e-02
## Achnanthes.minutissima.var..jackii -0.0130715 -1.650e-03 1.398e-02
## Adlafia.aquaeductae          0.2012964 1.163e-01 6.133e-02
## Adlafia.bryophila            0.0003356 9.425e-03 -2.578e-03

```

## Adlafia.minuscula	0.0087141	4.616e-02	1.449e-02
## Aneumastus.tusculus	0.0216243	3.067e-03	-2.742e-02
## Amphora.pediculus	0.0027523	1.546e-01	-2.393e-02
## Amphora.copulata	0.1317403	-1.479e-02	-1.815e-02
## Amphora.ovalis	0.0157000	3.055e-02	-8.596e-03
## Amphora.affinis	0.0103764	2.603e-03	5.126e-04
## Aulacoseira.alpigenica	-0.0350933	3.420e-02	-7.388e-02
## Aulacoseira.pfaffiana	-0.1038768	-2.563e-02	5.678e-03
## Aulacoseira.distans.var..humilis	-0.0017411	2.275e-03	-6.355e-03
## Aulacoseira.granulata	0.0110439	-2.298e-03	1.140e-03
## Aulacoseira.tenella	-0.2876765	-7.326e-02	6.983e-02
## Aulacoseira.tethera	-0.3043594	-7.048e-02	9.222e-02
## Aulacoseira.valida	-0.0064680	-5.279e-03	7.510e-03
## Aulacoseira.granulata.var..angustissima	0.0028444	9.197e-03	-3.717e-03
## Aulacoseira.ambigua	-0.0521353	-1.607e-02	1.077e-02
## Aulacoseira.sp4	0.0128790	3.349e-03	8.680e-03
## Brachysira.brebissonii	-0.6146421	-8.822e-02	6.492e-03
## Brachysira.intermedia	-0.2594749	-3.448e-02	1.654e-02
## Brachysira.neoexilis	-0.4748926	-6.889e-02	-4.596e-03
## Brachysira.neglectissima	-0.1274033	-1.503e-02	4.942e-02
## Brachysira.serians	-0.1156397	4.350e-03	2.788e-02
## Caloneis.silicula	0.0263167	2.234e-02	8.687e-03
## Caloneis.tenuis	0.4230219	-5.439e-02	5.618e-03
## Caloneis.bacilum	0.3678617	-2.090e-02	-2.391e-02
## Caloneis.bacilum.var..inflata	0.0039200	3.121e-03	3.609e-03
## Caloneis.ondulata	0.0833937	9.974e-03	-4.179e-03
## Caloneis.lancettula	-0.0030764	1.849e-02	5.899e-03
## Caloneis.fusus	0.0025840	1.385e-02	-1.227e-02
## Caloneis.alpestris	0.2111780	-1.243e-01	5.003e-02
## Caloneis.aerophila	0.0261384	1.429e-02	-8.858e-03
## Chamaepinnularia.mediocris	-0.1623401	5.260e-02	9.049e-02
## Chamaepinnularia.hassiac	0.0058872	1.951e-02	-2.667e-02
## Chamaepinnularia.submusciola	0.0587602	-9.100e-05	-2.368e-02
## Chamaepinnularia.soehrensii	-0.0162126	5.551e-02	1.236e-02
## Chamaepinnularia.begerii	-0.3517548	1.816e-03	3.272e-02
## Chamaepinnularia.musciola	-0.0165003	7.144e-02	-4.187e-04
## Chamaepinnularia.evanida	-0.0284626	7.980e-03	1.741e-02
## Cavinula.cocconeiformis	0.0174957	-1.435e-02	-7.760e-03
## Cocconeis.placentula	-0.0078120	2.674e-02	5.680e-03
## Cocconeis.lineata	-0.0057431	3.918e-02	6.706e-03
## Cocconeis.sp3	0.0016084	2.687e-02	1.162e-02
## Cyclotella.antiqua	0.0026750	4.124e-03	8.635e-04
## Cyclostephanos.dubius	0.0091311	-3.725e-03	-2.902e-03
## Cymbella.subcuspidata	-0.0018548	1.711e-02	-3.297e-03
## Cymbella.parva	0.1963593	-1.390e-01	-2.286e-02
## Cymbella.affinis.var..procera	-0.0131983	-2.919e-03	2.446e-03
## Cymbella.lancetulla	-0.0417135	-7.950e-02	5.470e-02
## Cymbella.excisa	-0.0081734	-8.059e-03	-1.520e-02
## Cymbella.cymbiformis	0.0238173	-3.970e-02	2.854e-03
## Cymbella.cleve.eulerge	0.0287037	-4.126e-02	1.732e-03
## Cymbella.lange.bertalotii	0.0169590	-2.066e-02	-8.718e-03
## Cymbella.aspera	0.0399436	1.758e-02	-1.412e-02
## Cymbella.romboidea	0.0081999	-7.739e-03	-3.738e-03
## Cymbopleura.acuta	0.0001980	1.596e-03	-6.707e-03

## Cymboppleura.naviculiformis	0.0267700	6.199e-02	-4.742e-02
## Cymboppleura.incerta	0.3563981	-6.807e-02	1.541e-02
## Cymboppleura.pyrenaica	-0.0090392	-1.359e-02	5.717e-03
## Cymboppleura.subaequalis	0.0833153	-6.910e-02	9.339e-02
## Cymboppleura.similiformis	0.0364178	-2.659e-02	-9.456e-03
## Cymboppleura.florentiniformis	0.0075329	-6.243e-03	-3.612e-05
## Cymboppleura.laponica	0.0564889	-1.046e-02	3.877e-02
## Cymboppleura.hybrida	0.0403431	1.422e-02	-3.617e-02
## Cymboppleura.sp3	0.0588089	-5.782e-02	-2.259e-02
## Cymboppleura.incerta.var..grunowii	0.0191485	-2.649e-02	-1.556e-02
## Cymboppleura.sublanceolata	-0.0057090	-3.168e-03	1.070e-03
## Cymboppleura.valaiseana	0.0070812	-1.327e-02	-9.124e-04
## Denticula.tenuis.var..frigida	0.0812862	-1.403e-01	8.873e-04
## Diploneis.oculata	0.0503238	8.806e-03	-3.249e-02
## Diploneis.parma	0.0714252	1.150e-02	2.259e-02
## Diploneis.krammeri	0.2685162	-3.190e-02	5.184e-02
## Diploneis.oblongella	0.1634287	1.175e-01	7.899e-02
## Diploneis.petersenii	0.0966823	-2.140e-02	-2.420e-03
## Diploneis.separanda	0.0144639	1.890e-02	7.372e-04
## Diploneis.minuta	0.0218200	1.751e-02	-2.364e-02
## Encyonema.neogracile	-0.2614780	3.756e-02	4.667e-03
## Encyonema.perpusillum	-0.1207422	-7.428e-02	7.900e-02
## Encyonema.silesiacum	0.0440969	-6.210e-02	2.439e-02
## Encyonema.silesiacum.var..altensis	0.0214447	-1.481e-02	-1.387e-02
## Encyonema.ventricosum	-0.0157589	-9.391e-03	-3.571e-02
## Encyonema.minutum	0.0326737	2.704e-02	8.844e-03
## Encyonema.lange.bertalotii	-0.0721552	-4.252e-02	3.040e-02
## Encyonema.vulgare	0.0554311	-6.312e-02	-4.274e-02
## Encyonema.hebredicum	-0.1134704	-3.053e-02	9.693e-03
## Encyonema.procerum	0.0033149	-1.592e-02	5.782e-03
## Encyonema.sp4	0.0166372	-3.399e-03	-1.597e-03
## Encyonema.sp5	-0.0173115	1.113e-02	8.556e-03
## Encyonema.sp13	-0.0098310	-5.750e-03	7.949e-03
## Encyonema.gaeumannii	0.0629020	-1.738e-02	3.354e-02
## Encyonema.alpinum	0.0286876	2.445e-03	-5.244e-03
## Encyonema.perminutum	0.0027214	1.238e-02	-8.001e-03
## Encyonema.sp20	-0.0042648	5.573e-03	-1.557e-02
## Encyonema.sp22	0.0248991	-9.597e-03	-1.294e-02
## Encyonopsis.cesatii	0.2663138	-8.291e-02	-6.229e-02
## Encyonopsis.krammeri	0.2245179	-1.282e-01	2.595e-02
## Encyonopsis.subminuta	0.1386325	-5.894e-02	6.959e-02
## Encyonopsis.minuta	-0.0090574	-2.108e-03	2.787e-03
## Encyonopsis.sp1	0.0170996	9.471e-03	-1.728e-02
## Encyonopsis.sp2	-0.0031995	-7.834e-03	5.004e-03
## Encyonopsis.sp4	-0.0354701	-3.596e-02	-5.453e-02
## Encyonopsis.sp14	0.0033149	-1.592e-02	5.782e-03
## Encyonopsis.sp15	0.0750305	-1.517e-01	7.422e-03
## Eolimna.minima	0.6428875	2.881e-01	6.963e-02
## Eolimna.sp1	0.1026465	-3.244e-02	-1.355e-02
## Eolimna.sp2	-0.0051661	3.235e-02	1.515e-02
## Eolimna.sp.N.5	-0.0167665	-3.202e-03	7.826e-03
## Epithemia.argus	0.3334851	-9.746e-02	1.776e-01
## Epithemia.frickei	0.0106531	-8.828e-03	-5.109e-05
## Eunotia.implicata	-0.0020516	6.816e-03	-6.098e-04

## Eunotia.naegeli	-0.0912618	4.241e-02	-9.210e-02
## Eunotia.bilunaris	-0.0154413	5.800e-02	-3.866e-02
## Eunotia.botiliformis	-0.0434873	-2.099e-03	6.461e-03
## Eunotia.cataractarum	-0.0217900	-1.195e-02	-1.581e-02
## Eunotia.exigua	-0.2108831	-5.837e-02	1.893e-02
## Eunotia.faba	-0.1486997	-6.233e-02	6.400e-02
## Eunotia.fallax	-0.3557092	-1.595e-02	-8.147e-02
## Eunotia.incisa	-0.7855492	-1.007e-01	-3.072e-02
## Eunotia.borealpina	-0.2560496	-4.694e-02	6.729e-03
## Eunotia.intermedia	-0.0908270	-2.388e-02	3.092e-02
## Eunotia.nymanniana	-0.3356352	7.076e-02	3.616e-02
## Eunotia.soleirolii	-0.0399882	-1.896e-02	1.761e-02
## Eunotia.minor	-0.1949657	-7.395e-03	-2.452e-02
## Eunotia.metensiae	-0.0004468	4.300e-02	1.195e-02
## Eunotia.neofallax	-0.0056943	3.551e-02	-1.581e-02
## Eunotia.rhomboidea	-0.1406717	-1.730e-01	6.768e-02
## Eunotia.gracilis	-0.1248251	-2.477e-02	-3.877e-02
## Eunotia.subarcuatoides	-0.0777187	-4.381e-02	2.853e-02
## Eunotia.tenella	-0.2645528	-7.346e-02	9.109e-03
## Eunotia.arcus	0.1673431	-1.392e-01	2.173e-02
## Eunotia.arcubus	-0.0030398	-5.880e-02	9.402e-03
## Eunotia.palatia	-0.0427014	-2.132e-02	-1.406e-02
## Eunotia.trinacria	-0.0588143	7.973e-03	6.431e-03
## Eunotia.curtagrunowii	-0.0040867	-4.030e-03	-7.601e-03
## Eunotia.mayamae	-0.0239709	-1.652e-03	-3.591e-02
## Eunotia.neocompacta.var..vixcompacta	-0.0263235	-9.830e-03	1.563e-03
## Eunotia.groenlandica	-0.2996724	2.898e-02	-2.999e-02
## Eunotia.major	-0.0245012	-3.132e-03	1.903e-02
## Eunotia.paludosa	-0.2034483	-2.305e-02	-4.891e-02
## Eunotia.superpaludosa	-0.0445359	3.844e-02	-1.181e-02
## Eunotia.lapponica	-0.2184061	-6.686e-02	6.272e-02
## Eunotia.panda	-0.0245660	2.560e-02	-5.980e-02
## Eunotia.ninae	-0.0054711	-2.916e-03	4.025e-03
## Eunotia.diodon	-0.0084017	-5.895e-03	2.418e-04
## Eunotia.islandica	-0.0014507	4.820e-03	-4.312e-04
## Eunotia.diadema	-0.0039787	-1.086e-02	-1.003e-02
## Eunotia.novaisiae.var..altopyrenaica	-0.0087623	-2.145e-02	1.370e-02
## Eunotia.ursamaioris	-0.0177431	-9.345e-04	9.675e-03
## Eunotia.paerupta	-0.0323359	1.106e-02	5.727e-03
## Eunotia.valida	-0.0813020	-1.469e-05	-2.076e-03
## Eunotia.glacialis	-0.0057090	-3.168e-03	1.070e-03
## Eunotia.meisteri	0.0125271	3.993e-02	9.225e-03
## Eucocconeis.laevis	0.1185183	-3.569e-02	-6.969e-02
## Eucocconeis.alpestris	0.0096294	-9.005e-03	-4.349e-02
## Eucocconeis.flexella	0.0209073	-3.417e-02	-2.587e-03
## Fallacia.vitrea	-0.1298781	2.142e-02	4.246e-02
## Fallacia.minuscula	0.0853299	2.989e-02	-6.723e-03
## Fallacia.insoabilis	0.0013687	1.216e-02	1.395e-02
## Fallacia.sp3	-0.0052663	4.010e-04	-1.393e-02
## Fragilaria.gracilis	-0.0128679	-1.155e-01	-3.707e-02
## Fragilaria.famelica	-0.0051820	1.110e-02	-6.095e-03
## Fragilaria.alpestris	0.0211991	6.416e-02	-1.480e-02
## Fragilaria.capucina	-0.0080823	-7.747e-04	5.859e-03
## Fragilaria.rumpens	-0.0071138	8.888e-03	3.297e-03

## Fragilaria.rinoi	-0.0000186	6.985e-02	-1.974e-02
## Fragilaria.capucina.var.gracilis	0.0163744	3.367e-02	5.530e-04
## Fragilaria.acidoclinata	0.0048598	4.244e-03	4.531e-03
## Fragilaria.austriaca	-0.0028897	-2.849e-03	-5.375e-03
## Fragilaria.nevadensis	-0.0061082	2.308e-02	1.847e-02
## Fragilaria.oldenburgiana	0.0012932	-1.121e-03	9.374e-03
## Fragilaria.sp	-0.0013502	4.102e-03	-1.457e-02
## Fragilaria.exiguiformis	-0.0575266	-8.030e-03	-3.741e-03
## Fragilariforma.visrescens	-0.1419162	-3.444e-02	-7.224e-02
## Fragilariforma.bicapitata	-0.0034592	3.036e-02	-1.748e-02
## Frustulia.crassinervia	-0.3483272	-3.476e-02	-1.916e-02
## Frustulia.vulgaris	0.0140868	4.597e-02	1.398e-02
## Frustulia.saxonica	-0.5684197	-8.062e-02	9.176e-02
## Gomphonema.cymbelliclinum	0.1144525	-8.834e-02	2.961e-02
## Gomphonema.gracile	-0.0096959	1.807e-01	2.654e-02
## Gomphonema.clavatum	-0.0177797	-1.778e-03	-1.116e-02
## Gomphonema.subclavatum	0.1066088	5.498e-02	2.710e-02
## Gomphonema.parvulum	0.4509684	-9.776e-02	1.018e-01
## Gomphonema.lateripunctatum	0.3214822	-6.643e-02	4.528e-02
## Gomphonema.micropus	0.0121510	9.468e-02	5.785e-02
## Gomphonema.exilissimum	0.1230895	1.336e-01	8.955e-03
## Gomphonema.varioireduncum	0.0519815	-7.027e-02	-3.494e-02
## Gomphonema.drutelingense	0.1399830	-2.379e-02	1.868e-02
## Gomphonema.productum	0.0006981	1.862e-02	1.157e-02
## Gomphonema.acidoclinatum	-0.0074649	1.854e-02	-3.577e-02
## Gomphonema.angustivalva	-0.0098556	1.136e-03	-9.989e-03
## Gomphonema.subtile	0.0200856	-1.896e-02	-9.157e-03
## Gomphonema.angustus	0.0119918	-1.461e-02	-6.165e-03
## Gomphonema.olivaceum.var..calcerum	0.0565268	-4.378e-03	-1.955e-02
## Gomphonema.herbridense	0.0236996	3.869e-02	1.442e-02
## Gomphonema.variscohercynicum	0.0298442	-8.386e-03	1.815e-02
## Gomphonema.sp6	0.0430089	-2.123e-02	2.646e-02
## Gomphonema.sp23	0.0167457	-2.320e-02	-1.433e-02
## Gomphonema.sp27	-0.0032486	-8.870e-03	-8.186e-03
## Gomphonema.sp28	0.0074124	-3.560e-02	1.293e-02
## Gomphonema.lagenula	0.0061482	3.921e-02	2.047e-02
## Greissleria.acceptata	-0.0006038	1.835e-03	-6.515e-03
## Halamphora.montana	-0.0034386	-4.407e-03	-7.462e-03
## Hannaea.arcus	-0.0127238	4.923e-02	-3.253e-03
## Hantzschia.amphioxys	0.0208103	1.241e-02	1.309e-02
## Hantzschia.calcifuga	-0.0093057	3.215e-02	1.574e-02
## Hantzschia.abundans	-0.0042756	8.969e-03	-9.214e-03
## Humidophila.perpusilla	-0.0193566	8.670e-02	3.070e-02
## Karayevia	-0.0041616	1.438e-02	7.041e-03
## Kobayasiella.micropunctata	-0.7681619	9.339e-02	-2.285e-01
## Kobayasiella.sp	-0.1625467	4.189e-02	5.383e-02
## Luticula.mutica	-0.0200839	2.531e-02	1.026e-02
## Mastogloia.lacustris	0.1605479	-8.214e-02	3.143e-01
## Mastogloia.grevillei	0.0357385	-8.280e-03	5.077e-02
## Meridion.circulare	0.2185106	-2.141e-01	-4.751e-02
## Meridion.anceps	0.0252712	6.634e-02	2.589e-02
## Meridion.circulare.var..constrictum	0.0120090	1.298e-01	4.142e-02
## Navicula.pseudoventralis	0.0007553	2.095e-02	1.221e-02
## Navicula.wendlingii	0.0089062	3.263e-02	-1.730e-02

## Navicula.phyllepta	0.0107626	2.622e-02	2.904e-03
## Navicula.radiosa	0.2724033	-7.172e-02	-1.915e-02
## Navicula.angusta	-0.0744911	-2.042e-02	-7.218e-03
## Navicula.cataracta.rheni	0.0007193	1.202e-02	5.197e-03
## Navicula.wallace	0.0138071	4.219e-02	2.485e-02
## Navicula.cryptofallax	0.0038090	1.155e-02	-3.041e-03
## Navicula.cryptocephala	0.0195447	1.521e-02	1.482e-02
## Navicula.erifuga	0.1943356	3.500e-02	-6.149e-03
## Navicula.exilis	0.5657171	-4.079e-02	-2.408e-02
## Navicula.simulata	0.0387835	3.195e-03	-6.537e-03
## Navicula.brockmanni	0.0073372	1.841e-03	3.625e-04
## Navicula.medioconvexa	0.0267011	1.293e-02	-2.042e-03
## Navicula.glomus	0.0084445	-2.132e-02	-1.033e-02
## Navicula.vilaplanii	0.0693123	-2.564e-02	1.614e-02
## Navicula.tridentula	0.0107962	4.427e-02	6.807e-03
## Navicula.associata	0.0170761	-2.964e-03	-9.803e-03
## Navicula.reichardtiana	0.0071254	-1.001e-02	2.721e-03
## Naviculadicta.cf..difficillima	0.0109898	6.408e-02	1.086e-02
## Naviculadicta.multiconfusa	-0.0058854	2.033e-02	9.957e-03
## Naviculadicta.sp.N.1	-0.0118145	-1.030e-02	-4.932e-03
## Neidium.alpinum	-0.0272114	3.018e-03	1.937e-02
## Neidium.bisulcatum	-0.0206125	-9.704e-03	-1.720e-03
## Neidium.fossum	-0.0089255	-1.661e-02	5.984e-03
## Neidium.affine.var..humerus	0.0097790	-1.474e-03	-1.861e-03
## Neidium.productum	0.0039200	3.121e-03	3.609e-03
## Neidium.sp2	-0.0571412	-7.982e-03	1.562e-02
## Neidium.ampliatum	-0.0103073	-3.331e-03	2.467e-04
## Neidiomorpha.binodiformis	0.0237646	3.165e-04	3.814e-02
## Nitzschia.acidoclinata	0.0043002	1.349e-01	-6.968e-02
## Nitzschia.palea.var..tenuirostris	-0.0202767	5.729e-03	-5.574e-02
## Nitzschia.perminuta	0.1340756	-4.423e-02	6.377e-03
## Nitzschia.gracilis	0.0004903	2.373e-02	7.498e-04
## Nitzschia.bryophila	-0.0357005	8.806e-02	4.117e-02
## Nitzschia.paleacea	0.1594843	2.186e-01	4.815e-02
## Nitzschia.cf..inconspicua	0.0065951	7.245e-03	4.473e-03
## Nitzschia.dissipata	-0.0036348	3.564e-02	-3.442e-03
## Nitzschia.dissipata.var..media	0.0250594	4.813e-03	4.455e-03
## Nitzschia.sinuata	0.1637370	-9.323e-02	2.425e-02
## Nitzschia.amphibia	0.1211567	4.310e-02	3.627e-02
## Nitzschia.bergii	0.0212725	1.721e-02	1.374e-02
## Nitzschia.fonticola	0.0077633	-1.168e-02	-5.426e-03
## Nitzschia.linearis	0.1073313	4.341e-02	-6.691e-03
## Nitzschia.palea	-0.0098469	2.490e-02	1.421e-02
## Nitzschia.perspicua	0.0117515	1.109e-02	-1.038e-02
## Nitzschia.filiformis	0.0093089	4.952e-02	-3.561e-02
## Nitzschia.soratensis	-0.0043847	1.010e-04	4.492e-03
## Nitzschia.sp.N.15	0.0170035	2.180e-02	-2.776e-03
## Nitzschia.exilis	-0.0050653	4.936e-03	-1.066e-02
## Nitzschia.capitellata	-0.0003186	-6.508e-03	-2.681e-03
## Nitzschia.sociabilis	0.0044939	1.881e-04	-1.334e-03
## Nitzschia.sigmoidea	0.0036162	1.153e-02	2.663e-03
## Nitzschia.acula	0.0150657	-1.249e-02	-7.225e-05
## Odonthidium.mesodon	0.1062342	6.989e-02	-7.731e-02
## Odonthidium.hyemale	-0.0072958	5.770e-03	-4.642e-03



## Odonthidium.neomaximum	0.0010359	1.009e-02	-3.881e-03
## Odonthidium.apiculatum	0.0079899	1.643e-02	2.699e-04
## Pinnularia.borealis	-0.0489068	1.565e-02	