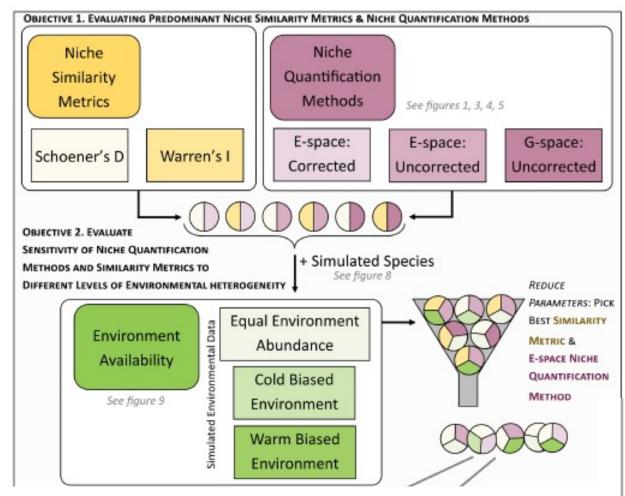


05 - Point Based

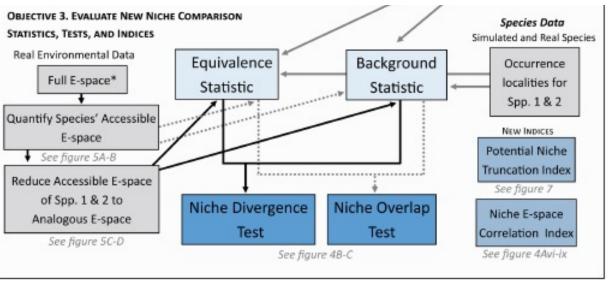
Shelly Gaynor University of Florida





Brown and Carnaval. 2019. A tale of two niche: methods, concepts, and evolution. Frontiers of Biogeography.

Climatic Niche



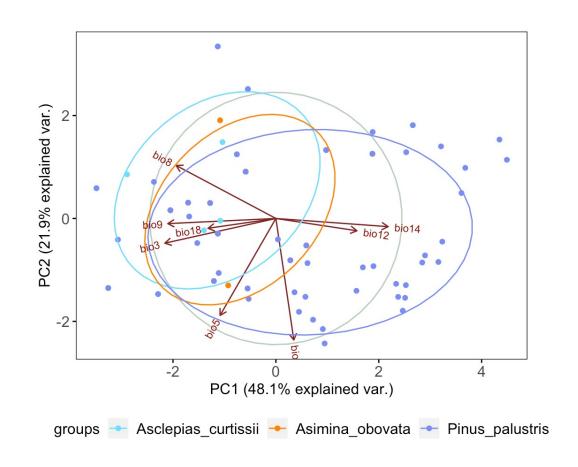
Ecological Niche

Realized Niche

 abiotic conditions that a species can occupy with the presence of biotic interactions

Fundamental Niche

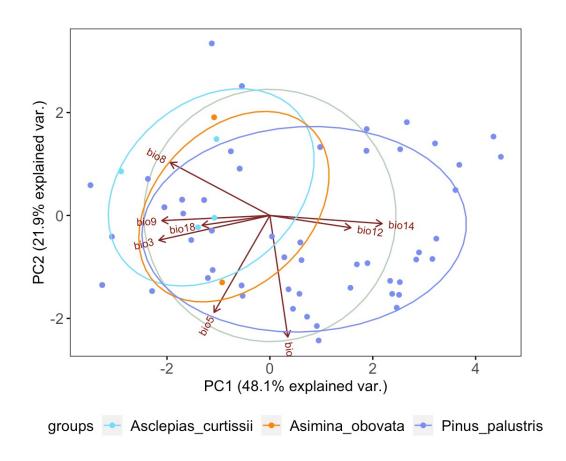
 abiotic conditions a species could potentially occupy in the absence of biotic interactions



Ecological Niche

Realized Niche

 abiotic conditions that a species can occupy with the presence of biotic interactions



Load Packages

```
library(gtools)
library(raster)
library(plyr)
library(dplyr)
library(tidyr)
library(ggplot2)
library(multcompView)
library(gridExtra)
library(ecospat)
library(dismo)
library(ade4)
```

