

Phasmid_morphospace - Habitat

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2024-07-24

Load Packages and data

Load necessary packages to run script

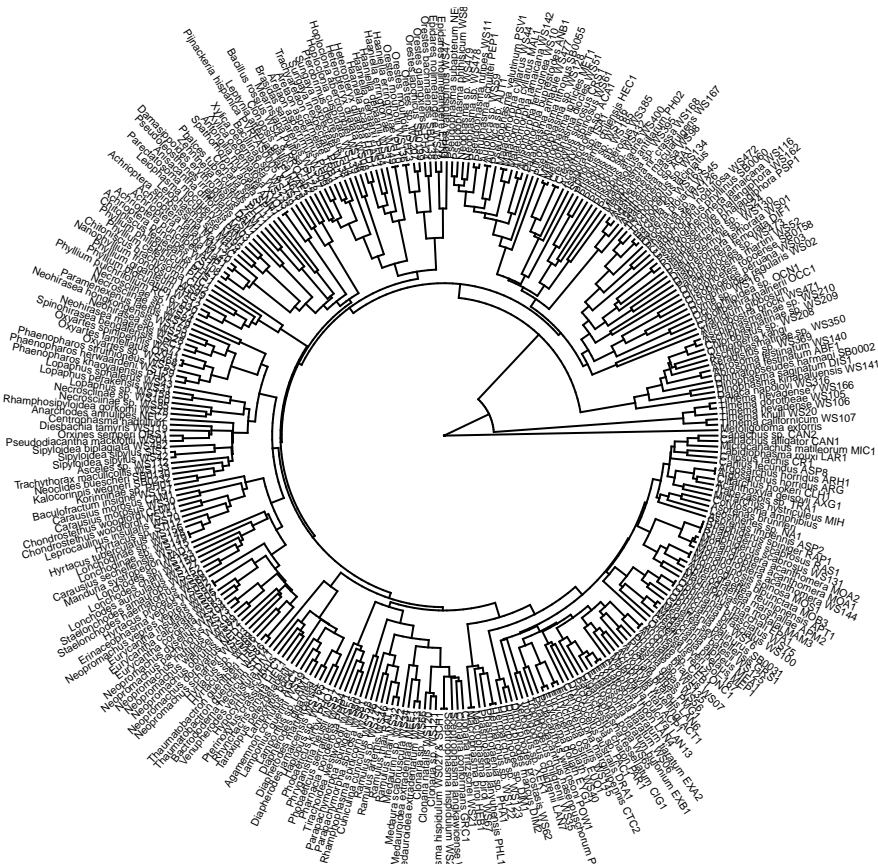
```
library(readxl)
library(ape)
library(phytools)
library(plyr)
library(hypervolume)
library(purrr)
library(geiger)
library(dispRity)
library(evobiR)
library(dynRB)
library(ggplot2)
library(cowplot)
library(randomForest)
library(caret)
library(reshape2)
library(dplyr)
library(ggpubr)
library(corrplot)
library(factoextra)

# Display versions of loaded packages
installed.packages()[names(sessionInfo())$otherPkgs], "Version"]
```

```
##      factoextra      corrplot      ggpubr      dplyr      reshape2      caret
##      "1.0.7"        "0.92"        "0.6.0"      "1.1.2"      "1.4.4"        "6.0-94"
##      lattice randomForest      cowplot      ggplot2      dynRB      evobiR
##      "0.21-8"      "4.7-1.1"      "1.1.1"      "3.4.2"      "0.18"        "1.1"
##      dispRity      geiger      purrr      hypervolume      Rcpp      plyr
##      "1.7.0"        "2.0.11"      "1.0.1"      "3.1.1"      "1.0.10"      "1.8.8"
##      phytools      maps      ape      readxl      formatR      tinytex
##      "1.5-1"        "3.4.1"        "5.7-1"      "1.4.2"      "1.14"        "0.49"
##      knitr
##      "1.43"
```

Load phylogenetic tree

```
tree <- read.nexus("Phasmid_convergence_MCCphylogeny_nexus") #load tree (NEXUS)
tree <- ladderize(tree, right = FALSE)
tree$tip.label <- gsub("'", "", tree$tip.label) #delete unwanted ' ' in tip labels
plotTree(tree, type = "fan", fsize = 0.3, lwd = 1)
```



Load morphological data

```
X <- read_excel("DatasetS1.xlsx", sheet = "pca_coordinates")
X <- subset(X, is.na(X$habitat_use) == F & is.na(X$pca_mix_dim1) ==
F)
```

Make dataset and phylogeny match

```
X <- as.data.frame(X)
# Attributing rownames to the dataset
rownames(X) <- X$taxon
## Cleaning both the data and the tree
cleaned_data <- clean.data(X, tree)
```

```
## Extracting the cleaned dataset and the cleaned tree
df <- cleaned_data$data
ttree <- cleaned_data$tree
name.check(ttree, df)
```

```
## [1] "OK"
```

```
# reorder data to match the order of tips in tree
X <- ReorderData(ttree, df, taxa.names = "row names")
```

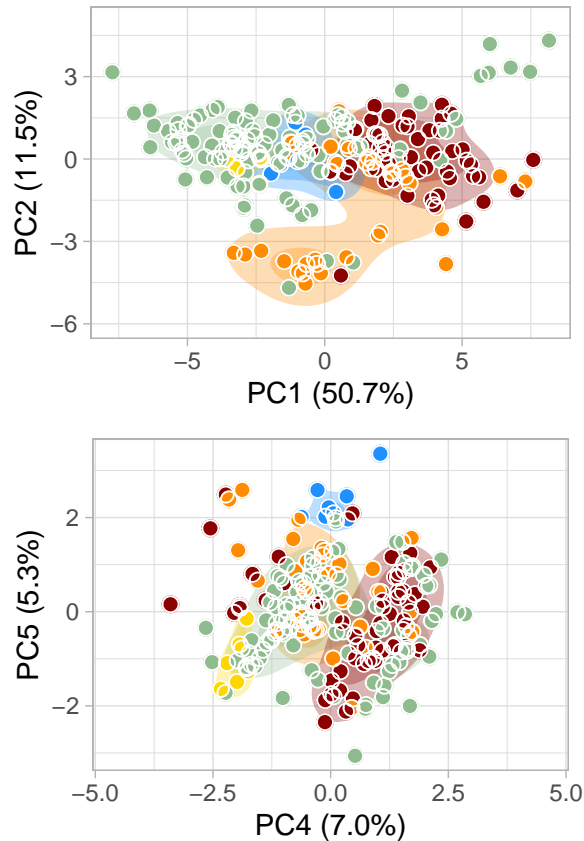
Hypervolumes in morphospace - Habitat

```
# Density
X$habitat2 <- factor(X$habitat_use, levels = c("laying on branches and leaves",
      "hanging from branches and leaves", "laying on leaf litter, logs, bark, trunk",
      "laying on palm leaves", "hanging from grass"))

p1 <- ggplot(data = X, aes(x = pca_mix_dim1, y = pca_mix_dim2,
      fill = habitat2)) + stat_density_2d(contour_var = "ndensity",
      geom = "polygon", breaks = c(1/3, 2/3), alpha = 0.3) + coord_fixed() +
      labs(x = "PC1 (50.7%)", y = "PC2 (11.5%)") + scale_fill_manual(values = c("darkorange",
      "darkseagreen", "darkred", "dodgerblue", "gold")) + scale_colour_manual(values = c("darkorange",
      "darkseagreen", "darkred", "dodgerblue", "gold")) + geom_point(aes(color = habitat2),
      size = 2.5, alpha = 1) + geom_point(colour = "white", shape = 1,
      size = 2.5, alpha = 1) + ylim(-6, 5) + theme_light() + theme(legend.position = "none")

p2 <- ggplot(data = X, aes(x = pca_mix_dim4, y = pca_mix_dim5,
      fill = habitat2)) + stat_density_2d(contour_var = "ndensity",
      geom = "polygon", breaks = c(1/3, 2/3), alpha = 0.3) + coord_fixed() +
      labs(x = "PC4 (7.0%)", y = "PC5 (5.3%)") + scale_fill_manual(values = c("darkorange",
      "darkseagreen", "darkred", "dodgerblue", "gold")) + scale_colour_manual(values = c("darkorange",
      "darkseagreen", "darkred", "dodgerblue", "gold")) + geom_point(aes(color = habitat2),
      size = 2.5, alpha = 1) + geom_point(colour = "white", shape = 1,
      size = 2.5, alpha = 1) + theme_light() + xlim(-4.6, 4.6) +
      theme(legend.position = "none")

plot_grid(p1, p2, ncol = 1)
```



Hypervolumes and overlaps - Habitat

Create appropriate dataset with 7 first dimensions of morphospace

```
Xrb <- matrix(nrow = length(X$habitat2), ncol = 8)
colnames(Xrb) <- c("habitat", "PC1", "PC2", "PC3", "PC4", "PC5",
  "PC6", "PC7")
Xrb <- as.data.frame(Xrb)
Xrb$habitat <- as.character(X$habitat2)
Xrb$PC1 <- as.numeric(X$pca_mix_dim1)
Xrb$PC2 <- as.numeric(X$pca_mix_dim2)
Xrb$PC3 <- as.numeric(X$pca_mix_dim3)
Xrb$PC4 <- as.numeric(X$pca_mix_dim4)
Xrb$PC5 <- as.numeric(X$pca_mix_dim5)
Xrb$PC6 <- as.numeric(X$pca_mix_dim6)
Xrb$PC7 <- as.numeric(X$pca_mix_dim7)
```

Using hypervolume package

```
Xrb_litter = Xrb[Xrb$habitat == "laying on leaf litter, logs, bark, trunk",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]
```

```

Xrb_hang = Xrb[Xrb$habitat == "hanging from branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]
Xrb_branch = Xrb[Xrb$habitat == "laying on branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]
Xrb_grass = Xrb[Xrb$habitat == "hanging from grass", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]
Xrb_palm = Xrb[Xrb$habitat == "laying on palm leaves", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]

hv1 = hypervolume_box(Xrb_litter, name = "litter")

```

```

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

```

```

hv2 = hypervolume_box(Xrb_hang, name = "hang")

```

```

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

```

```

hv3 = hypervolume_box(Xrb_branch, name = "branch")

```

```

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

```

```

hv4 = hypervolume_box(Xrb_grass, name = "grass")

```

```

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

```

```
hv5 = hypervolume_box(Xrb_palm, name = "palm")
```

```
##  
## Building tree...  
## done.  
## Ball query...  
##  
## done.  
## Binding random points... done.  
## Beginning volume calculation... done.
```

Calculate overlap/distance between pairs of habitats

```
hv <- list(hv1, hv2, hv3, hv4, hv5)  
jacc <- matrix(nrow = length(hv), ncol = length(hv))  
sorensen <- matrix(nrow = length(hv), ncol = length(hv))  
frac_unique_1 <- matrix(nrow = length(hv), ncol = length(hv))  
frac_unique_2 <- matrix(nrow = length(hv), ncol = length(hv))  
dist_min <- matrix(nrow = length(hv), ncol = length(hv))  
dist_centr <- matrix(nrow = length(hv), ncol = length(hv))  
  
for (i in c(1:length(hv))) {  
  for (j in c(1:length(hv))) {  
    if (j <= i) {  
    } else {  
      hv_set <- hypervolume_set(hv[[i]], hv[[j]], check.memory = FALSE)  
      stats <- hypervolume_overlap_statistics(hv_set)  
      centr <- hypervolume_distance(hv[[i]], hv[[j]], type = "centroid")  
      min <- hypervolume_distance(hv[[i]], hv[[j]], type = "minimum",  
        num.points.max = 500, check.memory = FALSE)  
  
      jacc[i, j] <- stats[1]  
      sorensen[i, j] <- stats[2]  
      frac_unique_1[i, j] <- stats[3]  
      frac_unique_2[i, j] <- stats[4]  
      dist_min[i, j] <- min  
      dist_centr[i, j] <- centr  
    }  
  }  
}
```

```
## Choosing num.points.max=442336 (use a larger value for more accuracy.)  
## Using minimum density of 1.429278  
## Retaining 35667 points in hv1 and 150805 points in hv2.  
## Beginning ball queries...  
##  
## Building tree...  
## done.  
## Ball query...  
##  
## done.
```

```
##
## Building tree...
## done.
## Ball query...
##
## done.
## Finished ball queries.
## Choosing num.points.max=442336 (use a larger value for more accuracy.)
## Using minimum density of 2.075205
## Retaining 51786 points in hv1 and 193381 points in hv2.
## Beginning ball queries...
##
## Building tree...
## done.
## Ball query...
##
## done.
##
## Building tree...
## done.
## Ball query...
##
## done.
## Finished ball queries.
## Choosing num.points.max=442336 (use a larger value for more accuracy.)
## Using minimum density of 6.992550
## Retaining 174499 points in hv1 and 951 points in hv2.
## Beginning ball queries...
##
## Building tree...
## done.
## Ball query...
##
## done.
##
## Building tree...
## done.
## Ball query...
##
## done.
## Finished ball queries.
## Choosing num.points.max=442336 (use a larger value for more accuracy.)
## Using minimum density of 6.992550
## Retaining 174499 points in hv1 and 6654 points in hv2.
## Beginning ball queries...
##
## Building tree...
## done.
## Ball query...
##
## done.
##
## Building tree...
## done.
```

```
## Ball query...
##
## done.
## Finished ball queries.
## Choosing num.points.max=442336 (use a larger value for more accuracy.)
## Using minimum density of 1.429278
## Retaining 150805 points in hv1 and 133189 points in hv2.
## Beginning ball queries...
##
## Building tree...
## done.
## Ball query...
##
## done.
##
## Building tree...
## done.
## Ball query...
##
## done.
## Finished ball queries.
## Choosing num.points.max=442336 (use a larger value for more accuracy.)
## Using minimum density of 1.429278
## Retaining 150805 points in hv1 and 194 points in hv2.
## Beginning ball queries...
##
## Building tree...
## done.
## Ball query...
##
## done.
##
## Building tree...
## done.
## Ball query...
##
## done.
## Finished ball queries.
## Choosing num.points.max=442336 (use a larger value for more accuracy.)
## Using minimum density of 1.429278
## Retaining 150805 points in hv1 and 1360 points in hv2.
## Beginning ball queries...
##
## Building tree...
## done.
## Ball query...
##
## done.
##
## Building tree...
## done.
## Ball query...
##
## done.
```



```
## Finished ball queries.
## Choosing num.points.max=442336 (use a larger value for more accuracy.)
## Using minimum density of 2.075205
## Retaining 193381 points in hv1 and 282 points in hv2.
## Beginning ball queries...
##
## Building tree...
## done.
## Ball query...
##
## done.
##
## Building tree...
## done.
## Ball query...
##
## done.
## Finished ball queries.
## Choosing num.points.max=442336 (use a larger value for more accuracy.)
## Using minimum density of 2.075205
## Retaining 193381 points in hv1 and 1974 points in hv2.
## Beginning ball queries...
##
## Building tree...
## done.
## Ball query...
##
## done.
##
## Building tree...
## done.
## Ball query...
##
## done.
## Finished ball queries.
## Choosing num.points.max=442336 (use a larger value for more accuracy.)
## Using minimum density of 257.356596
## Retaining 35015 points in hv1 and 244925 points in hv2.
## Beginning ball queries...
##
## Building tree...
## done.
## Ball query...
##
## done.
##
## Building tree...
## done.
## Ball query...
##
## done.
## Finished ball queries.
```

```

jacc <- t(jacc)
sorensen <- t(sorensen)
frac_unique_1 <- t(frac_unique_1)
frac_unique_2 <- t(frac_unique_2)
dist_centra <- t(dist_centra)
dist_min <- t(dist_min)

```

Plot heatmaps

```

theme_change <- theme(plot.title = element_text(family = "Helvetica",
  face = "bold", size = (15), hjust = 0.5), plot.subtitle = element_text(family = "Helvetica",
  size = (10), hjust = 0.5), plot.background = element_blank(),
  panel.grid.minor = element_blank(), panel.grid.major = element_blank(),
  panel.background = element_blank(), panel.border = element_blank(),
  axis.line = element_blank(), axis.ticks = element_blank(),
  axis.text.x = element_text(colour = "black", size = rel(1.5),
    angle = 35, hjust = 1), axis.text.y = element_text(colour = "black",
    size = rel(1.5)), axis.title.x = element_blank(), axis.title.y = element_blank())
col1 <- colorRampPalette(c("navyblue", "dodgerblue", "white",
  "red", "darkred")) #define color gradient

matrix <- jacc
row.names(matrix) <- c("litter", "hang", "branch", "grass", "palm")
colnames(matrix) <- c("litter", "hang", "branch", "grass", "palm")

lev <- ordered(c("palm", "grass", "branch", "hang", "litter"))
melt_mat <- melt(matrix)
p1 <- ggplot(melt_mat, aes(x = ordered(Var1, levels = lev), y = Var2,
  fill = value)) + geom_tile() + scale_fill_gradientn(colours = col1(8),
  breaks = c(0, 0.25, 0.5), guide = "colorbar", limits = c(0,
  0.5)) + labs(title = "Jaccard similarity") + theme_change

matrix <- sorensen
row.names(matrix) <- c("litter", "hang", "branch", "grass", "palm")
colnames(matrix) <- c("litter", "hang", "branch", "grass", "palm")
melt_mat <- melt(matrix)
p2 <- ggplot(melt_mat, aes(x = ordered(Var1, levels = lev), y = Var2,
  fill = value)) + geom_tile() + scale_fill_gradientn(colours = col1(8),
  breaks = c(0, 0.25, 0.5), guide = "colorbar", limits = c(0,
  0.5)) + labs(title = "Sorensen similarity") + theme_change

matrix1 <- frac_unique_1
matrix2 <- t(frac_unique_2)
row.names(matrix1) <- c("litter", "hang", "branch", "grass",
  "palm")
colnames(matrix1) <- c("litter", "hang", "branch", "grass", "palm")
row.names(matrix2) <- c("litter", "hang", "branch", "grass",
  "palm")
colnames(matrix2) <- c("litter", "hang", "branch", "grass", "palm")
melt_mat1 <- na.omit(melt(matrix1))