2.928e-02
## Pinnularia.sinistra	-0.0228720	-1.205e-02	-2.076e-02
## Pinnularia.grunowii	-0.0004506	-9.204e-03	-3.792e-03
## Pinnularia.krammeri	-0.0874705	-5.117e-03	-4.443e-02
## Pinnularia.stomatophora	0.0046517	1.050e-01	-5.915e-02
## Pinnularia.subcapitata	-0.0990612	1.019e-02	5.169e-02
## Pinnularia.subcapitata.var..elongata	-0.0184654	6.037e-02	-1.058e-02
## Pinnularia.subcapitata.var..subrostrata	-0.0166103	1.357e-03	3.875e-03
## Pinnularia.subinterrupta	-0.0114301	-2.528e-03	2.118e-03
## Pinnularia.viridis	0.0190776	2.798e-02	3.017e-03
## Pinnularia.viridiformis	0.0016871	5.886e-02	2.172e-02
## Pinnularia.divergentissima	-0.0160347	2.213e-02	2.514e-03
## Pinnularia.microstauron	-0.0750869	9.752e-03	4.308e-02
## Pinnularia.microstauron.var..nonfasciata	-0.0111321	1.994e-02	3.532e-02
## Pinnularia.microstauron.var..angusta	-0.1689063	2.113e-02	6.204e-02
## Pinnularia.subanglica	-0.0058729	-6.195e-03	-1.402e-03
## Pinnularia.gibba	-0.0128120	-6.328e-03	8.397e-03
## Pinnularia.gibbiformis	-0.0408745	-7.977e-03	-4.980e-03
## Pinnularia.marchica	-0.0071634	6.981e-03	-1.508e-02
## Pinnularia.anglica	-0.0238925	-7.439e-03	1.961e-02
## Pinnularia.rupestris	-0.0436835	2.378e-03	3.029e-03
## Pinnularia.isselana	0.0500648	1.550e-01	4.703e-02
## Pinnularia.ivaloensis	0.0113012	5.126e-02	3.732e-02
## Pinnularia.rhombarea	-0.0156064	1.018e-02	4.965e-03
## Pinnularia.brebissonii	-0.0203898	-1.114e-02	-2.026e-02
## Pinnularia.brebissonii.var..minuta	0.0356203	2.397e-02	2.613e-03
## Pinnularia.pseudogibba	-0.0266734	-8.278e-03	1.437e-02
## Pinnularia.decrescens	-0.0258572	-5.709e-03	-5.458e-03
## Pinnularia.flexuosa	-0.0132678	9.818e-03	-1.096e-03
## Pinnularia.obscura	-0.0087715	5.017e-02	3.519e-02
## Pinnularia.subcommutata.var..nonfasciata	-0.0041061	-6.598e-04	8.659e-03
## Pinnularia.frequentis	-0.0217164	-6.596e-03	-4.133e-03
## Pinnularia.schoenfelderi	-0.0078550	4.428e-04	1.809e-03
## Pinnularia.perirrorata	-0.0497630	2.935e-02	4.737e-02
## Pinnularia.schimanskii	-0.0090574	-2.108e-03	2.787e-03
## Pinnularia.divergens.var..sublinearis	-0.0692965	-2.646e-02	-3.260e-02
## Pinnularia.appendiculata.var..amaniana	-0.0378627	-1.027e-02	3.919e-03
## Pinnularia.silvatica	-0.0195198	6.744e-02	3.302e-02
## Pinnularia.kneuckeri	-0.0064680	-5.279e-03	7.510e-03
## Pinnularia.nodosa	0.0130812	6.877e-02	4.786e-04
## Placoneis.paraelginensis	0.0124733	2.867e-02	6.914e-03
## Placoneis.ignorata	0.0494801	1.072e-02	-4.501e-03
## Placoneis.abiskoensis	0.0233981	1.459e-02	1.301e-02
## Placoneis.sp2	0.0075092	-1.917e-03	-2.622e-03
## Placoneis.porifera	-0.0069690	8.029e-04	-7.064e-03
## Planothidium.lanceolatum	0.0629073	2.267e-01	4.270e-02
## Planothidium.frequentissimum	0.0599237	1.051e-01	2.452e-02
## Planothidium.victorii	0.0258800	2.765e-02	2.548e-02
## Planothidium.hinzianun	0.0188219	1.644e-02	1.755e-02
## Planothidium.bipororum	0.0048598	4.244e-03	4.531e-03
## Psammothidium.helveticum	-0.0069921	-3.880e-03	1.310e-03
## Psammothidium.subatomoides	-0.1605916	-4.440e-02	5.131e-02

## Psammothidium.bioretii	-0.0076455	-7.539e-03	-1.422e-02
## Psammothidium.lauenburgianum	0.0489320	1.947e-04	-3.675e-02
## Pseudostaurosira.microstriata	0.2213931	1.825e-02	-5.043e-02
## Pseudostaurosira.brevistriata.var..inflata	-0.0011897	1.736e-02	2.420e-04
## Pseudostaurosira.alvareziae	-0.0126238	-4.080e-03	3.021e-04
## Pseudostaurosira.pseudconstruens	-0.0159929	-1.644e-02	-4.184e-02
## Pseudostaurosira.venter	-0.0269907	3.110e-03	-2.736e-02
## Punctastriata.lancetulla	0.0050532	1.039e-02	1.707e-04
## Reimeria.sinuata	0.0030781	-2.178e-02	-2.153e-02
## Rhopalodia.gibba	0.3643313	2.418e-03	6.870e-02
## Rhopalodia.rupestris	0.0956295	1.843e-01	1.904e-02
## Sellaphora.pupula.urban	0.0095575	2.703e-02	8.973e-03
## Sellaphora.pseudopupula	0.0001332	2.135e-02	1.669e-02
## Sellaphora.blackfordensis	0.0200578	-3.419e-03	-3.685e-02
## Sellaphora.radiosa	0.0070022	2.048e-03	8.305e-03
## Sellaphora.laevissima	0.0273155	-1.845e-02	4.804e-03
## Sellaphora.stroemii	0.0163746	-1.942e-02	6.574e-03
## Sellaphora.pupula	0.0160127	-2.071e-02	5.688e-02
## Sellaphora.pupula.cf.auldreekie	-0.0002253	-4.602e-03	-1.896e-03
## Sellaphora.sp1	0.0184938	-2.910e-02	-5.649e-03
## Sellaphora.bacillum	0.0204178	-8.329e-03	-6.490e-03
## Stauroneis.gracilis	0.0034482	4.094e-03	-1.421e-02
## Stauroneis.smithii	0.0285642	7.620e-03	-1.718e-04
## Stauroneis.bovbjergii	0.0318066	5.784e-02	1.757e-02
## Stauroneis.reichardtii	-0.0009574	7.796e-03	7.973e-03
## Stauroneis.separanda	0.0246282	2.263e-03	2.183e-03
## Stauroneis.kriegeri	0.0235154	1.008e-02	1.561e-02
## Staurosira.confusa	0.2057100	-1.992e-02	-5.214e-02
## Staurosira.construens.var..venter	0.1669647	1.003e-02	-3.535e-01
## Staurosira.cf..construens.var..binodis	0.0146896	8.234e-03	1.135e-02
## Staurosirella.pinnata	0.0507922	8.742e-02	-5.195e-02
## Staurosirella.leptostauron	0.0059672	5.084e-02	-1.594e-02
## Staurosirella.oldenburgiana	0.0082876	8.074e-02	-3.105e-02
## Staurosirella.sp4	0.0010359	1.009e-02	-3.881e-03
## Steпноpterobia.delicatissima	-0.0738793	3.824e-03	4.036e-02
## Steпноpterobia.curvula	-0.0144027	6.678e-04	3.349e-03
## Surirella.linearis	-0.0298372	2.211e-03	4.531e-02
## Surirella.roba	-0.0022624	-5.539e-03	3.538e-03
## Surirella.biseriata	-0.0077654	-3.733e-04	2.892e-03
## Tabellaria.flocculosa	-0.4698307	-1.201e-01	7.714e-02
## Tabellaria.ventricosa	-0.0406648	-4.688e-03	2.558e-03
## Tabellaria.fenestrata	-0.0065393	6.373e-03	-1.377e-02
## Tetracyclus.rupestris	-0.0173353	-2.175e-03	6.834e-03
## Ulnaria.danica	0.1338102	5.722e-02	2.879e-02
## Ulnaria.ulna	0.0035732	7.347e-03	1.207e-04
##	CAP4	MDS1	MDS2
## Achnanthidium.minutisimum	0.1009844	6.349e-01	-2.128e-01
## Achnanthidium.caledonicum	0.0148235	-1.945e-02	1.965e-03
## Achnanthidium.rivulare	0.0020828	1.702e-02	-7.507e-04
## Achnanthidium.lineare	0.0346239	5.313e-02	-1.779e-02
## Achnanthidium.delmontii	0.0309494	1.029e-01	7.537e-02
## Achnanthidium.macrocephalum	-0.0086945	6.033e-03	1.315e-02
## Achnanthidium.sp7	0.0007957	-1.968e-03	-6.664e-04
## Achnanthidium.sp12	-0.0174601	1.481e-03	3.586e-03

## Achnanthidium.affine	-0.0040601	1.189e-02	-2.099e-05
## Achnanthidium.trinode	-0.0089430	-5.941e-03	-7.293e-05
## Achnanthes.lapidosa.var..lapidosa	0.0181284	3.891e-02	2.084e-02
## Achnanthes.minutissima.var..jackii	0.0094137	-4.028e-03	-1.492e-02
## Adlafia.aquaeductae	0.0412155	7.075e-03	-6.330e-02
## Adlafia.bryophila	-0.0002621	4.743e-03	-1.055e-02
## Adlafia.minuscula	-0.0132656	2.360e-02	2.125e-02
## Aneumastus.tusculus	-0.0026101	-4.673e-05	-3.802e-03
## Amphora.pediculus	-0.0158612	1.244e-01	9.451e-02
## Amphora.copulata	0.0426272	1.102e-01	-5.476e-02
## Amphora.ovalis	-0.0021221	2.785e-02	8.769e-03
## Amphora.affinis	0.0098348	7.826e-04	-2.410e-03
## Aulacoseira.alpigenica	-0.0416490	-4.454e-02	-1.336e-03
## Aulacoseira.pfaffiana	-0.0270367	1.651e-02	4.416e-02
## Aulacoseira.distans.var..humilis	-0.0034273	-7.920e-03	-8.907e-03
## Aulacoseira.granulata	-0.0027835	-4.970e-02	-1.670e-02
## Aulacoseira.tenella	0.0087008	-2.751e-03	-9.172e-03
## Aulacoseira.tethera	0.0551983	5.290e-03	-3.989e-02
## Aulacoseira.valida	0.0070802	-6.951e-03	5.014e-03
## Aulacoseira.granulata.var..angustissima	-0.0031699	2.753e-03	-5.625e-03
## Aulacoseira.ambigua	-0.0122651	-2.310e-02	2.559e-02
## Aulacoseira.sp4	-0.0057589	1.413e-03	-5.844e-03
## Brachysira.brebissonii	0.0892084	-5.529e-01	3.636e-03
## Brachysira.intermedia	0.0637191	-3.064e-01	2.777e-02
## Brachysira.neoexilis	-0.0246814	-2.279e-01	1.066e-01
## Brachysira.neglectissima	0.0560493	-1.097e-01	-6.302e-02
## Brachysira.serians	0.0183423	-3.599e-02	1.289e-01
## Caloneis.silicula	0.0072829	4.452e-03	-3.765e-03
## Caloneis.tenuis	-0.0188958	-1.053e-01	-4.499e-01
## Caloneis.bacilum	0.0263351	4.717e-02	-1.032e-01
## Caloneis.bacilum.var..inflata	0.0033118	1.027e-02	-4.590e-03
## Caloneis.ondulata	0.0072036	-4.682e-02	-1.064e-02
## Caloneis.lancettula	0.0173203	2.882e-02	-3.967e-02
## Caloneis.fusus	-0.0129304	-1.333e-02	-1.494e-02
## Caloneis.alpestris	0.0013245	3.156e-02	-2.845e-02
## Caloneis.aerophila	0.0106956	-1.062e-02	-1.318e-02
## Chamaepinnularia.mediocris	0.0340240	2.904e-02	1.542e-01
## Chamaepinnularia.hassiacca	0.0025818	-7.595e-03	-3.927e-02
## Chamaepinnularia.submusciola	0.0120872	9.352e-03	-1.938e-02
## Chamaepinnularia.soehrensii	0.0106163	2.659e-02	3.113e-02
## Chamaepinnularia.begerii	-0.0248053	-9.277e-02	7.239e-02
## Chamaepinnularia.musciola	0.0083892	-4.585e-02	-8.918e-02
## Chamaepinnularia.evanida	0.0245959	-2.548e-02	-4.658e-02
## Cavinula.cocconeiformis	0.0161579	2.653e-02	-1.590e-03
## Cocconeis.placentula	0.0049762	1.261e-02	1.532e-02
## Cocconeis.lineata	0.0053258	2.568e-02	1.345e-02
## Cocconeis.sp3	0.0124133	1.416e-02	1.011e-02
## Cyclotella.antiqua	0.0041930	5.069e-03	-6.539e-03
## Cyclostephanos.dubius	0.0044089	-5.361e-03	2.189e-03
## Cymbella.subcuspidata	-0.0057695	-1.398e-03	1.330e-02
## Cymbella.parva	0.0304514	4.945e-02	1.266e-01
## Cymbella.affinis.var..procera	0.0002074	6.973e-03	1.308e-02
## Cymbella.lancetulla	0.0385873	5.149e-03	8.900e-02
## Cymbella.excisa	-0.0059875	1.682e-02	-6.586e-05

## Cymbella.cymbiformis	0.0019939	7.672e-03	1.260e-02
## Cymbella.cleve.eulerge	0.0043080	5.457e-03	1.399e-02
## Cymbella.lange.bertalotii	0.0062177	3.482e-03	-1.983e-03
## Cymbella.aspera	0.0042342	1.004e-02	-3.376e-02
## Cymbella.romboidea	0.0031803	-3.055e-04	4.485e-03
## Cymbopleura.acuta	-0.0029023	-1.265e-02	6.864e-03
## Cymbopleura.naviculiformis	-0.0139000	1.449e-02	-3.137e-02
## Cymbopleura.incerta	-0.0322015	5.494e-02	-1.139e-01
## Cymbopleura.pyrenaica	0.0100808	1.499e-02	1.945e-02
## Cymbopleura.subaequalis	-0.1050895	-1.178e-01	-1.393e-01
## Cymbopleura.similiformis	0.0174661	7.026e-03	-4.773e-04
## Cymbopleura.florentiniformis	0.0022847	-1.304e-03	2.575e-03
## Cymbopleura.laponica	-0.0308178	-2.058e-02	4.497e-03
## Cymbopleura.hybrida	0.0060563	2.185e-02	-1.613e-02
## Cymbopleura.sp3	0.0278030	2.561e-03	3.184e-02
## Cymbopleura.incerta.var..grunowii	-0.0012065	-9.012e-03	-6.275e-03
## Cymbopleura.sublanceolata	0.0014270	-2.408e-03	-4.088e-03
## Cymbopleura.valaiseana	-0.0006061	1.853e-03	-1.444e-05
## Denticula.tenuis.var..frigida	0.0360574	3.522e-02	2.448e-02
## Diploneis.oculata	0.0028679	5.233e-02	-1.043e-01
## Diploneis.parma	-0.0127879	-3.654e-03	-2.310e-02
## Diploneis.krammeri	0.0013011	5.884e-02	-7.726e-02
## Diploneis.oblongella	-0.1308776	1.878e-02	-1.435e-01
## Diploneis.petersenii	-0.0039916	1.546e-02	-2.998e-02
## Diploneis.separanda	-0.0146202	3.591e-03	-1.472e-02
## Diploneis.minuta	-0.0192396	1.336e-02	-2.583e-02
## Encyonema.neogracile	0.0304122	-2.845e-01	-9.810e-02
## Encyonema.perpusilum	0.1308728	-1.352e-01	-2.065e-01
## Encyonema.silesiacum	0.0396646	6.721e-02	-1.053e-02
## Encyonema.silesiacum.var..altensis	0.0249948	3.076e-02	3.457e-02
## Encyonema.ventricosum	0.0137884	5.878e-02	5.768e-02
## Encyonema.minutum	-0.0609502	-7.853e-03	-3.225e-02
## Encyonema.lange.bertalotii	0.0389534	-3.151e-02	-1.971e-01
## Encyonema.vulgare	-0.0276564	-7.426e-03	-7.827e-02
## Encyonema.hebredicum	-0.0004125	2.263e-02	4.654e-02
## Encyonema.procerum	-0.0033859	6.652e-03	2.602e-03
## Encyonema.sp4	-0.0021172	6.719e-02	-6.162e-02
## Encyonema.sp5	0.0143899	-9.616e-03	-3.253e-02
## Encyonema.sp13	0.0099688	-7.075e-03	-1.546e-02
## Encyonema.gaeumannii	-0.0318766	-1.702e-02	2.335e-03
## Encyonema.alpinum	0.0102574	-3.530e-02	-2.105e-02
## Encyonema.perminutum	0.0136776	-4.005e-02	3.965e-02
## Encyonema.sp20	-0.0083950	-1.940e-02	-2.182e-02
## Encyonema.sp22	0.0133091	-1.368e-03	2.568e-02
## Encyonopsis.cesatii	0.0619856	6.174e-03	-9.311e-02
## Encyonopsis.krammeri	-0.1069296	2.352e-03	-5.765e-02
## Encyonopsis.subminuta	-0.1026932	8.852e-05	-1.005e-01
## Encyonopsis.minuta	-0.0008938	3.422e-03	-1.410e-03
## Encyonopsis.sp1	-0.0225207	3.126e-03	-1.803e-02
## Encyonopsis.sp2	0.0103496	-8.644e-03	2.366e-03
## Encyonopsis.sp4	-0.0228522	-2.756e-02	8.498e-03
## Encyonopsis.sp14	-0.0033859	6.652e-03	2.602e-03
## Encyonopsis.sp15	0.0012202	4.204e-02	3.189e-02
## Eolimna.minima	0.1979454	4.883e-01	-1.123e-01

## Eolimna.sp1	0.0281719	7.225e-02	-3.447e-02
## Eolimna.sp2	0.0121898	2.001e-02	9.409e-03
## Eolimna.sp.N.5	0.0040228	4.394e-03	5.235e-03
## Epithemia.argus	-0.1606460	-1.228e-01	-1.508e-02
## Epithemia.frickei	0.0032311	-1.844e-03	3.642e-03
## Eunotia.implicata	0.0020029	-1.311e-02	-8.408e-03
## Eunotia.naegelii	-0.0528897	-1.594e-01	7.884e-02
## Eunotia.bilunaris	-0.0359491	-4.546e-02	-1.888e-02
## Eunotia.botiliformis	-0.0051724	-1.241e-02	2.690e-02
## Eunotia.cataractarum	-0.0251810	1.027e-02	-9.440e-03
## Eunotia.exigua	0.0091770	-1.083e-01	1.909e-02
## Eunotia.faba	0.0738462	-7.088e-02	8.159e-03
## Eunotia.fallax	-0.1032709	1.904e-02	2.974e-01
## Eunotia.incisa	0.0633466	-7.468e-01	-3.975e-01
## Eunotia.borealpina	0.0487974	-2.556e-01	-2.012e-01
## Eunotia.intermedia	0.0185404	-1.355e-02	1.100e-02
## Eunotia.nymanniana	-0.0159585	-1.092e-01	2.751e-01
## Eunotia.soleirolii	0.0139122	-1.935e-02	7.108e-03
## Eunotia.minor	0.0199641	-1.331e-01	-1.069e-01
## Eunotia.metensiae	0.0082155	8.864e-03	-3.005e-03
## Eunotia.neofallax	0.0003264	-3.254e-02	-4.866e-02
## Eunotia.rhomboidea	0.1446609	-1.883e-01	6.615e-02
## Eunotia.gracilis	-0.0113435	-1.210e-01	3.410e-03
## Eunotia.subarcuatoides	0.0452833	-4.773e-02	3.601e-02
## Eunotia.tenella	-0.0131542	-1.460e-01	1.255e-02
## Eunotia.arcus	-0.0450283	-1.545e-02	-9.710e-02
## Eunotia.arcubus	0.0007498	-1.263e-02	-1.556e-02
## Eunotia.palatia	0.0125932	-3.840e-02	-7.248e-02
## Eunotia.trinacria	-0.0136425	-3.142e-02	3.361e-02
## Eunotia.curtagrunowii	-0.0029937	8.412e-03	-3.293e-05
## Eunotia.mayamae	-0.0386664	-3.380e-02	1.849e-02
## Eunotia.neocompacta.var..vixcompacta	0.0011864	-2.749e-03	-2.989e-04
## Eunotia.groenlandica	-0.0191905	-1.824e-01	3.713e-02
## Eunotia.major	0.0119815	-9.207e-03	-1.467e-02
## Eunotia.paludosa	-0.0623364	-2.591e-02	1.502e-01
## Eunotia.superpaludosa	0.0178440	-8.029e-02	-8.676e-02
## Eunotia.lapponica	0.0086849	-2.363e-03	1.445e-01
## Eunotia.panda	-0.0191060	-1.092e-01	8.678e-03
## Eunotia.ninae	0.0043590	-6.494e-03	2.042e-03
## Eunotia.diodon	0.0126571	-2.655e-02	-4.598e-03
## Eunotia.islandica	0.0014163	-9.271e-03	-5.945e-03
## Eunotia.diadema	-0.0044355	4.050e-03	-9.110e-03
## Eunotia.novaisiae.var..altopyrenaica	0.0283436	-2.367e-02	6.479e-03
## Eunotia.ursamaioris	0.0091757	-1.396e-02	-8.664e-03
## Eunotia.paerupta	0.0141319	-5.899e-02	-9.984e-03
## Eunotia.valida	0.0126509	-3.024e-02	2.789e-03
## Eunotia.glacialis	0.0014270	-2.408e-03	-4.088e-03
## Eunotia.meisteri	0.0026899	-2.026e-02	2.302e-02
## Eucocconeis.laevis	-0.0255796	7.878e-02	-9.898e-03
## Eucocconeis.alpestris	-0.0098239	-1.071e-02	1.201e-04
## Eucocconeis.flexella	0.0033100	7.451e-03	8.244e-03
## Fallacia.vitrea	-0.0006895	1.180e-02	1.335e-01
## Fallacia.minuscula	0.0625095	2.536e-02	6.602e-02
## Fallacia.insoabilis	0.0026010	1.451e-02	5.875e-03

## Fallacia.sp3	-0.0050627	-3.070e-02	1.126e-02
## Fragilaria.gracilis	0.1036432	-1.039e-02	-6.827e-02
## Fragilaria.famelica	-0.0218134	1.171e-02	7.314e-03
## Fragilaria.alpestris	-0.0325245	5.411e-02	4.965e-02
## Fragilaria.capucina	0.0065162	