Load Datafiles

Load functions

This function is from vqv/ggbiplot.

```
source("functions/ggbiplot_copy.R")
```

Load data file

```
alldf <- read.csv("data/cleaning_demo/maxent_ready/diapensiac
eae_maxentready_20230605.csv")</pre>
```

Load Raster Layers

_

```
list <- list.files("data/climate_processing/PresentLayers/al
l", full.names = TRUE, recursive = FALSE)
list <- mixedsort(sort(list))
envtStack <- stack(list)</pre>
```

Preparing Data

Extract value for each point

For each occurence record, extract the value for each bioclim variables using the package raster.

```
ptExtracted <- raster::extract(envtStack, alldf[2:3])</pre>
```

Convert to data frame

```
ptExtracteddf <- as.data.frame(ptExtracted)</pre>
```

Add species name

Drop any NA

Create two dataframes.

```
data.bioclim <- ptExtracteddf[, 1:8]
data.species <- ptExtracteddf[, 9]</pre>
```

Using only the bioclim columns to run the principal components analysis.

```
data.pca <- prcomp(data.bioclim, scale. = TRUE)
```

```
loadings <- data.pca$rotation
summary(loadings)</pre>
```

```
loadings_relative_A <- t(t(abs(loadings))/rowSums(t(abs(loadings))))*100
loadings_relative_A</pre>
```

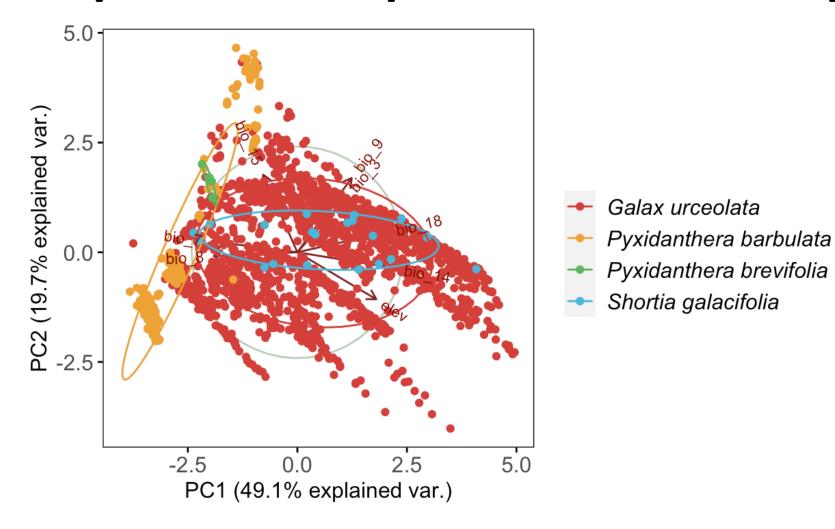
```
PC1
                        PC2 PC3
                                            PC4
                                                      PC5
                                                               PC6
                                                                         PC7
## bio 3 8.591746 18.290740 15.598618 26.50100048 15.693465 2.892773 10.826188
## bio 7 15.106599 3.605906 20.544960 4.56019510 8.196901 6.586901 27.743419
## bio 8 14.713687 1.268164 13.920838 0.01157387 24.676885 22.390720 4.151105
## bio 9 9.541759 24.350510 8.759713 16.03534080 7.639386 25.535124 3.884780
## bio 14 16.784509 5.136389 1.759000 16.24120193 10.151969 6.787089 7.464726
## bio 15 6.003487 25.442930 21.469601 8.28251442 9.927169 9.084507 6.348032
## bio 18 15.380721 6.468852 14.248091 7.35172686 17.912054 10.264436 18.683458
## elev 13.877493 15.436508 3.699179 21.01644654 5.802170 16.458450 20.898292
##
               PC8
## bio 3 0.2683966
## bio 7 10.5755001
## bio 8 6.4386942
## bio 9 2.3740267
## bio 14 31.8174754
## bio 15 18.6736844
## bio 18 19.1865553
## elev 10.6656673
```

Set theme

First, I made a theme to change the background of the plot. Next, I changed the plot margins and the text size.