```

```

melt_mat2 <- na.omit(melt(matrix2))
melt_mat <- rbind(melt_mat1, melt_mat2)
p3 <- ggplot(melt_mat, aes(x = ordered(Var1, levels = lev), y = Var2,
  fill = 1 - value)) + geom_tile() + scale_fill_gradientn(colours = col1(8),
  breaks = c(0, 0.5, 1), guide = "colorbar", limits = c(0,
  1)) + labs(title = "Volume of pairwise shared components divided by volume of y") +
  theme_change

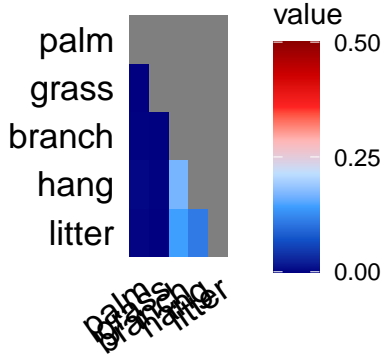
matrix <- dist_cent
row.names(matrix) <- c("litter", "hang", "branch", "grass", "palm")
colnames(matrix) <- c("litter", "hang", "branch", "grass", "palm")
melt_mat <- melt(matrix)
p4 <- ggplot(melt_mat, aes(x = ordered(Var1, levels = lev), y = Var2,
  fill = value)) + geom_tile() + scale_fill_gradientn(colours = col1(8),
  breaks = c(4, 5, 6), guide = "colorbar") + labs(title = "Centroid distance") +
  theme_change

matrix <- dist_min
row.names(matrix) <- c("litter", "hang", "branch", "grass", "palm")
colnames(matrix) <- c("litter", "hang", "branch", "grass", "palm")
melt_mat <- melt(matrix)
p5 <- ggplot(melt_mat, aes(x = ordered(Var1, levels = lev), y = Var2,
  fill = value)) + geom_tile() + scale_fill_gradientn(colours = col1(8),
  breaks = c(0.4, 0.8, 1.2), guide = "colorbar") + labs(title = "Minimum distance") +
  theme_change

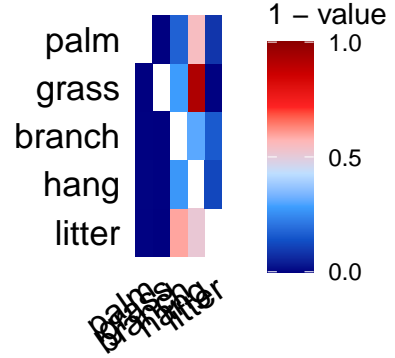
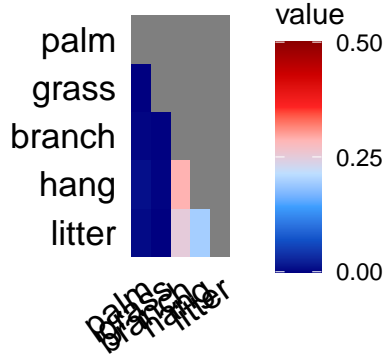
plot_grid(p1, p2, p3, p4, p5, label = "AUTO", nrow = 2)

```

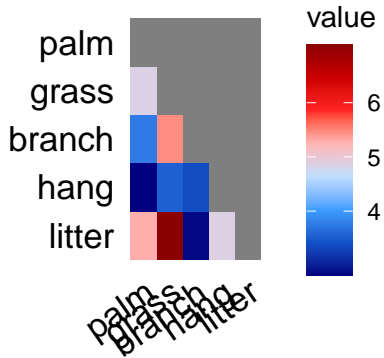
Jaccard similarity



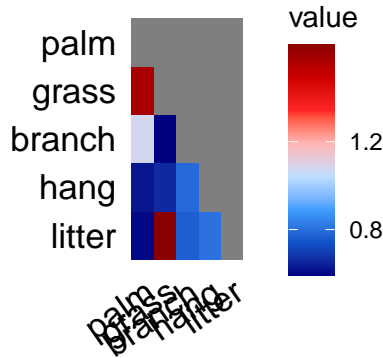
Solovese of pairwise shared components divi



Centroid distance



Minimum distance



Unsing dynRB package

Calculate hypervolume size and overlap

```
# Size and pairwise overlap
r <- dynRB_VPa(Xrb)
```

```
## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
## [1] 6
## [1] 7
## [1] 8
## [1] 9
## [1] 10
## [1] 11
## [1] 12
## [1] 13
## [1] 14
## [1] 15
## [1] 16
```

```

## [1] 17
## [1] 18
## [1] 19
## [1] 20
## [1] 21
## [1] 22
## [1] 23
## [1] 24
## [1] 25
##
##                                     V1
## 1      hanging from branches and leaves
## 2      hanging from grass
## 3      laying on branches and leaves
## 4 laying on leaf litter, logs, bark, trunk
## 5      laying on palm leaves
## 6      hanging from branches and leaves
## 7      hanging from grass
## 8      laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10     laying on palm leaves
## 11     hanging from branches and leaves
## 12     hanging from grass
## 13     laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15     laying on palm leaves
## 16     hanging from branches and leaves
## 17     hanging from grass
## 18     laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20     laying on palm leaves
## 21     hanging from branches and leaves
## 22     hanging from grass
## 23     laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25     laying on palm leaves
##
##                                     V2      port_prod  vol_V1_prod
## 1      hanging from branches and leaves 1.000000e+00 8.026471e-02
## 2      hanging from branches and leaves 2.440322e-05 6.373489e-06
## 3      hanging from branches and leaves 9.552701e-02 2.708696e-02
## 4      hanging from branches and leaves 3.985398e-02 9.275388e-03
## 5      hanging from branches and leaves 0.000000e+00 6.429999e-05
## 6      hanging from grass 3.411457e-01 8.026471e-02
## 7      hanging from grass 1.000000e+00 6.373489e-06
## 8      hanging from grass 5.028213e-02 2.708696e-02
## 9      hanging from grass 0.000000e+00 9.275388e-03
## 10     hanging from grass 0.000000e+00 6.429999e-05
## 11     laying on branches and leaves 2.377482e-01 8.026471e-02
## 12     laying on branches and leaves 8.395322e-06 6.373489e-06
## 13     laying on branches and leaves 1.000000e+00 2.708696e-02
## 14     laying on branches and leaves 8.144047e-02 9.275388e-03
## 15     laying on branches and leaves 2.329100e-04 6.429999e-05
## 16 laying on leaf litter, logs, bark, trunk 3.149566e-01 8.026471e-02
## 17 laying on leaf litter, logs, bark, trunk 0.000000e+00 6.373489e-06
## 18 laying on leaf litter, logs, bark, trunk 2.559732e-01 2.708696e-02

```

```

## 19 laying on leaf litter, logs, bark, trunk 1.000000e+00 9.275388e-03
## 20 laying on leaf litter, logs, bark, trunk 2.242259e-05 6.429999e-05
## 21          laying on palm leaves 0.000000e+00 8.026471e-02
## 22          laying on palm leaves 0.000000e+00 6.373489e-06
## 23          laying on palm leaves 1.256811e-01 2.708696e-02
## 24          laying on palm leaves 6.106268e-03 9.275388e-03
## 25          laying on palm leaves 1.000000e+00 6.429999e-05
##      vol_V2_prod
## 1  8.026471e-02
## 2  8.026471e-02
## 3  8.026471e-02
## 4  8.026471e-02
## 5  8.026471e-02
## 6  6.373489e-06
## 7  6.373489e-06
## 8  6.373489e-06
## 9  6.373489e-06
## 10 6.373489e-06
## 11 2.708696e-02
## 12 2.708696e-02
## 13 2.708696e-02
## 14 2.708696e-02
## 15 2.708696e-02
## 16 9.275388e-03
## 17 9.275388e-03
## 18 9.275388e-03
## 19 9.275388e-03
## 20 9.275388e-03
## 21 6.429999e-05
## 22 6.429999e-05
## 23 6.429999e-05
## 24 6.429999e-05
## 25 6.429999e-05

```

```
r$result
```

```

##                                     V1
## 1          hanging from branches and leaves
## 2                hanging from grass
## 3          laying on branches and leaves
## 4 laying on leaf litter, logs, bark, trunk
## 5                laying on palm leaves
## 6          hanging from branches and leaves
## 7                hanging from grass
## 8          laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10               laying on palm leaves
## 11          hanging from branches and leaves
## 12               hanging from grass
## 13          laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15               laying on palm leaves
## 16          hanging from branches and leaves
## 17               hanging from grass

```

```

## 18          laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20          laying on palm leaves
## 21          hanging from branches and leaves
## 22          hanging from grass
## 23          laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25          laying on palm leaves
##
##          V2      port_prod port_mean port_gmean
## 1      hanging from branches and leaves 1.000000e+00 1.0000000 1.00000000
## 2      hanging from branches and leaves 2.440322e-05 0.2785816 0.13636491
## 3      hanging from branches and leaves 9.552701e-02 0.7055622 0.68074627
## 4      hanging from branches and leaves 3.985398e-02 0.6805866 0.56350439
## 5      hanging from branches and leaves 0.000000e+00 0.3491063 0.00000000
## 6          hanging from grass 3.411457e-01 0.8905466 0.52079172
## 7          hanging from grass 1.000000e+00 1.0000000 1.00000000
## 8          hanging from grass 5.028213e-02 0.6071049 0.38025652
## 9          hanging from grass 0.000000e+00 0.7916137 0.00000000
## 10         hanging from grass 0.000000e+00 0.4176534 0.00000000
## 11         laying on branches and leaves 2.377482e-01 0.7929074 0.76188259
## 12         laying on branches and leaves 8.395322e-06 0.2022186 0.11357210
## 13         laying on branches and leaves 1.000000e+00 1.0000000 1.00000000
## 14         laying on branches and leaves 8.144047e-02 0.6704278 0.66073281
## 15         laying on branches and leaves 2.329100e-04 0.3734930 0.18744212
## 16 laying on leaf litter, logs, bark, trunk 3.149566e-01 0.8730048 0.75080560
## 17 laying on leaf litter, logs, bark, trunk 0.000000e+00 0.3390452 0.00000000
## 18 laying on leaf litter, logs, bark, trunk 2.559732e-01 0.7884876 0.78182337
## 19 laying on leaf litter, logs, bark, trunk 1.000000e+00 1.0000000 1.00000000
## 20 laying on leaf litter, logs, bark, trunk 2.242259e-05 0.3216908 0.07940308
## 21         laying on palm leaves 0.000000e+00 0.8509387 0.00000000
## 22         laying on palm leaves 0.000000e+00 0.3071580 0.00000000
## 23         laying on palm leaves 1.256811e-01 0.8178359 0.44802457
## 24         laying on palm leaves 6.106268e-03 0.5905056 0.17040009
## 25         laying on palm leaves 1.000000e+00 1.0000000 1.00000000
##
##      vol_V1_prod vol_V1_mean vol_V1_gmean  vol_V2_prod vol_V2_mean vol_V2_gmean
## 1  8.026471e-02  0.9664312  0.9626571 8.026471e-02  0.9664312  0.9626571
## 2  6.373489e-06  0.4744041  0.4397935 8.026471e-02  0.9664312  0.9626571
## 3  2.708696e-02  0.9111737  0.9103507 8.026471e-02  0.9664312  0.9626571
## 4  9.275388e-03  0.8471250  0.8335036 8.026471e-02  0.9664312  0.9626571
## 5  6.429999e-05  0.5704378  0.5544605 8.026471e-02  0.9664312  0.9626571
## 6  8.026471e-02  0.9664312  0.9626571 6.373489e-06  0.4744041  0.4397935
## 7  6.373489e-06  0.4744041  0.4397935 6.373489e-06  0.4744041  0.4397935
## 8  2.708696e-02  0.9111737  0.9103507 6.373489e-06  0.4744041  0.4397935
## 9  9.275388e-03  0.8471250  0.8335036 6.373489e-06  0.4744041  0.4397935
## 10 6.429999e-05  0.5704378  0.5544605 6.373489e-06  0.4744041  0.4397935
## 11 8.026471e-02  0.9664312  0.9626571 2.708696e-02  0.9111737  0.9103507
## 12 6.373489e-06  0.4744041  0.4397935 2.708696e-02  0.9111737  0.9103507
## 13 2.708696e-02  0.9111737  0.9103507 2.708696e-02  0.9111737  0.9103507
## 14 9.275388e-03  0.8471250  0.8335036 2.708696e-02  0.9111737  0.9103507
## 15 6.429999e-05  0.5704378  0.5544605 2.708696e-02  0.9111737  0.9103507
## 16 8.026471e-02  0.9664312  0.9626571 9.275388e-03  0.8471250  0.8335036
## 17 6.373489e-06  0.4744041  0.4397935 9.275388e-03  0.8471250  0.8335036
## 18 2.708696e-02  0.9111737  0.9103507 9.275388e-03  0.8471250  0.8335036
## 19 9.275388e-03  0.8471250  0.8335036 9.275388e-03  0.8471250  0.8335036

```

```
## 20 6.429999e-05 0.5704378 0.5544605 9.275388e-03 0.8471250 0.8335036
## 21 8.026471e-02 0.9664312 0.9626571 6.429999e-05 0.5704378 0.5544605
## 22 6.373489e-06 0.4744041 0.4397935 6.429999e-05 0.5704378 0.5544605
## 23 2.708696e-02 0.9111737 0.9103507 6.429999e-05 0.5704378 0.5544605
## 24 9.275388e-03 0.8471250 0.8335036 6.429999e-05 0.5704378 0.5544605
## 25 6.429999e-05 0.5704378 0.5544605 6.429999e-05 0.5704378 0.5544605
```

Quantification of overlaps per dimension

```
r1 <- dynRB_Pn(Xrb)
```

```
## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
## [1] 6
## [1] 7
## [1] 8
## [1] 9
## [1] 10
## [1] 11
## [1] 12
## [1] 13
## [1] 14
## [1] 15
## [1] 16
## [1] 17
## [1] 18
## [1] 19
## [1] 20
## [1] 21
## [1] 22
## [1] 23
## [1] 24
## [1] 25
##
## V1
## 1 hanging from branches and leaves
## 2 hanging from grass
## 3 laying on branches and leaves
## 4 laying on leaf litter, logs, bark, trunk
## 5 laying on palm leaves
## 6 hanging from branches and leaves
## 7 hanging from grass
## 8 laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10 laying on palm leaves
## 11 hanging from branches and leaves
## 12 hanging from grass
## 13 laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15 laying on palm leaves
## 16 hanging from branches and leaves
## 17 hanging from grass
## 18 laying on branches and leaves
```


##		V2	PC1	PC2	PC3
## 19	laying on leaf litter, logs, bark, trunk				
## 20	laying on palm leaves				
## 21	hanging from branches and leaves				
## 22	hanging from grass				
## 23	laying on branches and leaves				
## 24	laying on leaf litter, logs, bark, trunk				
## 25	laying on palm leaves				
## 1	hanging from branches and leaves	1.000000000	1.000000000	1.000000000	
## 2	hanging from branches and leaves	0.210093186	0.15419043	0.06797012	
## 3	hanging from branches and leaves	0.245567579	0.18422464	0.83133033	
## 4	hanging from branches and leaves	0.090450812	0.47179467	0.13086272	
## 5	hanging from branches and leaves	0.230951054	0.26416445	0.36483523	
## 6	hanging from grass	0.754957567	0.49773293	0.29872106	
## 7	hanging from grass	1.000000000	1.000000000	1.000000000	
## 8	hanging from grass	0.022221437	0.36840968	0.54328816	
## 9	hanging from grass	0.000000000	0.96703375	0.82078817	
## 10	hanging from grass	0.033996642	0.67273909	0.07913739	
## 11	laying on branches and leaves	0.336640512	0.16731929	0.56598094	
## 12	laying on branches and leaves	0.006520454	0.08255016	0.11941625	
## 13	laying on branches and leaves	1.000000000	1.000000000	1.000000000	
## 14	laying on branches and leaves	0.218087538	0.19757597	0.27186404	
## 15	laying on branches and leaves	0.201298722	0.12316184	0.36586459	
## 16	laying on leaf litter, logs, bark, trunk	0.183762469	0.50901019	0.20111997	
## 17	laying on leaf litter, logs, bark, trunk	0.000000000	0.27460778	0.28621764	
## 18	laying on leaf litter, logs, bark, trunk	0.303436493	0.32532605	0.45220830	
## 19	laying on leaf litter, logs, bark, trunk	1.000000000	1.000000000	1.000000000	
## 20	laying on leaf litter, logs, bark, trunk	0.003746752	0.44542313	0.09191161	
## 21	laying on palm leaves	0.605950491	0.47956676	0.59938087	
## 22	laying on palm leaves	0.026421728	0.36539124	0.03243976	
## 23	laying on palm leaves	0.416935434	0.34024043	0.84317380	
## 24	laying on palm leaves	0.007461446	0.87495960	0.12397693	
## 25	laying on palm leaves	1.000000000	1.000000000	1.000000000	
##		PC4	PC5	PC6	PC7
## 1		1.000000000	1.000000000	1.000000000	1.000000000
## 2		0.007245944	0.147382935	0.11493125	0.47281327
## 3		0.538143134	0.531894990	0.71956128	0.26642829
## 4		0.481805738	0.888464818	0.74639091	0.57809040
## 5		0.202545765	0.000000000	0.57861936	0.08817817
## 6		0.060941341	0.339478548	0.30148918	0.93812826
## 7		1.000000000	1.000000000	1.000000000	1.000000000
## 8		0.055257466	0.093784584	0.17467979	0.17518207
## 9		0.129375534	0.385738150	0.23646719	0.75493422
## 10		0.000000000	0.000000000	0.14318899	0.11801872
## 11		0.856507649	0.482495802	0.77047317	0.26797356
## 12		0.008246031	0.036707930	0.08404161	0.12277058
## 13		1.000000000	1.000000000	1.000000000	1.000000000
## 14		0.427191240	0.573826192	0.76873067	0.20261539
## 15		0.229707180	0.012052704	0.74612937	0.02214705
## 16		0.555625433	0.770764887	0.89744298	0.93180583
## 17		0.016146077	0.151655583	0.11882992	0.59934011
## 18		0.326731016	0.575980672	0.86729906	0.25902412
## 19		1.000000000	1.000000000	1.000000000	1.000000000
## 20		0.149927635	0.003067077	0.70551432	0.06270849

```

## 21 0.865627804 0.000000000 0.68811003 0.19163046
## 22 0.000000000 0.000000000 0.08289027 0.19765878
## 23 0.601310676 0.036089895 0.81467072 0.06650936
## 24 0.523962305 0.010433731 0.69380825 0.10825052
## 25 1.000000000 1.000000000 1.00000000 1.00000000