3.973e-03	-2.095e-02
## Fragilaria.rumpens	0.0067187	-3.897e-03	-2.868e-02
## Fragilaria.rinoi	-0.0208614	4.608e-02	3.938e-02
## Fragilaria.capucina.var.gracilis	-0.0172011	1.535e-02	2.281e-02
## Fragilaria.acidoclinata	-0.0017142	3.064e-03	4.416e-03
## Fragilaria.austriaca	-0.0021169	5.948e-03	-2.328e-05
## Fragilaria.nevadensis	0.0165378	1.034e-02	4.279e-04
## Fragilaria.oldenburgiana	-0.0041171	2.119e-03	-5.724e-03
## Fragilaria.sp	-0.0070529	-9.271e-03	2.087e-02
## Fragilaria.exiguiformis	-0.0009193	-2.088e-02	7.978e-04
## Fragilariforma.visrescens	-0.1447318	5.320e-02	-2.213e-02
## Fragilariforma.bicapitata	-0.0222069	2.753e-02	2.417e-02
## Frustulia.crassinervia	0.0548194	-3.336e-01	-4.100e-02
## Frustulia.vulgaris	0.0036752	1.473e-02	2.596e-02
## Frustulia.saxonica	0.0526345	-1.111e-01	3.162e-01
## Gomphonema.cymbelliclinum	0.0851035	9.708e-02	3.273e-02
## Gomphonema.gracile	0.0435236	-3.701e-01	-2.787e-01
## Gomphonema.clavatum	-0.0104681	9.493e-03	2.663e-03
## Gomphonema.subclavatum	0.0246050	7.249e-02	-4.547e-02
## Gomphonema.parvulum	0.0411983	2.304e-01	-1.637e-01
## Gomphonema.lateripunctatum	-0.0818305	3.756e-02	-8.515e-02
## Gomphonema.micropus	0.0301240	8.050e-02	5.026e-02
## Gomphonema.exilissimum	-0.0116520	3.941e-02	1.208e-02
## Gomphonema.varioireduncum	0.0234108	2.837e-02	2.746e-02
## Gomphonema.drutelingense	-0.0255297	9.268e-02	-4.657e-02
## Gomphonema.productum	0.0029799	1.274e-02	7.873e-03
## Gomphonema.acidoclinatum	-0.0194074	6.234e-02	1.445e-02
## Gomphonema.angustivalva	-0.0097124	9.717e-03	5.512e-03
## Gomphonema.subtile	0.0077901	-7.482e-04	1.099e-02
## Gomphonema.angustus	0.0043966	2.462e-03	-1.402e-03
## Gomphonema.olivaceum.var..calcerum	0.0579879	5.619e-02	-5.147e-02
## Gomphonema.herbridense	-0.0124931	-1.368e-02	4.336e-03
## Gomphonema.variscohercynicum	-0.0251895	-6.060e-04	-7.814e-03
## Gomphonema.sp6	-0.0520116	8.747e-03	-1.823e-02
## Gomphonema.sp23	0.0090784	2.653e-02	-2.099e-02
## Gomphonema.sp27	-0.0036216	3.307e-03	-7.438e-03
## Gomphonema.sp28	-0.0075711	1.487e-02	5.819e-03
## Gomphonema.lagenula	0.0044459	1.314e-02	1.918e-02
## Greissleria.acceptata	-0.0031542	-4.146e-03	9.332e-03
## Halamphora.montana	-0.0028587	-4.033e-03	1.290e-03
## Hannaea.arcus	0.0090199	-3.284e-03	-2.420e-03
## Hantzschia.amphioxys	-0.0022736	3.450e-04	9.810e-03
## Hantzschia.calcifuga	0.0104962	2.163e-02	7.731e-03
## Hantzschia.abundans	-0.0113718	2.714e-03	-3.360e-03
## Humidophila.perpusilla	0.0252293	4.193e-02	1.030e-02
## Karayevia	0.0046941	9.674e-03	3.457e-03
## Kobayasiella.micropunctata	-0.2197208	-1.212e-01	2.914e-01
## Kobayasiella.sp	0.0193818	3.147e-02	1.394e-01
## Luticula.mutica	0.0069272	2.542e-02	1.147e-02
## Mastogloia.lacustris	-0.3567568	-5.023e-02	-1.671e-02

## Mastogloia.grevillei	-0.0560398	-1.296e-02	-7.119e-03
## Meridion.circulare	0.0806978	1.672e-01	1.993e-01
## Meridion.anceps	0.0026390	6.551e-02	4.231e-02
## Meridion.circulare.var..constrictum	0.0311100	7.479e-02	2.433e-02
## Navicula.pseudoventralis	0.0225071	3.830e-02	1.561e-02
## Navicula.wendlingii	-0.0106454	1.771e-02	-1.721e-02
## Navicula.phyllepta	-0.0067307	8.522e-04	1.660e-02
## Navicula.radiosa	0.0463060	9.873e-02	-4.513e-02
## Navicula.angusta	0.0069316	-3.654e-02	-6.321e-02
## Navicula.cataracta.rheni	0.0055514	6.331e-03	4.520e-03
## Navicula.wallace	0.0104888	2.425e-02	2.277e-02
## Navicula.cryptofallax	-0.0063698	-5.880e-03	-1.026e-02
## Navicula.cryptocephala	-0.0129760	-1.344e-02	1.038e-02
## Navicula.erifuga	0.0386696	9.106e-02	-2.589e-02
## Navicula.exilis	0.0462368	2.414e-01	-2.034e-01
## Navicula.simulata	0.0049091	5.820e-02	2.458e-03
## Navicula.brockmanni	0.0069542	5.534e-04	-1.704e-03
## Navicula.medioconvexa	-0.0052001	2.430e-02	-2.453e-02
## Navicula.glomus	0.0083184	2.627e-02	-1.398e-02
## Navicula.vilaplanii	-0.0157456	-9.853e-03	4.940e-03
## Navicula.tridentula	0.0039345	-2.611e-03	1.219e-02
## Navicula.associata	0.0092723	1.051e-02	-1.645e-02
## Navicula.reichardtiana	0.0098268	6.455e-03	3.438e-03
## Naviculadicta.cf..difficillima	0.0100345	8.514e-03	3.137e-02
## Naviculadicta.multiconfusa	0.0066384	1.368e-02	4.890e-03
## Naviculadicta.sp.N.1	0.0030061	-1.071e-02	-9.107e-03
## Neidium.alpinum	0.0207852	1.893e-02	2.256e-02
## Neidium.bisulcatum	-0.0053855	1.407e-02	1.167e-02
## Neidium.fossum	0.0143480	-1.208e-02	-3.577e-03
## Neidium.affine.var..humerus	0.0081281	8.400e-03	-9.949e-03
## Neidium.productum	0.0033118	1.027e-02	-4.590e-03
## Neidium.sp2	-0.0022574	3.265e-02	5.957e-02
## Neidium.ampliatum	-0.0001203	-1.707e-04	1.894e-03
## Neidiomorpha.binodiformis	-0.0274137	1.461e-02	-6.533e-03
## Nitzschia.acidoclinata	0.0319687	-5.956e-02	-9.891e-02
## Nitzschia.palea.var..tenuirostris	-0.0313590	-1.085e-02	4.429e-02
## Nitzschia.perminuta	0.0838986	1.242e-01	-1.505e-01
## Nitzschia.gracilis	0.0003299	-2.017e-02	1.919e-02
## Nitzschia.bryophila	0.0434806	7.553e-02	-7.666e-03
## Nitzschia.paleacea	0.0307105	-9.461e-02	-1.076e-01
## Nitzschia.cf..inconspicua	0.0075048	1.534e-02	-1.113e-02
## Nitzschia.dissipata	0.0005797	-2.457e-02	1.563e-02
## Nitzschia.dissipata.var..media	-0.0032106	-2.698e-03	-4.062e-02
## Nitzschia.sinuata	-0.0247438	5.342e-03	-1.320e-02
## Nitzschia.amphibia	-0.0703997	2.542e-03	-7.921e-02
## Nitzschia.bergii	0.0329746	2.007e-02	9.526e-03
## Nitzschia.fonticola	0.0011142	4.105e-03	3.619e-03
## Nitzschia.linearis	0.0356809	4.542e-02	-6.918e-03
## Nitzschia.palea	0.0096134	1.842e-02	6.980e-03
## Nitzschia.perspicua	0.0230915	6.479e-02	-2.769e-02
## Nitzschia.filiformis	0.0465377	4.429e-02	-1.069e-02
## Nitzschia.soratensis	0.0043973	-6.916e-03	-1.916e-03
## Nitzschia.sp.N.15	0.0022098	7.909e-03	-1.491e-02
## Nitzschia.exilis	-0.0060115	-6.429e-03	-1.928e-04

## Nitzschia.capitellata	0.0021538	1.393e-02	-8.164e-03
## Nitzschia.sociabilis	0.0040495	2.573e-03	-3.312e-03
## Nitzschia.sigmoidea	0.0007765	-5.847e-03	6.646e-03
## Nitzschia.acula	0.0045695	-2.607e-03	5.151e-03
## Odonthidium.mesodon	0.0972307	1.668e-01	2.950e-01
## Odonthidium.hyemale	-0.0124431	1.009e-02	7.552e-03
## Odonthidium.neomaximum	-0.0019831	6.313e-03	5.279e-03
## Odonthidium.apiculatum	-0.0083933	7.490e-03	1.113e-02
## Pinnularia.borealis	0.0211015	6.454e-03	-1.617e-03
## Pinnularia.sinistra	-0.0247521	7.136e-04	-1.699e-03
## Pinnularia.grunowii	0.0030460	1.970e-02	-1.155e-02
## Pinnularia.krammeri	-0.0346633	-2.875e-02	1.732e-02
## Pinnularia.stomatophora	-0.0276228	-6.216e-02	-5.353e-02
## Pinnularia.subcapitata	0.0357987	1.903e-03	1.719e-02
## Pinnularia.subcapitata.var..elongata	-0.0324872	-3.808e-02	5.134e-02
## Pinnularia.subcapitata.var..subrostrata	0.0006802	9.206e-03	2.474e-02
## Pinnularia.subinterrupta	0.0001796	6.038e-03	1.132e-02
## Pinnularia.viridis	0.0061264	-1.897e-02	8.163e-03
## Pinnularia.viridiformis	-0.0630016	3.525e-02	-5.480e-02
## Pinnularia.divergentissima	-0.0093154	9.153e-03	-1.982e-02
## Pinnularia.microstauron	0.0335047	3.458e-02	2.738e-02
## Pinnularia.microstauron.var..nonfasciata	-0.0327595	6.407e-02	-4.032e-02
## Pinnularia.microstauron.var..angusta	-0.0451834	-7.645e-02	1.170e-01
## Pinnularia.subanglica	0.0035227	1.090e-02	-8.362e-03
## Pinnularia.gibba	0.0058565	-6.910e-04	-2.092e-04
## Pinnularia.gibbiformis	-0.0082304	-2.272e-03	1.482e-02
## Pinnularia.marchica	-0.0085016	-9.091e-03	-2.726e-04
## Pinnularia.anglica	0.0150647	-1.515e-02	5.967e-03
## Pinnularia.rupestris	-0.0069296	2.794e-02	5.431e-02
## Pinnularia.isselana	0.0166801	2.992e-02	3.117e-04
## Pinnularia.ivaloensis	-0.0300571	-2.286e-02	1.849e-02
## Pinnularia.rhombarea	0.0012729	9.757e-03	1.133e-02
## Pinnularia.brebissonii	-0.0115507	2.218e-02	-2.516e-03
## Pinnularia.brebissonii.var..minuta	-0.0382435	9.705e-03	2.283e-02
## Pinnularia.pseudogibba	0.0060571	5.962e-03	1.183e-02
## Pinnularia.decrescens	0.0012732	-1.145e-02	2.910e-02
## Pinnularia.flexuosa	0.0025705	-2.320e-02	-4.195e-04
## Pinnularia.obscura	0.0274045	1.723e-02	4.292e-02
## Pinnularia.subcommutata.var..nonfasciata	-0.0048140	6.622e-03	-2.567e-03
## Pinnularia.frequentis	0.0098764	8.786e-03	-2.603e-02
## Pinnularia.schoenfelderii	0.0019359	-4.286e-03	-2.801e-04
## Pinnularia.perirrorata	0.0364560	-2.023e-02	4.556e-02
## Pinnularia.schimanskii	-0.0008938	3.422e-03	-1.410e-03
## Pinnularia.divergens.var..sublinearis	-0.0278173	-4.471e-02	-3.401e-02
## Pinnularia.appendiculata.var..amaniana	-0.0009797	-1.692e-02	-3.381e-02
## Pinnularia.silvatica	0.0220171	4.537e-02	1.622e-02
## Pinnularia.kneuckeri	0.0070802	-6.951e-03	5.014e-03
## Pinnularia.nodosa	0.0017261	-4.024e-03	-1.216e-02
## Placoneis.paraelginensis	0.0035278	2.980e-02	-5.028e-03
## Placoneis.ignorata	0.0344179	2.921e-02	-1.776e-02
## Placoneis.abiskoensis	-0.0227072	3.967e-05	-1.673e-02
## Placoneis.sp2	0.0078799	1.088e-02	-6.114e-03
## Placoneis.porifera	-0.0068677	6.871e-03	3.897e-03
## Planothidium.lanceolatum	0.0699712	1.490e-01	1.030e-01



## Planothidium.frequentissimum	0.0273612	4.980e-02	6.343e-02
## Planothidium.victorii	0.0104969	3.415e-02	-8.294e-03
## Planothidium.hinzianun	-0.0066389	1.187e-02	1.710e-02
## Planothidium.bipororum	-0.0017142	3.064e-03	4.416e-03
## Psammothidium.helveticum	0.0017478	-2.949e-03	-5.006e-03
## Psammothidium.subatomoides	0.0598047	-2.044e-02	2.772e-02
## Psammothidium.bioretii	-0.0056008	1.574e-02	-6.160e-05
## Psammothidium.lauenburgianum	0.0252974	-1.985e-02	-8.903e-02
## Pseudostaurosira.microstriata	-0.0373407	7.835e-02	-9.480e-02
## Pseudostaurosira.brevistriata.var..inflata	-0.0091202	1.146e-02	5.776e-03
## Pseudostaurosira.alvareziae	-0.0001474	-2.091e-04	2.320e-03
## Pseudostaurosira.pseudconstruens	-0.0220347	2.415e-02	6.287e-03
## Pseudostaurosira.venter	-0.0265984	2.661e-02	1.509e-02
## Punctastriata.lancetulla	-0.0053084	4.737e-03	7.041e-03
## Reimeria.sinuata	-0.0271792	3.078e-02	3.912e-02
## Rhopalodia.gibba	-0.1291798	-9.407e-02	-1.346e-01
## Rhopalodia.rupestris	-0.1029585	-1.055e-01	-2.431e-01
## Sellaphora.pupula.urban	0.0199343	2.066e-02	-2.513e-03
## Sellaphora.pseudopupula	0.0109183	3.996e-02	1.593e-02
## Sellaphora.blackfordensis	0.0030897	3.164e-02	-2.961e-02
## Sellaphora.radiosa	-0.0055441	4.527e-03	-1.637e-03
## Sellaphora.laevissima	0.0057259	3.458e-02	-3.066e-05
## Sellaphora.stroemii	-0.0019198	9.410e-03	1.150e-02
## Sellaphora.pupula	-0.0622375	1.919e-03	-3.705e-03
## Sellaphora.pupula.cf.auldreekie	0.0015230	9.848e-03	-5.773e-03
## Sellaphora.sp1	0.0022517	4.362e-03	-1.012e-03
## Sellaphora.bacillum	0.0098587	-1.199e-02	4.894e-03
## Stauroneis.gracilis	-0.0054487	1.412e-02	-6.928e-04
## Stauroneis.smithii	0.0091445	5.710e-02	-4.078e-02
## Stauroneis.bovbjergii	0.0136502	4.652e-02	-1.287e-02
## Stauroneis.reichardtii	0.0057830	-1.076e-02	1.019e-02
## Stauroneis.separanda	0.0161365	1.049e-02	-3.836e-03
## Stauroneis.kriegeri	-0.0178242	4.692e-03	-1.241e-02
## Staurosira.confusa	0.0833156	1.083e-01	-4.892e-02
## Staurosira.construens.var..venter	0.0721717	7.119e-01	-4.441e-01
## Staurosira.cf..construens.var..binodis	-0.0184271	-1.134e-04	-1.661e-02
## Staurosirella.pinnata	-0.0529851	1.373e-01	1.470e-02
## Staurosirella.leptostauron	-0.0154023	5.460e-02	2.147e-02
## Staurosirella.oldenburgiana	-0.0158650	5.050e-02	4.223e-02
## Staurosirella.sp4	-0.0019831	6.313e-03	5.279e-03
## Stepanopteroberia.delicatissima	0.0247233	-1.606e-02	5.424e-02
## Stepanopteroberia.curvula	0.0023141	-3.310e-03	1.841e-02
## Surirella.linearis	0.0067616	8.862e-03	1.608e-02
## Surirella.roba	0.0073183	-6.112e-03	1.673e-03
## Surirella.biseriata	-0.0007889	4.814e-03	8.548e-03
## Tabellaria.flocculosa	0.1096533	-3.774e-01	1.983e-02
## Tabellaria.ventricosa	0.0058342	-1.879e-02	1.220e-02
## Tabellaria.fenestrata	-0.0077608	-8.299e-03	-2.489e-04
## Tetracyclus.rupestris	0.0080454	-1.031e-02	-8.173e-03
## Ulnaria.danica	0.0353092	5.790e-02	5.898e-03
## Ulnaria.ulna	-0.0037536	3.350e-03	4.979e-03
##			
##			
## Site scores (weighted sums of species scores)			

##		CAP1	CAP2	CAP3	CAP4	MDS1	MDS2
##	BAI01	-0.87401	-0.29982	-0.072676	0.69386	-0.253855	-0.409982
##	BAI04	-1.05407	-0.53636	0.848129	0.44281	-0.017995	0.190002
##	BAI06	-0.94961	-0.36757	0.216179	0.05076	0.519755	0.927336
##	BAI07	-0.87075	0.09519	-0.129325	0.13519	0.360704	-0.141406
##	BPU04	-0.20829	2.58228	0.788654	0.05274	-0.871736	0.942653
##	BSC03	0.92926	-1.45133	2.256423	-1.86282	-0.885722	-0.010345
##	BSC04	0.80853	0.87552	1.429616	-0.57696	-0.545068	0.098636
##	BSC05	0.88567	-1.03157	2.630528	-3.59975	0.086267	-0.158439
##	BV01	-1.24628	-1.06396	1.574258	1.31488	-0.968096	0.289696
##	BV02	-0.78909	-0.76817	-0.220622	1.94224	-0.527407	-1.096393
##	BValp01	-1.12720	-0.92037	0.795963	1.23470	-0.072846	-0.020977
##	BValp02	-0.93503	-0.78817	0.897028	0.97449	-0.911247	0.237265
##	BValp03	-1.17908	-0.50115	1.617889	0.94369	-0.732728	0.502892
##	CBR02	0.41432	2.98803	0.757075	1.70428	0.943794	0.641044
##	CBR03	0.06211	2.28185	0.653496	1.71310	1.442211	0.490398
##	CON02	-0.72735	0.47432	0.209137	-0.05062	-1.382130	-0.843262
##	CON03	-0.47537	-0.01601	-0.142070	-0.80283	-1.030837	-1.105952
##	CON06	-1.08159	0.11201	0.189537	-0.79241	-1.455517	0.419849
##	CR01	-0.81773	-0.04124	-0.014198	-1.42668	0.528563	-0.265130
##	CR06	-0.86004	0.05242	-1.365233	-1.86070	0.740444	0.314611
##	CR09	-0.63301	0.32281	-2.789819	-1.88652	1.024394	0.552805
##	ERT05	0.78923	-1.73018	0.869705	-0.13238	0.701223	0.261009
##	ES03	0.87069	-0.59279	-1.640272	-0.20901	0.522311	-0.593333
##	ES10	0.93532	-0.64631	-1.567506	-0.71076	-0.002463	-0.190669
##	FRE03	0.77651	0.45657	0.015917	1.28379	0.456796	0.626290
##	GMG01	0.22197	1.32494	-2.281662	-0.37163	0.499396	0.706154
##	GMG06	0.42861	2.10658	-1.012327	0.45303	0.941136	0.748707
##	HOR01	0.93123	0.32952	-0.534047	0.33378	0.082497	-0.241665
##	HOR04	-1.01820	-0.28440	1.082180	0.02700	-1.031014	-0.271755
##	HOR05	-0.69137	0.27007	0.529590	0.43226	-0.245153	-0.863661
##	HOR06	-0.09585	-0.11487	-0.275246	-0.38772	0.418815	-2.100926
##	LLA07	0.73118	-1.10333	-0.855568	0.83279	0.152243	1.549072
##	MA_alp01	-0.41689	-0.36278	-0.189231	0.28929	-0.425182	0.129374
##	MA_alp02	0.03622	-0.03209	-2.198717	0.58597	0.886794	-0.003303
##	MAR01	0.82250	-0.58149	-1.148773	0.30098	1.468160	-0.818767
##	MAR02	0.93924	-0.25902	-0.599448	0.30712	0.923027	-0.630025
##	MAR03	0.82143	0.67547	-1.214494	0.18191	1.026251	-0.980965
##	MAR04	0.94883	-1.25495	-0.533233	0.48902	-0.045539	0.636156
##	MAR05	1.02375	-1.13575	0.138053	0.66673	0.183513	-0.099442
##	MAR06	0.76235	-0.74866	-0.877015	-0.27833	0.195361	-0.001448
##	MAR09	0.99484	0.06559	-0.342004	0.37075	0.383607	-0.469806
##	MAR10	1.07703	-1.56516	1.047017	-0.17422	-0.194345	0.365294
##	MT02	0.85288	1.31899	1.228996	-0.38449	0.499991	-1.058538
##	MTL03	0.80107	-1.18003	-1.337822	1.19608	0.680488	0.344813
##	MUN02	-1.14493	-0.40503	0.131821	0.27987	-0.899363	-0.200745
##	MUN03	-1.11593	0.04411	0.223913	-0.65643	-0.578090	-0.450198
##	PA01	0.99838	0.62475	-0.418520	1.09457	0.755767	-0.927530
##	PA02	0.96891	-0.29273	-0.824985	0.54779	1.082601	-0.460399
##	PA05	1.04550	0.58133	-0.043125	1.45238	0.699502	-0.237219
##	RAT07	-0.23442	0.32347	-2.237779	1.02952	0.348580	-0.746035
##	SO03	-1.02164	-0.52294	1.805362	-0.39656	-0.027875	1.247895
##	SO04	-1.16016	-0.40772	0.860248	-0.21664	-0.454019	0.846195