Set color palette

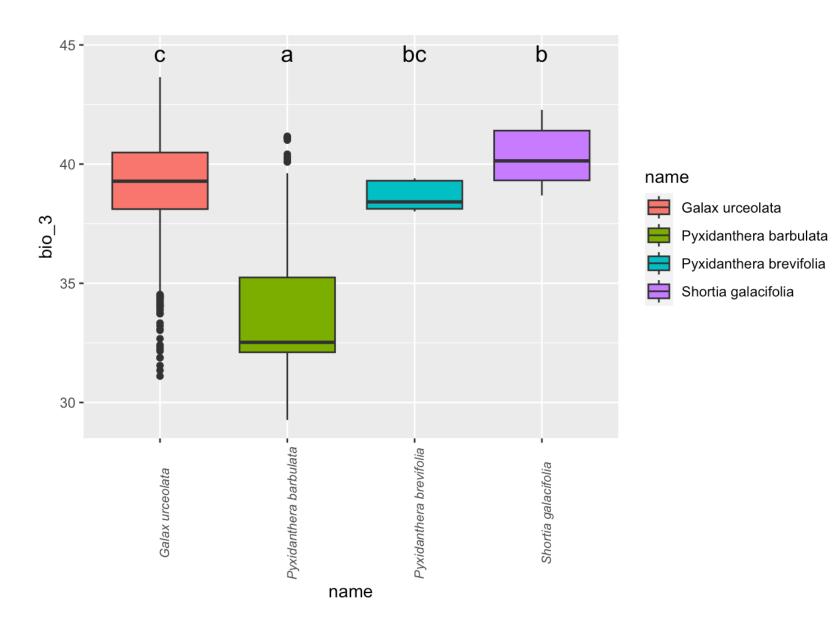
```
pal <- c("#D43F3AFF", "#EEA236FF", "#5CB85CFF", "#46B8DAFF")
```



Simple function to run an ANOVA and a post-hoc Tukey-HSD test

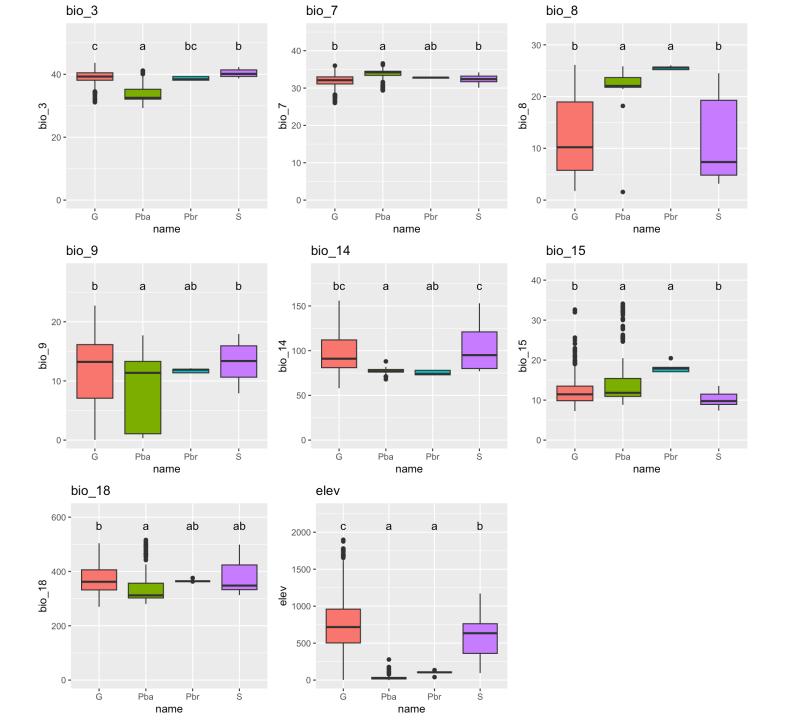
```
stat.test <- function(dataframe, x = "name", y) {
   bioaov <- aov(as.formula(paste0(y,"~",x)), data = dataframe)
   TH <- TukeyHSD(bioaov, "name")
   m <- multcompLetters(TH$name[,4])
   y.val <- as.numeric(max(dataframe[[y]]) + 1)
   groups <- data.frame(groups = m$Letters, name = names(m$Letters), y.val = rep(y.val, 4))
   return(groups)
}</pre>
```