```

```
r1$result
```

```

##                               V1
## 1      hanging from branches and leaves
## 2                               hanging from grass
## 3      laying on branches and leaves
## 4 laying on leaf litter, logs, bark, trunk
## 5                               laying on palm leaves
## 6      hanging from branches and leaves
## 7                               hanging from grass
## 8      laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10                              laying on palm leaves
## 11      hanging from branches and leaves
## 12                              hanging from grass
## 13      laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15                              laying on palm leaves
## 16      hanging from branches and leaves
## 17                              hanging from grass
## 18      laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20                              laying on palm leaves
## 21      hanging from branches and leaves
## 22                              hanging from grass
## 23      laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25                              laying on palm leaves
##                               V2          PC1          PC2          PC3
## 1      hanging from branches and leaves 1.000000000 1.00000000 1.00000000
## 2      hanging from branches and leaves 0.210093186 0.15419043 0.06797012
## 3      hanging from branches and leaves 0.245567579 0.18422464 0.83133033
## 4      hanging from branches and leaves 0.090450812 0.47179467 0.13086272
## 5      hanging from branches and leaves 0.230951054 0.26416445 0.36483523
## 6      hanging from grass 0.754957567 0.49773293 0.29872106
## 7      hanging from grass 1.000000000 1.00000000 1.00000000
## 8      hanging from grass 0.022221437 0.36840968 0.54328816
## 9      hanging from grass 0.000000000 0.96703375 0.82078817
## 10     hanging from grass 0.033996642 0.67273909 0.07913739
## 11     laying on branches and leaves 0.336640512 0.16731929 0.56598094
## 12     laying on branches and leaves 0.006520454 0.08255016 0.11941625
## 13     laying on branches and leaves 1.000000000 1.00000000 1.00000000
## 14     laying on branches and leaves 0.218087538 0.19757597 0.27186404
## 15     laying on branches and leaves 0.201298722 0.12316184 0.36586459
## 16 laying on leaf litter, logs, bark, trunk 0.183762469 0.50901019 0.20111997
## 17 laying on leaf litter, logs, bark, trunk 0.000000000 0.27460778 0.28621764
## 18 laying on leaf litter, logs, bark, trunk 0.303436493 0.32532605 0.45220830
## 19 laying on leaf litter, logs, bark, trunk 1.000000000 1.00000000 1.00000000

```

```

## 20 laying on leaf litter, logs, bark, trunk 0.003746752 0.44542313 0.09191161
## 21          laying on palm leaves 0.605950491 0.47956676 0.59938087
## 22          laying on palm leaves 0.026421728 0.36539124 0.03243976
## 23          laying on palm leaves 0.416935434 0.34024043 0.84317380
## 24          laying on palm leaves 0.007461446 0.87495960 0.12397693
## 25          laying on palm leaves 1.000000000 1.00000000 1.00000000
##          PC4          PC5          PC6          PC7
## 1  1.000000000 1.000000000 1.000000000 1.000000000
## 2  0.007245944 0.147382935 0.11493125 0.47281327
## 3  0.538143134 0.531894990 0.71956128 0.26642829
## 4  0.481805738 0.888464818 0.74639091 0.57809040
## 5  0.202545765 0.000000000 0.57861936 0.08817817
## 6  0.060941341 0.339478548 0.30148918 0.93812826
## 7  1.000000000 1.000000000 1.000000000 1.000000000
## 8  0.055257466 0.093784584 0.17467979 0.17518207
## 9  0.129375534 0.385738150 0.23646719 0.75493422
## 10 0.000000000 0.000000000 0.14318899 0.11801872
## 11 0.856507649 0.482495802 0.77047317 0.26797356
## 12 0.008246031 0.036707930 0.08404161 0.12277058
## 13 1.000000000 1.000000000 1.000000000 1.000000000
## 14 0.427191240 0.573826192 0.76873067 0.20261539
## 15 0.229707180 0.012052704 0.74612937 0.02214705
## 16 0.555625433 0.770764887 0.89744298 0.93180583
## 17 0.016146077 0.151655583 0.11882992 0.59934011
## 18 0.326731016 0.575980672 0.86729906 0.25902412
## 19 1.000000000 1.000000000 1.000000000 1.000000000
## 20 0.149927635 0.003067077 0.70551432 0.06270849
## 21 0.865627804 0.000000000 0.68811003 0.19163046
## 22 0.000000000 0.000000000 0.08289027 0.19765878
## 23 0.601310676 0.036089895 0.81467072 0.06650936
## 24 0.523962305 0.010433731 0.69380825 0.10825052
## 25 1.000000000 1.000000000 1.000000000 1.000000000

```

```
# Quantification of sizes per dimension
```

```
r2 <- dynRB_Vn(Xrb)
```

```

## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
##          V1          PC1          PC2          PC3
## 1  hanging from branches and leaves 0.5426586 0.3553761 0.3919268
## 2          hanging from grass 0.1527619 0.1012398 0.1061405
## 3  laying on branches and leaves 0.4630451 0.7970514 0.5705280
## 4 laying on leaf litter, logs, bark, trunk 0.3249917 0.3540151 0.3055534
## 5          laying on palm leaves 0.1761936 0.1666608 0.2198220
##          PC4          PC5          PC6          PC7
## 1 0.60290402 0.4079075 0.4665972 0.3995189
## 2 0.07439702 0.2128673 0.1579574 0.2024307
## 3 0.38070971 0.5448182 0.4385814 0.3718108
## 4 0.54008980 0.5011369 0.3935210 0.2508620
## 5 0.14165851 0.1392975 0.4032588 0.1448645

```

```
r2$result
```

```
##          V1          PC1          PC2          PC3
## 1 hanging from branches and leaves 0.5426586 0.3553761 0.3919268
## 2 hanging from grass 0.1527619 0.1012398 0.1061405
## 3 laying on branches and leaves 0.4630451 0.7970514 0.5705280
## 4 laying on leaf litter, logs, bark, trunk 0.3249917 0.3540151 0.3055534
## 5 laying on palm leaves 0.1761936 0.1666608 0.2198220
##          PC4          PC5          PC6          PC7
## 1 0.60290402 0.4079075 0.4665972 0.3995189
## 2 0.07439702 0.2128673 0.1579574 0.2024307
## 3 0.38070971 0.5448182 0.4385814 0.3718108
## 4 0.54008980 0.5011369 0.3935210 0.2508620
## 5 0.14165851 0.1392975 0.4032588 0.1448645
```

Plot heatmaps

```
# Functions to define color gradients
signif.floor <- function(x, n) {
  pow <- floor(log10(abs(x))) + 1 - n
  y <- floor(x/10^pow) * 10^pow
  # handle the x = 0 case
  y[x == 0] <- 0
  y
}
signif.ceiling <- function(x, n) {
  pow <- floor(log10(abs(x))) + 1 - n
  y <- ceiling(x/10^pow) * 10^pow
  # handle the x = 0 case
  y[x == 0] <- 0
  y
}

# Plot
theme_change <- theme(plot.title = element_text(family = "Helvetica",
  face = "bold", size = (15), hjust = 0.5), plot.subtitle = element_text(family = "Helvetica",
  size = (10), hjust = 0.5), plot.background = element_blank(),
  panel.grid.minor = element_blank(), panel.grid.major = element_blank(),
  panel.background = element_blank(), panel.border = element_blank(),
  axis.line = element_blank(), axis.ticks = element_blank(),
  axis.text.x = element_text(colour = "black", size = rel(1.5),
  angle = 35, hjust = 1), axis.text.y = element_text(colour = "black",
  size = rel(1.5)), axis.title.x = element_blank(), axis.title.y = element_blank())
result <- r$result
```

Overall overlap (product)

```
Overlap <- as.numeric(ifelse(result$V1 == result$V2, "NA", result$port_prod))
Result2 <- cbind(result, Overlap)
```

```

breaks <- seq(min(Result2$Overlap, na.rm = T), max(Result2$Overlap,
  na.rm = T), by = 0.1)
col1 <- colorRampPalette(c("navyblue", "dodgerblue", "white",
  "red", "darkred")) #define color gradient

Result2$V1 <- gsub("hanging from branches and leaves", "hang",
  Result2$V1)
Result2$V1 <- gsub("hanging from grass", "grass", Result2$V1)
Result2$V1 <- gsub("laying on branches and leaves", "branch",
  Result2$V1)
Result2$V1 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V1)
Result2$V1 <- gsub("laying on palm leaves", "palm", Result2$V1)
Result2$V2 <- gsub("hanging from branches and leaves", "hang",
  Result2$V2)
Result2$V2 <- gsub("hanging from grass", "grass", Result2$V2)
Result2$V2 <- gsub("laying on branches and leaves", "branch",
  Result2$V2)
Result2$V2 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V2)
Result2$V2 <- gsub("laying on palm leaves", "palm", Result2$V2)

lev <- ordered(c("palm", "grass", "branch", "hang", "litter"))
lev2 <- ordered(c("litter", "hang", "branch", "grass", "palm"))

p6 <- ggplot(Result2, aes(x = ordered(V1, levels = lev), y = ordered(V2,
  levels = lev2))) + geom_tile(data = subset(Result2, !is.na(Overlap)),
  aes(fill = Overlap), color = "black") + geom_tile(data = subset(Result2,
  is.na(Overlap)), fill = "lightgrey", color = "black") + scale_fill_gradientn(colours = col1(8),
  breaks = round(breaks, 1), guide = "colorbar", limits = c(signif.floor(min(breaks),
  1), signif.ceiling(max(breaks), 1))) + labs(title = "Overall overlap",
  subtitle = "Aggregation = product") + theme_change

```

Overall overlap (arithmetic mean)

```

Overlap <- as.numeric(ifelse(result$V1 == result$V2, "NA", result$port_mean))
Result2 <- cbind(result, Overlap)
breaks <- seq(min(Result2$Overlap, na.rm = T) - 0.05, max(Result2$Overlap +
  0.05, na.rm = T), by = 0.05)
col1 <- colorRampPalette(c("navyblue", "dodgerblue", "white",
  "red", "darkred")) #define color gradient
Result2$V1 <- gsub("hanging from branches and leaves", "hang",
  Result2$V1)
Result2$V1 <- gsub("hanging from grass", "grass", Result2$V1)
Result2$V1 <- gsub("laying on branches and leaves", "branch",
  Result2$V1)
Result2$V1 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V1)
Result2$V1 <- gsub("laying on palm leaves", "palm", Result2$V1)
Result2$V2 <- gsub("hanging from branches and leaves", "hang",
  Result2$V2)

```

```

Result2$V2 <- gsub("hanging from grass", "grass", Result2$V2)
Result2$V2 <- gsub("laying on branches and leaves", "branch",
  Result2$V2)
Result2$V2 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V2)
Result2$V2 <- gsub("laying on palm leaves", "palm", Result2$V2)
lev <- ordered(c("palm", "grass", "branch", "hang", "litter"))
lev2 <- ordered(c("litter", "hang", "branch", "grass", "palm"))

p11 <- ggplot(Result2, aes(x = ordered(V1, levels = lev), y = ordered(V2,
  levels = lev2))) + geom_tile(data = subset(Result2, !is.na(Overlap)),
  aes(fill = Overlap), color = "black") + geom_tile(data = subset(Result2,
  is.na(Overlap)), fill = "lightgrey", color = "black") + scale_fill_gradientn(colours = col1(8),
  breaks = round(breaks, 2), guide = "colorbar", limits = c(signif.floor(min(breaks),
  2), signif.ceiling(max(breaks), 2))) + labs(title = "Overall overlap",
  subtitle = "Aggregation = mean") + theme_change

```

Overall overlap (geometric mean)

```

Overlap <- as.numeric(ifelse(result$V1 == result$V2, "NA", result$port_gmean))
Result2 <- cbind(result, Overlap)
breaks <- seq(min(Result2$Overlap, na.rm = T) - 0.05, max(Result2$Overlap +
  0.05, na.rm = T), by = 0.05)
col1 <- colorRampPalette(c("navyblue", "dodgerblue", "white",
  "red", "darkred")) #define color gradient
Result2$V1 <- gsub("hanging from branches and leaves", "hang",
  Result2$V1)
Result2$V1 <- gsub("hanging from grass", "grass", Result2$V1)
Result2$V1 <- gsub("laying on branches and leaves", "branch",
  Result2$V1)
Result2$V1 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V1)
Result2$V1 <- gsub("laying on palm leaves", "palm", Result2$V1)
Result2$V2 <- gsub("hanging from branches and leaves", "hang",
  Result2$V2)
Result2$V2 <- gsub("hanging from grass", "grass", Result2$V2)
Result2$V2 <- gsub("laying on branches and leaves", "branch",
  Result2$V2)
Result2$V2 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V2)
Result2$V2 <- gsub("laying on palm leaves", "palm", Result2$V2)
lev <- ordered(c("palm", "grass", "branch", "hang", "litter"))
lev2 <- ordered(c("litter", "hang", "branch", "grass", "palm"))

p12 <- ggplot(Result2, aes(x = ordered(V1, levels = lev), y = ordered(V2,
  levels = lev2))) + geom_tile(data = subset(Result2, !is.na(Overlap)),
  aes(fill = Overlap), color = "black") + geom_tile(data = subset(Result2,
  is.na(Overlap)), fill = "lightgrey", color = "black") + scale_fill_gradientn(colours = col1(8),
  breaks = round(breaks, 2), guide = "colorbar", limits = c(signif.floor(min(breaks),
  2), signif.ceiling(max(breaks), 2))) + labs(title = "Overall overlap",
  subtitle = "Aggregation = geometric mean") + theme_change

```

Overlap per dimension

```
result <- r1$result
Overlap <- as.numeric(iffelse(result$V1 == result$V2, "NA", result$PC1))
Result2 <- cbind(result, Overlap)
breaks <- seq(min(Result2$Overlap, na.rm = T) - 0.05, max(Result2$Overlap,
  na.rm = T) + 0.1, by = 0.1)
col1 <- colorRampPalette(c("navyblue", "dodgerblue", "white",
  "red", "darkred")) #define color gradient
Result2$V1 <- gsub("hanging from branches and leaves", "hang",
  Result2$V1)
Result2$V1 <- gsub("hanging from grass", "grass", Result2$V1)
Result2$V1 <- gsub("laying on branches and leaves", "branch",
  Result2$V1)
Result2$V1 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V1)
Result2$V1 <- gsub("laying on palm leaves", "palm", Result2$V1)
Result2$V2 <- gsub("hanging from branches and leaves", "hang",
  Result2$V2)
Result2$V2 <- gsub("hanging from grass", "grass", Result2$V2)
Result2$V2 <- gsub("laying on branches and leaves", "branch",
  Result2$V2)
Result2$V2 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V2)
Result2$V2 <- gsub("laying on palm leaves", "palm", Result2$V2)
lev <- ordered(c("palm", "grass", "branch", "hang", "litter"))
lev2 <- ordered(c("litter", "hang", "branch", "grass", "palm"))
p13 <- ggplot(Result2, aes(x = ordered(V1, levels = lev), y = ordered(V2,
  levels = lev2))) + geom_tile(data = subset(Result2, !is.na(Overlap)),
  aes(fill = Overlap), color = "black") + geom_tile(data = subset(Result2,
  is.na(Overlap)), fill = "lightgrey", color = "black") + scale_fill_gradientn(colours = col1(8),
  breaks = round(breaks, 2), guide = "colorbar", limits = c(signif.floor(min(breaks),
  2), signif.ceiling(max(breaks), 2))) + labs(title = "Overlap PC1",
  subtitle = "Relative body width") + theme_change
```