```

## S005      -0.98016 -1.13812  1.340075  1.32751 -0.451830 -0.028094
## S008      -0.89939 -0.10946  0.843967 -1.34301  0.717703  1.212363
## S009      -0.97074  0.66933 -0.008562 -2.12221  0.613792  1.569220
## SON02      0.49388  0.70939  0.763528 -1.37520 -0.619808 -1.029349
## SON03      0.78320  0.81415 -0.466427  0.33513  0.410478 -0.797852
## T29       -0.40446 -0.56061  0.171893  0.14149 -1.180814 -1.263410
## T46       -1.21053 -0.16654 -0.301968 -0.61589 -1.886342  0.973530
## T52       -0.74169  0.48459 -1.569999 -0.63146 -0.553330 -0.015787
## T54       -0.70531  0.71933 -1.295267 -0.81751 -0.618143  1.323686
## TAU03      0.69868  1.29146  0.623602 -0.15537  0.477184 -0.164147
## TAU04      1.00123 -0.10634  0.016015 -1.51769 -0.101061 -0.712215
## TAU06      0.88209  0.54871  0.604269 -0.59414 -0.932412 -0.335826
## TOR01      0.96189 -0.45705 -0.488317  0.89284 -0.072121  1.287613
## TOR02      0.94164  0.40133  1.836191 -0.10442 -0.799260  0.310437
##
##
## Site constraints (linear combinations of constraining variables)
##
##           CAP1      CAP2      CAP3      CAP4      MDS1      MDS2
## BAI01     -0.54680 -0.320421  0.120147  0.16187 -0.253855 -0.409982
## BAI04     -0.98720 -0.336868  0.027708 -0.01365 -0.017995  0.190002
## BAI06     -0.89385 -0.208765  0.194278  0.01664  0.519755  0.927336
## BAI07     -0.86749 -0.213156  0.313045 -0.10138  0.360704 -0.141406
## BPU04      0.48982  1.648799  0.423034  0.12457 -0.871736  0.942653
## BSC03      1.25730  0.105609  1.109475 -1.43461 -0.885722 -0.010345
## BSC04      1.38213 -0.110779  1.241699 -1.59646 -0.545068  0.098636
## BSC05      0.65395 -0.893198  2.724268 -3.01027  0.086267 -0.158439
## BV01      -0.74106 -0.417019  0.639457  0.69926 -0.968096  0.289696
## BV02      -0.66580 -0.411171  0.631355  0.79958 -0.527407 -1.096393
## BValp01   -1.22711 -0.639983  0.943232  0.66432 -0.072846 -0.020977
## BValp02   -0.30644 -0.792249  0.562026  1.17398 -0.911247  0.237265
## BValp03   -0.61949 -0.533832  0.843562  0.80312 -0.732728  0.502892
## CBR02      0.09743  1.718683  0.825608  0.89054  0.943794  0.641044
## CBR03     -0.56369  2.056350  1.118456  0.75301  1.442211  0.490398
## CON02     -0.19650  0.689309 -0.068500  0.22720 -1.382130 -0.843262
## CON03     -0.26379  0.820206 -0.163803  0.32146 -1.030837 -1.105952
## CON06     -0.43376  0.743667 -0.094316  0.14058 -1.455517  0.419849
## CR01      -1.18078 -0.363136 -0.711696 -1.62278  0.528563 -0.265130
## CR06      -1.16423  0.154320 -1.030419 -1.21701  0.740444  0.314611
## CR09      -0.94395  0.114840 -1.122062 -1.10169  1.024394  0.552805
## ERT05      0.31750 -1.610180  0.649449 -0.38407  0.701223  0.261009
## ES03      0.52594  0.037767 -1.359878 -0.61899  0.522311 -0.593333
## ES10      1.03556  0.155077 -1.539910 -0.14803 -0.002463 -0.190669
## FRE03      0.65826  0.606988  0.719775 -0.27498  0.456796  0.626290
## GMG01      0.48399  1.050788  0.019171 -0.60214  0.499396  0.706154
## GMG06      0.14032  1.443468 -0.616556 -0.31813  0.941136  0.748707
## HOR01      0.99383  0.263276  0.057581  1.11557  0.082497 -0.241665
## HOR04     -0.59391  0.014446  0.713543  0.70541 -1.031014 -0.271755
## HOR05     -0.72282 -0.096363  0.906720  0.61650 -0.245153 -0.863661
## HOR06     -0.77410 -0.078345  0.658127  0.73915  0.418815 -2.100926
## LLA07      0.92236 -0.600488 -0.351020  0.64559  0.152243  1.549072
## MA_alp01  -0.32934 -0.445724 -0.838166 -0.32427 -0.425182  0.129374
## MA_alp02  -0.39141 -0.407516 -0.853832 -0.33959  0.886794 -0.003303
## MAR01     -0.03051 -0.658191 -0.301175  0.24431  1.468160 -0.818767

```

```

## MAR02      0.45537 -0.871974 -0.501725  0.36904  0.923027 -0.630025
## MAR03      0.09415 -0.626539 -0.444654  0.33562  1.026251 -0.980965
## MAR04      1.11068 -1.106807 -0.593868  0.51017 -0.045539  0.636156
## MAR05      0.81214 -1.044928 -0.489628  0.35264  0.183513 -0.099442
## MAR06      0.67822 -1.341944 -0.102481 -0.06875  0.195361 -0.001448
## MAR09      0.60870  0.026896 -0.211830  0.64961  0.383607 -0.469806
## MAR10      1.02032 -0.892821 -0.005738  0.36651 -0.194345  0.365294
## MT02       0.03214  0.953157 -0.289580 -0.02973  0.499991 -1.058538
## MTL03      0.68245 -1.012354  0.305622  1.11467  0.680488  0.344813
## MUN02     -0.81766 -0.688773 -0.478063 -0.35131 -0.899363 -0.200745
## MUN03     -1.04018 -0.465114 -0.088335 -0.45027 -0.578090 -0.450198
## PA01       0.36233  0.589802  0.137170  0.67263  0.755767 -0.927530
## PA02       0.37545  0.315646  0.405391  0.37566  1.082601 -0.460399
## PA05       0.56176  0.597770  0.085265  0.89503  0.699502 -0.237219
## RAT07     -0.31114 -0.896996 -0.919507 -0.41080  0.348580 -0.746035
## S003      -0.80895  0.069044  0.190685  0.18564 -0.027875  1.247895
## S004      -0.80683 -0.002139  0.262283  0.10869 -0.454019  0.846195
## S005      -0.75233  0.044785  0.203161  0.21960 -0.451830 -0.028094
## S008      -1.05182 -0.053388  0.459341 -0.12655  0.717703  1.212363
## S009      -1.00617  0.086809  0.275303  0.04880  0.613792  1.569220
## SON02      0.36482  1.167970 -0.341586 -0.72254 -0.619808 -1.029349
## SON03      0.38527  1.315404 -0.590447 -0.50851  0.410478 -0.797852
## T29       -0.23583  0.325430 -1.009521 -0.54979 -1.180814 -1.263410
## T46       0.02682  0.228298 -1.065479 -0.46558 -1.886342  0.973530
## T52       -0.39612  0.407603 -0.977997 -0.55677 -0.553330 -0.015787
## T54       -0.08179  0.262391 -1.034919 -0.50598 -0.618143  1.323686
## TAU03      0.67065  0.207088  0.932812 -0.62888  0.477184 -0.164147
## TAU04      0.89668  0.126647  0.865189 -0.56790 -0.101061 -0.712215
## TAU06      1.22730  0.810832 -0.179258  0.54236 -0.932412 -0.335826
## TOR01      1.19238 -0.485287 -0.726924  0.75484 -0.072121  1.287613
## TOR02      1.23681 -0.532719 -0.461065  0.70727 -0.799260  0.310437
##
##
## Biplot scores for constraining variables
##
##          CAP1      CAP2      CAP3      CAP4 MDS1 MDS2
## GDD    0.4603  0.73934  0.3330 -0.36149    0    0
## pH     0.9729 -0.19236  0.1270 -0.01572    0    0
## Prec -0.3996 -0.61446  0.1162  0.67025    0    0
## Mg     0.5184 -0.06498  0.5818 -0.62332    0    0

```

## 5.2 dbRDA COI

```

analysis_coi <- capscale(datos.trans_coi ~ WTD+GDD+Rad+
                        Sphagnum+Brown_mosses+Acrocarp_mosses+Liverworts+
                        Bryophytes+Vascular_plants+
                        pH+Prec+K+Mg+Mn+Na+P+S+Si+Al+Ca+Fe, Metadata,
                        dist="bray")

anova (analysis_coi) # 0.001 *** - it is significant

## Permutation test for capscale under reduced model
## Permutation: free
## Number of permutations: 999

```

```
##
## Model: capscale(formula = datos.trans_coi ~ WTD + GDD + Rad + Sphagnum + Brown_mosses + Acrocarp_mos
##           Df SumOfSqs      F Pr(>F)
## Model      20    11.871 1.9747 0.001 ***
## Residual   45     13.526
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

adjR2.tbrda <- RsquareAdj (analysis_coi)$adj.r.squared

#Forward selection
sel.fs <- forward.sel (Y = datos.trans_coi , X=x, adjR2thresh = adjR2.tbrda)

## Testing variable 1
## Testing variable 2
## Testing variable 3
## Testing variable 4
## Testing variable 5
## Procedure stopped (alpha criteria): pvalue for variable 5 is 0.066000 (> 0.050000)

sel.fs

##   variables order      R2      R2Cum AdjR2Cum      F pvalue
## 1      pH      5 0.13085803 0.1308580 0.1172777 9.635841 0.001
## 2      GDD      2 0.04252872 0.1733868 0.1471451 3.241309 0.001
## 3      Mg     16 0.03102715 0.2044139 0.1659178 2.417945 0.001
## 4 Sphagnum      7 0.02054400 0.2249579 0.1741355 1.616923 0.015

analysis_coi_fs <- capscale(datos.trans_coi ~ GDD+Sphagnum+Acrocarp_mosses+pH+Mg, Metadata,
                           dist="bray")
summary(analysis_coi_fs)

##
## Call:
## capscale(formula = datos.trans_coi ~ GDD + Sphagnum + Acrocarp_mosses +      pH + Mg, data = Metadata
##
## Partitioning of squared Bray distance:
##           Inertia Proportion
## Total      25.398      1.0000
## Constrained    7.185      0.2829
## Unconstrained 18.213      0.7171
##
## Eigenvalues, and their contribution to the squared Bray distance
##
## Importance of components:
##           CAP1      CAP2      CAP3      CAP4      CAP5      MDS1      MDS2
## Eigenvalue      4.4027 1.24660 0.70723 0.4673 0.3607 1.8134 1.68262
## Proportion Explained 0.1734 0.04908 0.02785 0.0184 0.0142 0.0714 0.06625
## Cumulative Proportion 0.1734 0.22244 0.25028 0.2687 0.2829 0.3543 0.42054
##           MDS3      MDS4      MDS5      MDS6      MDS7      MDS8      MDS9
## Eigenvalue      1.26544 1.03129 0.93590 0.92400 0.75287 0.70368 0.66080
## Proportion Explained 0.04983 0.04061 0.03685 0.03638 0.02964 0.02771 0.02602
## Cumulative Proportion 0.47036 0.51097 0.54782 0.58420 0.61384 0.64155 0.66757
##           MDS10 MDS11 MDS12 MDS13 MDS14 MDS15 MDS16
## Eigenvalue      0.59792 0.5689 0.52509 0.45130 0.44753 0.41441 0.39243
## Proportion Explained 0.02354 0.0224 0.02067 0.01777 0.01762 0.01632 0.01545
```

```

## Cumulative Proportion 0.69111 0.7135 0.73418 0.75195 0.76958 0.78589 0.80134
##                               MDS17   MDS18   MDS19   MDS20   MDS21   MDS22   MDS23
## Eigenvalue              0.37408 0.33719 0.32731 0.30811 0.29042 0.27883 0.246369
## Proportion Explained    0.01473 0.01328 0.01289 0.01213 0.01144 0.01098 0.009701
## Cumulative Proportion 0.81607 0.82935 0.84224 0.85437 0.86580 0.87678 0.886482
##                               MDS24   MDS25   MDS26   MDS27   MDS28   MDS29
## Eigenvalue              0.239425 0.234143 0.215198 0.203572 0.191179 0.170403
## Proportion Explained    0.009427 0.009219 0.008473 0.008015 0.007527 0.006709
## Cumulative Proportion 0.895910 0.905129 0.913602 0.921617 0.929145 0.935854
##                               MDS30   MDS31   MDS32   MDS33   MDS34   MDS35
## Eigenvalue              0.165421 0.14553 0.143360 0.132297 0.127186 0.111257
## Proportion Explained    0.006513 0.00573 0.005645 0.005209 0.005008 0.004381
## Cumulative Proportion 0.942368 0.94810 0.953742 0.958951 0.963959 0.968340
##                               MDS36   MDS37   MDS38   MDS39   MDS40   MDS41
## Eigenvalue              0.101734 0.093381 0.085384 0.080554 0.069432 0.064398
## Proportion Explained    0.004006 0.003677 0.003362 0.003172 0.002734 0.002536
## Cumulative Proportion 0.972345 0.976022 0.979384 0.982556 0.985290 0.987825
##                               MDS42   MDS43   MDS44   MDS45   MDS46   MDS47
## Eigenvalue              0.060831 0.058324 0.054718 0.03200 0.029599 0.0227173
## Proportion Explained    0.002395 0.002296 0.002154 0.00126 0.001165 0.0008945
## Cumulative Proportion 0.990220 0.992517 0.994671 0.99593 0.997096 0.9979910
##                               MDS48   MDS49   MDS50   MDS51   MDS52
## Eigenvalue              0.0177755 0.0146970 0.0080199 0.0061544 0.0043779
## Proportion Explained    0.0006999 0.0005787 0.0003158 0.0002423 0.0001724
## Cumulative Proportion 0.9986909 0.9992695 0.9995853 0.9998276 1.0000000
##
## Accumulated constrained eigenvalues
## Importance of components:
##                               CAP1   CAP2   CAP3   CAP4   CAP5
## Eigenvalue              4.4027 1.2466 0.70723 0.46728 0.36072
## Proportion Explained    0.6128 0.1735 0.09844 0.06504 0.05021
## Cumulative Proportion 0.6128 0.7863 0.88475 0.94979 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 6.332038
##
##
## Species scores
##
##                               CAP1   CAP2   CAP3   CAP4   CAP5
## BOGC_000020776 0.0175587 0.0357572 1.048e-01 1.585e-01 -1.223e-01
## BOGC_000020780 0.0042488 -0.0149412 8.153e-02 1.348e-01 -6.107e-02
## BOGC_000020797 0.0505245 -0.0124669 8.645e-02 1.420e-01 -9.599e-02
## BOGC_000020810 0.0065229 0.0352416 1.703e-02 -4.092e-03 -9.069e-04
## BOGC_000020819 0.0160957 0.0029675 3.647e-02 6.296e-02 -5.178e-02
## BOGC_000020822 0.0099156 -0.0038268 2.206e-02 5.843e-02 -3.969e-02
## BOGC_000020825 0.0013962 0.0155460 1.326e-02 3.637e-02 -1.913e-02
## BOGC_000020848 0.0619115 0.0215046 3.782e-02 -1.420e-03 -3.815e-02
## BOGC_000020882 0.0154871 -0.0126095 2.567e-02 5.556e-02 -2.962e-02
## BOGC_000020914 0.0060681 0.0038194 -1.693e-04 -2.219e-03 -4.778e-03
## BOGC_000020929 0.0084993 -0.0004968 7.211e-03 -3.145e-03 -1.070e-02
## BOGC_000020937 0.0184691 0.0150642 8.112e-03 -3.130e-03 -7.442e-03

```

```

## BOGC_000020972 0.0410823 0.0081916 2.499e-02 -1.512e-04 -2.939e-02
## BOGC_000020985 0.0060681 0.0038194 -1.693e-04 -2.219e-03 -4.778e-03
## BOGC_000020991 -0.0253147 0.0001597 -4.104e-03 4.596e-02 -9.210e-03
## BOGC_000021028 0.0098875 0.0096627 8.351e-03 8.124e-06 -6.851e-04
## BOGC_000021098 0.0060681 0.0038194 -1.693e-04 -2.219e-03 -4.778e-03
## BOGC_000021157 -0.0324978 -0.0054832 8.450e-03 -2.900e-02 1.263e-02
## BOGC_000021403 0.0084993 -0.0004968 7.211e-03 -3.145e-03 -1.070e-02
## BOGC_000021455 -0.0420401 -0.0383465 1.063e-02 7.356e-02 -5.698e-03
## BOGC_000021624 0.0020057 -0.0143110 1.581e-02 3.308e-02 -2.327e-02
## BOGC_000021665 0.0060681 0.0038194 -1.693e-04 -2.219e-03 -4.778e-03
## BOGC_000021824 -0.0146145 0.0025142 -4.378e-03 1.779e-02 1.985e-02
## BOGC_000022034 0.0159556 0.0134821 8.182e-03 -2.211e-03 -5.463e-03
## BOGC_000022881 -0.0090837 0.0032993 1.146e-02 -4.752e-03 -9.299e-03
## BOGC_000022923 -0.0221910 -0.0025878 -2.782e-03 3.407e-02 -3.134e-03
## BOGC_000024889 0.0098875 0.0096627 8.351e-03 8.124e-06 -6.851e-04
## BOGC_000032443 0.0084993 -0.0004968 7.211e-03 -3.145e-03 -1.070e-02
## BOGC_000038860 0.0133261 0.0347690 1.483e-02 -5.421e-03 -8.262e-04
## BOGC_000069470 0.0054032 0.0448537 -4.139e-03 -1.300e-02 -5.612e-03
## BOGC_000085702 -0.0794294 0.0461834 -2.053e-02 4.002e-02 2.706e-02
## BOGC_000092530 0.0060099 -0.0003513 5.099e-03 -2.224e-03 -7.566e-03
## BOGC_000092629 0.0244222 0.0109105 1.963e-02 1.344e-03 -1.081e-02
## BOGC_000116990 0.0060681 0.0038194 -1.693e-04 -2.219e-03 -4.778e-03
## BOGC_000118115 0.0145347 0.0012478 1.128e-02 1.336e-03 -1.012e-02
## BOGC_000178751 0.2382797 0.0727539 1.090e-01 -2.335e-02 8.544e-02
## BOGC_000258122 -0.7975160 0.0704498 -5.430e-03 2.411e-01 -7.497e-02
## BOGC_000259147 -0.1656989 -0.0369813 2.668e-02 -3.575e-02 8.509e-02
## BOGC_000259233 -0.1804643 -0.0783535 1.012e-01 -1.725e-01 2.958e-01
## BOGC_000259352 -0.2513324 -0.0871199 4.330e-02 -1.092e-01 5.875e-02
## BOGC_000261697 -0.1301092 -0.0713310 1.770e-02 -9.736e-02 2.741e-02
## BOGC_000264231 -1.1091638 -0.1658197 1.517e-01 -4.195e-01 -2.142e-01
## BOGC_000265243 -0.0379827 -0.0155599 -2.605e-03 -2.550e-02 1.171e-02
## BOGC_000307139 -0.0408425 -0.0274593 1.383e-02 -3.359e-02 3.594e-02
## BOGC_000309494 -0.1102091 -0.0351019 1.600e-02 -3.753e-03 2.720e-02
## BOGC_000361366 -0.2687766 -0.0869469 -2.539e-02 2.183e-02 -3.900e-03
## BOGC_000361410 -0.4901158 -0.1135083 3.444e-02 -1.893e-01 -1.045e-01
## BOGC_000361518 -0.1348645 -0.0351650 6.823e-03 -3.264e-03 -2.338e-02
## BOGC_000361521 -0.1013488 -0.0252048 3.433e-02 -2.700e-02 -4.209e-02
## BOGC_000361629 -0.0643333 -0.0258978 1.824e-02 -5.529e-02 -2.195e-02
## BOGC_000362154 -0.1324738 -0.0557283 1.166e-03 8.024e-03 -2.684e-03
## BOGC_000363872 -0.0382732 0.1189944 1.140e-01 -3.282e-02 8.339e-02
## BOGC_000364792 -0.0181482 -0.0144422 7.251e-03 -1.496e-02 -4.036e-03
## BOGC_000396048 -0.0604031 -0.0190157 1.876e-02 -5.403e-02 -2.279e-02
## BOGC_000437136 -0.0169798 -0.0080730 3.480e-04 -1.134e-02 9.134e-03
## BOGC_000456027 -0.0771571 -0.0325415 -3.249e-02 9.373e-02 6.475e-03
## BOGC_000462072 -0.2743890 0.0858969 -4.601e-02 2.160e-01 1.381e-01
## BOGC_000550505 -0.1630744 -0.0484999 1.149e-02 5.224e-02 -3.390e-02
## BOGC_000550512 -0.4302298 -0.1500639 6.419e-02 7.725e-03 1.231e-01
## BOGC_000550639 -0.0758589 -0.0257986 1.363e-02 8.253e-03 9.723e-04
## BOGC_000579958 0.0060099 -0.0003513 5.099e-03 -2.224e-03 -7.566e-03
## BOGC_000636436 0.0031983 0.0181306 -1.126e-03 -2.525e-03 -9.062e-03
## BOGC_000636639 0.0650779 0.1088769 4.689e-02 -4.944e-02 -1.020e-01
## BOGC_000636714 -0.0395398 -0.0047547 2.263e-02 -6.041e-02 -4.383e-02
## BOGC_000637323 0.0281992 -0.0059164 2.168e-02 -4.787e-03 -2.388e-02
## BOGC_000638034 -0.0040898 0.0086430 2.875e-03 -5.555e-03 -7.158e-03

```