Run for bio_3 only



Loop through all variables

```
variablelist <- colnames(ptExtracteddf)[1:8]</pre>
plotlist <- list()</pre>
maxstats <- c()</pre>
statsout <- c()
for(i in 1:8){
  vname <- variablelist[i]</pre>
  maxstats[i] <- as.numeric(max(ptExtracteddf[[vname]]) + 1)</pre>
  statsout[[i]] <- stat.test(dataframe = ptExtracteddf, y = vname)</pre>
  plotlist[[i]] <- ggplot(ptExtracteddf, aes(x = name, y = .data[[vname]])) +</pre>
                     geom_boxplot(aes(fill = name)) +
                     scale_colour_manual(name = 'Species', values = pal) +
                     geom text(data = statsout[[i]],
                                mapping = aes(x = name,
                                              y = (y.val*1.1),
                                              label = groups),
                                size = 4, inherit.aes = FALSE) +
                     scale x discrete(labels = c('G', 'Pba', 'Pbr', 'S')) +
                     ggtitle(label = vname) +
                     ylab(vname) +
                     theme(legend.position = "none") +
                    ylim(0, (maxstats[i]*1.2))
gridExtra::grid.arrange(grobs = plotlist)
```



Set up background points

```
bgl <- randomPoints(mask = envtStack, n = 1000, p = alldf[,3:2])

## Warning in .couldBeLonLat(x, warnings = warnings): CRS is NA. Assuming it is
## longitude/latitude

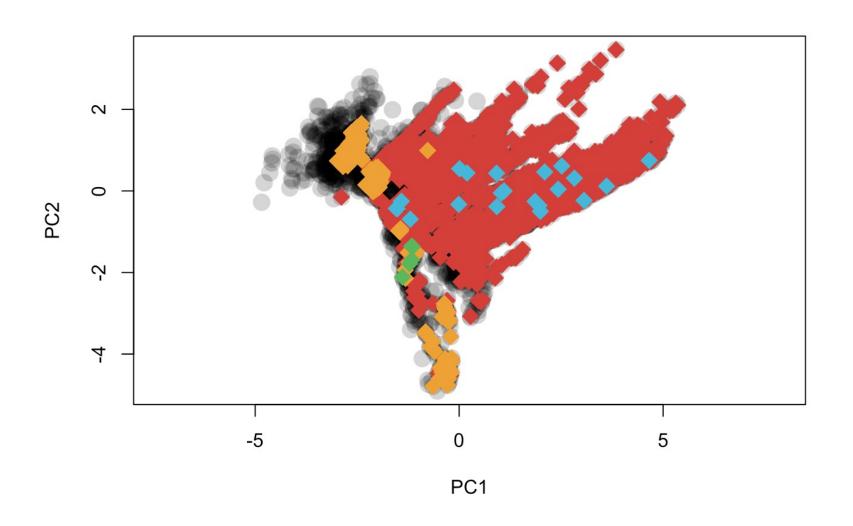
bgl.env <- raster::extract(envtStack, bgl)
bgl.env <- data.frame(bgl.env)
allpt.bioclim <- rbind(bgl.env, data.bioclim)</pre>
```

dudi.PCA to reduce variables

Pull out scores for each species

```
pl.score <- suprow(pca.env, dplyr::filter(ptExtracteddf, name == "Galax urceolat
a")[, 1:8])$li
p2.score <- suprow(pca.env, dplyr::filter(ptExtracteddf, name == "Pyxidanthera ba
rbulata")[, 1:8])$li
p3.score <- suprow(pca.env, dplyr::filter(ptExtracteddf, name == "Pyxidanthera br
evifolia")[, 1:8])$li
p4.score <- suprow(pca.env, dplyr::filter(ptExtracteddf, name == "Shortia galacif
olia")[, 1:8])$li
scores.clim <- pca.env$li</pre>
```