PC1 (Relative body width)

```
Overlap <- as.numeric(iffelse(result$V1 == result$V2, "NA", result$PC2))
Result2 <- cbind(result, Overlap)
breaks <- seq(min(Result2$Overlap, na.rm = T) - 0.05, max(Result2$Overlap,
  na.rm = T) + 0.1, by = 0.1)
col1 <- colorRampPalette(c("navyblue", "dodgerblue", "white",
  "red", "darkred")) #define color gradient
Result2$V1 <- gsub("hanging from branches and leaves", "hang",
  Result2$V1)
Result2$V1 <- gsub("hanging from grass", "grass", Result2$V1)
Result2$V1 <- gsub("laying on branches and leaves", "branch",
  Result2$V1)
Result2$V1 <- gsub("laying on leaf litter, logs, bark, trunk",
```

```

    "litter", Result2$V1)
Result2$V1 <- gsub("laying on palm leaves", "palm", Result2$V1)
Result2$V2 <- gsub("hanging from branches and leaves", "hang",
  Result2$V2)
Result2$V2 <- gsub("hanging from grass", "grass", Result2$V2)
Result2$V2 <- gsub("laying on branches and leaves", "branch",
  Result2$V2)
Result2$V2 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V2)
Result2$V2 <- gsub("laying on palm leaves", "palm", Result2$V2)
lev <- ordered(c("palm", "grass", "branch", "hang", "litter"))
lev2 <- ordered(c("litter", "hang", "branch", "grass", "palm"))
p14 <- ggplot(Result2, aes(x = ordered(V1, levels = lev), y = ordered(V2,
  levels = lev2))) + geom_tile(data = subset(Result2, !is.na(Overlap)),
  aes(fill = Overlap), color = "black") + geom_tile(data = subset(Result2,
  is.na(Overlap)), fill = "lightgrey", color = "black") + scale_fill_gradientn(colours = col1(8),
  breaks = round(breaks, 2), guide = "colorbar", limits = c(signif.floor(min(breaks),
  2), signif.ceiling(max(breaks), 2))) + labs(title = "Overlap PC2",
  subtitle = "Relative wing size") + theme_change

```

PC2 (Relative wing size)

```

Overlap <- as.numeric(ifelse(result$V1 == result$V2, "NA", result$PC3))
Result2 <- cbind(result, Overlap)
breaks <- seq(min(Result2$Overlap, na.rm = T) - 0.05, max(Result2$Overlap,
  na.rm = T) + 0.1, by = 0.1)
col1 <- colorRampPalette(c("navyblue", "dodgerblue", "white",
  "red", "darkred")) #define color gradient
Result2$V1 <- gsub("hanging from branches and leaves", "hang",
  Result2$V1)
Result2$V1 <- gsub("hanging from grass", "grass", Result2$V1)
Result2$V1 <- gsub("laying on branches and leaves", "branch",
  Result2$V1)
Result2$V1 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V1)
Result2$V1 <- gsub("laying on palm leaves", "palm", Result2$V1)
Result2$V2 <- gsub("hanging from branches and leaves", "hang",
  Result2$V2)
Result2$V2 <- gsub("hanging from grass", "grass", Result2$V2)
Result2$V2 <- gsub("laying on branches and leaves", "branch",
  Result2$V2)
Result2$V2 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V2)
Result2$V2 <- gsub("laying on palm leaves", "palm", Result2$V2)
lev <- ordered(c("palm", "grass", "branch", "hang", "litter"))
lev2 <- ordered(c("litter", "hang", "branch", "grass", "palm"))
p15 <- ggplot(Result2, aes(x = ordered(V1, levels = lev), y = ordered(V2,
  levels = lev2))) + geom_tile(data = subset(Result2, !is.na(Overlap)),
  aes(fill = Overlap), color = "black") + geom_tile(data = subset(Result2,
  is.na(Overlap)), fill = "lightgrey", color = "black") + scale_fill_gradientn(colours = col1(8),

```



```
breaks = round(breaks, 2), guide = "colorbar", limits = c(signif.floor(min(breaks),
  2), signif.ceiling(max(breaks), 2))) + labs(title = "Overlap PC3",
  subtitle = "Relative body height") + theme_change
```

PC3 (Relative body height)

```
Overlap <- as.numeric(ifelse(result$V1 == result$V2, "NA", result$PC4))
Result2 <- cbind(result, Overlap)
breaks <- seq(min(Result2$Overlap, na.rm = T) - 0.05, max(Result2$Overlap,
  na.rm = T) + 0.1, by = 0.1)
col1 <- colorRampPalette(c("navyblue", "dodgerblue", "white",
  "red", "darkred")) #define color gradient
Result2$V1 <- gsub("hanging from branches and leaves", "hang",
  Result2$V1)
Result2$V1 <- gsub("hanging from grass", "grass", Result2$V1)
Result2$V1 <- gsub("laying on branches and leaves", "branch",
  Result2$V1)
Result2$V1 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V1)
Result2$V1 <- gsub("laying on palm leaves", "palm", Result2$V1)
Result2$V2 <- gsub("hanging from branches and leaves", "hang",
  Result2$V2)
Result2$V2 <- gsub("hanging from grass", "grass", Result2$V2)
Result2$V2 <- gsub("laying on branches and leaves", "branch",
  Result2$V2)
Result2$V2 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V2)
Result2$V2 <- gsub("laying on palm leaves", "palm", Result2$V2)
lev <- ordered(c("palm", "grass", "branch", "hang", "litter"))
lev2 <- ordered(c("litter", "hang", "branch", "grass", "palm"))
p16 <- ggplot(Result2, aes(x = ordered(V1, levels = lev), y = ordered(V2,
  levels = lev2))) + geom_tile(data = subset(Result2, !is.na(Overlap)),
  aes(fill = Overlap), color = "black") + geom_tile(data = subset(Result2,
  is.na(Overlap)), fill = "lightgrey", color = "black") + scale_fill_gradientn(colours = col1(8),
  breaks = round(breaks, 2), guide = "colorbar", limits = c(signif.floor(min(breaks),
  2), signif.ceiling(max(breaks), 2))) + labs(title = "Overlap PC4",
  subtitle = "Spininess") + theme_change
```

PC4 (Spininess)

```
Overlap <- as.numeric(ifelse(result$V1 == result$V2, "NA", result$PC5))
Result2 <- cbind(result, Overlap)
breaks <- seq(min(Result2$Overlap, na.rm = T) - 0.05, max(Result2$Overlap,
  na.rm = T) + 0.1, by = 0.1)
col1 <- colorRampPalette(c("navyblue", "dodgerblue", "white",
  "red", "darkred")) #define color gradient
Result2$V1 <- gsub("hanging from branches and leaves", "hang",
```

```

Result2$V1)
Result2$V1 <- gsub("hanging from grass", "grass", Result2$V1)
Result2$V1 <- gsub("laying on branches and leaves", "branch",
  Result2$V1)
Result2$V1 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V1)
Result2$V1 <- gsub("laying on palm leaves", "palm", Result2$V1)
Result2$V2 <- gsub("hanging from branches and leaves", "hang",
  Result2$V2)
Result2$V2 <- gsub("hanging from grass", "grass", Result2$V2)
Result2$V2 <- gsub("laying on branches and leaves", "branch",
  Result2$V2)
Result2$V2 <- gsub("laying on leaf litter, logs, bark, trunk",
  "litter", Result2$V2)
Result2$V2 <- gsub("laying on palm leaves", "palm", Result2$V2)
lev <- ordered(c("palm", "grass", "branch", "hang", "litter"))
lev2 <- ordered(c("litter", "hang", "branch", "grass", "palm"))
p17 <- ggplot(Result2, aes(x = ordered(V1, levels = lev), y = ordered(V2,
  levels = lev2))) + geom_tile(data = subset(Result2, !is.na(Overlap)),
  aes(fill = Overlap), color = "black") + geom_tile(data = subset(Result2,
  is.na(Overlap)), fill = "lightgrey", color = "black") + scale_fill_gradientn(colours = col1(8),
  breaks = round(breaks, 2), guide = "colorbar", limits = c(signif.floor(min(breaks),
  2), signif.ceiling(max(breaks), 2))) + labs(title = "Overlap PC5",
  subtitle = "Relative head size") + theme_change

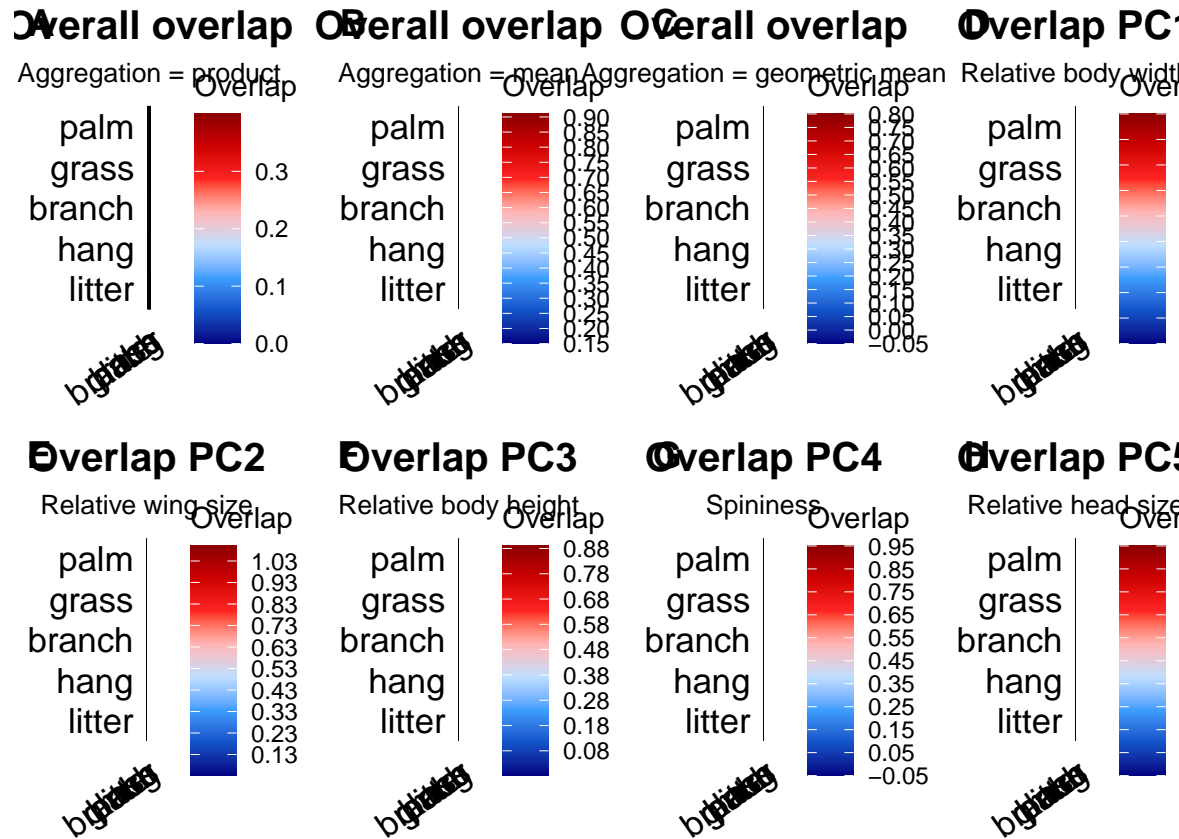
```

PC5 (Relative head size)

```

plot_grid(p6, p11, p12, p13, p14, p15, p16, p17, nrow = 2, labels = "AUTO")

```



Plot all dimensions

Hypervolume size - Habitat

```

result <- r2$result
hs <- subset(result, result$V1 == "hanging from branches and leaves")
ht <- subset(result, result$V1 == "hanging from grass")
ll <- subset(result, result$V1 == "laying on leaf litter, logs, bark, trunk")
lp <- subset(result, result$V1 == "laying on palm leaves")
lb <- subset(result, result$V1 == "laying on branches and leaves")

hs <- as.data.frame(t(hs[, -1]))
hs$habitat <- "hanging from branches and leaves"
hs$PC <- row.names(hs)
colnames(hs) <- c("range", "habitat", "PC")

ht <- as.data.frame(t(ht[, -1]))
ht$habitat <- "hanging from grass"
ht$PC <- row.names(ht)
colnames(ht) <- c("range", "habitat", "PC")

ll <- as.data.frame(t(ll[, -1]))
ll$habitat <- "laying on leaf litter, logs, bark, trunk"
ll$PC <- row.names(ll)

```

```

colnames(ll) <- c("range", "habitat", "PC")

lp <- as.data.frame(t(lp[, -1]))
lp$habitat <- "laying on palm leaves"
lp$PC <- row.names(lp)
colnames(lp) <- c("range", "habitat", "PC")

lb <- as.data.frame(t(lb[, -1]))
lb$habitat <- "laying on branches and leaves"
lb$PC <- row.names(lb)
colnames(lb) <- c("range", "habitat", "PC")

hab <- rbind(hs, ht, ll, lp, lb)

hab$PC <- factor(hab$PC, levels = c("PC1", "PC2", "PC3", "PC4",
  "PC5", "PC6", "PC7"))
p <- ggplot(hab, aes(x = PC, y = range)) + geom_bar(aes(fill = habitat),
  stat = "identity", position = position_dodge(0.6), width = 0.5) +
  scale_fill_manual(values = c("darkseagreen", "gold", "darkorange",
    "darkred", "dodgerblue")) + ylab("Occupied range") +
  xlab("") + #guides(fill=F)+ xlab("") + #guides(fill=F)+
  theme_classic()

```

Relative hypervolume size on each PC axis

Full hypervolume size

```

result <- r$result
vol <- as.data.frame(matrix(nrow = 5, ncol = 2))
colnames(vol) <- c("volume", "habitat")
vol$habitat <- c("hanging from branches and leaves", "hanging from grass",
  "laying on branches and leaves", "laying on leaf litter, logs, bark, trunk",
  "laying on palm leaves")
# product
vol[, 1] <- c(result[1:5, 6])
p1 <- ggplot(vol, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Full niche hypervolume") +
  # xlab('')+
  coord_cartesian(ylim = c(-1e-04, 0.1)) + labs(title = "Product") +
  scale_x_discrete(labels = NULL, breaks = NULL) + guides(fill = F) +
  theme_classic()

```

Product

```

vol[, 1] <- c(result[1:5, 7])
p2 <- ggplot(vol, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Full niche hypervolume") +
  ylab("Full niche hypervolume") + xlab("") + labs(title = "Mean") +
  scale_x_discrete(labels = NULL, breaks = NULL) + guides(fill = F) +
  theme_classic()

```

Arithmetic mean

```

vol[, 1] <- c(result[1:5, 8])
p3 <- ggplot(vol, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Full niche hypervolume") +
  ylab("Full niche hypervolume") + xlab("") + labs(title = "Geometric mean") +
  scale_x_discrete(labels = NULL, breaks = NULL) + guides(fill = F) +
  theme_classic()

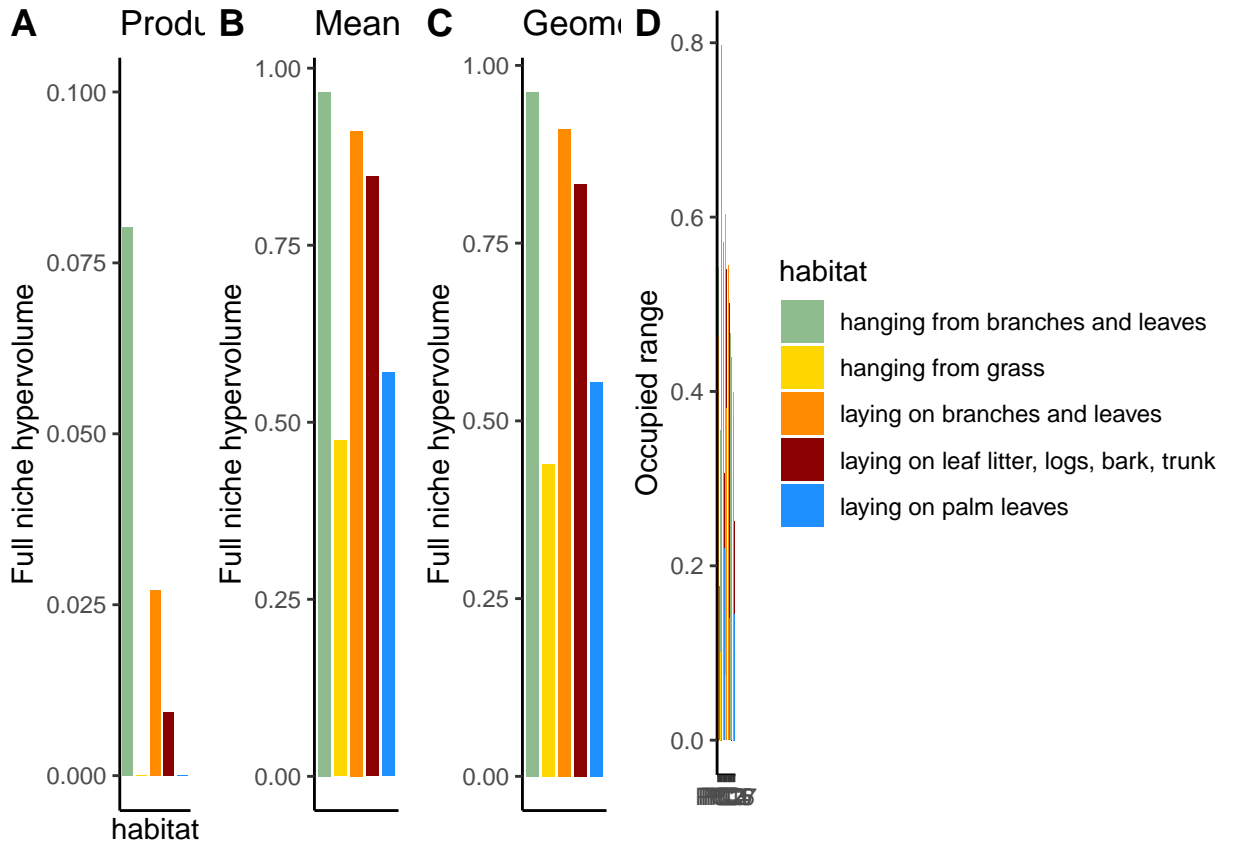
```