```

## BOGC_000639559 -0.0848335 -0.0224135 3.056e-02 -9.526e-03 -4.097e-02
## BOGC_000639602 0.1700908 0.0119758 8.085e-02 -1.179e-01 6.955e-02
## BOGC_000654903 -0.7485403 0.0796972 -5.015e-02 1.493e-01 8.910e-02
## BOGC_000684921 0.3233258 0.1781678 8.326e-02 -5.885e-02 -3.977e-02
## BOGC_000685554 0.0209912 0.1027646 -1.406e-02 5.669e-02 -1.339e-02
## BOGC_000685680 0.1008314 0.0327015 -1.644e-02 -1.749e-03 -5.312e-02
## BOGC_000686480 0.0597903 0.0514636 2.298e-02 -3.821e-03 -1.740e-02
## BOGC_000687729 0.1413255 0.2179237 2.357e-02 -3.578e-02 6.541e-02
## BOGC_000687916 0.0264511 0.0119560 1.868e-02 2.107e-03 -1.074e-02
## BOGC_000688079 0.0799492 0.0264116 -3.383e-03 -9.116e-03 -4.234e-02
## BOGC_000688986 0.0343197 0.0196439 5.842e-03 -4.527e-03 -1.392e-02
## BOGC_000689006 -0.0004645 0.0149226 2.135e-03 -3.842e-03 1.009e-04
## BOGC_000689180 0.1213531 0.0959419 3.334e-02 -1.125e-02 -3.118e-02
## BOGC_000690007 0.1749883 -0.0379657 2.486e-02 6.647e-03 -6.598e-02
## BOGC_000825479 -0.0713223 0.0085917 -5.756e-03 2.406e-02 1.833e-02
## BOGC_000947790 0.3291288 -0.0450685 1.624e-01 3.798e-02 8.086e-02
## BOGC_000947841 0.1228334 0.1366207 3.321e-02 -3.298e-02 5.935e-02
## BOGC_000948108 0.0638996 0.1356595 -4.412e-03 -1.396e-02 9.792e-03
## BOGC_000948129 0.0240785 0.0385601 5.726e-05 2.734e-04 5.513e-02
## BOGC_000948159 0.0426288 0.0597538 -9.762e-03 -3.932e-04 9.091e-02
## BOGC_000948651 0.4129909 0.1683149 -1.572e-01 3.278e-05 1.016e-01
## BOGC_000970548 0.3048037 -0.1015218 1.320e-01 -1.970e-02 -6.290e-02
## BOGC_000970727 0.9747389 -0.6659184 1.682e-01 -2.205e-02 -9.848e-02
## BOGC_000970738 0.0861283 -0.0231817 5.885e-02 3.818e-03 1.968e-03
## BOGC_000971042 0.2113168 -0.0993360 6.994e-02 -1.296e-02 7.310e-03
## BOGC_000971228 0.0060681 0.0038194 -1.693e-04 -2.219e-03 -4.778e-03
## BOGC_000971379 0.1703139 0.0008061 2.385e-02 -4.064e-03 2.449e-02
## BOGC_000971407 0.0669129 -0.0207746 3.655e-02 2.894e-03 -2.315e-02
## BOGC_000971482 -0.0106873 -0.0187607 1.708e-02 8.156e-04 -1.331e-02
## BOGC_000971911 -0.0208549 -0.0074751 -1.063e-02 2.316e-02 2.094e-03
## BOGC_000972558 0.0586266 -0.0131200 4.316e-02 -2.136e-02 -1.743e-02
## BOGC_000972757 0.0331272 0.0045866 2.065e-02 4.197e-03 -2.010e-02
## BOGC_000973046 0.0410134 -0.0225653 3.641e-02 -1.561e-02 -1.462e-02
## BOGC_000973565 0.0099539 0.0094795 -1.029e-02 -9.524e-04 -2.671e-05
## BOGC_000974385 0.0085816 0.0054014 -2.394e-04 -3.138e-03 -6.757e-03
## BOGC_000974819 0.0099539 0.0094795 -1.029e-02 -9.524e-04 -2.671e-05
## BOGC_000975801 0.0082331 -0.0130670 9.486e-03 -1.193e-02 -1.364e-02
## BOGC_000976058 0.0408183 0.0647340 -3.525e-02 -8.774e-05 -9.734e-03
## BOGC_000976714 0.0079870 -0.0130742 8.506e-03 -2.024e-03 -2.680e-03
## BOGC_000987689 0.0058669 -0.0068434 5.445e-03 -7.296e-03 -8.753e-03
## BOGC_001001027 0.0004279 0.0229803 1.820e-03 -3.049e-03 1.617e-03
## BOGC_001021386 0.0712074 0.1215348 -4.871e-02 -1.719e-02 -3.564e-02
## BOGC_001033584 -0.1135594 -0.0316841 1.855e-02 1.255e-02 6.561e-02
## BOGC_001033666 0.0298236 0.0925299 -1.938e-02 -1.666e-02 -2.500e-02
## BOGC_001033669 0.0438231 0.0060675 2.732e-02 5.552e-03 -2.660e-02
## BOGC_001037523 0.0098875 0.0096627 8.351e-03 8.124e-06 -6.851e-04
## BOGC_001037897 0.0623187 -0.0189608 4.003e-02 3.338e-03 -3.454e-02
## BOGC_001038127 0.0098875 0.0096627 8.351e-03 8.124e-06 -6.851e-04
## BOGC_001038483 -0.0088226 0.0130912 1.073e-02 -9.459e-03 -1.506e-02
## BOGC_001040176 0.0178241 0.0242030 -2.352e-02 -4.475e-03 -3.806e-03
## BOGC_001040813 0.0625026 -0.0065697 -4.401e-02 2.725e-03 1.206e-02
## BOGC_001041328 0.0269056 0.0065871 -5.507e-03 7.197e-04 -6.336e-03
## BOGC_001062949 0.5102123 0.5698487 5.888e-02 -8.180e-02 -2.584e-02
## BOGC_001087197 0.0306191 0.0530619 1.865e-02 -3.382e-03 -3.103e-02

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## BOGC_001101225 0.0896008 0.0375982 1.004e-02 -1.449e-02 -2.949e-02
## BOGC_001101492 0.0385042 -0.0149659 -5.900e-03 -2.265e-02 -1.887e-02
## BOGC_001215101 -0.0219699 -0.0211582 3.662e-03 6.161e-02 -5.138e-03
## BOGC_001239041 -0.0110434 -0.0007564 5.993e-03 -1.262e-02 5.113e-02
## BOGC_001273770 0.0104095 -0.0006084 8.832e-03 -3.852e-03 -1.310e-02
## BOGC_001275273 -0.0415758 -0.0159206 6.540e-03 -2.457e-02 -7.827e-03
## BOGC_001330265 -0.1685266 0.0104557 9.465e-02 -1.657e-01 -1.266e-01
## BOGC_001379896 -0.1534765 -0.0249887 5.210e-02 -1.026e-01 -6.534e-02
## BOGC_001381033 -0.0062154 -0.0071383 4.624e-03 -7.144e-03 -2.781e-03
## BOGC_001417378 -0.0278346 0.2195389 2.140e-02 -5.920e-02 1.566e-01
## BOGC_001446969 -0.0280394 -0.0135592 1.361e-02 -2.920e-02 4.446e-02
## BOGC_001447024 0.0055980 0.1190870 4.791e-03 -3.538e-02 4.073e-02
## BOGC_001447168 -0.0099167 0.0134724 8.363e-03 -1.370e-02 -2.951e-03
## BOGC_001447265 -0.0179172 -0.0164681 1.133e-02 -1.844e-02 -7.912e-03
## BOGC_001447380 -0.1188009 -0.0644181 9.558e-03 1.040e-02 2.165e-02
## BOGC_001452586 -0.0073147 -0.0067231 4.626e-03 -7.529e-03 -3.230e-03
## BOGC_001510909 -0.0274273 0.0192836 2.284e-02 -2.997e-02 2.878e-02
## BOGC_001515965 0.0169794 -0.0049978 1.017e-02 1.943e-03 1.025e-02
## BOGC_001624382 -0.0521513 0.0082625 1.716e-02 5.393e-02 -3.209e-02
## BOGC_001624966 -0.0324792 0.0265869 7.695e-03 4.112e-03 1.537e-01
## BOGC_001655608 -0.0847046 0.0559235 1.902e-02 1.370e-01 -5.674e-02
## BOGC_001655669 0.1111243 -0.0307722 4.933e-02 -5.469e-03 3.552e-02
## BOGC_001655696 0.0007411 0.0398031 3.153e-03 -5.281e-03 2.800e-03
## BOGC_001655701 -0.0020635 0.3663447 4.638e-02 4.961e-02 -8.333e-02
## BOGC_001655818 0.0009697 0.0229753 1.099e-03 -2.964e-03 -1.123e-02
## BOGC_001655831 0.0157790 0.0462708 -1.502e-02 -1.704e-02 6.397e-02
## BOGC_001655856 -0.3031266 -0.1102259 2.845e-02 -7.476e-02 1.487e-01
## BOGC_001656003 0.0167008 0.0376559 1.963e-02 -2.562e-03 1.213e-01
## BOGC_001656145 -0.0181655 0.1069824 3.987e-03 8.089e-02 -4.395e-02
## BOGC_001657949 0.0176283 -0.0076422 -1.948e-02 -9.568e-03 8.152e-04
## BOGC_001658165 -0.0092247 0.0604423 3.835e-05 -1.117e-02 -7.239e-03
## BOGC_001658626 -0.0270834 -0.0214414 3.246e-03 -2.042e-02 -1.054e-03
## BOGC_001659251 0.0182519 0.0471834 -2.170e-02 -7.524e-03 -2.189e-03
## BOGC_001659791 0.0004279 0.0229803 1.820e-03 -3.049e-03 1.617e-03
## BOGC_001675403 0.0143565 0.0403473 -1.363e-02 -2.188e-03 2.663e-02
## BOGC_001772854 0.0084993 -0.0004968 7.211e-03 -3.145e-03 -1.070e-02
## BOGC_001863642 0.0889192 0.0285822 6.695e-02 -4.053e-02 -8.455e-02
## BOGC_001863644 0.0411104 0.0035293 3.191e-02 3.778e-03 -2.863e-02
## BOGC_001863701 0.0333605 -0.0111693 -3.828e-02 -3.083e-03 -1.198e-02
## BOGC_001863724 0.0145347 0.0012478 1.128e-02 1.336e-03 -1.012e-02
## BOGC_001863770 0.0145347 0.0012478 1.128e-02 1.336e-03 -1.012e-02
## BOGC_001916701 0.1173555 -0.0172052 -1.435e-01 -6.524e-04 -3.579e-03
## BOGC_001987145 -0.0981702 -0.0139817 6.165e-03 -4.721e-02 -1.224e-02
## BOGC_002015445 -0.0218950 -0.0104384 -2.294e-03 5.034e-03 -1.315e-03
## BOGC_002026671 -0.0183499 -0.0016448 -7.129e-04 -1.326e-02 -3.544e-03
## BOGC_002120333 0.3993742 0.0509656 -2.798e-02 -2.930e-02 1.590e-01
## BOGC_002120600 -0.0773739 0.0234873 -8.703e-03 2.032e-02 3.724e-02
## BOGC_002188612 -0.1108083 -0.0069560 -1.749e-02 1.635e-01 1.535e-02
## BOGC_002251140 -0.0156914 -0.0018298 -1.968e-03 2.409e-02 -2.216e-03
## BOGC_002251203 -0.0313828 -0.0036597 -3.935e-03 4.818e-02 -4.432e-03
## BOGC_002362160 -0.0085175 -0.0012359 1.585e-03 -7.625e-03 -3.758e-03
## BOGC_002471041 -0.0157172 0.0067348 7.861e-03 5.959e-02 -1.676e-02
## BOGC_002473720 -0.0084508 0.0045215 2.835e-03 1.678e-03 -5.464e-03
## BOGC_002475336 -0.0017865 0.0014378 5.519e-03 2.641e-02 -9.358e-03

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## BOGC_002513491 -0.1392019 -0.0437979 -1.233e-02 3.793e-03 6.784e-02
## BOGC_002659394 -0.0031849 -0.0477728 2.750e-02 -4.883e-02 -4.986e-02
## BOGC_002839632 0.1928533 -0.1033193 -9.401e-02 -6.506e-03 -3.514e-02
## BOGC_002840418 0.0776012 -0.0412458 -4.766e-02 3.180e-03 -2.102e-02
## BOGC_002840474 0.2171734 -0.1044820 -7.488e-02 2.442e-03 -5.952e-02
## BOGC_002840484 0.0571898 -0.0513680 -5.653e-02 -4.058e-03 6.632e-04
## BOGC_002842748 0.0838724 -0.0276368 -7.003e-02 6.009e-03 -6.418e-03
## BOGC_002844734 0.0881621 0.0074687 -7.418e-02 3.255e-03 -1.296e-02
## BOGC_002848214 0.1165827 -0.0665716 -2.695e-02 4.623e-03 -4.591e-03
## BOGC_002849342 0.0869808 -0.0468685 -1.521e-01 -4.836e-03 -1.008e-02
## BOGC_002855787 0.0199037 -0.0026839 -3.009e-02 9.591e-04 -4.347e-03
## BOGC_002862016 0.0298577 0.0067956 -4.039e-02 6.658e-06 -4.374e-03
## BOGC_002868046 0.0281481 -0.0037956 -4.256e-02 1.356e-03 -6.148e-03
## BOGC_002868289 0.0339327 0.0034200 -9.150e-03 -9.026e-04 -5.699e-03
## BOGC_002871816 0.0641973 0.0609732 7.220e-03 -9.897e-02 -3.388e-02
## BOGC_002873710 0.2249499 -0.0773868 -7.894e-01 -7.687e-02 -2.185e-02
## BOGC_002874421 -0.0155439 -0.0030425 3.787e-03 -6.585e-03 -1.330e-03
## BOGC_002885634 -0.0197530 0.1409356 -4.757e-04 -2.562e-02 -3.395e-02
## BOGC_002886501 0.2042832 -0.0203225 6.286e-02 3.630e-02 -3.153e-02
## BOGC_002887012 0.0967284 0.0218930 1.985e-02 6.445e-04 -4.797e-02
## BOGC_002887121 0.0274863 -0.0272485 1.917e-02 3.064e-04 1.476e-03
## BOGC_002887157 0.0194993 -0.0141744 1.067e-02 2.331e-03 4.156e-03
## BOGC_002887259 -0.0002012 -0.0106628 5.615e-03 -5.077e-03 -3.976e-03
## BOGC_002888876 0.1813895 -0.1668639 5.267e-02 3.005e-03 3.099e-02
## BOGC_002889770 0.9943020 -0.3623471 3.042e-01 5.675e-02 8.896e-02
## BOGC_002890109 0.0792010 -0.0679806 5.499e-02 -5.517e-03 3.545e-02
## BOGC_002898141 0.0661590 -0.0542454 1.277e-02 3.877e-03 -2.052e-02
## BOGC_002902738 0.0758408 -0.0088822 -1.539e-01 -3.794e-03 -6.172e-03
## BOGC_002903258 0.0158224 0.0487082 2.594e-03 -2.379e-02 -4.226e-02
## BOGC_002911856 0.0466324 -0.0417169 1.833e-02 -1.711e-04 -1.569e-02
## BOGC_002923505 -0.0342184 -0.0101481 -7.587e-03 -2.318e-02 -2.968e-03
## BOGC_002942886 0.0370373 0.0051280 2.309e-02 4.693e-03 -2.248e-02
## BOGC_002956160 0.2454644 0.2245980 5.183e-03 -6.919e-03 4.403e-02
## BOGC_002961744 0.0004279 0.0229803 1.820e-03 -3.049e-03 1.617e-03
## BOGC_002962205 0.0312325 0.0244844 3.130e-02 -4.612e-03 -8.313e-03
## BOGC_002962271 0.1039111 0.0156327 -5.286e-02 4.103e-03 5.898e-03
## BOGC_002963593 0.0169794 -0.0049978 1.017e-02 1.943e-03 1.025e-02
## BOGC_002995588 0.0164329 -0.0279398 1.407e-04 -2.116e-03 -8.070e-04
## BOGC_002999314 0.0165636 0.0022933 1.033e-02 2.099e-03 -1.005e-02
## BOGC_003009107 0.0468135 0.0255462 -2.609e-02 -1.185e-03 -6.389e-03
## BOGC_003179898 0.0460043 -0.0033924 -6.966e-02 1.889e-03 -1.125e-02
## BOGC_003190776 -0.0003485 -0.0184684 9.725e-03 -8.793e-03 -6.886e-03
## BOGC_003207030 -0.0722120 -0.0335439 -5.573e-03 -1.744e-02 -7.494e-03
## BOGC_003208049 0.0031983 0.0181306 -1.126e-03 -2.525e-03 -9.062e-03
## BOGC_003292662 0.0547035 0.0065067 3.940e-02 -2.174e-02 1.537e-02
## BOGC_003292739 0.0028793 0.0160600 -5.630e-03 -2.992e-03 -7.803e-03
## BOGC_003293701 0.0060423 0.0130745 -9.866e-05 -8.318e-03 -1.372e-02
## BOGC_003318885 0.0019704 -0.0096378 7.863e-03 -4.087e-03 -2.689e-03
## BOGC_003326108 0.0448828 0.0851115 1.050e-02 -9.120e-03 8.796e-03
## BOGC_003326121 0.0012101 0.0649981 5.149e-03 -8.623e-03 4.573e-03
## BOGC_003326158 0.0686035 0.4287341 -5.825e-02 -3.125e-02 -7.561e-02
## BOGC_003326162 0.0191499 0.1089185 -1.155e-02 -1.248e-02 6.019e-03
## BOGC_003326307 0.0004279 0.0229803 1.820e-03 -3.049e-03 1.617e-03
## BOGC_003326805 0.0095989 0.0997710 -1.760e-03 -1.151e-02 -3.334e-03

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## BOGC_003327323 -0.0134736 0.0624981 -1.766e-03 -1.274e-02 -1.185e-02
## BOGC_003342732 -0.0104452 -0.0027959 -9.621e-04 -8.556e-03 -2.589e-03
## BOGC_003359556 -0.0081809 -0.0113324 -1.007e-02 1.160e-02 3.406e-04
## BOGC_003403105 0.0144770 0.0351201 -1.188e-02 -4.523e-03 -1.284e-02
## BOGC_003405745 0.0070528 0.0393389 -1.379e-02 -7.329e-03 -1.911e-02
## BOGC_003447297 0.0136359 -0.0160704 7.859e-03 -1.534e-04 -4.827e-03
## BOGC_003465220 0.0312909 -0.0263300 -1.363e-02 -4.697e-03 -1.391e-02
## BOGC_003474538 0.0023754 0.0562777 2.692e-03 -7.261e-03 -2.751e-02
## BOGC_003474566 0.0027429 0.0649839 3.109e-03 -8.384e-03 -3.176e-02
## BOGC_003474609 0.0022342 0.1284085 -3.451e-03 -4.920e-03 -2.252e-02
## BOGC_003474639 -0.0088600 0.0509185 -1.437e-03 -9.820e-03 -2.075e-02
## BOGC_003474685 0.0016796 0.0397943 1.904e-03 -5.134e-03 -1.945e-02
## BOGC_003474692 0.0016796 0.0397943 1.904e-03 -5.134e-03 -1.945e-02
## BOGC_003474758 -0.0170257 0.0483991 -4.393e-03 -1.187e-02 -1.650e-02
## BOGC_003474872 0.0009697 0.0229753 1.099e-03 -2.964e-03 -1.123e-02
## BOGC_003475128 0.0071329 0.0250840 -3.567e-03 7.382e-03 4.270e-04
## BOGC_003475597 0.0005053 0.0378979 3.234e-03 -6.806e-03 -1.113e-02
## BOGC_003476712 0.0109237 0.0324548 -9.193e-03 -3.917e-03 -1.126e-02
## BOGC_003477699 0.0159497 0.0560896 -7.976e-03 1.651e-02 9.549e-04
## BOGC_003580614 -0.0002012 -0.0106628 5.615e-03 -5.077e-03 -3.976e-03
## BOGC_003610125 0.0138854 -0.0074240 -9.116e-03 -5.073e-03 -3.941e-04
## BOGC_003642549 0.0593901 -0.1046073 7.524e-04 -1.380e-02 1.428e-02
## BOGC_003642698 0.0061696 -0.0236819 -9.723e-03 -4.241e-03 2.454e-03
## BOGC_003643564 0.0079870 -0.0130742 8.506e-03 -2.024e-03 -2.680e-03
## BOGC_003648679 0.0296543 0.0331911 -1.607e-02 -8.548e-03 -1.487e-02
## BOGC_003714221 0.0199037 -0.0026839 -3.009e-02 9.591e-04 -4.347e-03
## BOGC_003801294 0.1234576 -0.0379657 -7.728e-02 -1.061e-02 2.438e-02
## BOGC_003801311 0.0195803 -0.0079272 1.378e-02 2.816e-03 -7.720e-03
## BOGC_003801417 0.0099539 0.0094795 -1.029e-02 -9.524e-04 -2.671e-05
## BOGC_003848294 0.0427340 -0.0035768 -1.219e-02 -7.091e-03 -1.013e-02
## BOGC_003848301 0.0304744 0.0990597 -6.185e-02 -1.790e-02 7.447e-03
## BOGC_003848594 -0.0098298 0.0279433 -2.536e-03 -6.856e-03 -9.525e-03
## BOGC_003848736 0.0141613 0.0209951 -2.026e-02 -5.792e-03 5.357e-03
## BOGC_003851301 0.0103420 0.0042938 -1.583e-02 -1.379e-03 3.717e-03
## BOGC_003943434 -0.0335325 0.0344523 2.357e-02 2.296e-01 -4.674e-02
## BOGC_003982489 -0.0008459 0.0028844 9.836e-03 -5.746e-03 7.596e-02
## BOGC_004017409 0.0099075 -0.0008204 8.260e-03 -6.420e-04 -7.611e-03
## BOGC_004033098 0.0186888 0.0101711 5.797e-03 3.094e-03 -1.217e-02
## BOGC_004332832 0.0017692 -0.0203005 1.348e-02 -9.163e-03 -6.664e-03
## BOGC_004334495 0.0128332 0.0255395 -1.592e-02 -3.945e-03 -7.830e-03
## BOGC_004471641 -0.0094019 0.0509236 -7.157e-04 -9.904e-03 -7.908e-03
## BOGC_004555822 -0.0261351 0.0403627 -1.055e-02 -9.631e-03 -1.720e-02
## BOGC_004585250 -0.0037659 0.0003404 3.852e-03 5.722e-02 -9.035e-03
## BOGC_004614042 -0.0002012 -0.0106628 5.615e-03 -5.077e-03 -3.976e-03
## BOGC_004719261 0.0116198 -0.0197564 9.947e-05 -1.496e-03 -5.706e-04
## BOGC_004866528 -0.0170257 0.0483991 -4.393e-03 -1.187e-02 -1.650e-02
## BOGC_004897287 -0.0240780 0.0684467 -6.212e-03 -1.679e-02 -2.333e-02
## BOGC_005229678 0.0368831 -0.0076817 -1.992e-02 2.902e-03 5.901e-03
## BOGC_005445377 0.0099539 0.0094795 -1.029e-02 -9.524e-04 -2.671e-05
## MDS1
## BOGC_000020776 1.104e-01
## BOGC_000020780 5.865e-02
## BOGC_000020797 4.975e-02
## BOGC_000020810 3.265e-02