Visualize



Kernel density estimates

Create occurrence density grids based on the ordination data.

```
z1 <- ecospat.grid.clim.dyn(scores.clim, scores.clim, p1.score, R = 100)
z2 <- ecospat.grid.clim.dyn(scores.clim, scores.clim, p2.score, R = 100)
z3 <- ecospat.grid.clim.dyn(scores.clim, scores.clim, p3.score, R = 100)
z4 <- ecospat.grid.clim.dyn(scores.clim, scores.clim, p4.score, R = 100)
z1ist <- list(z1, z2, z3, z4)</pre>
```

Niche Overlap

Schoener's D ranges from 0 to 1. 0 represents no similarity between niche space. 1 represents completely identical niche space.

```
overlapD <- matrix(ncol = 2, nrow = 7)
n <- 1
for(i in 1:3){
   for(j in 2:4){
      if(i != j){
         overlapD[n, 1]<- paste0("z", i, "-", "z", j)
         overlapD[n, 2]<- ecospat.niche.overlap(zlist[[i]], zlist[[j]], cor = TRUE)$D
      n <- n + 1
      }
}

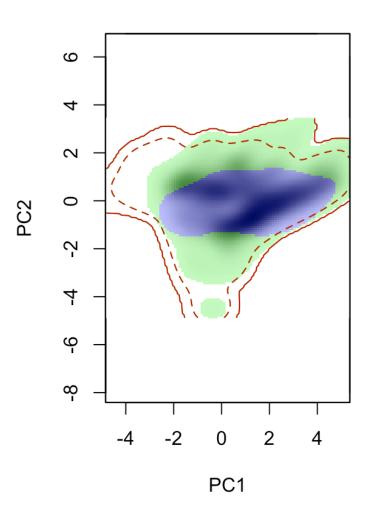
overlapDdf <- data.frame(overlapD)
overlapDdf</pre>
```

Niche Overlap

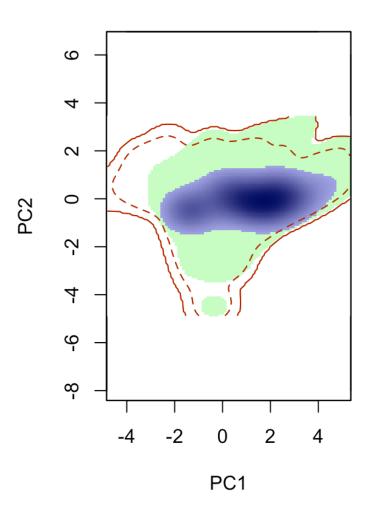
Niche Overlap Visualization

Niche Overlap

Niche Overlap - Z1 top



Niche Overlap - Z4 top



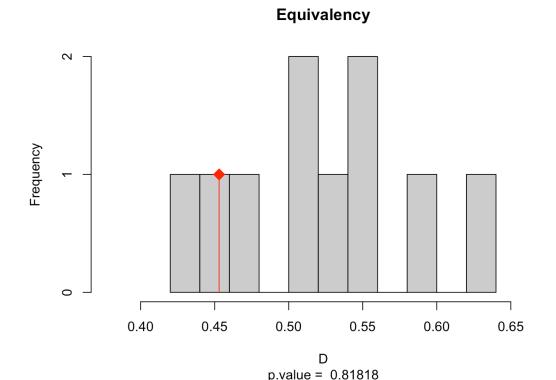
Niche Equivalency Test

Niche Equivalency Test

Based on Warren et al. 2008 - Are the two niche identical?

Hypothesis test for D, null based on randomization. H1: the niche overlap is higher than expected by chance (or when randomized).

```
eq.test <- ecospat.niche.equivalency.test(z1, z4, rep = 10)
ecospat.plot.overlap.test(eq.test, "D", "Equivalency")</pre>
```



Niche Similarity Test

Based on Warren et al. 2008 - Are the two niche similar?

Can one species' niche predict the occurrences of a second species better than expected by chance?

```
sim.test <- ecospat.niche.similarity.test(z1, z4, rep = 10, rand.type=2)
ecospat.plot.overlap.test(sim.test, "D", "Similarity")</pre>
```