Geometric mean

```

plot_grid(p1, p2, p3, p, nrow = 1, rel_widths = (c(1, 1, 1, 3)),
  labels = "AUTO")

```



Plots

Hypervolume size - Clade

Using the DynRB package

```

Xrc <- matrix(nrow = length(X$clade), ncol = 8)
colnames(Xrc) <- c("clade", "PC1", "PC2", "PC3", "PC4", "PC5",
  "PC6", "PC7")
Xrc <- as.data.frame(Xrc)
Xrc$clade <- as.character(X$clade)
Xrc$PC1 <- as.numeric(X$pca_mix_dim1)
Xrc$PC2 <- as.numeric(X$pca_mix_dim2)
Xrc$PC3 <- as.numeric(X$pca_mix_dim3)
Xrc$PC4 <- as.numeric(X$pca_mix_dim4)
Xrc$PC5 <- as.numeric(X$pca_mix_dim5)
Xrc$PC6 <- as.numeric(X$pca_mix_dim6)
Xrc$PC7 <- as.numeric(X$pca_mix_dim7)

# Size and pairwise overlap
rc <- dynRB_VPa(Xrc)

```

```

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```

	V1	V2	port_prod
## 1	African/Malagasy clade	African/Malagasy clade	1.000000e+00
## 2	Agathemeridae	African/Malagasy clade	0.000000e+00
## 3	Aschiphasmatinae	African/Malagasy clade	3.275021e-04
## 4	Clitumninae/Pachymorphinae	African/Malagasy clade	9.763066e-03
## 5	Diapheromerinae	African/Malagasy clade	5.359756e-02
## 6	European clade	African/Malagasy clade	0.000000e+00
## 7	Heteropteryginae	African/Malagasy clade	1.407811e-03
## 8	Lanceocercata	African/Malagasy clade	2.382017e-01
## 9	Lonchodinae	African/Malagasy clade	9.646317e-02
## 10	Necrosiinae	African/Malagasy clade	1.714932e-01
## 11	Palophinae/Cladomorphinae	African/Malagasy clade	1.622733e-02
## 12	Pharnaciini	African/Malagasy clade	2.117387e-05
## 13	Phylliinae	African/Malagasy clade	0.000000e+00
## 14	Pseudophasmatinae	African/Malagasy clade	5.408692e-03
## 15	Stephanacridini	African/Malagasy clade	9.882564e-08
## 16	Timematinae	African/Malagasy clade	0.000000e+00

## 17	African/Malagasy clade	Agathemeridae	0.000000e+00
## 18	Agathemeridae	Agathemeridae	1.000000e+00
## 19	Aschiphasmatinae	Agathemeridae	0.000000e+00
## 20	Clitumninae/Pachymorphinae	Agathemeridae	0.000000e+00
## 21	Diapheromerinae	Agathemeridae	0.000000e+00
## 22	European clade	Agathemeridae	0.000000e+00
## 23	Heteropteryginae	Agathemeridae	0.000000e+00
## 24	Lanceocercata	Agathemeridae	0.000000e+00
## 25	Lonchodinae	Agathemeridae	0.000000e+00
## 26	Necrosiinae	Agathemeridae	0.000000e+00
## 27	Palophinae/Cladomorphinae	Agathemeridae	0.000000e+00
## 28	Pharnaciini	Agathemeridae	0.000000e+00
## 29	Phylliinae	Agathemeridae	0.000000e+00
## 30	Pseudophasmatinae	Agathemeridae	0.000000e+00
## 31	Stephanacridini	Agathemeridae	0.000000e+00
## 32	Timematinae	Agathemeridae	0.000000e+00
## 33	African/Malagasy clade	Aschiphasmatinae	2.056023e-02
## 34	Agathemeridae	Aschiphasmatinae	0.000000e+00
## 35	Aschiphasmatinae	Aschiphasmatinae	1.000000e+00
## 36	Clitumninae/Pachymorphinae	Aschiphasmatinae	0.000000e+00
## 37	Diapheromerinae	Aschiphasmatinae	7.943962e-03
## 38	European clade	Aschiphasmatinae	0.000000e+00
## 39	Heteropteryginae	Aschiphasmatinae	7.914129e-03
## 40	Lanceocercata	Aschiphasmatinae	2.672427e-01
## 41	Lonchodinae	Aschiphasmatinae	5.320893e-03
## 42	Necrosiinae	Aschiphasmatinae	5.734753e-02
## 43	Palophinae/Cladomorphinae	Aschiphasmatinae	1.176894e-03
## 44	Pharnaciini	Aschiphasmatinae	0.000000e+00
## 45	Phylliinae	Aschiphasmatinae	0.000000e+00
## 46	Pseudophasmatinae	Aschiphasmatinae	1.520808e-01
## 47	Stephanacridini	Aschiphasmatinae	0.000000e+00
## 48	Timematinae	Aschiphasmatinae	0.000000e+00
## 49	African/Malagasy clade	Clitumninae/Pachymorphinae	4.412046e-01
## 50	Agathemeridae	Clitumninae/Pachymorphinae	0.000000e+00
## 51	Aschiphasmatinae	Clitumninae/Pachymorphinae	0.000000e+00
## 52	Clitumninae/Pachymorphinae	Clitumninae/Pachymorphinae	1.000000e+00
## 53	Diapheromerinae	Clitumninae/Pachymorphinae	3.765548e-01
## 54	European clade	Clitumninae/Pachymorphinae	2.483243e-05
## 55	Heteropteryginae	Clitumninae/Pachymorphinae	0.000000e+00
## 56	Lanceocercata	Clitumninae/Pachymorphinae	3.184679e-01
## 57	Lonchodinae	Clitumninae/Pachymorphinae	6.143690e-01
## 58	Necrosiinae	Clitumninae/Pachymorphinae	2.081693e-01
## 59	Palophinae/Cladomorphinae	Clitumninae/Pachymorphinae	1.837032e-02
## 60	Pharnaciini	Clitumninae/Pachymorphinae	3.702183e-05
## 61	Phylliinae	Clitumninae/Pachymorphinae	0.000000e+00
## 62	Pseudophasmatinae	Clitumninae/Pachymorphinae	0.000000e+00
## 63	Stephanacridini	Clitumninae/Pachymorphinae	0.000000e+00
## 64	Timematinae	Clitumninae/Pachymorphinae	0.000000e+00
## 65	African/Malagasy clade	Diapheromerinae	2.417317e-01
## 66	Agathemeridae	Diapheromerinae	0.000000e+00
## 67	Aschiphasmatinae	Diapheromerinae	2.754138e-04
## 68	Clitumninae/Pachymorphinae	Diapheromerinae	4.029842e-02
## 69	Diapheromerinae	Diapheromerinae	1.000000e+00
## 70	European clade	Diapheromerinae	1.586802e-05

## 71	Heteropteryginae	Diapheromerinae	2.830717e-03
## 72	Lanceocercata	Diapheromerinae	4.064139e-01
## 73	Lonchodinae	Diapheromerinae	1.361778e-01
## 74	Necrosiinae	Diapheromerinae	1.531475e-01
## 75	Palophinae/Cladomorphinae	Diapheromerinae	3.099556e-02
## 76	Pharnaciini	Diapheromerinae	0.000000e+00
## 77	Phylliinae	Diapheromerinae	0.000000e+00
## 78	Pseudophasmatinae	Diapheromerinae	1.642537e-02
## 79	Stephanacridini	Diapheromerinae	2.676573e-07
## 80	Timematinae	Diapheromerinae	0.000000e+00
## 81	African/Malagasy clade	European clade	0.000000e+00
## 82	Agathemeridae	European clade	0.000000e+00
## 83	Aschiphasmatinae	European clade	0.000000e+00
## 84	Clitumninae/Pachymorphinae	European clade	7.994354e-03
## 85	Diapheromerinae	European clade	7.078366e-02
## 86	European clade	European clade	1.000000e+00
## 87	Heteropteryginae	European clade	0.000000e+00
## 88	Lanceocercata	European clade	0.000000e+00
## 89	Lonchodinae	European clade	0.000000e+00
## 90	Necrosiinae	European clade	0.000000e+00
## 91	Palophinae/Cladomorphinae	European clade	0.000000e+00
## 92	Pharnaciini	European clade	0.000000e+00
## 93	Phylliinae	European clade	0.000000e+00
## 94	Pseudophasmatinae	European clade	0.000000e+00
## 95	Stephanacridini	European clade	0.000000e+00
## 96	Timematinae	European clade	0.000000e+00
## 97	African/Malagasy clade	Heteropteryginae	3.033358e-02
## 98	Agathemeridae	Heteropteryginae	0.000000e+00
## 99	Aschiphasmatinae	Heteropteryginae	1.603868e-03
## 100	Clitumninae/Pachymorphinae	Heteropteryginae	0.000000e+00
## 101	Diapheromerinae	Heteropteryginae	1.478070e-02
## 102	European clade	Heteropteryginae	0.000000e+00
## 103	Heteropteryginae	Heteropteryginae	1.000000e+00
## 104	Lanceocercata	Heteropteryginae	4.422066e-01
## 105	Lonchodinae	Heteropteryginae	1.238987e-01
## 106	Necrosiinae	Heteropteryginae	4.397350e-02
## 107	Palophinae/Cladomorphinae	Heteropteryginae	4.555807e-02
## 108	Pharnaciini	Heteropteryginae	0.000000e+00
## 109	Phylliinae	Heteropteryginae	0.000000e+00
## 110	Pseudophasmatinae	Heteropteryginae	1.541589e-02
## 111	Stephanacridini	Heteropteryginae	0.000000e+00
## 112	Timematinae	Heteropteryginae	0.000000e+00
## 113	African/Malagasy clade	Lanceocercata	4.966182e-02
## 114	Agathemeridae	Lanceocercata	0.000000e+00
## 115	Aschiphasmatinae	Lanceocercata	9.062417e-04
## 116	Clitumninae/Pachymorphinae	Lanceocercata	1.598793e-03
## 117	Diapheromerinae	Lanceocercata	2.034120e-02
## 118	European clade	Lanceocercata	0.000000e+00
## 119	Heteropteryginae	Lanceocercata	5.179991e-03
## 120	Lanceocercata	Lanceocercata	1.000000e+00
## 121	Lonchodinae	Lanceocercata	2.146974e-02
## 122	Necrosiinae	Lanceocercata	1.001263e-01
## 123	Palophinae/Cladomorphinae	Lanceocercata	1.335272e-02
## 124	Pharnaciini	Lanceocercata	2.139655e-06

## 125	Phylliinae	Lanceocercata	0.000000e+00
## 126	Pseudophasmatinae	Lanceocercata	2.560251e-02
## 127	Stephanacridini	Lanceocercata	7.896274e-09
## 128	Timematinae	Lanceocercata	0.000000e+00
## 129	African/Malagasy clade	Lonchodinae	4.241710e-01
## 130	Agathemeridae	Lonchodinae	0.000000e+00
## 131	Aschiphasmatinae	Lonchodinae	2.947715e-04
## 132	Clitumninae/Pachymorphinae	Lonchodinae	6.774484e-02
## 133	Diapheromerinae	Lonchodinae	1.459946e-01
## 134	European clade	Lonchodinae	0.000000e+00
## 135	Heteropteryginae	Lonchodinae	3.223264e-02
## 136	Lanceocercata	Lonchodinae	4.421265e-01
## 137	Lonchodinae	Lonchodinae	1.000000e+00
## 138	Necrosiinae	Lonchodinae	1.917005e-01
## 139	Palophinae/Cladomorphinae	Lonchodinae	6.468927e-02
## 140	Pharnaciini	Lonchodinae	9.699221e-06
## 141	Phylliinae	Lonchodinae	0.000000e+00
## 142	Pseudophasmatinae	Lonchodinae	0.000000e+00
## 143	Stephanacridini	Lonchodinae	0.000000e+00
## 144	Timematinae	Lonchodinae	0.000000e+00
## 145	African/Malagasy clade	Necrosiinae	2.151400e-01
## 146	Agathemeridae	Necrosiinae	0.000000e+00
## 147	Aschiphasmatinae	Necrosiinae	1.319990e-03
## 148	Clitumninae/Pachymorphinae	Necrosiinae	8.310035e-03
## 149	Diapheromerinae	Necrosiinae	6.335222e-02
## 150	European clade	Necrosiinae	0.000000e+00
## 151	Heteropteryginae	Necrosiinae	5.241305e-03
## 152	Lanceocercata	Necrosiinae	6.719758e-01
## 153	Lonchodinae	Necrosiinae	6.712249e-02
## 154	Necrosiinae	Necrosiinae	1.000000e+00
## 155	Palophinae/Cladomorphinae	Necrosiinae	2.926076e-02
## 156	Pharnaciini	Necrosiinae	3.304761e-06
## 157	Phylliinae	Necrosiinae	0.000000e+00
## 158	Pseudophasmatinae	Necrosiinae	1.018411e-01
## 159	Stephanacridini	Necrosiinae	0.000000e+00
## 160	Timematinae	Necrosiinae	0.000000e+00
## 161	African/Malagasy clade	Palophinae/Cladomorphinae	2.537918e-01
## 162	Agathemeridae	Palophinae/Cladomorphinae	0.000000e+00
## 163	Aschiphasmatinae	Palophinae/Cladomorphinae	1.836290e-04
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## 168	Lanceocercata	Palophinae/Cladomorphinae	8.255467e-01
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## 174	Pseudophasmatinae	Palophinae/Cladomorphinae	1.749771e-02
## 175	Stephanacridini	Palophinae/Cladomorphinae	5.311823e-08
## 176	Timematinae	Palophinae/Cladomorphinae	0.000000e+00
## 177	African/Malagasy clade	Pharnaciini	3.691901e-01
## 178	Agathemeridae	Pharnaciini	0.000000e+00

## 179	Aschiphasmatinae	Pharnaciini	0.000000e+00
## 180	Clitumninae/Pachymorphinae	Pharnaciini	1.634723e-02
## 181	Diapheromerinae	Pharnaciini	0.000000e+00
## 182	European clade	Pharnaciini	0.000000e+00
## 183	Heteropteryginae	Pharnaciini	0.000000e+00
## 184	Lanceocercata	Pharnaciini	1.695196e-01
## 185	Lonchodinae	Pharnaciini	3.869827e-02
## 186	Necroschiinae	Pharnaciini	3.150878e-02
## 187	Palophinae/Cladomorphinae	Pharnaciini	0.000000e+00
## 188	Pharnaciini	Pharnaciini	1.000000e+00
## 189	Phylliinae	Pharnaciini	0.000000e+00
## 190	Pseudophasmatinae	Pharnaciini	0.000000e+00
## 191	Stephanacridini	Pharnaciini	0.000000e+00
## 192	Timematinae	Pharnaciini	0.000000e+00
## 193	African/Malagasy clade	Phylliinae	0.000000e+00
## 194	Agathemeridae	Phylliinae	0.000000e+00
## 195	Aschiphasmatinae	Phylliinae	0.000000e+00
## 196	Clitumninae/Pachymorphinae	Phylliinae	0.000000e+00
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## 198	European clade	Phylliinae	0.000000e+00
## 199	Heteropteryginae	Phylliinae	0.000000e+00
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## 201	Lonchodinae	Phylliinae	0.000000e+00
## 202	Necroschiinae	Phylliinae	0.000000e+00
## 203	Palophinae/Cladomorphinae	Phylliinae	0.000000e+00
## 204	Pharnaciini	Phylliinae	0.000000e+00
## 205	Phylliinae	Phylliinae	1.000000e+00
## 206	Pseudophasmatinae	Phylliinae	0.000000e+00
## 207	Stephanacridini	Phylliinae	0.000000e+00
## 208	Timematinae	Phylliinae	0.000000e+00
## 209	African/Malagasy clade	Pseudophasmatinae	2.004219e-02
## 210	Agathemeridae	Pseudophasmatinae	0.000000e+00
## 211	Aschiphasmatinae	Pseudophasmatinae	6.938575e-03
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## 213	Diapheromerinae	Pseudophasmatinae	1.528486e-02
## 214	European clade	Pseudophasmatinae	0.000000e+00
## 215	Heteropteryginae	Pseudophasmatinae	3.183924e-03
## 216	Lanceocercata	Pseudophasmatinae	3.404328e-01
## 217	Lonchodinae	Pseudophasmatinae	0.000000e+00
## 218	Necroschiinae	Pseudophasmatinae	1.842550e-01
## 219	Palophinae/Cladomorphinae	Pseudophasmatinae	4.601739e-03
## 220	Pharnaciini	Pseudophasmatinae	0.000000e+00
## 221	Phylliinae	Pseudophasmatinae	0.000000e+00
## 222	Pseudophasmatinae	Pseudophasmatinae	1.000000e+00
## 223	Stephanacridini	Pseudophasmatinae	0.000000e+00
## 224	Timematinae	Pseudophasmatinae	0.000000e+00
## 225	African/Malagasy clade	Stephanacridini	3.811380e-01
## 226	Agathemeridae	Stephanacridini	0.000000e+00
## 227	Aschiphasmatinae	Stephanacridini	0.000000e+00
## 228	Clitumninae/Pachymorphinae	Stephanacridini	0.000000e+00
## 229	Diapheromerinae	Stephanacridini	2.217445e-01
## 230	European clade	Stephanacridini	0.000000e+00
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## 232	Lanceocercata	Stephanacridini	2.505661e-01