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```
## B0GC_000020819 2.542e-02
## B0GC_000020822 4.520e-02
## B0GC_000020825 2.005e-02
## B0GC_000020848 2.188e-02
## B0GC_000020882 1.982e-02
## B0GC_000020914 7.406e-03
## B0GC_000020929 -1.447e-03
## B0GC_000020937 1.570e-02
## B0GC_000020972 1.055e-02
## B0GC_000020985 7.406e-03
## B0GC_000020991 1.111e-02
## B0GC_000021028 5.230e-03
## B0GC_000021098 7.406e-03
## B0GC_000021157 -1.777e-03
## B0GC_000021403 -1.447e-03
## B0GC_000021455 2.133e-02
## B0GC_000021624 6.021e-03
## B0GC_000021665 7.406e-03
## B0GC_000021824 2.326e-03
## B0GC_000022034 1.264e-02
## B0GC_000022881 1.085e-03
## B0GC_000022923 2.616e-03
## B0GC_000024889 5.230e-03
## B0GC_000032443 -1.447e-03
## B0GC_000038860 3.496e-02
## B0GC_000069470 3.551e-02
## B0GC_000085702 9.292e-02
## B0GC_000092530 -1.023e-03
## B0GC_000092629 5.265e-03
## B0GC_000116990 7.406e-03
## B0GC_000118115 3.524e-05
## B0GC_000178751 8.303e-02
## B0GC_000258122 -3.597e-01
## B0GC_000259147 4.740e-02
## B0GC_000259233 -8.271e-02
## B0GC_000259352 -1.725e-01
## B0GC_000261697 -1.020e-01
## B0GC_000264231 -9.948e-01
## B0GC_000265243 -6.716e-04
## B0GC_000307139 -1.537e-02
## B0GC_000309494 -6.292e-02
## B0GC_000361366 -7.459e-02
## B0GC_000361410 -3.509e-01
## B0GC_000361518 -4.473e-02
## B0GC_000361521 -1.110e-01
## B0GC_000361629 -6.723e-02
## B0GC_000362154 -4.116e-02
## B0GC_000363872 -2.961e-01
## B0GC_000364792 -1.046e-02
## B0GC_000396048 -2.658e-02
## B0GC_000437136 2.834e-03
## B0GC_000456027 4.609e-02
## B0GC_000462072 -4.943e-02
## B0GC_000550505 -8.571e-02
```

## B0GC\_000550512 -1.455e-01  
## B0GC\_000550639 6.944e-03  
## B0GC\_000579958 -1.023e-03  
## B0GC\_000636436 1.717e-02  
## B0GC\_000636639 1.215e-01  
## B0GC\_000636714 -3.511e-02  
## B0GC\_000637323 3.702e-02  
## B0GC\_000638034 -5.355e-03  
## B0GC\_000639559 2.599e-03  
## B0GC\_000639602 3.192e-01  
## B0GC\_000654903 4.531e-03  
## B0GC\_000684921 4.784e-01  
## B0GC\_000685554 4.706e-02  
## B0GC\_000685680 8.543e-03  
## B0GC\_000686480 4.608e-02  
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## B0GC\_000687916 -2.036e-04  
## B0GC\_000688079 9.197e-02  
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## B0GC\_000948129 -4.431e-03  
## B0GC\_000948159 2.852e-03  
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## B0GC\_000970548 2.357e-01  
## B0GC\_000970727 1.353e-01  
## B0GC\_000970738 2.377e-02  
## B0GC\_000971042 1.332e-01  
## B0GC\_000971228 7.406e-03  
## B0GC\_000971379 1.273e-02  
## B0GC\_000971407 8.343e-03  
## B0GC\_000971482 4.411e-03  
## B0GC\_000971911 1.599e-02  
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## B0GC\_000972757 -1.087e-02  
## B0GC\_000973046 1.134e-01  
## B0GC\_000973565 7.348e-03  
## B0GC\_000974385 1.047e-02  
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## B0GC\_000975801 5.334e-02  
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## B0GC\_000987689 3.216e-02  
## B0GC\_001001027 1.015e-02  
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## B0GC\_001033584 2.712e-02  
## B0GC\_001033666 7.291e-02  
## B0GC\_001033669 -1.438e-02  
## B0GC\_001037523 5.230e-03

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## B0GC_001037897 3.180e-03
## B0GC_001038127 5.230e-03
## B0GC_001038483 1.194e-03
## B0GC_001040176 3.715e-02
## B0GC_001040813 -4.667e-03
## B0GC_001041328 8.694e-03
## B0GC_001062949 6.077e-01
## B0GC_001087197 3.029e-02
## B0GC_001101225 1.303e-01
## B0GC_001101492 1.466e-01
## B0GC_001215101 1.380e-02
## B0GC_001239041 -7.334e-03
## B0GC_001273770 -1.772e-03
## B0GC_001275273 -1.337e-02
## B0GC_001330265 -1.311e-01
## B0GC_001379896 -1.295e-01
## B0GC_001381033 -1.316e-02
## B0GC_001417378 -7.008e-03
## B0GC_001446969 -8.169e-03
## B0GC_001447024 7.782e-02
## B0GC_001447168 1.280e-02
## B0GC_001447265 4.581e-03
## B0GC_001447380 5.026e-02
## B0GC_001452586 1.870e-03
## B0GC_001510909 -3.680e-02
## B0GC_001515965 -3.874e-03
## B0GC_001624382 -9.637e-04
## B0GC_001624966 4.081e-02
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## B0GC_001655669 4.050e-02
## B0GC_001655696 1.759e-02
## B0GC_001655701 2.517e-01
## B0GC_001655818 1.332e-02
## B0GC_001655831 4.119e-02
## B0GC_001655856 8.756e-02
## B0GC_001656003 -1.192e-02
## B0GC_001656145 2.931e-02
## B0GC_001657949 5.947e-02
## B0GC_001658165 3.167e-02
## B0GC_001658626 1.802e-02
## B0GC_001659251 4.731e-02
## B0GC_001659791 1.015e-02
## B0GC_001675403 7.794e-03
## B0GC_001772854 -1.447e-03
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## B0GC_001863644 9.966e-05
## B0GC_001863701 1.986e-04
## B0GC_001863724 3.524e-05
## B0GC_001863770 3.524e-05
## B0GC_001916701 -5.510e-02
## B0GC_001987145 -8.800e-02
## B0GC_002015445 -7.837e-03
## B0GC_002026671 -1.685e-02
## B0GC_002120333 -6.433e-03
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## B0GC\_002120600 -5.531e-02  
## B0GC\_002188612 1.392e-02  
## B0GC\_002251140 1.850e-03  
## B0GC\_002251203 3.700e-03  
## B0GC\_002362160 -1.952e-02  
## B0GC\_002471041 -1.434e-02  
## B0GC\_002473720 1.613e-03  
## B0GC\_002475336 -8.040e-03  
## B0GC\_002513491 -4.080e-02  
## B0GC\_002659394 -6.753e-02  
## B0GC\_002839632 -3.825e-02  
## B0GC\_002840418 -1.668e-02  
## B0GC\_002840474 -1.782e-02  
## B0GC\_002840484 2.520e-03  
## B0GC\_002842748 -3.532e-02  
## B0GC\_002844734 -2.290e-02  
## B0GC\_002848214 -1.963e-02  
## B0GC\_002849342 -2.297e-02  
## B0GC\_002855787 -9.416e-03  
## B0GC\_002862016 -2.068e-03  
## B0GC\_002868046 -1.332e-02  
## B0GC\_002868289 -6.909e-03  
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## B0GC\_002886501 -6.537e-02  
## B0GC\_002887012 3.355e-02  
## B0GC\_002887121 1.654e-02  
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## B0GC\_002888876 7.114e-04  
## B0GC\_002889770 -3.610e-02  
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## B0GC\_002903258 1.552e-02  
## B0GC\_002911856 1.375e-02  
## B0GC\_002923505 -1.146e-02  
## B0GC\_002942886 -1.215e-02  
## B0GC\_002956160 2.672e-03  
## B0GC\_002961744 1.015e-02  
## B0GC\_002962205 2.754e-02  
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## B0GC\_002963593 -3.874e-03  
## B0GC\_002995588 1.232e-02  
## B0GC\_002999314 -5.434e-03  
## B0GC\_003009107 2.339e-02  
## B0GC\_003179898 -2.768e-02  
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## B0GC\_003292662 1.556e-01  
## B0GC\_003292739 3.598e-03

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## B0GC_003293701 3.116e-02
## B0GC_003318885 2.607e-02
## B0GC_003326108 4.315e-02
## B0GC_003326121 2.872e-02
## B0GC_003326158 1.727e-01
## B0GC_003326162 4.944e-02
## B0GC_003326307 1.015e-02
## B0GC_003326805 4.456e-02
## B0GC_003327323 3.464e-02
## B0GC_003342732 -9.837e-03
## B0GC_003359556 1.153e-02
## B0GC_003403105 3.163e-02
## B0GC_003405745 8.813e-03
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## B0GC_003474538 3.263e-02
## B0GC_003474566 3.768e-02
## B0GC_003474609 3.797e-02
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## B0GC_003474685 2.307e-02
## B0GC_003474692 2.307e-02
## B0GC_003474758 2.999e-02
## B0GC_003474872 1.332e-02
## B0GC_003475128 -8.869e-03
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## B0GC_003476712 2.067e-02
## B0GC_003477699 -1.983e-02
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## B0GC_003801417 7.348e-03
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## B0GC_003848594 1.732e-02
## B0GC_003848736 3.947e-02
## B0GC_003851301 1.413e-02
## B0GC_003943434 -5.273e-02
## B0GC_003982489 3.540e-03
## B0GC_004017409 4.329e-03
## B0GC_004033098 -9.761e-03
## B0GC_004332832 5.082e-02
## B0GC_004334495 1.095e-02
## B0GC_004471641 2.747e-02
## B0GC_004555822 -1.110e-02
## B0GC_004585250 -1.662e-02
## B0GC_004614042 2.475e-02
## B0GC_004719261 8.708e-03
## B0GC_004866528 2.999e-02

```



```

## BOGC_004897287 4.241e-02
## BOGC_005229678 -1.329e-02
## BOGC_005445377 7.348e-03
##
##
## Site scores (weighted sums of species scores)
##
##          CAP1      CAP2      CAP3      CAP4      CAP5      MDS1
## BAI01    -0.8514 -0.531708 -0.014648  0.506768  1.086803 -0.228068
## BAI04    -1.0369 -1.136306 -0.237236 -0.392449  0.004671 -0.169814
## BAI06    -1.1863 -0.750191 -0.088532 -0.629135 -0.627428 -0.681917
## BAI07    -1.1544 -0.504794 -0.237894  0.123895 -0.609981 -0.338059
## BPU04    -0.2177  2.069399 -0.173381  1.462899  0.488734 -0.640650
## BSC03     0.8349 -1.383023 -3.834971 -0.078437 -0.533750 -1.037683
## BSC04     1.0030 -0.544298 -1.877488 -0.045075  0.243773 -0.680186
## BSC05     0.3748 -0.767215 -4.663746 -0.188575 -0.398322  0.009936
## BV01     -0.5250 -0.365057  0.385555  2.593956 -0.696794  0.445837
## BV02     -0.9495 -0.312301  0.515652 -1.491350  1.574127 -0.377769
## BValp01  -0.9952 -0.739960  0.212306 -0.465101  2.189868  0.204741
## BValp02  -0.7679 -0.440342  0.347381 -0.492296  2.735037 -0.869503
## BValp03  -1.2078 -0.762765 -0.055109 -1.362672 -1.222239 -1.150651
## CBR02     0.3227  2.243847  0.118622 -0.189039 -0.211897  0.962329
## CBR03     0.3116  2.300239 -0.562719 -0.236681  0.843224  1.250801
## CON02    -0.6592  0.659215 -0.005137 -1.216503 -1.036545 -0.386846
## CON03    -0.6329  0.332743  0.448686 -2.408732  0.209823 -0.815829
## CON06    -0.9015 -0.424177 -0.222096 -1.510176 -1.241704 -1.448959
## CR01     -0.8516 -0.201913 -0.547759  2.424907  1.156801  1.154797
## CR06     -1.1990 -0.279654 -0.393736  1.544835 -0.498068 -0.125198
## CR09     -0.9266 -0.063583 -0.300341  2.538498 -0.332938  0.183434
## ERT05     1.0246 -1.454521  0.451374 -0.470996  0.566695  0.970514
## ES03      0.0124 -0.145255  0.032492  2.770231 -1.414757 -0.073922
## ES10      0.5630  0.225873  0.360047  1.824550 -1.727200  0.002545
## FRE03     0.8844  1.182769  0.030723 -0.466361  1.162091  0.530796
## GMG01     0.5260  2.155593  0.102370 -1.070286  1.166147  0.281522
## GMG06     0.2492  2.133748 -0.012785 -0.300527 -0.178286  0.733454
## HOR01     1.0263 -0.194781  0.539489  0.304280 -0.855446 -0.392523
## HOR04    -0.9821 -0.429245  0.295027 -2.307606 -1.108985 -1.409771
## HOR05    -0.9856 -0.495712  0.070920 -1.070699  0.445351 -0.710576
## HOR06    -0.2718  0.134509  0.036368 -0.586281 -0.759639  0.427260
## LLA07     1.1899 -0.943293  0.338613  0.206716  1.367798 -0.279829
## MA_alp01 -1.0648 -0.322562  0.187306 -1.341810 -1.295633 -0.950393
## MA_alp02 -0.5409  0.036284  0.608376 -1.120648 -0.363453  0.135089
## MAR01     0.7116  0.078586  0.595839 -0.496321 -0.390313  1.787840
## MAR02     1.2602 -0.009454  1.020416 -0.714589 -0.073883  1.638504
## MAR03     0.7693  1.127553  0.717650 -0.743485 -1.084524  1.883422
## MAR04     1.2249 -1.645131  0.775919  0.542966  0.425081 -0.443883
## MAR05     1.2664 -0.971272  0.573804  0.222234 -0.659749  0.496425
## MAR06     0.9993 -0.360500  0.019385 -0.708193 -0.534850  0.629071
## MAR09     1.0684 -0.124055  0.387895  0.060412  0.606198  0.312695
## MAR10     1.2364 -1.171931  0.276870  0.372063  0.737746 -0.241136
## MT02      0.4589  1.709301  0.290501 -0.025571 -1.561846  1.408001
## MTL03     1.0997 -1.233715  0.089968  0.009731  0.728287 -0.012329
## MUN02    -0.8573  0.005428 -0.067124 -0.100395  0.810503  0.650797
## MUN03    -0.3591 -0.439112 -0.316214 -1.109843  0.912638  0.910832

```

## PA01	0.9584	0.839069	0.288224	-0.047541	2.555138	-0.601638
## PA02	0.6044	-0.037620	0.997169	1.775298	-1.156131	0.535012
## PA05	0.7990	0.303719	0.948124	1.493401	-1.189808	0.377813
## RAT07	-0.3847	0.461250	0.376354	-0.919269	3.991216	0.180803
## S003	-1.2625	-0.428210	-0.196354	-0.404915	-0.982540	-0.925852
## S004	-1.2131	-0.517744	0.016968	-0.698500	-1.320374	-0.860795
## S005	-1.0676	-0.583896	0.228797	-0.727163	-0.160487	-0.860702
## S008	-0.7019	-0.027284	-0.667846	2.225659	1.635071	0.168011
## S009	-0.8839	-0.146811	-0.143985	2.830501	-0.038129	0.133635
## SON02	0.4128	1.254163	0.008703	-0.375408	0.005282	0.259911
## SON03	0.6198	1.818637	0.443672	-0.412280	-0.531447	1.240539
## T29	-0.4539	-0.440232	-0.008028	-0.598819	-1.246323	-0.890186
## T46	-0.4593	-0.854976	0.209977	-0.110265	0.152147	-0.410659
## T52	-0.5936	0.045273	0.111724	0.912814	0.009618	0.116509
## T54	-0.5621	0.532856	0.254690	1.562047	-1.506873	-0.849087
## TAU03	0.7976	1.421607	-0.038686	-0.395397	-0.927993	1.020541
## TAU04	0.9650	-0.111000	-0.912151	-0.277250	0.984489	-0.665605
## TAU06	0.7741	1.057475	-0.024706	-0.722514	-0.153111	-0.705134
## TOR01	1.0857	-1.266488	0.716722	0.511000	0.143478	-0.603437
## TOR02	1.2727	-0.567047	1.171997	0.209492	-0.306390	-0.204831
##						
##						
## Site constraints (linear combinations of constraining variables)						
##						
##	CAP1	CAP2	CAP3	CAP4	CAP5	MDS1
## BAI01	-0.51490	-0.44551	0.058510	1.1289911	-0.007724	-0.228068
## BAI04	-1.02635	-0.41302	0.048049	-0.8421119	-0.076443	-0.169814
## BAI06	-0.92977	-0.28686	-0.050262	-0.7969047	-0.114342	-0.681917
## BAI07	-0.90759	-0.29318	-0.225486	-0.7369748	-0.090637	-0.338059
## BPU04	0.46342	1.77512	-0.259674	0.5748006	0.031943	-0.640650
## BSC03	1.16010	0.02853	-1.973084	0.0414987	-0.381674	-1.037683
## BSC04	1.29312	-0.18993	-2.190910	0.0746821	-0.325175	-0.680186
## BSC05	0.53181	-0.94337	-4.209647	-0.4202472	0.415433	0.009936
## BV01	-0.63834	-0.66961	0.119220	2.1453669	-0.171870	0.445837
## BV02	-0.58954	-0.51102	0.263935	-0.5826428	-0.150938	-0.377769
## BValp01	-1.10316	-0.57130	0.025338	-0.8828362	0.683225	0.204741
## BValp02	-0.04740	-0.51071	0.652558	-0.3713116	1.271895	-0.869503
## BValp03	-0.52911	-0.64580	0.125518	-0.6324144	-0.152763	-1.150651
## CBR02	0.06300	1.62589	0.080019	-0.2308189	-0.839967	0.962329
## CBR03	-0.63863	1.97746	-0.184638	-0.5338310	-0.712488	1.250801
## CON02	-0.26571	0.61164	0.209318	-0.4325866	-0.535412	-0.386846
## CON03	-0.33847	0.74314	0.347461	-0.4338973	-0.572613	-0.815829
## CON06	-0.52143	0.67153	0.174078	-0.5288763	-0.482667	-1.448959
## CR01	-1.35492	-0.52899	-0.774169	1.8032160	0.156623	1.154797
## CR06	-1.35294	-0.01645	-0.259665	1.7589382	-0.094136	-0.125198
## CR09	-1.09360	0.20815	-0.053137	0.0382889	0.488360	0.183434
## ERT05	0.40083	-1.67589	-0.707902	-0.3302462	0.183561	0.970514
## ES03	0.39046	-0.02486	0.371220	-0.1731788	-0.565927	-0.073922
## ES10	0.94431	0.08830	0.821325	0.1040195	-0.757151	0.002545
## FRE03	0.64670	0.67084	-0.749308	-0.0741636	-0.001998	0.530796
## GMG01	0.46636	1.42763	-0.496255	-0.0852058	0.995976	0.281522
## GMG06	0.02780	1.62625	0.132533	-0.2374090	0.120942	0.733454
## HOR01	1.07612	0.16229	0.751840	0.1634139	-0.751926	-0.392523
## HOR04	-0.55337	-0.08746	0.115401	-0.5937289	-0.281082	-1.409771

```

## HOR05      -0.67862 -0.19786 -0.070046 -0.6662670 -0.193694 -0.710576
## HOR06      -0.71409 -0.15224  0.194964 -0.3625554 -0.070343  0.427260
## LLA07       1.10313 -0.35368  0.740574  0.1512928  0.766579 -0.279829
## MA_alp01    -0.40381 -0.50515  0.336652 -0.5562651 -0.208023 -0.950393
## MA_alp02    -0.47523 -0.47577  0.336818 -0.5862926 -0.241608  0.135089
## MAR01      -0.01307 -0.75457  0.408768 -0.3952996 -0.297375  1.787840
## MAR02       0.51891 -0.92522  0.619266 -0.1576339 -0.200497  1.638504
## MAR03       0.12801 -0.68203  0.572469 -0.3182157 -0.201104  1.883422
## MAR04       1.26685 -1.00308  0.776581  0.1814927  0.310886 -0.443883
## MAR05       0.88591 -1.13726  0.572163 -0.0119487 -0.361071  0.496425
## MAR06       0.75493 -1.39810  0.007242 -0.1165213 -0.042681  0.629071
## MAR09       0.64368 -0.05806  0.601361 -0.0499902 -0.569316  0.312695
## MAR10       1.10981 -0.99117  0.240069  0.0779529 -0.396364 -0.241136
## MT02        -0.03018  1.05603  0.155413 -0.2991298  0.007544  1.408001
## MTL03       0.90894 -0.93612  0.679974  0.0217324  0.353236 -0.012329
## MUN02       -0.87979 -0.75867  0.118158 -0.7950329 -0.039404  0.650797
## MUN03      -1.11290 -0.52139 -0.208990 -0.8709884  0.055933  0.910832
## PA01        0.64888  1.45519  0.532788  0.0674939  2.917523 -0.601638
## PA02        0.39424  0.27029 -0.012323 -0.1727929 -0.357376  0.535012
## PA05        0.64238  0.68380  0.608008  0.0006326 -0.051244  0.377813
## RAT07       -0.03886  0.14433  0.506337 -0.3163961  4.017912  0.180803
## S003        -0.84555 -0.01204  0.069054 -0.7057902 -0.221481 -0.925852
## S004        -0.84299 -0.08231 -0.036701 -0.7299257 -0.187473 -0.860795
## S005        -0.77443 -0.06678  0.071352 -0.1361030 -0.235772 -0.860702
## S008        -0.94949  0.17793 -0.318736  1.3855011  1.485112  0.168011
## S009       -1.01945 -0.12949 -0.143241  1.8757413 -0.165764  0.133635
## SON02       0.18706  1.13652 -0.409875 -0.2329915 -0.583683  0.259911
## SON03       0.20779  1.28305 -0.081941 -0.1965839 -0.677875  1.240539
## T29         -0.38844  0.26625  0.237539 -0.4016442 -0.428456 -0.890186
## T46         -0.08207  0.07195  0.284140  1.4541714 -0.494944 -0.410659
## T52         -0.54904  0.31998  0.206372  0.1306345 -0.408721  0.116509
## T54         -0.17301  0.01703  0.198321  3.1506752 -0.477885 -0.849087
## TAU03       0.67191  0.30386 -1.152754 -0.1073725  0.278012  1.020541
## TAU04       0.91519  0.22920 -1.072424  0.0002935  0.267897 -0.665605
## TAU06       1.21419  0.71978  0.422049  0.2409344 -0.910421 -0.705134
## TOR01       1.27211 -0.56099  1.003461  0.2192914 -0.577454 -0.603437
## TOR02       1.43829 -0.23600  0.844948  0.3140704  0.858371 -0.204831

```

```
##
```

```
##
```

```
## Biplot scores for constraining variables
```

```
##
```

```

##              CAP1      CAP2      CAP3      CAP4      CAP5 MDS1
## GDD           0.3830  0.72169 -0.53958  0.01863 -0.202448    0
## Sphagnum      -0.4045 -0.05841 -0.05115  0.91120 -0.005755    0
## Acrocarp_mosses 0.2187  0.22634  0.15217  0.01733  0.936734    0
## pH            0.9786 -0.14370 -0.14180 -0.02512 -0.031773    0
## Mg            0.4779 -0.05854 -0.87578 -0.03028  0.015781    0

```

### 5.3 dbRDA 18S

```

analysis_18S <- capscale(datos.trans_18s ~ WTD+GDD+Rad+
                        Sphagnum+Brown_mosses+Acrocarp_mosses+Liverworts+
                        Bryophytes+Vascular_plants+

```

```

        pH+Prec+K+Mg+Mn+Na+P+S+Si+Al+Ca+Fe, Metadata,
        dist="bray")

anova (analisis_18S) # 0.001 *** - it is significant

## Permutation test for capscale under reduced model
## Permutation: free
## Number of permutations: 999
##
## Model: capscale(formula = datos.trans_18s ~ WTD + GDD + Rad + Sphagnum + Brown_mosses + Acrocarp_mos
##           Df SumOfSqs      F Pr(>F)
## Model      20   12.266 2.1522 0.001 ***
## Residual   45   12.824
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

adjR2.tbrda <- RsquareAdj (analisis_18S)$adj.r.squared

#Forward selection
sel.fs <- forward.sel (Y = datos.trans_18s , X=x, adjR2thresh = adjR2.tbrda)

## Testing variable 1
## Testing variable 2
## Testing variable 3
## Testing variable 4
## Testing variable 5
## Testing variable 6
## Testing variable 7
## Procedure stopped (alpha criteria): pvalue for variable 7 is 0.059000 (> 0.050000)
sel.fs

##      variables order      R2      R2Cum AdjR2Cum      F pvalue
## 1          pH        5 0.16224603 0.1622460 0.1491561 12.394744 0.001
## 2          GDD        2 0.03065437 0.1929004 0.1672782 2.392797 0.001
## 3          Mg       16 0.03368428 0.2265847 0.1891614 2.700264 0.001
## 4          S       20 0.02192105 0.2485057 0.1992274 1.779367 0.009
## 5          Al       13 0.02064942 0.2691552 0.2082514 1.695251 0.018
## 6 Brown_mosses      8 0.02016456 0.2893197 0.2170471 1.674043 0.025

analisis_18S_fs <- capscale(datos.trans_18s ~ GDD+Brown_mosses+
        pH+Mg+S+Al, Metadata,
        dist="bray")
summary(analisis_18S_fs)

##
## Call:
## capscale(formula = datos.trans_18s ~ GDD + Brown_mosses + pH +      Mg + S + Al, data = Metadata, di
##
## Partitioning of squared Bray distance:
##           Inertia Proportion
## Total          25.090      1.0000
## Constrained      7.831      0.3121
## Unconstrained   17.259      0.6879
##
## Eigenvalues, and their contribution to the squared Bray distance

```

```

##
## Importance of components:
##          CAP1      CAP2      CAP3      CAP4      CAP5      CAP6      MDS1
## Eigenvalue      4.9978 0.93221 0.70318 0.4668 0.41549 0.31539 2.38455
## Proportion Explained 0.1992 0.03715 0.02803 0.0186 0.01656 0.01257 0.09504
## Cumulative Proportion 0.1992 0.23635 0.26438 0.2830 0.29954 0.31211 0.40715
##          MDS2      MDS3      MDS4      MDS5      MDS6      MDS7      MDS8
## Eigenvalue      1.50026 1.27526 0.99324 0.94692 0.83755 0.74233 0.7177
## Proportion Explained 0.05979 0.05083 0.03959 0.03774 0.03338 0.02959 0.0286
## Cumulative Proportion 0.46694 0.51777 0.55736 0.59510 0.62848 0.65807 0.6867
##          MDS9      MDS10     MDS11     MDS12     MDS13     MDS14     MDS15
## Eigenvalue      0.66776 0.59447 0.56685 0.55109 0.4716 0.4115 0.38611
## Proportion Explained 0.02661 0.02369 0.02259 0.02196 0.0188 0.0164 0.01539
## Cumulative Proportion 0.71329 0.73698 0.75957 0.78154 0.8003 0.8167 0.83212
##          MDS16     MDS17     MDS18     MDS19     MDS20     MDS21     MDS22
## Eigenvalue      0.37322 0.36271 0.32699 0.31226 0.25718 0.247352 0.221183
## Proportion Explained 0.01488 0.01446 0.01303 0.01245 0.01025 0.009859 0.008816
## Cumulative Proportion 0.84700 0.86146 0.87449 0.88693 0.89718 0.907043 0.915859
##          MDS23     MDS24     MDS25     MDS26     MDS27     MDS28
## Eigenvalue      0.207610 0.194180 0.181324 0.164013 0.153597 0.148513
## Proportion Explained 0.008275 0.007739 0.007227 0.006537 0.006122 0.005919
## Cumulative Proportion 0.924133 0.931872 0.939099 0.945636 0.951758 0.957677
##          MDS29     MDS30     MDS31     MDS32     MDS33     MDS34
## Eigenvalue      0.133441 0.127228 0.11266 0.098672 0.090015 0.083941
## Proportion Explained 0.005318 0.005071 0.00449 0.003933 0.003588 0.003346
## Cumulative Proportion 0.962996 0.968067 0.97256 0.976490 0.980077 0.983423
##          MDS35     MDS36     MDS37     MDS38     MDS39     MDS40
## Eigenvalue      0.075881 0.070130 0.065369 0.050504 0.043239 0.033202
## Proportion Explained 0.003024 0.002795 0.002605 0.002013 0.001723 0.001323
## Cumulative Proportion 0.986447 0.989242 0.991848 0.993861 0.995584 0.996907
##          MDS41     MDS42     MDS43     MDS44     MDS45
## Eigenvalue      0.029699 0.0230934 0.0151629 0.0066079 0.0030304
## Proportion Explained 0.001184 0.0009204 0.0006043 0.0002634 0.0001208
## Cumulative Proportion 0.998091 0.9990115 0.9996159 0.9998792 1.0000000
##
## Accumulated constrained eigenvalues
## Importance of components:
##          CAP1      CAP2      CAP3      CAP4      CAP5      CAP6
## Eigenvalue      4.9978 0.9322 0.7032 0.4668 0.41549 0.31539
## Proportion Explained 0.6382 0.1190 0.0898 0.0596 0.05306 0.04027
## Cumulative Proportion 0.6382 0.7573 0.8471 0.9067 0.95973 1.00000
##
## Scaling 2 for species and site scores
## * Species are scaled proportional to eigenvalues
## * Sites are unscaled: weighted dispersion equal on all dimensions
## * General scaling constant of scores: 6.265441
##
##
## Species scores
##
##          CAP1      CAP2      CAP3      CAP4      CAP5
## BOGS_000001575 -0.0159081 -1.583e-02 -0.0038474 2.105e-02 0.0053668
## BOGS_000014580 1.6355569 -2.769e-01 0.1430701 2.028e-01 -0.2966855
## BOGS_000032386 0.0373785 -1.527e-01 0.0876460 -1.493e-01 -0.0449055

```

```

## BOGS_000124623 -0.0953194 -7.424e-02 0.0896508 -2.009e-01 0.0142152
## BOGS_000124699 -0.0787855 -9.913e-02 0.1180751 -2.080e-01 0.0231789
## BOGS_000124963 -0.4969998 -1.286e-01 -0.0919830 2.160e-01 -0.0984183
## BOGS_000125182 -0.1266540 -7.467e-03 0.0359470 1.433e-01 0.0057309
## BOGS_000125502 -0.3810045 8.194e-02 -0.0191037 1.096e-01 0.0372034
## BOGS_000125634 -0.3645461 -1.808e-01 0.0290376 2.102e-01 0.2552819
## BOGS_000126337 0.1645930 3.520e-01 0.1339655 -9.989e-02 -0.0124073
## BOGS_000128470 -0.0165875 -1.267e-02 0.0017029 -2.528e-02 -0.0166570
## BOGS_000128660 -0.0338323 -1.786e-02 -0.0237607 5.512e-03 -0.0373837
## BOGS_000129899 -0.0435841 -3.581e-02 -0.0083019 2.193e-02 0.0113557
## BOGS_000133938 -0.0727224 -8.542e-02 0.0467702 -1.360e-02 0.0426781
## BOGS_000133943 -0.3679997 -1.558e-01 -0.0353010 2.522e-01 0.0752949
## BOGS_000133948 -0.2251566 -6.413e-03 0.0041926 7.090e-02 -0.0547385
## BOGS_000133951 -0.3138212 -1.290e-01 -0.0383730 1.398e-01 0.0227596
## BOGS_000134033 -0.0764681 -4.903e-03 0.0168169 2.136e-02 -0.0638262
## BOGS_000134036 -0.1316104 -5.293e-02 -0.0187819 -1.371e-02 -0.0001365
## BOGS_000134154 -0.0816966 -6.859e-02 0.0362660 3.863e-02 -0.0259764
## BOGS_000134551 -0.0148807 7.971e-03 -0.0143934 -8.449e-03 -0.0164715
## BOGS_000134630 -0.9619571 -4.349e-01 0.4915483 -2.447e-01 0.0787167
## BOGS_000134681 0.0102084 1.490e-03 0.0004054 1.232e-02 -0.0041097
## BOGS_000137869 -0.1465513 1.112e-01 -0.0474575 6.133e-02 0.0173064
## BOGS_000138001 -0.0157447 2.225e-03 -0.0053039 -1.855e-03 -0.0100130
## BOGS_000150046 0.3362282 -1.220e-01 0.0748470 5.555e-02 -0.0829329
## BOGS_000150326 -0.0315441 8.341e-03 -0.0594525 -2.453e-04 -0.0046834
## BOGS_000158824 -0.1321093 2.359e-01 0.1617799 1.805e-01 -0.0157793
## BOGS_000159330 -0.0031423 -7.183e-03 0.0095841 3.340e-02 0.0030498
## BOGS_000167353 -0.0829925 4.760e-03 0.2230978 6.256e-02 0.0148922
## BOGS_000172272 0.0155195 5.316e-02 0.2014013 -1.075e-02 0.0564820
## BOGS_000172334 0.0766842 2.671e-02 0.0484855 -5.487e-02 -0.0221851
## BOGS_000172384 0.0011519 9.965e-02 -0.0332039 -2.772e-02 -0.0249220
## BOGS_000172410 0.1710965 1.179e-01 0.0137264 -1.216e-01 -0.0724037
## BOGS_000172574 0.8300646 -9.638e-02 -0.1168474 5.926e-02 0.0051221
## BOGS_000172793 -0.0302729 3.180e-02 0.0047499 5.829e-04 -0.0203494
## BOGS_000172990 0.0088801 -8.592e-03 -0.0211879 1.496e-02 -0.0224488
## BOGS_000173013 0.6656924 -2.400e-01 -0.1887115 -1.713e-02 -0.0552868
## BOGS_000174702 -0.0108031 -2.671e-03 0.0426531 -3.618e-03 0.0233194
## BOGS_000174886 0.0684342 9.941e-02 0.1169502 -5.497e-02 -0.0309138
## BOGS_000175854 0.0798191 -1.078e-02 0.0034903 -7.376e-02 0.1182791
## BOGS_000176094 0.2873769 3.026e-02 -0.1384003 4.474e-02 0.0923237
## BOGS_000177666 -0.0101410 5.335e-03 0.0133188 1.858e-02 0.0045179
## BOGS_000178089 0.0589215 9.217e-02 0.0745987 1.094e-02 0.0053176
## BOGS_000178967 0.3072325 -8.342e-02 -0.0114538 -1.199e-02 0.0967460
## BOGS_000179838 0.3315982 4.897e-02 -0.1053489 -1.715e-02 -0.0132266
## BOGS_000180056 -0.0303416 4.739e-01 0.1272296 -1.142e-01 -0.0945444
## BOGS_000180074 0.1033396 4.802e-02 -0.0240043 4.095e-02 -0.0844022
## BOGS_000180301 0.0678965 2.258e-01 0.0476032 3.240e-03 -0.0331458
## BOGS_000181801 0.0603076 1.315e-01 -0.0110315 1.361e-02 0.0150298
## BOGS_000185116 -0.0023618 3.718e-02 -0.0179053 -1.620e-02 -0.0192177
## BOGS_000186619 0.0824496 7.347e-02 -0.0150976 4.151e-02 -0.0495753
## BOGS_000187217 0.2151682 1.584e-02 0.0335349 -1.779e-01 0.0172974
## BOGS_000188077 0.2354279 -2.509e-01 -0.0864533 -4.554e-02 -0.0834416
## BOGS_000199997 -1.1802014 1.518e-01 -0.5199404 -2.864e-01 -0.2981061
## BOGS_000200166 -0.1373714 1.935e-02 -0.0804811 -7.175e-03 -0.0177261
## BOGS_000201530 -0.0256004 7.167e-03 -0.0033443 -1.347e-02 -0.0103460

```

```

## BOGS_000203482 -0.2081443 4.549e-02 -0.1139720 -4.694e-02 -0.0771790
## BOGS_000206295 0.0444263 9.691e-03 0.0518922 -1.187e-05 -0.0149348
## BOGS_000208231 -0.0224381 7.754e-05 0.0019762 1.981e-03 -0.0015086
## BOGS_000209030 -0.0224381 7.754e-05 0.0019762 1.981e-03 -0.0015086
## BOGS_000211135 0.2050077 -1.077e-01 -0.1101917 9.763e-02 0.0088878
## BOGS_000211172 0.0359558 1.042e-03 0.0289559 1.509e-02 -0.0205404
## BOGS_000211188 0.1082769 -3.166e-03 0.0818346 1.367e-02 -0.0485183
## BOGS_000211221 0.1338924 1.223e-01 0.1219795 -5.184e-02 0.0271329
## BOGS_000211531 0.2228546 -7.643e-03 0.0409081 1.350e-02 0.0490815
## BOGS_000211710 0.0061349 -2.388e-02 -0.0093109 7.603e-04 -0.0223764
## BOGS_000213701 0.0824385 1.950e-02 0.0143495 -3.783e-02 -0.0014666
## BOGS_000213947 0.1505750 -4.409e-02 0.0963472 -1.394e-02 -0.0239491
## BOGS_000214439 0.1082492 3.659e-02 -0.0443354 -1.046e-01 -0.0331264
## BOGS_000214547 0.0309951 1.949e-02 0.0228490 1.274e-02 -0.0052961
## BOGS_000214733 0.1717460 8.816e-02 0.0238472 -2.334e-02 0.0249703
## BOGS_000215413 0.0680593 -1.331e-02 -0.0155564 5.402e-02 -0.0403202
## BOGS_000215414 0.0417930 -3.386e-02 -0.0087886 -7.881e-03 -0.0205497
## BOGS_000215489 0.0102084 1.490e-03 0.0004054 1.232e-02 -0.0041097
## BOGS_000215518 0.0098094 7.919e-02 -0.0414926 -1.738e-02 -0.0297336
## BOGS_000215595 0.0374016 1.352e-02 -0.0058613 2.611e-02 -0.0079318
## BOGS_000216005 0.0219969 -4.175e-03 0.0270455 -5.625e-03 -0.0015842
## BOGS_000216464 0.0102084 1.490e-03 0.0004054 1.232e-02 -0.0041097
## BOGS_000218184 0.0102084 1.490e-03 0.0004054 1.232e-02 -0.0041097
## BOGS_000218580 0.0429837 3.147e-02 -0.0347320 -1.281e-02 -0.0106574
## BOGS_000219289 0.0183334 -2.365e-02 0.0073961 -8.891e-03 -0.0019471
## BOGS_000220652 0.0332361 -1.799e-03 0.0065313 2.183e-02 -0.0162412
## BOGS_000221316 0.0096993 1.027e-02 0.0157668 1.242e-02 -0.0222447
## BOGS_000223957 0.0206995 6.580e-03 -0.0154460 1.703e-02 0.0174175
## BOGS_000224298 0.0115383 5.198e-02 -0.0283849 -5.522e-03 -0.0156653
## BOGS_000224378 0.0235729 -2.289e-02 -0.0042503 1.039e-02 -0.0006625
## BOGS_000225378 0.0754904 9.958e-03 -0.1021506 -3.925e-02 0.1359517
## BOGS_000246158 -0.0530552 -4.710e-03 -0.0257416 9.978e-03 -0.0252652
## BOGS_000246536 -0.2071546 -6.872e-02 0.1338301 5.941e-02 -0.0233133
## BOGS_000247056 0.0030166 5.704e-02 0.0344037 1.956e-02 -0.0171014
## BOGS_000247573 -0.0135360 -1.235e-03 0.0065229 -1.381e-02 -0.0033975
## BOGS_000260111 -0.0248952 -2.877e-02 0.0115691 -1.702e-03 -0.0005485
## BOGS_000265594 0.0528863 4.844e-02 -0.0674630 -3.932e-02 0.1126918
## BOGS_000265762 0.0907464 -4.558e-02 -0.0400008 3.439e-02 0.0112536
## BOGS_000266220 -0.0535327 -5.837e-02 -0.0300120 2.748e-02 -0.0535435
## BOGS_000267187 -0.0063136 -1.089e-02 -0.0060653 2.343e-03 -0.0131857
## BOGS_000268867 0.0058755 2.529e-02 0.0026067 4.674e-03 0.0135135
## BOGS_000269084 -0.0084754 -1.155e-02 -0.0041892 6.244e-03 -0.0093522
## BOGS_000269350 -0.0214627 -3.356e-02 -0.0060970 -3.723e-02 -0.0173623
## BOGS_000269461 0.0880618 -3.829e-02 -0.0441019 1.741e-02 -0.0621419
## BOGS_000287071 -0.0220169 6.416e-03 -0.0114704 -1.468e-03 -0.0195821
## BOGS_000287072 0.0016892 3.568e-02 0.0259220 7.652e-03 0.0008883
## BOGS_000287443 0.0531788 3.425e-02 0.0455760 3.568e-03 0.0194507
## BOGS_000290336 0.1455306 4.283e-01 0.0139356 1.195e-01 0.0333401
## BOGS_000290621 -0.0491325 1.596e-02 -0.0480332 -2.907e-02 -0.0127911
## BOGS_000290638 -0.0055365 -1.786e-02 0.0064679 -1.958e-02 -0.0037486
## BOGS_000290815 -0.0198340 6.255e-03 -0.0181868 -1.085e-02 -0.0052714
## BOGS_000291713 -0.0292985 9.706e-03 -0.0298464 -1.822e-02 -0.0075197
## BOGS_000299781 0.2910763 -1.501e-01 -0.0278937 -1.639e-02 -0.0210659
## BOGS_000309758 0.0219969 -4.175e-03 0.0270455 -5.625e-03 -0.0015842

```

```

## BOGS_000309837 0.0219969 -4.175e-03 0.0270455 -5.625e-03 -0.0015842
## BOGS_000313064 -0.0257521 6.929e-02 0.0381500 3.707e-02 0.0194492
## BOGS_000313142 -0.0750410 -2.429e-02 -0.0044155 8.996e-02 0.0619937
## BOGS_000322074 -0.1888583 -2.268e-02 -0.0754930 2.401e-02 -0.0845677
## BOGS_000322220 -0.0655036 -1.478e-02 -0.0341683 3.114e-02 0.0003730
## BOGS_000322357 -0.0214337 -9.148e-03 0.0205149 -1.801e-02 0.0160299
## BOGS_000323061 -0.0129092 -1.243e-02 0.0174935 -1.053e-02 0.0126775
## BOGS_000326807 -0.0198340 6.255e-03 -0.0181868 -1.085e-02 -0.0052714
## BOGS_000328696 -0.0198633 -1.858e-03 -0.0089520 2.344e-03 -0.0016636
## BOGS_000334667 -0.0332743 1.782e-02 -0.0321847 -1.889e-02 -0.0368315
## BOGS_000346557 -0.0143415 7.544e-03 0.0188357 2.627e-02 0.0063893
## BOGS_000366488 -0.0231578 -9.077e-03 -0.0102573 3.299e-03 0.0003263
## BOGS_000378733 -0.0096034 4.229e-03 0.0114417 8.191e-03 -0.0205725
## BOGS_000392882 -0.0202108 -4.997e-03 0.0797966 -6.768e-03 0.0436266
## BOGS_000393583 -0.0042570 1.228e-02 0.0196131 -5.338e-03 -0.0034741
## BOGS_000418510 0.0521120 1.931e-02 -0.0722763 2.645e-02 0.0461495
## BOGS_000419105 0.0722470 -2.058e-02 -0.0418841 6.658e-03 0.0301814
## BOGS_000421133 0.0481318 -2.000e-02 -0.0659401 -1.741e-02 0.0530734
## BOGS_000423554 0.1616901 -9.063e-02 -0.0396692 -1.877e-02 0.0087594
## BOGS_000426320 0.0197277 -7.438e-03 -0.0097724 -3.176e-03 -0.0121341
## BOGS_000426631 0.0344353 -9.204e-03 0.0131702 -1.834e-02 -0.0135621
## BOGS_000428278 0.0483800 -5.526e-03 -0.0449765 -7.387e-02 -0.0127072
## BOGS_000430467 0.0421262 -1.870e-02 0.0118043 1.330e-02 -0.0298609
## BOGS_000430575 0.0235015 -1.272e-03 0.0046183 1.543e-02 -0.0114843
## BOGS_000432024 0.0239364 1.214e-02 -0.0279155 1.032e-02 0.0215059
## BOGS_000432338 0.0344306 2.343e-02 0.0248609 2.724e-02 -0.0156131
## BOGS_000433089 0.0087701 5.481e-02 -0.0040084 -1.115e-03 -0.0261774
## BOGS_000433683 0.0257999 -1.064e-02 -0.0009470 3.265e-03 -0.0129506
## BOGS_000437337 -0.0080599 4.116e-03 0.0066925 1.556e-03 -0.0104533
## BOGS_000443271 0.0009562 -9.688e-03 0.0061192 -2.358e-02 -0.0041340
## BOGS_000448391 0.0684672 1.722e-01 0.0825299 7.292e-02 -0.0050572
## BOGS_000448407 0.0162397 1.520e-01 0.0832196 4.724e-02 -0.0147502
## BOGS_000448810 0.0011945 2.523e-02 0.0183296 5.411e-03 0.0006281
## BOGS_000455988 0.0574244 -1.832e-03 -0.0865099 -2.506e-02 0.0979677
## BOGS_000458031 0.1694168 1.047e-01 -0.3054069 -1.349e-01 0.4119308
## BOGS_000458037 0.0709612 4.664e-02 -0.1437536 -8.703e-02 0.2106736
## BOGS_000459244 0.0338511 1.716e-02 -0.0394785 1.460e-02 0.0304140
## BOGS_000460736 0.0139001 1.480e-02 -0.0104796 1.068e-02 0.0035525
## BOGS_000461069 -0.0135136 1.272e-01 0.0649603 5.559e-02 0.0070211
## BOGS_000461246 0.0091002 2.870e-02 0.0148133 1.529e-02 -0.0191313
## BOGS_000474205 -0.0205713 8.617e-02 0.0090708 1.002e-02 -0.0054651
## BOGS_000479425 0.0159566 -4.759e-02 -0.0209771 -5.106e-02 0.0289075
## BOGS_000481274 0.0112830 -3.365e-02 -0.0148331 -3.611e-02 0.0204407
## BOGS_000485412 0.0515999 -2.128e-02 -0.0018941 6.529e-03 -0.0259011
## BOGS_000495343 0.0811577 2.677e-02 0.0259760 2.257e-02 0.0466472
## BOGS_000495386 0.0235729 -2.289e-02 -0.0042503 1.039e-02 -0.0006625
## BOGS_000495417 0.0103419 4.377e-03 0.0079133 1.324e-02 -0.0090667
## BOGS_000499417 0.0298340 1.593e-02 -0.0276608 -5.090e-02 0.0832348
## BOGS_000510152 0.0131381 5.656e-02 0.0058288 1.045e-02 0.0302171
## BOGS_000527403 0.0158312 2.988e-02 0.0074076 1.745e-02 0.0129442
## BOGS_000527436 0.0197277 -7.438e-03 -0.0097724 -3.176e-03 -0.0121341
## BOGS_000533827 -0.0344042 -3.218e-03 -0.0155054 4.060e-03 -0.0028814
## BOGS_000564122 0.0197277 -7.438e-03 -0.0097724 -3.176e-03 -0.0121341
## BOGS_000591711 0.0035420 -1.379e-02 -0.0053756 4.390e-04 -0.0129190