## 233	Lonchodinae	Stephanacridini	0.000000e+00
## 234	Necrosciinae	Stephanacridini	0.000000e+00
## 235	Palophinae/Cladomorphinae	Stephanacridini	1.971288e-02
## 236	Pharnaciini	Stephanacridini	0.000000e+00
## 237	Phylliinae	Stephanacridini	0.000000e+00
## 238	Pseudophasmatinae	Stephanacridini	0.000000e+00
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## 240	Timematinae	Stephanacridini	0.000000e+00
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## 242	Agathemeridae	Timematinae	0.000000e+00
## 243	Aschiphasmatinae	Timematinae	0.000000e+00
## 244	Clitumninae/Pachymorphinae	Timematinae	0.000000e+00
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## 248	Lanceocercata	Timematinae	0.000000e+00
## 249	Lonchodinae	Timematinae	0.000000e+00
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## 251	Palophinae/Cladomorphinae	Timematinae	0.000000e+00
## 252	Pharnaciini	Timematinae	0.000000e+00
## 253	Phylliinae	Timematinae	0.000000e+00
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## 255	Stephanacridini	Timematinae	0.000000e+00
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## 6	7.722328e-07	8.953343e-03	
## 7	4.382212e-04	8.953343e-03	
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226 4.963308e-08 2.572885e-09
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## 246 7.722328e-07 2.861688e-09
## 247 4.382212e-04 2.861688e-09
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## 249 1.688916e-03 2.861688e-09
## 250 4.767779e-03 2.861688e-09
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## 253 3.673400e-05 2.861688e-09
## 254 2.642797e-03 2.861688e-09
## 255 2.572885e-09 2.861688e-09
## 256 2.861688e-09 2.861688e-09

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rc\$result

##	V1	V2	port_prod
## 1	African/Malagasy clade	African/Malagasy clade	1.000000e+00
## 2	Agathemeridae	African/Malagasy clade	0.000000e+00
## 3	Aschiphasmatinae	African/Malagasy clade	3.275021e-04
## 4	Clitumninae/Pachymorphinae	African/Malagasy clade	9.763066e-03
## 5	Diapheromerinae	African/Malagasy clade	5.359756e-02
## 6	European clade	African/Malagasy clade	0.000000e+00
## 7	Heteropteryginae	African/Malagasy clade	1.407811e-03
## 8	Lanceocercata	African/Malagasy clade	2.382017e-01
## 9	Lonchodinae	African/Malagasy clade	9.646317e-02
## 10	Necrosiinae	African/Malagasy clade	1.714932e-01
## 11	Palophinae/Cladomorphinae	African/Malagasy clade	1.622733e-02
## 12	Pharnaciini	African/Malagasy clade	2.117387e-05
## 13	Phylliinae	African/Malagasy clade	0.000000e+00
## 14	Pseudophasmatinae	African/Malagasy clade	5.408692e-03
## 15	Stephanacridini	African/Malagasy clade	9.882564e-08
## 16	Timematinae	African/Malagasy clade	0.000000e+00
## 17	African/Malagasy clade	Agathemeridae	0.000000e+00
## 18	Agathemeridae	Agathemeridae	1.000000e+00
## 19	Aschiphasmatinae	Agathemeridae	0.000000e+00
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## 21	Diapheromerinae	Agathemeridae	0.000000e+00
## 22	European clade	Agathemeridae	0.000000e+00
## 23	Heteropteryginae	Agathemeridae	0.000000e+00
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## 25	Lonchodinae	Agathemeridae	0.000000e+00
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## 27	Palophinae/Cladomorphinae	Agathemeridae	0.000000e+00
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## 29	Phylliinae	Agathemeridae	0.000000e+00
## 30	Pseudophasmatinae	Agathemeridae	0.000000e+00
## 31	Stephanacridini	Agathemeridae	0.000000e+00
## 32	Timematinae	Agathemeridae	0.000000e+00
## 33	African/Malagasy clade	Aschiphasmatinae	2.056023e-02
## 34	Agathemeridae	Aschiphasmatinae	0.000000e+00
## 35	Aschiphasmatinae	Aschiphasmatinae	1.000000e+00
## 36	Clitumninae/Pachymorphinae	Aschiphasmatinae	0.000000e+00
## 37	Diapheromerinae	Aschiphasmatinae	7.943962e-03
## 38	European clade	Aschiphasmatinae	0.000000e+00
## 39	Heteropteryginae	Aschiphasmatinae	7.914129e-03

## 40	Lanceocercata	Aschiphasmatinae	2.672427e-01
## 41	Lonchodinae	Aschiphasmatinae	5.320893e-03
## 42	Necrosiinae	Aschiphasmatinae	5.734753e-02
## 43	Palophinae/Cladomorphinae	Aschiphasmatinae	1.176894e-03
## 44	Pharnaciini	Aschiphasmatinae	0.000000e+00
## 45	Phylliinae	Aschiphasmatinae	0.000000e+00
## 46	Pseudophasmatinae	Aschiphasmatinae	1.520808e-01
## 47	Stephanacridini	Aschiphasmatinae	0.000000e+00
## 48	Timematinae	Aschiphasmatinae	0.000000e+00
## 49	African/Malagasy clade	Clitumninae/Pachymorphinae	4.412046e-01
## 50	Agathemeridae	Clitumninae/Pachymorphinae	0.000000e+00
## 51	Aschiphasmatinae	Clitumninae/Pachymorphinae	0.000000e+00
## 52	Clitumninae/Pachymorphinae	Clitumninae/Pachymorphinae	1.000000e+00
## 53	Diapheromerinae	Clitumninae/Pachymorphinae	3.765548e-01
## 54	European clade	Clitumninae/Pachymorphinae	2.483243e-05
## 55	Heteropteryginae	Clitumninae/Pachymorphinae	0.000000e+00
## 56	Lanceocercata	Clitumninae/Pachymorphinae	3.184679e-01
## 57	Lonchodinae	Clitumninae/Pachymorphinae	6.143690e-01
## 58	Necrosiinae	Clitumninae/Pachymorphinae	2.081693e-01
## 59	Palophinae/Cladomorphinae	Clitumninae/Pachymorphinae	1.837032e-02
## 60	Pharnaciini	Clitumninae/Pachymorphinae	3.702183e-05
## 61	Phylliinae	Clitumninae/Pachymorphinae	0.000000e+00
## 62	Pseudophasmatinae	Clitumninae/Pachymorphinae	0.000000e+00
## 63	Stephanacridini	Clitumninae/Pachymorphinae	0.000000e+00
## 64	Timematinae	Clitumninae/Pachymorphinae	0.000000e+00
## 65	African/Malagasy clade	Diapheromerinae	2.417317e-01
## 66	Agathemeridae	Diapheromerinae	0.000000e+00
## 67	Aschiphasmatinae	Diapheromerinae	2.754138e-04
## 68	Clitumninae/Pachymorphinae	Diapheromerinae	4.029842e-02
## 69	Diapheromerinae	Diapheromerinae	1.000000e+00
## 70	European clade	Diapheromerinae	1.586802e-05
## 71	Heteropteryginae	Diapheromerinae	2.830717e-03
## 72	Lanceocercata	Diapheromerinae	4.064139e-01
## 73	Lonchodinae	Diapheromerinae	1.361778e-01
## 74	Necrosiinae	Diapheromerinae	1.531475e-01
## 75	Palophinae/Cladomorphinae	Diapheromerinae	3.099556e-02
## 76	Pharnaciini	Diapheromerinae	0.000000e+00
## 77	Phylliinae	Diapheromerinae	0.000000e+00
## 78	Pseudophasmatinae	Diapheromerinae	1.642537e-02
## 79	Stephanacridini	Diapheromerinae	2.676573e-07
## 80	Timematinae	Diapheromerinae	0.000000e+00
## 81	African/Malagasy clade	European clade	0.000000e+00
## 82	Agathemeridae	European clade	0.000000e+00
## 83	Aschiphasmatinae	European clade	0.000000e+00
## 84	Clitumninae/Pachymorphinae	European clade	7.994354e-03
## 85	Diapheromerinae	European clade	7.078366e-02
## 86	European clade	European clade	1.000000e+00
## 87	Heteropteryginae	European clade	0.000000e+00
## 88	Lanceocercata	European clade	0.000000e+00
## 89	Lonchodinae	European clade	0.000000e+00
## 90	Necrosiinae	European clade	0.000000e+00
## 91	Palophinae/Cladomorphinae	European clade	0.000000e+00
## 92	Pharnaciini	European clade	0.000000e+00
## 93	Phylliinae	European clade	0.000000e+00

## 94	Pseudophasmatinae	European clade	0.000000e+00
## 95	Stephanacridini	European clade	0.000000e+00
## 96	Timematinae	European clade	0.000000e+00
## 97	African/Malagasy clade	Heteropteryginae	3.033358e-02
## 98	Agathemeridae	Heteropteryginae	0.000000e+00
## 99	Aschiphasmatinae	Heteropteryginae	1.603868e-03
## 100	Clitumninae/Pachymorphinae	Heteropteryginae	0.000000e+00
## 101	Diapheromerinae	Heteropteryginae	1.478070e-02
## 102	European clade	Heteropteryginae	0.000000e+00
## 103	Heteropteryginae	Heteropteryginae	1.000000e+00
## 104	Lanceocercata	Heteropteryginae	4.422066e-01
## 105	Lonchodinae	Heteropteryginae	1.238987e-01
## 106	Necrosiinae	Heteropteryginae	4.397350e-02
## 107	Palophinae/Cladomorphinae	Heteropteryginae	4.555807e-02
## 108	Pharnaciini	Heteropteryginae	0.000000e+00
## 109	Phylliinae	Heteropteryginae	0.000000e+00
## 110	Pseudophasmatinae	Heteropteryginae	1.541589e-02
## 111	Stephanacridini	Heteropteryginae	0.000000e+00
## 112	Timematinae	Heteropteryginae	0.000000e+00
## 113	African/Malagasy clade	Lanceocercata	4.966182e-02
## 114	Agathemeridae	Lanceocercata	0.000000e+00
## 115	Aschiphasmatinae	Lanceocercata	9.062417e-04
## 116	Clitumninae/Pachymorphinae	Lanceocercata	1.598793e-03
## 117	Diapheromerinae	Lanceocercata	2.034120e-02
## 118	European clade	Lanceocercata	0.000000e+00
## 119	Heteropteryginae	Lanceocercata	5.179991e-03
## 120	Lanceocercata	Lanceocercata	1.000000e+00
## 121	Lonchodinae	Lanceocercata	2.146974e-02
## 122	Necrosiinae	Lanceocercata	1.001263e-01
## 123	Palophinae/Cladomorphinae	Lanceocercata	1.335272e-02
## 124	Pharnaciini	Lanceocercata	2.139655e-06
## 125	Phylliinae	Lanceocercata	0.000000e+00
## 126	Pseudophasmatinae	Lanceocercata	2.560251e-02
## 127	Stephanacridini	Lanceocercata	7.896274e-09
## 128	Timematinae	Lanceocercata	0.000000e+00
## 129	African/Malagasy clade	Lonchodinae	4.241710e-01
## 130	Agathemeridae	Lonchodinae	0.000000e+00
## 131	Aschiphasmatinae	Lonchodinae	2.947715e-04
## 132	Clitumninae/Pachymorphinae	Lonchodinae	6.774484e-02
## 133	Diapheromerinae	Lonchodinae	1.459946e-01
## 134	European clade	Lonchodinae	0.000000e+00
## 135	Heteropteryginae	Lonchodinae	3.223264e-02
## 136	Lanceocercata	Lonchodinae	4.421265e-01
## 137	Lonchodinae	Lonchodinae	1.000000e+00
## 138	Necrosiinae	Lonchodinae	1.917005e-01
## 139	Palophinae/Cladomorphinae	Lonchodinae	6.468927e-02
## 140	Pharnaciini	Lonchodinae	9.699221e-06
## 141	Phylliinae	Lonchodinae	0.000000e+00
## 142	Pseudophasmatinae	Lonchodinae	0.000000e+00
## 143	Stephanacridini	Lonchodinae	0.000000e+00
## 144	Timematinae	Lonchodinae	0.000000e+00
## 145	African/Malagasy clade	Necrosiinae	2.151400e-01
## 146	Agathemeridae	Necrosiinae	0.000000e+00
## 147	Aschiphasmatinae	Necrosiinae	1.319990e-03

## 148	Clitumninae/Pachymorphinae	Necrosciinae	8.310035e-03
## 149	Diapheromerinae	Necrosciinae	6.335222e-02
## 150	European clade	Necrosciinae	0.000000e+00
## 151	Heteropteryginae	Necrosciinae	5.241305e-03
## 152	Lanceocercata	Necrosciinae	6.719758e-01
## 153	Lonchodinae	Necrosciinae	6.712249e-02
## 154	Necrosciinae	Necrosciinae	1.000000e+00
## 155	Palophinae/Cladomorphinae	Necrosciinae	2.926076e-02
## 156	Pharnaciini	Necrosciinae	3.304761e-06
## 157	Phylliinae	Necrosciinae	0.000000e+00
## 158	Pseudophasmatinae	Necrosciinae	1.018411e-01
## 159	Stephanacridini	Necrosciinae	0.000000e+00
## 160	Timematinae	Necrosciinae	0.000000e+00
## 161	African/Malagasy clade	Palophinae/Cladomorphinae	2.537918e-01
## 162	Agathemeridae	Palophinae/Cladomorphinae	0.000000e+00
## 163	Aschiphasmatinae	Palophinae/Cladomorphinae	1.836290e-04
## 164	Clitumninae/Pachymorphinae	Palophinae/Cladomorphinae	5.747381e-03
## 165	Diapheromerinae	Palophinae/Cladomorphinae	1.178798e-01
## 166	European clade	Palophinae/Cladomorphinae	0.000000e+00
## 167	Heteropteryginae	Palophinae/Cladomorphinae	3.560793e-02
## 168	Lanceocercata	Palophinae/Cladomorphinae	8.255467e-01
## 169	Lonchodinae	Palophinae/Cladomorphinae	2.064705e-01
## 170	Necrosciinae	Palophinae/Cladomorphinae	2.443124e-01
## 171	Palophinae/Cladomorphinae	Palophinae/Cladomorphinae	1.000000e+00
## 172	Pharnaciini	Palophinae/Cladomorphinae	0.000000e+00
## 173	Phylliinae	Palophinae/Cladomorphinae	0.000000e+00
## 174	Pseudophasmatinae	Palophinae/Cladomorphinae	1.749771e-02
## 175	Stephanacridini	Palophinae/Cladomorphinae	5.311823e-08
## 176	Timematinae	Palophinae/Cladomorphinae	0.000000e+00
## 177	African/Malagasy clade	Pharnaciini	3.691901e-01
## 178	Agathemeridae	Pharnaciini	0.000000e+00
## 179	Aschiphasmatinae	Pharnaciini	0.000000e+00
## 180	Clitumninae/Pachymorphinae	Pharnaciini	1.634723e-02
## 181	Diapheromerinae	Pharnaciini	0.000000e+00
## 182	European clade	Pharnaciini	0.000000e+00
## 183	Heteropteryginae	Pharnaciini	0.000000e+00
## 184	Lanceocercata	Pharnaciini	1.695196e-01
## 185	Lonchodinae	Pharnaciini	3.869827e-02
## 186	Necrosciinae	Pharnaciini	3.150878e-02
## 187	Palophinae/Cladomorphinae	Pharnaciini	0.000000e+00
## 188	Pharnaciini	Pharnaciini	1.000000e+00
## 189	Phylliinae	Pharnaciini	0.000000e+00
## 190	Pseudophasmatinae	Pharnaciini	0.000000e+00
## 191	Stephanacridini	Pharnaciini	0.000000e+00
## 192	Timematinae	Pharnaciini	0.000000e+00
## 193	African/Malagasy clade	Phylliinae	0.000000e+00
## 194	Agathemeridae	Phylliinae	0.000000e+00
## 195	Aschiphasmatinae	Phylliinae	0.000000e+00
## 196	Clitumninae/Pachymorphinae	Phylliinae	0.000000e+00
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## 198	European clade	Phylliinae	0.000000e+00
## 199	Heteropteryginae	Phylliinae	0.000000e+00
## 200	Lanceocercata	Phylliinae	0.000000e+00
## 201	Lonchodinae	Phylliinae	0.000000e+00

## 202	Necrosciinae	Phylliinae	0.000000e+00
## 203	Palophinae/Cladomorphinae	Phylliinae	0.000000e+00
## 204	Pharnaciini	Phylliinae	0.000000e+00
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## 207	Stephanacridini	Phylliinae	0.000000e+00