```



```

## BOGS_000640427 -0.0198340 6.255e-03 -0.0181868 -1.085e-02 -0.0052714
## CAP6
## BOGS_000001575 -9.892e-03
## BOGS_000014580 3.615e-02
## BOGS_000032386 7.012e-03
## BOGS_000124623 1.482e-02
## BOGS_000124699 -6.775e-04
## BOGS_000124963 9.338e-02
## BOGS_000125182 1.773e-02
## BOGS_000125502 -7.298e-02
## BOGS_000125634 -3.222e-01
## BOGS_000126337 1.076e-01
## BOGS_000128470 7.045e-03
## BOGS_000128660 2.899e-02
## BOGS_000129899 -1.609e-02
## BOGS_000133938 -4.209e-02
## BOGS_000133943 -9.179e-02
## BOGS_000133948 6.410e-02
## BOGS_000133951 -6.948e-02
## BOGS_000134033 1.183e-01
## BOGS_000134036 -3.877e-02
## BOGS_000134154 3.896e-02
## BOGS_000134551 5.666e-03
## BOGS_000134630 4.076e-02
## BOGS_000134681 1.606e-02
## BOGS_000137869 -9.157e-02
## BOGS_000138001 1.170e-02
## BOGS_000150046 -2.585e-02
## BOGS_000150326 1.855e-02
## BOGS_000158824 1.321e-01
## BOGS_000159330 -2.616e-03
## BOGS_000167353 1.763e-01
## BOGS_000172272 2.743e-02
## BOGS_000172334 -3.907e-02
## BOGS_000172384 -7.064e-02
## BOGS_000172410 -7.016e-02
## BOGS_000172574 5.717e-02
## BOGS_000172793 9.870e-03
## BOGS_000172990 3.584e-02
## BOGS_000173013 -1.019e-01
## BOGS_000174702 -1.502e-02
## BOGS_000174886 1.929e-02
## BOGS_000175854 3.917e-02
## BOGS_000176094 1.045e-01
## BOGS_000177666 7.472e-04
## BOGS_000178089 4.985e-02
## BOGS_000178967 1.056e-01
## BOGS_000179838 1.724e-02
## BOGS_000180056 -6.193e-02
## BOGS_000180074 -4.949e-02
## BOGS_000180301 -1.184e-01
## BOGS_000181801 7.787e-02
## BOGS_000185116 -2.570e-02
## BOGS_000186619 -2.793e-03

```

## BOGS\_000187217 -1.546e-01  
## BOGS\_000188077 -3.009e-02  
## BOGS\_000199997 9.972e-02  
## BOGS\_000200166 -3.534e-02  
## BOGS\_000201530 2.383e-02  
## BOGS\_000203482 2.983e-02  
## BOGS\_000206295 -1.579e-03  
## BOGS\_000208231 1.213e-02  
## BOGS\_000209030 1.213e-02  
## BOGS\_000211135 4.401e-02  
## BOGS\_000211172 1.870e-02  
## BOGS\_000211188 1.414e-02  
## BOGS\_000211221 -5.493e-02  
## BOGS\_000211531 3.474e-02  
## BOGS\_000211710 5.415e-03  
## BOGS\_000213701 -3.903e-02  
## BOGS\_000213947 1.360e-02  
## BOGS\_000214439 -1.877e-01  
## BOGS\_000214547 1.614e-02  
## BOGS\_000214733 2.040e-02  
## BOGS\_000215413 1.710e-02  
## BOGS\_000215414 2.318e-02  
## BOGS\_000215489 1.606e-02  
## BOGS\_000215518 -3.278e-02  
## BOGS\_000215595 9.602e-03  
## BOGS\_000216005 -2.769e-03  
## BOGS\_000216464 1.606e-02  
## BOGS\_000218184 1.606e-02  
## BOGS\_000218580 -4.454e-02  
## BOGS\_000219289 -6.417e-03  
## BOGS\_000220652 -3.023e-03  
## BOGS\_000221316 -8.066e-03  
## BOGS\_000223957 3.198e-02  
## BOGS\_000224298 -1.396e-02  
## BOGS\_000224378 3.771e-03  
## BOGS\_000225378 5.284e-02  
## BOGS\_000246158 1.232e-02  
## BOGS\_000246536 6.154e-02  
## BOGS\_000247056 3.085e-02  
## BOGS\_000247573 3.244e-03  
## BOGS\_000260111 -1.026e-03  
## BOGS\_000265594 8.475e-02  
## BOGS\_000265762 4.177e-02  
## BOGS\_000266220 5.905e-02  
## BOGS\_000267187 1.580e-02  
## BOGS\_000268867 -1.923e-02  
## BOGS\_000269084 1.043e-02  
## BOGS\_000269350 -2.470e-03  
## BOGS\_000269461 -5.294e-02  
## BOGS\_000287071 1.712e-02  
## BOGS\_000287072 1.092e-03  
## BOGS\_000287443 -3.634e-02  
## BOGS\_000290336 -1.271e-01  
## BOGS\_000290621 -1.017e-02

## BOGS\_000290638 8.038e-03  
## BOGS\_000290815 -1.109e-02  
## BOGS\_000291713 9.245e-04  
## BOGS\_000299781 -7.770e-02  
## BOGS\_000309758 -2.769e-03  
## BOGS\_000309837 -2.769e-03  
## BOGS\_000313064 -1.644e-02  
## BOGS\_000313142 -8.213e-02  
## BOGS\_000322074 7.555e-02  
## BOGS\_000322220 -3.229e-02  
## BOGS\_000322357 -1.210e-02  
## BOGS\_000323061 -1.578e-02  
## BOGS\_000326807 -1.109e-02  
## BOGS\_000328696 -1.281e-03  
## BOGS\_000334667 1.267e-02  
## BOGS\_000346557 1.057e-03  
## BOGS\_000366488 5.478e-03  
## BOGS\_000378733 4.227e-02  
## BOGS\_000392882 -2.810e-02  
## BOGS\_000393583 5.999e-03  
## BOGS\_000418510 2.215e-02  
## BOGS\_000419105 -4.162e-03  
## BOGS\_000421133 1.448e-02  
## BOGS\_000423554 -6.918e-02  
## BOGS\_000426320 -2.160e-02  
## BOGS\_000426631 -1.884e-02  
## BOGS\_000428278 -1.067e-01  
## BOGS\_000430467 9.611e-05  
## BOGS\_000430575 -2.137e-03  
## BOGS\_000432024 1.173e-02  
## BOGS\_000432338 -2.250e-03  
## BOGS\_000433089 -1.085e-02  
## BOGS\_000433683 -1.934e-02  
## BOGS\_000437337 2.231e-02  
## BOGS\_000443271 -5.517e-03  
## BOGS\_000448391 -4.110e-02  
## BOGS\_000448407 2.476e-02  
## BOGS\_000448810 7.724e-04  
## BOGS\_000455988 4.285e-02  
## BOGS\_000458031 1.512e-01  
## BOGS\_000458037 7.335e-02  
## BOGS\_000459244 1.659e-02  
## BOGS\_000460736 1.174e-02  
## BOGS\_000461069 2.872e-04  
## BOGS\_000461246 1.231e-02  
## BOGS\_000474205 -3.733e-02  
## BOGS\_000479425 -1.678e-03  
## BOGS\_000481274 -1.187e-03  
## BOGS\_000485412 -3.868e-02  
## BOGS\_000495343 8.493e-03  
## BOGS\_000495386 3.771e-03  
## BOGS\_000495417 1.412e-02  
## BOGS\_000499417 2.530e-02  
## BOGS\_000510152 -4.300e-02

```

## BOGS_000527403 2.339e-02
## BOGS_000527436 -2.160e-02
## BOGS_000533827 -2.218e-03
## BOGS_000564122 -2.160e-02
## BOGS_000591711 3.126e-03
## BOGS_000640427 -1.109e-02
##
##
## Site scores (weighted sums of species scores)
##
##          CAP1      CAP2      CAP3      CAP4      CAP5      CAP6
## BAI01      -0.92025  0.56100 -0.616868  0.54431 -0.39622  0.29568
## BAI04      -1.05760 -1.39732  0.831820 -1.11803  0.94806 -1.07492
## BAI06      -0.99777 -0.64651 -0.918815  1.87935  1.15869 -2.45426
## BAI07      -0.93808  0.58122 -2.823528 -1.84945 -2.08962  0.35431
## BPU04      -0.26945  2.45543  0.183040  0.16926  0.99883 -2.06653
## BSC03       0.76491  0.06550 -1.922362 -0.46094  3.52794  1.90387
## BSC04       0.86486 -0.63403 -1.427617  0.26605  1.30782  0.94207
## BSC05       0.52714  0.77166 -1.727635 -1.44677  4.31536  1.95126
## BV01       -1.10609 -1.46111  0.423890  0.68465  0.43108 -0.38625
## BV02       -0.62363 -1.52334  2.072417 -0.07802  1.38434 -0.29604
## BValp01    -0.83195 -0.98018  1.195059 -1.68807  0.91209 -0.76314
## BValp02    -0.55100 -0.74788  2.094357 -2.87386  1.02742  0.25297
## BValp03    -0.88837 -0.30806 -0.366212  1.99047  0.60328 -0.78927
## CBR02       0.21099  2.32487  1.042500  1.85391 -0.12440 -0.92354
## CBR03       0.03451  3.08572  0.435193  0.95586  0.97664 -0.80287
## CON02      -0.56034 -1.35546  1.394415 -0.93105  0.54648  1.27016
## CON03      -0.61451  0.31288  1.837358 -1.08814 -0.18834  1.16936
## CON06      -0.72951  0.14895  0.259379  2.16954  0.75364  0.01192
## CR01       -1.23719  0.06815 -2.252773 -1.19370 -1.03292 -0.84071
## CR06       -1.16680  0.43249 -2.616364 -0.13993 -1.88815  0.10567
## CR09       -0.90094  0.56572 -2.792046 -2.07261 -1.95554  0.36329
## ERT05       0.75904 -0.82324 -0.519305 -0.35898 -0.49191 -1.89336
## ES03        0.95238 -1.09606 -0.308546  0.75330 -1.38280 -0.43531
## ES10        0.99376 -0.22569  0.514519 -0.40924 -0.36564  0.15612
## FRE03       0.89134  0.44979 -0.542620 -0.11056 -1.16467 -0.75895
## GMG01       0.55310  1.77375  1.165176  0.04059 -0.31190 -0.92839
## GMG06       0.41895  1.18242  1.118270  0.86090 -1.21757  0.39425
## HOR01       0.89217 -0.11920 -0.176303  0.90787 -0.96938  0.08432
## HOR04      -0.97153 -0.62518  0.687905 -1.02314 -0.48631  0.81976
## HOR05      -0.82896  0.55542  0.141146  1.96150  0.62749  0.08320
## HOR06      -0.33226  0.63715  1.717109 -1.28542  1.58574  1.39894
## LLA07       0.96400 -1.21621 -0.103541 -0.88615 -0.72287 -0.56489
## MA_alp01   -0.38446 -0.40338 -0.232408  1.69545  0.91656  1.82747
## MA_alp02    0.14924 -0.47073  0.422283  0.72367 -1.17237  1.73669
## MAR01       0.71872 -0.26491  0.625333 -1.14003 -1.48098  0.07256
## MAR02       1.11875 -1.05130 -0.699726 -0.26683 -0.83808 -0.41239
## MAR03       0.86095  0.23044  0.121914 -0.22030 -0.49978  0.54713
## MAR04       0.94522 -1.31960 -0.452348  0.23067 -0.40216 -1.14306
## MAR05       1.15505 -0.90704 -1.336503  0.58531  0.35007  0.28612
## MAR06       0.97679 -0.03070  0.043493  0.56690 -1.09814  0.59844
## MAR09       1.02540  0.11731 -0.177000 -0.15680 -0.34383 -0.83923
## MAR10       1.07025 -1.03630 -0.611178  0.51399 -0.53694 -0.03943
## MT02        0.32994  2.42940 -0.001545 -1.40450 -0.12696 -0.83906

```

## MTL03	0.83113	-1.36661	-0.436519	-1.06612	-0.12929	-0.44790
## MUN02	-0.84762	-0.81928	2.091768	-0.29744	1.05015	0.83513
## MUN03	-0.81676	-0.04527	1.517657	1.44381	0.75902	1.75647
## PA01	0.86234	-0.13894	-0.199608	0.55483	-1.02555	-0.04525
## PA02	0.90264	-0.62590	-0.313245	0.88806	-0.89531	0.34043
## PA05	0.83804	-0.24395	-0.068003	0.94246	-1.73423	0.80151
## RAT07	-0.82627	0.18419	1.032988	-1.22268	0.30874	-0.87562
## S003	-1.27530	-0.20121	-2.445890	-0.27848	-1.89744	0.11141
## S004	-1.07607	-0.26619	-2.497319	0.59005	-1.56265	0.06166
## S005	-1.08197	-0.05270	-0.849496	-0.83318	-1.12979	1.00989
## S008	-0.93964	-0.29976	-0.061814	2.57149	2.12514	-3.31798
## S009	-1.27993	0.01436	-2.453453	-0.54713	-1.39829	-0.40762
## SON02	0.58385	1.57833	1.392418	-0.47459	-0.14641	0.39152
## SON03	0.74179	1.01365	0.938103	1.01226	-0.12275	0.54169
## T29	-0.49089	-0.09358	2.819168	-1.67551	1.68208	1.23940
## T46	-1.02346	-1.42751	1.894719	1.14168	1.14537	0.72842
## T52	-0.67269	-0.11990	1.947951	0.21401	0.74902	1.43040
## T54	-0.18948	0.36924	1.161985	0.87057	1.34164	1.31337
## TAU03	0.69833	0.88606	0.251146	0.66747	-0.36226	-0.16793
## TAU04	1.02797	0.21820	-0.450173	-0.77436	0.59733	0.28146
## TAU06	0.82924	0.68884	0.502790	-0.48180	0.48922	0.83223
## TOR01	1.03641	0.19591	-0.944706	0.35181	-0.04661	-1.65724
## TOR02	0.90156	0.44619	-0.531798	-0.74823	-0.88119	-3.12940
##						
##						
## Site constraints (linear combinations of constraining variables)						
##						
##	CAP1	CAP2	CAP3	CAP4	CAP5	CAP6
## BAI01	-0.56575	-0.207353	-0.68420	0.21501	-0.38757	-0.185340
## BAI04	-1.01578	-0.445764	1.06242	-0.92271	0.92682	-0.635753
## BAI06	-1.03072	-0.268218	-0.34535	0.62336	0.41876	-0.925172
## BAI07	-0.93997	0.304803	-0.94185	-0.55582	-0.30478	-0.582649
## BPU04	0.38501	2.261693	-0.69687	-0.07906	-0.68655	-0.817184
## BSC03	1.13439	0.591353	-1.44568	0.52879	1.24344	0.616203
## BSC04	1.23484	0.470531	-1.87151	0.67730	1.33415	0.581744
## BSC05	0.49433	0.441355	-1.74679	-1.89987	3.54512	0.984641
## BV01	-0.75391	-0.771235	-0.19925	1.07822	0.31030	-0.519553
## BV02	-0.61179	-0.605837	0.90594	-0.53945	0.73299	-0.828708
## BValp01	-1.00637	-0.339191	0.61284	-2.20655	0.02873	0.395963
## BValp02	-0.08089	-1.005991	0.83359	-0.86211	-0.01072	0.072416
## BValp03	-0.51741	-0.738267	-0.04230	0.24993	0.11891	0.040162
## CBR02	0.18626	1.463941	0.73930	0.53801	-0.25398	0.341254
## CBR03	-0.61022	1.688013	0.98785	0.94944	0.56226	-0.431848
## CON02	-0.18117	0.599455	0.27035	0.15588	-0.38699	0.173920
## CON03	-0.20175	0.598500	1.01571	-0.27340	-0.20087	0.315089
## CON06	-0.48060	0.259943	0.68975	0.95150	0.26122	0.039242
## CR01	-1.38851	0.472944	-1.54567	-0.93335	-0.43478	0.048554
## CR06	-1.21325	0.349222	-0.17319	-0.68991	-0.59819	1.251584
## CR09	-1.06338	0.003778	0.10234	0.10145	-0.08723	0.637291
## ERT05	0.37811	-1.159474	-0.54318	-1.30772	0.83569	-0.044068
## ES03	0.49287	-0.060840	0.84014	-0.54927	-0.05838	0.102556
## ES10	1.04247	-0.203432	1.40062	-0.28810	-0.09159	-0.145421
## FRE03	0.65875	0.720994	-0.54271	0.54687	0.20540	0.616539
## GMG01	0.27845	1.232424	0.13500	0.23940	0.78133	-1.010075

```

## GMG06      0.05661  1.229262  0.94925  0.27714  0.03632  0.040566
## HOR01      1.11378 -0.061979  0.23917  0.79049 -0.66400 -0.112251
## HOR04     -0.52391 -0.309786  0.04274  0.72881  0.02678  0.217506
## HOR05     -0.66564 -0.064016 -0.32782  0.08490 -0.06801 -0.047561
## HOR06     -0.64150 -0.060167  0.33780 -0.70724 -0.19644  0.170354
## LLA07      0.93493 -0.362430 -0.50609 -0.16269 -0.70158 -1.134504
## MA_alp01  -0.29921 -0.530826 -0.31411  0.12002 -0.76238  0.829818
## MA_alp02  -0.40166 -0.562926 -0.21695  0.31981 -0.54073  0.547869
## MAR01      0.04532 -0.472057  0.31690 -1.20771 -0.23902 -0.289772
## MAR02      0.51460 -0.975563 -0.69836  0.60431 -0.85291  0.105611
## MAR03      0.16786 -0.671904 -0.27839  0.02248 -0.74696  0.164184
## MAR04      1.18623 -1.179690  0.33774 -0.43617 -0.10022 -0.736300
## MAR05      0.88798 -1.304814 -0.56413  0.87424 -0.59266  0.241695
## MAR06      0.73293 -1.371485 -0.87567  0.45341 -0.12128  0.339874
## MAR09      0.71480 -0.177368  0.65053 -0.13161 -0.31555 -0.046885
## MAR10      1.11716 -1.115549 -0.22011  0.53242 -0.03831  0.198075
## MT02      -0.11193  1.811592 -0.92727 -0.82970 -1.11114 -1.349931
## MTL03      0.86885 -1.152154  0.38303 -0.45539 -0.11258 -0.337012
## MUN02     -0.81511 -0.697950 -0.38639 -0.08552 -0.36442  0.534699
## MUN03     -1.09749 -0.442275 -0.53120  0.16899  0.01887  0.287708
## PA01       0.49012  0.213267  0.40981  0.67810 -0.52422  0.741366
## PA02       0.48379  0.072605  0.02099  0.63087 -0.23762  0.843440
## PA05       0.63283  0.503728 -0.13444  0.63091 -0.92514 -0.122415
## RAT07     -0.26239 -0.870095  0.33496 -1.00314 -0.21674  0.422149
## S003      -0.74617  0.108429 -0.27468 -0.09501 -0.57893  0.614577
## S004      -0.84659 -0.047734 -0.31350  0.40916 -0.02254 -0.125495
## S005      -0.70522  0.388406 -0.74540 -0.43276 -0.95236  0.297574
## S008      -1.42555 -0.785937  0.25837  2.52032  2.26837 -2.305607
## S009      -1.10190  0.263551 -0.52813  0.13023 -0.07256 -0.626381
## SON02      0.35514  1.011512  0.94412 -0.13798  0.35779  0.959695
## SON03      0.34413  1.180835  0.45165  0.40204 -0.28930  0.720018
## T29       -0.36203 -0.092037  1.56193 -0.13102  0.95339 -0.557892
## T46       -0.10530 -0.247489  0.35096  1.20967  0.12469 -0.097167
## T52       -0.38197  0.200537  0.34659  0.07970 -0.60439  1.171946
## T54       -0.07315  0.005534  0.24595  0.33986 -0.58508  1.047881
## TAU03      0.69366  0.226722 -0.56562  0.61672  0.71209  1.187685
## TAU04      0.91956  0.334852  0.31430 -0.70719  1.26739  0.343956
## TAU06      1.35427  0.356626  2.17434 -0.58698  0.62336 -0.003186
## TOR01      1.22271 -0.518539 -0.04905  0.16721 -0.74878 -1.015826
## TOR02      1.10546  0.513966 -1.03117 -1.42951 -0.90667 -3.211444
##
##
## Biplot scores for constraining variables
##
##          CAP1      CAP2      CAP3      CAP4      CAP5      CAP6
## GDD          0.4031  0.80559 -0.009221  0.17415  0.3613  0.16608
## Brown_mosses 0.3442 -0.24890  0.352859 -0.71939  0.3044 -0.29139
## pH           0.9698 -0.09510 -0.161393 -0.03139  0.1438 -0.05257
## Mg           0.4668  0.19531 -0.479487 -0.11380  0.6738  0.21701
## S            0.3143  0.40165 -0.540436 -0.15873 -0.1490 -0.63279
## Al          -0.5982 -0.07545  0.171565  0.47210  0.4220 -0.45396

```