## 208	Timematinae	Phylliinae	0.000000e+00
## 209	African/Malagasy clade	Pseudophasmatinae	2.004219e-02
## 210	Agathemeridae	Pseudophasmatinae	0.000000e+00
## 211	Aschiphasmatinae	Pseudophasmatinae	6.938575e-03
## 212	Clitumninae/Pachymorphinae	Pseudophasmatinae	0.000000e+00
## 213	Diapheromerinae	Pseudophasmatinae	1.528486e-02
## 214	European clade	Pseudophasmatinae	0.000000e+00
## 215	Heteropteryginae	Pseudophasmatinae	3.183924e-03
## 216	Lanceocercata	Pseudophasmatinae	3.404328e-01
## 217	Lonchodinae	Pseudophasmatinae	0.000000e+00
## 218	Necrosciinae	Pseudophasmatinae	1.842550e-01
## 219	Palophinae/Cladomorphinae	Pseudophasmatinae	4.601739e-03
## 220	Pharnaciini	Pseudophasmatinae	0.000000e+00
## 221	Phylliinae	Pseudophasmatinae	0.000000e+00
## 222	Pseudophasmatinae	Pseudophasmatinae	1.000000e+00
## 223	Stephanacridini	Pseudophasmatinae	0.000000e+00
## 224	Timematinae	Pseudophasmatinae	0.000000e+00
## 225	African/Malagasy clade	Stephanacridini	3.811380e-01
## 226	Agathemeridae	Stephanacridini	0.000000e+00
## 227	Aschiphasmatinae	Stephanacridini	0.000000e+00
## 228	Clitumninae/Pachymorphinae	Stephanacridini	0.000000e+00
## 229	Diapheromerinae	Stephanacridini	2.217445e-01
## 230	European clade	Stephanacridini	0.000000e+00
## 231	Heteropteryginae	Stephanacridini	0.000000e+00
## 232	Lanceocercata	Stephanacridini	2.505661e-01
## 233	Lonchodinae	Stephanacridini	0.000000e+00
## 234	Necrosciinae	Stephanacridini	0.000000e+00
## 235	Palophinae/Cladomorphinae	Stephanacridini	1.971288e-02
## 236	Pharnaciini	Stephanacridini	0.000000e+00
## 237	Phylliinae	Stephanacridini	0.000000e+00
## 238	Pseudophasmatinae	Stephanacridini	0.000000e+00
## 239	Stephanacridini	Stephanacridini	1.000000e+00
## 240	Timematinae	Stephanacridini	0.000000e+00
## 241	African/Malagasy clade	Timematinae	0.000000e+00
## 242	Agathemeridae	Timematinae	0.000000e+00
## 243	Aschiphasmatinae	Timematinae	0.000000e+00
## 244	Clitumninae/Pachymorphinae	Timematinae	0.000000e+00
## 245	Diapheromerinae	Timematinae	0.000000e+00
## 246	European clade	Timematinae	0.000000e+00
## 247	Heteropteryginae	Timematinae	0.000000e+00
## 248	Lanceocercata	Timematinae	0.000000e+00
## 249	Lonchodinae	Timematinae	0.000000e+00
## 250	Necrosciinae	Timematinae	0.000000e+00
## 251	Palophinae/Cladomorphinae	Timematinae	0.000000e+00
## 252	Pharnaciini	Timematinae	0.000000e+00
## 253	Phylliinae	Timematinae	0.000000e+00
## 254	Pseudophasmatinae	Timematinae	0.000000e+00
## 255	Stephanacridini	Timematinae	0.000000e+00

##	256	Timematinae			Timematinae 1.000000e+00		
##		port_mean	port_gmean	vol_V1_prod	vol_V1_mean	vol_V1_gmean	vol_V2_prod
##	1	1.00000000	1.00000000	8.953343e-03	0.8719701	0.8489090	8.953343e-03
##	2	0.09565504	0.00000000	4.963308e-08	0.2997245	0.2647680	8.953343e-03
##	3	0.35826539	0.29373149	1.227698e-04	0.6074820	0.5936156	8.953343e-03
##	4	0.51522336	0.50529197	1.850251e-04	0.6339217	0.6076417	8.953343e-03
##	5	0.65173472	0.63371262	2.005731e-03	0.7496253	0.7451235	8.953343e-03
##	6	0.19619718	0.00000000	7.722328e-07	0.4082043	0.3566448	8.953343e-03
##	7	0.52079523	0.28140411	4.382212e-04	0.6769434	0.6571513	8.953343e-03
##	8	0.81067593	0.80604853	4.127894e-02	0.9290058	0.9252524	8.953343e-03
##	9	0.71797742	0.71219998	1.688916e-03	0.7756071	0.7427276	8.953343e-03
##	10	0.76316052	0.76584858	4.767779e-03	0.8081557	0.8007580	8.953343e-03
##	11	0.59318272	0.42573781	5.551642e-04	0.6796223	0.6760241	8.953343e-03
##	12	0.24275438	0.15976768	8.530158e-07	0.4297822	0.3608378	8.953343e-03
##	13	0.18061747	0.00000000	3.673400e-05	0.5327732	0.5279986	8.953343e-03
##	14	0.57368225	0.41393725	2.642797e-03	0.7775782	0.7686443	8.953343e-03
##	15	0.14893707	0.07383753	2.572885e-09	0.2891650	0.1879559	8.953343e-03
##	16	0.04217356	0.00000000	2.861688e-09	0.2030123	0.1903365	8.953343e-03
##	17	0.42399979	0.00000000	8.953343e-03	0.8719701	0.8489090	4.963308e-08
##	18	1.00000000	1.00000000	4.963308e-08	0.2997245	0.2647680	4.963308e-08
##	19	0.78302864	0.00000000	1.227698e-04	0.6074820	0.5936156	4.963308e-08
##	20	0.42312095	0.00000000	1.850251e-04	0.6339217	0.6076417	4.963308e-08
##	21	0.27793721	0.00000000	2.005731e-03	0.7496253	0.7451235	4.963308e-08
##	22	0.13231575	0.00000000	7.722328e-07	0.4082043	0.3566448	4.963308e-08
##	23	0.34425538	0.00000000	4.382212e-04	0.6769434	0.6571513	4.963308e-08
##	24	0.64330618	0.00000000	4.127894e-02	0.9290058	0.9252524	4.963308e-08
##	25	0.38503379	0.00000000	1.688916e-03	0.7756071	0.7427276	4.963308e-08
##	26	0.57353826	0.00000000	4.767779e-03	0.8081557	0.8007580	4.963308e-08
##	27	0.28344468	0.00000000	5.551642e-04	0.6796223	0.6760241	4.963308e-08
##	28	0.26983037	0.00000000	8.530158e-07	0.4297822	0.3608378	4.963308e-08
##	29	0.44372166	0.00000000	3.673400e-05	0.5327732	0.5279986	4.963308e-08
##	30	0.60839403	0.00000000	2.642797e-03	0.7775782	0.7686443	4.963308e-08
##	31	0.34470745	0.00000000	2.572885e-09	0.2891650	0.1879559	4.963308e-08
##	32	0.25516235	0.00000000	2.861688e-09	0.2030123	0.1903365	4.963308e-08
##	33	0.61457914	0.51790487	8.953343e-03	0.8719701	0.8489090	1.227698e-04
##	34	0.28448978	0.00000000	4.963308e-08	0.2997245	0.2647680	1.227698e-04
##	35	1.00000000	1.00000000	1.227698e-04	0.6074820	0.5936156	1.227698e-04
##	36	0.53490405	0.00000000	1.850251e-04	0.6339217	0.6076417	1.227698e-04
##	37	0.53547345	0.25987854	2.005731e-03	0.7496253	0.7451235	1.227698e-04
##	38	0.18869901	0.00000000	7.722328e-07	0.4082043	0.3566448	1.227698e-04
##	39	0.54524989	0.37180675	4.382212e-04	0.6769434	0.6571513	1.227698e-04
##	40	0.79499891	0.72820519	4.127894e-02	0.9290058	0.9252524	1.227698e-04
##	41	0.56988908	0.31858374	1.688916e-03	0.7756071	0.7427276	1.227698e-04
##	42	0.71987038	0.51415133	4.767779e-03	0.8081557	0.8007580	1.227698e-04
##	43	0.35697675	0.25558347	5.551642e-04	0.6796223	0.6760241	1.227698e-04
##	44	0.20904125	0.00000000	8.530158e-07	0.4297822	0.3608378	1.227698e-04
##	45	0.34654904	0.00000000	3.673400e-05	0.5327732	0.5279986	1.227698e-04
##	46	0.79365000	0.70726576	2.642797e-03	0.7775782	0.7686443	1.227698e-04
##	47	0.17081942	0.00000000	2.572885e-09	0.2891650	0.1879559	1.227698e-04
##	48	0.16458961	0.00000000	2.861688e-09	0.2030123	0.1903365	1.227698e-04
##	49	0.86452879	0.84546589	8.953343e-03	0.8719701	0.8489090	1.850251e-04
##	50	0.14730561	0.00000000	4.963308e-08	0.2997245	0.2647680	1.850251e-04
##	51	0.51920183	0.00000000	1.227698e-04	0.6074820	0.5936156	1.850251e-04
##	52	1.00000000	1.00000000	1.850251e-04	0.6339217	0.6076417	1.850251e-04

## 53	0.83173780	0.84152538	2.005731e-03	0.7496253	0.7451235	1.850251e-04
## 54	0.31174604	0.12664835	7.722328e-07	0.4082043	0.3566448	1.850251e-04
## 55	0.66789358	0.00000000	4.382212e-04	0.6769434	0.6571513	1.850251e-04
## 56	0.82117178	0.79779736	4.127894e-02	0.9290058	0.9252524	1.850251e-04
## 57	0.93230950	0.92801576	1.688916e-03	0.7756071	0.7427276	1.850251e-04
## 58	0.80496576	0.78156374	4.767779e-03	0.8081557	0.8007580	1.850251e-04
## 59	0.60015923	0.36301853	5.551642e-04	0.6796223	0.6760241	1.850251e-04
## 60	0.33459139	0.14942114	8.530158e-07	0.4297822	0.3608378	1.850251e-04
## 61	0.36526830	0.00000000	3.673400e-05	0.5327732	0.5279986	1.850251e-04
## 62	0.73846672	0.00000000	2.642797e-03	0.7775782	0.7686443	1.850251e-04
## 63	0.16606841	0.00000000	2.572885e-09	0.2891650	0.1879559	1.850251e-04
## 64	0.10101288	0.00000000	2.861688e-09	0.2030123	0.1903365	1.850251e-04
## 65	0.80574493	0.78672774	8.953343e-03	0.8719701	0.8489090	2.005731e-03
## 66	0.07241822	0.00000000	4.963308e-08	0.2997245	0.2647680	2.005731e-03
## 67	0.38647235	0.16416728	1.227698e-04	0.6074820	0.5936156	2.005731e-03
## 68	0.64356672	0.62675104	1.850251e-04	0.6339217	0.6076417	2.005731e-03
## 69	1.00000000	1.00000000	2.005731e-03	0.7496253	0.7451235	2.005731e-03
## 70	0.33087842	0.10794184	7.722328e-07	0.4082043	0.3566448	2.005731e-03
## 71	0.60376020	0.18343600	4.382212e-04	0.6769434	0.6571513	2.005731e-03
## 72	0.87818287	0.85852629	4.127894e-02	0.9290058	0.9252524	2.005731e-03
## 73	0.77673786	0.74298898	1.688916e-03	0.7756071	0.7427276	2.005731e-03
## 74	0.76535217	0.75820181	4.767779e-03	0.8081557	0.8007580	2.005731e-03
## 75	0.64951188	0.49216837	5.551642e-04	0.6796223	0.6760241	2.005731e-03
## 76	0.31702570	0.00000000	8.530158e-07	0.4297822	0.3608378	2.005731e-03
## 77	0.27722012	0.00000000	3.673400e-05	0.5327732	0.5279986	2.005731e-03
## 78	0.66514112	0.50319810	2.642797e-03	0.7775782	0.7686443	2.005731e-03
## 79	0.18573180	0.07943700	2.572885e-09	0.2891650	0.1879559	2.005731e-03
## 80	0.04297645	0.00000000	2.861688e-09	0.2030123	0.1903365	2.005731e-03
## 81	0.64533422	0.00000000	8.953343e-03	0.8719701	0.8489090	7.722328e-07
## 82	0.08497548	0.00000000	4.963308e-08	0.2997245	0.2647680	7.722328e-07
## 83	0.36515546	0.00000000	1.227698e-04	0.6074820	0.5936156	7.722328e-07
## 84	0.59768771	0.28552542	1.850251e-04	0.6339217	0.6076417	7.722328e-07
## 85	0.78883298	0.35240351	2.005731e-03	0.7496253	0.7451235	7.722328e-07
## 86	1.00000000	1.00000000	7.722328e-07	0.4082043	0.3566448	7.722328e-07
## 87	0.76497352	0.00000000	4.382212e-04	0.6769434	0.6571513	7.722328e-07
## 88	0.88919034	0.00000000	4.127894e-02	0.9290058	0.9252524	7.722328e-07
## 89	0.68770644	0.00000000	1.688916e-03	0.7756071	0.7427276	7.722328e-07
## 90	0.66178012	0.00000000	4.767779e-03	0.8081557	0.8007580	7.722328e-07
## 91	0.79954453	0.00000000	5.551642e-04	0.6796223	0.6760241	7.722328e-07
## 92	0.22012136	0.00000000	8.530158e-07	0.4297822	0.3608378	7.722328e-07
## 93	0.43010285	0.00000000	3.673400e-05	0.5327732	0.5279986	7.722328e-07
## 94	0.55566232	0.00000000	2.642797e-03	0.7775782	0.7686443	7.722328e-07
## 95	0.19142019	0.00000000	2.572885e-09	0.2891650	0.1879559	7.722328e-07
## 96	0.28419246	0.00000000	2.861688e-09	0.2030123	0.1903365	7.722328e-07
## 97	0.74035865	0.43468585	8.953343e-03	0.8719701	0.8489090	4.382212e-04
## 98	0.09565192	0.00000000	4.963308e-08	0.2997245	0.2647680	4.382212e-04
## 99	0.45321521	0.30725445	1.227698e-04	0.6074820	0.5936156	4.382212e-04
## 100	0.57340016	0.00000000	1.850251e-04	0.6339217	0.6076417	4.382212e-04
## 101	0.70439480	0.23148670	2.005731e-03	0.7496253	0.7451235	4.382212e-04
## 102	0.37022465	0.00000000	7.722328e-07	0.4082043	0.3566448	4.382212e-04
## 103	1.00000000	1.00000000	4.382212e-04	0.6769434	0.6571513	4.382212e-04
## 104	0.89050540	0.78563198	4.127894e-02	0.9290058	0.9252524	4.382212e-04
## 105	0.74460498	0.62598685	1.688916e-03	0.7756071	0.7427276	4.382212e-04
## 106	0.75326830	0.40963520	4.767779e-03	0.8081557	0.8007580	4.382212e-04

## 107	0.71836109	0.47172529	5.551642e-04	0.6796223	0.6760241	4.382212e-04
## 108	0.21205214	0.00000000	8.530158e-07	0.4297822	0.3608378	4.382212e-04
## 109	0.28073423	0.00000000	3.673400e-05	0.5327732	0.5279986	4.382212e-04
## 110	0.56381171	0.49039377	2.642797e-03	0.7775782	0.7686443	4.382212e-04
## 111	0.14678988	0.00000000	2.572885e-09	0.2891650	0.1879559	4.382212e-04
## 112	0.10021431	0.00000000	2.861688e-09	0.2030123	0.1903365	4.382212e-04
## 113	0.68749707	0.64313083	8.953343e-03	0.8719701	0.8489090	4.127894e-02
## 114	0.10684879	0.00000000	4.963308e-08	0.2997245	0.2647680	4.127894e-02
## 115	0.37724081	0.32751765	1.227698e-04	0.6074820	0.5936156	4.127894e-02
## 116	0.40372926	0.37978880	1.850251e-04	0.6339217	0.6076417	4.127894e-02
## 117	0.57190394	0.55274661	2.005731e-03	0.7496253	0.7451235	4.127894e-02
## 118	0.22522506	0.00000000	7.722328e-07	0.4082043	0.3566448	4.127894e-02
## 119	0.49362989	0.41678233	4.382212e-04	0.6769434	0.6571513	4.127894e-02
## 120	1.00000000	1.00000000	4.127894e-02	0.9290058	0.9252524	4.127894e-02
## 121	0.60525812	0.57240512	1.688916e-03	0.7756071	0.7427276	4.127894e-02
## 122	0.73823901	0.71485396	4.767779e-03	0.8081557	0.8007580	4.127894e-02
## 123	0.54468534	0.53399278	5.551642e-04	0.6796223	0.6760241	4.127894e-02
## 124	0.20215501	0.12562655	8.530158e-07	0.4297822	0.3608378	4.127894e-02
## 125	0.15533517	0.00000000	3.673400e-05	0.5327732	0.5279986	4.127894e-02
## 126	0.61546961	0.58243394	2.642797e-03	0.7775782	0.7686443	4.127894e-02
## 127	0.15239463	0.02571976	2.572885e-09	0.2891650	0.1879559	4.127894e-02
## 128	0.05861035	0.00000000	2.861688e-09	0.2030123	0.1903365	4.127894e-02
## 129	0.85911311	0.86488468	8.953343e-03	0.8719701	0.8489090	1.688916e-03
## 130	0.10014971	0.00000000	4.963308e-08	0.2997245	0.2647680	1.688916e-03
## 131	0.40181388	0.21513715	1.227698e-04	0.6074820	0.5936156	1.688916e-03
## 132	0.67437227	0.67723506	1.850251e-04	0.6339217	0.6076417	1.688916e-03
## 133	0.73637816	0.72896413	2.005731e-03	0.7496253	0.7451235	1.688916e-03
## 134	0.26888913	0.00000000	7.722328e-07	0.4082043	0.3566448	1.688916e-03
## 135	0.63374100	0.51323035	4.382212e-04	0.6769434	0.6571513	1.688916e-03
## 136	0.86727195	0.87100818	4.127894e-02	0.9290058	0.9252524	1.688916e-03
## 137	1.00000000	1.00000000	1.688916e-03	0.7756071	0.7427276	1.688916e-03
## 138	0.77906065	0.77945959	4.767779e-03	0.8081557	0.8007580	1.688916e-03
## 139	0.64724897	0.66713032	5.551642e-04	0.6796223	0.6760241	1.688916e-03
## 140	0.25529330	0.12975580	8.530158e-07	0.4297822	0.3608378	1.688916e-03
## 141	0.24567222	0.00000000	3.673400e-05	0.5327732	0.5279986	1.688916e-03
## 142	0.65574934	0.00000000	2.642797e-03	0.7775782	0.7686443	1.688916e-03
## 143	0.13259606	0.00000000	2.572885e-09	0.2891650	0.1879559	1.688916e-03
## 144	0.05933493	0.00000000	2.861688e-09	0.2030123	0.1903365	1.688916e-03
## 145	0.81401467	0.78931414	8.953343e-03	0.8719701	0.8489090	4.767779e-03
## 146	0.13509521	0.00000000	4.963308e-08	0.2997245	0.2647680	4.767779e-03
## 147	0.43006913	0.30421051	1.227698e-04	0.6074820	0.5936156	4.767779e-03
## 148	0.52128884	0.48671141	1.850251e-04	0.6339217	0.6076417	4.767779e-03
## 149	0.64294468	0.63534350	2.005731e-03	0.7496253	0.7451235	4.767779e-03
## 150	0.21923334	0.00000000	7.722328e-07	0.4082043	0.3566448	4.767779e-03
## 151	0.53057183	0.29777400	4.382212e-04	0.6769434	0.6571513	4.767779e-03
## 152	0.92449207	0.92543341	4.127894e-02	0.9290058	0.9252524	4.767779e-03
## 153	0.70295721	0.66473267	1.688916e-03	0.7756071	0.7427276	4.767779e-03
## 154	1.00000000	1.00000000	4.767779e-03	0.8081557	0.8007580	4.767779e-03
## 155	0.59042519	0.57890695	5.551642e-04	0.6796223	0.6760241	4.767779e-03
## 156	0.22013247	0.12590122	8.530158e-07	0.4297822	0.3608378	4.767779e-03
## 157	0.18717082	0.00000000	3.673400e-05	0.5327732	0.5279986	4.767779e-03
## 158	0.72206146	0.71300354	2.642797e-03	0.7775782	0.7686443	4.767779e-03
## 159	0.16999767	0.00000000	2.572885e-09	0.2891650	0.1879559	4.767779e-03
## 160	0.07497167	0.00000000	2.861688e-09	0.2030123	0.1903365	4.767779e-03

## 161	0.85659175	0.62666256	8.953343e-03	0.8719701	0.8489090	5.551642e-04
## 162	0.08311792	0.00000000	4.963308e-08	0.2997245	0.2647680	5.551642e-04
## 163	0.29269237	0.20049948	1.227698e-04	0.6074820	0.5936156	5.551642e-04
## 164	0.52368121	0.30747012	1.850251e-04	0.6339217	0.6076417	5.551642e-04
## 165	0.76253546	0.58077919	2.005731e-03	0.7496253	0.7451235	5.551642e-04
## 166	0.39224894	0.00000000	7.722328e-07	0.4082043	0.3566448	5.551642e-04
## 167	0.70600443	0.45289733	4.382212e-04	0.6769434	0.6571513	5.551642e-04
## 168	0.96736915	0.96539672	4.127894e-02	0.9290058	0.9252524	5.551642e-04
## 169	0.77984111	0.79396577	1.688916e-03	0.7756071	0.7427276	5.551642e-04
## 170	0.82805209	0.80502215	4.767779e-03	0.8081557	0.8007580	5.551642e-04
## 171	1.00000000	1.00000000	5.551642e-04	0.6796223	0.6760241	5.551642e-04
## 172	0.25109417	0.00000000	8.530158e-07	0.4297822	0.3608378	5.551642e-04
## 173	0.23928399	0.00000000	3.673400e-05	0.5327732	0.5279986	5.551642e-04
## 174	0.62283249	0.42590247	2.642797e-03	0.7775782	0.7686443	5.551642e-04
## 175	0.16674623	0.04398215	2.572885e-09	0.2891650	0.1879559	5.551642e-04
## 176	0.10282328	0.00000000	2.861688e-09	0.2030123	0.1903365	5.551642e-04
## 177	0.79645883	0.63368905	8.953343e-03	0.8719701	0.8489090	8.530158e-07
## 178	0.20328804	0.00000000	4.963308e-08	0.2997245	0.2647680	8.530158e-07
## 179	0.40247741	0.00000000	1.227698e-04	0.6074820	0.5936156	8.530158e-07
## 180	0.59614565	0.34785016	1.850251e-04	0.6339217	0.6076417	8.530158e-07
## 181	0.70146885	0.00000000	2.005731e-03	0.7496253	0.7451235	8.530158e-07
## 182	0.19495195	0.00000000	7.722328e-07	0.4082043	0.3566448	8.530158e-07
## 183	0.45108355	0.00000000	4.382212e-04	0.6769434	0.6571513	8.530158e-07
## 184	0.76277510	0.60355829	4.127894e-02	0.9290058	0.9252524	8.530158e-07
## 185	0.66434159	0.41156452	1.688916e-03	0.7756071	0.7427276	8.530158e-07
## 186	0.64522171	0.45228801	4.767779e-03	0.8081557	0.8007580	8.530158e-07
## 187	0.50593241	0.00000000	5.551642e-04	0.6796223	0.6760241	8.530158e-07
## 188	1.00000000	1.00000000	8.530158e-07	0.4297822	0.3608378	8.530158e-07
## 189	0.27164700	0.00000000	3.673400e-05	0.5327732	0.5279986	8.530158e-07
## 190	0.50835738	0.00000000	2.642797e-03	0.7775782	0.7686443	8.530158e-07
## 191	0.19754170	0.00000000	2.572885e-09	0.2891650	0.1879559	8.530158e-07
## 192	0.04832299	0.00000000	2.861688e-09	0.2030123	0.1903365	8.530158e-07
## 193	0.38120117	0.00000000	8.953343e-03	0.8719701	0.8489090	3.673400e-05
## 194	0.18699386	0.00000000	4.963308e-08	0.2997245	0.2647680	3.673400e-05
## 195	0.42607617	0.00000000	1.227698e-04	0.6074820	0.5936156	3.673400e-05
## 196	0.48915997	0.00000000	1.850251e-04	0.6339217	0.6076417	3.673400e-05
## 197	0.46722243	0.00000000	2.005731e-03	0.7496253	0.7451235	3.673400e-05
## 198	0.31711227	0.00000000	7.722328e-07	0.4082043	0.3566448	3.673400e-05
## 199	0.40154052	0.00000000	4.382212e-04	0.6769434	0.6571513	3.673400e-05
## 200	0.40987606	0.00000000	4.127894e-02	0.9290058	0.9252524	3.673400e-05
## 201	0.43434253	0.00000000	1.688916e-03	0.7756071	0.7427276	3.673400e-05
## 202	0.38559674	0.00000000	4.767779e-03	0.8081557	0.8007580	3.673400e-05
## 203	0.34840358	0.00000000	5.551642e-04	0.6796223	0.6760241	3.673400e-05
## 204	0.19253977	0.00000000	8.530158e-07	0.4297822	0.3608378	3.673400e-05
## 205	1.00000000	1.00000000	3.673400e-05	0.5327732	0.5279986	3.673400e-05
## 206	0.44270192	0.00000000	2.642797e-03	0.7775782	0.7686443	3.673400e-05
## 207	0.03314362	0.00000000	2.572885e-09	0.2891650	0.1879559	3.673400e-05
## 208	0.18315471	0.00000000	2.861688e-09	0.2030123	0.1903365	3.673400e-05
## 209	0.64990886	0.47896411	8.953343e-03	0.8719701	0.8489090	2.642797e-03
## 210	0.15574178	0.00000000	4.963308e-08	0.2997245	0.2647680	2.642797e-03
## 211	0.53271241	0.45685615	1.227698e-04	0.6074820	0.5936156	2.642797e-03
## 212	0.54446119	0.00000000	1.850251e-04	0.6339217	0.6076417	2.642797e-03
## 213	0.60830138	0.46911436	2.005731e-03	0.7496253	0.7451235	2.642797e-03
## 214	0.20469019	0.00000000	7.722328e-07	0.4082043	0.3566448	2.642797e-03

##	215	0.45173272	0.37530333	4.382212e-04	0.6769434	0.6571513	2.642797e-03
##	216	0.82691150	0.82345156	4.127894e-02	0.9290058	0.9252524	2.642797e-03
##	217	0.63945650	0.00000000	1.688916e-03	0.7756071	0.7427276	2.642797e-03
##	218	0.78050033	0.77457291	4.767779e-03	0.8081557	0.8007580	2.642797e-03
##	219	0.48807226	0.34486986	5.551642e-04	0.6796223	0.6760241	2.642797e-03
##	220	0.18172291	0.00000000	8.530158e-07	0.4297822	0.3608378	2.642797e-03
##	221	0.24481701	0.00000000	3.673400e-05	0.5327732	0.5279986	2.642797e-03
##	222	1.00000000	1.00000000	2.642797e-03	0.7775782	0.7686443	2.642797e-03
##	223	0.14673465	0.00000000	2.572885e-09	0.2891650	0.1879559	2.642797e-03
##	224	0.09846820	0.00000000	2.861688e-09	0.2030123	0.1903365	2.642797e-03
##	225	0.74140541	0.64652175	8.953343e-03	0.8719701	0.8489090	2.572885e-09
##	226	0.45294328	0.00000000	4.963308e-08	0.2997245	0.2647680	2.572885e-09
##	227	0.46030577	0.00000000	1.227698e-04	0.6074820	0.5936156	2.572885e-09
##	228	0.55585014	0.00000000	1.850251e-04	0.6339217	0.6076417	2.572885e-09
##	229	0.72025291	0.56524277	2.005731e-03	0.7496253	0.7451235	2.572885e-09
##	230	0.30308938	0.00000000	7.722328e-07	0.4082043	0.3566448	2.572885e-09
##	231	0.52987910	0.00000000	4.382212e-04	0.6769434	0.6571513	2.572885e-09
##	232	0.97108214	0.30093175	4.127894e-02	0.9290058	0.9252524	2.572885e-09
##	233	0.59810016	0.00000000	1.688916e-03	0.7756071	0.7427276	2.572885e-09
##	234	0.74839329	0.00000000	4.767779e-03	0.8081557	0.8007580	2.572885e-09
##	235	0.60633058	0.27150565	5.551642e-04	0.6796223	0.6760241	2.572885e-09
##	236	0.31791306	0.00000000	8.530158e-07	0.4297822	0.3608378	2.572885e-09
##	237	0.09498876	0.00000000	3.673400e-05	0.5327732	0.5279986	2.572885e-09
##	238	0.62762097	0.00000000	2.642797e-03	0.7775782	0.7686443	2.572885e-09
##	239	1.00000000	1.00000000	2.572885e-09	0.2891650	0.1879559	2.572885e-09
##	240	0.02816310	0.00000000	2.861688e-09	0.2030123	0.1903365	2.572885e-09
##	241	0.32017518	0.00000000	8.953343e-03	0.8719701	0.8489090	2.861688e-09
##	242	0.33398733	0.00000000	4.963308e-08	0.2997245	0.2647680	2.861688e-09
##	243	0.71708422	0.00000000	1.227698e-04	0.6074820	0.5936156	2.861688e-09
##	244	0.42304777	0.00000000	1.850251e-04	0.6339217	0.6076417	2.861688e-09
##	245	0.28569573	0.00000000	2.005731e-03	0.7496253	0.7451235	2.861688e-09
##	246	0.58685911	0.00000000	7.722328e-07	0.4082043	0.3566448	2.861688e-09
##	247	0.49562410	0.00000000	4.382212e-04	0.6769434	0.6571513	2.861688e-09
##	248	0.54405949	0.00000000	4.127894e-02	0.9290058	0.9252524	2.861688e-09
##	249	0.34198320	0.00000000	1.688916e-03	0.7756071	0.7427276	2.861688e-09
##	250	0.55176550	0.00000000	4.767779e-03	0.8081557	0.8007580	2.861688e-09
##	251	0.51578326	0.00000000	5.551642e-04	0.6796223	0.6760241	2.861688e-09
##	252	0.11608263	0.00000000	8.530158e-07	0.4297822	0.3608378	2.861688e-09
##	253	0.62626694	0.00000000	3.673400e-05	0.5327732	0.5279986	2.861688e-09
##	254	0.60436649	0.00000000	2.642797e-03	0.7775782	0.7686443	2.861688e-09
##	255	0.02883505	0.00000000	2.572885e-09	0.2891650	0.1879559	2.861688e-09
##	256	1.00000000	1.00000000	2.861688e-09	0.2030123	0.1903365	2.861688e-09
##		vol_V2_mean	vol_V2_gmean				
##	1	0.8719701	0.8489090				
##	2	0.8719701	0.8489090				
##	3	0.8719701	0.8489090				
##	4	0.8719701	0.8489090				
##	5	0.8719701	0.8489090				
##	6	0.8719701	0.8489090				
##	7	0.8719701	0.8489090				
##	8	0.8719701	0.8489090				
##	9	0.8719701	0.8489090				
##	10	0.8719701	0.8489090				
##	11	0.8719701	0.8489090				

## 12	0.8719701	0.8489090
## 13	0.8719701	0.8489090
## 14	0.8719701	0.8489090
## 15	0.8719701	0.8489090
## 16	0.8719701	0.8489090
## 17	0.2997245	0.2647680
## 18	0.2997245	0.2647680
## 19	0.2997245	0.2647680
## 20	0.2997245	0.2647680
## 21	0.2997245	0.2647680
## 22	0.2997245	0.2647680
## 23	0.2997245	0.2647680
## 24	0.2997245	0.2647680
## 25	0.2997245	0.2647680
## 26	0.2997245	0.2647680
## 27	0.2997245	0.2647680
## 28	0.2997245	0.2647680
## 29	0.2997245	0.2647680
## 30	0.2997245	0.2647680
## 31	0.2997245	0.2647680
## 32	0.2997245	0.2647680
## 33	0.6074820	0.5936156
## 34	0.6074820	0.5936156
## 35	0.6074820	0.5936156
## 36	0.6074820	0.5936156
## 37	0.6074820	0.5936156
## 38	0.6074820	0.5936156
## 39	0.6074820	0.5936156
## 40	0.6074820	0.5936156
## 41	0.6074820	0.5936156
## 42	0.6074820	0.5936156
## 43	0.6074820	0.5936156
## 44	0.6074820	0.5936156
## 45	0.6074820	0.5936156
## 46	0.6074820	0.5936156
## 47	0.6074820	0.5936156
## 48	0.6074820	0.5936156
## 49	0.6339217	0.6076417
## 50	0.6339217	0.6076417
## 51	0.6339217	0.6076417
## 52	0.6339217	0.6076417
## 53	0.6339217	0.6076417
## 54	0.6339217	0.6076417
## 55	0.6339217	0.6076417
## 56	0.6339217	0.6076417
## 57	0.6339217	0.6076417
## 58	0.6339217	0.6076417
## 59	0.6339217	0.6076417
## 60	0.6339217	0.6076417
## 61	0.6339217	0.6076417
## 62	0.6339217	0.6076417
## 63	0.6339217	0.6076417
## 64	0.6339217	0.6076417
## 65	0.7496253	0.7451235

## 66	0.7496253	0.7451235
## 67	0.7496253	0.7451235
## 68	0.7496253	0.7451235
## 69	0.7496253	0.7451235
## 70	0.7496253	0.7451235
## 71	0.7496253	0.7451235
## 72	0.7496253	0.7451235
## 73	0.7496253	0.7451235
## 74	0.7496253	0.7451235
## 75	0.7496253	0.7451235
## 76	0.7496253	0.7451235
## 77	0.7496253	0.7451235
## 78	0.7496253	0.7451235
## 79	0.7496253	0.7451235
## 80	0.7496253	0.7451235
## 81	0.4082043	0.3566448
## 82	0.4082043	0.3566448
## 83	0.4082043	0.3566448
## 84	0.4082043	0.3566448
## 85	0.4082043	0.3566448
## 86	0.4082043	0.3566448
## 87	0.4082043	0.3566448
## 88	0.4082043	0.3566448
## 89	0.4082043	0.3566448
## 90	0.4082043	0.3566448
## 91	0.4082043	0.3566448
## 92	0.4082043	0.3566448
## 93	0.4082043	0.3566448
## 94	0.4082043	0.3566448
## 95	0.4082043	0.3566448
## 96	0.4082043	0.3566448
## 97	0.6769434	0.6571513
## 98	0.6769434	0.6571513
## 99	0.6769434	0.6571513
## 100	0.6769434	0.6571513
## 101	0.6769434	0.6571513
## 102	0.6769434	0.6571513
## 103	0.6769434	0.6571513
## 104	0.6769434	0.6571513
## 105	0.6769434	0.6571513
## 106	0.6769434	0.6571513
## 107	0.6769434	0.6571513
## 108	0.6769434	0.6571513
## 109	0.6769434	0.6571513
## 110	0.6769434	0.6571513
## 111	0.6769434	0.6571513
## 112	0.6769434	0.6571513
## 113	0.9290058	0.9252524
## 114	0.9290058	0.9252524
## 115	0.9290058	0.9252524
## 116	0.9290058	0.9252524
## 117	0.9290058	0.9252524
## 118	0.9290058	0.9252524
## 119	0.9290058	0.9252524

## 120	0.9290058	0.9252524
## 121	0.9290058	0.9252524
## 122	0.9290058	0.9252524
## 123	0.9290058	0.9252524
## 124	0.9290058	0.9252524
## 125	0.9290058	0.9252524
## 126	0.9290058	0.9252524
## 127	0.9290058	0.9252524
## 128	0.9290058	0.9252524
## 129	0.7756071	0.7427276
## 130	0.7756071	0.7427276
## 131	0.7756071	0.7427276
## 132	0.7756071	0.7427276
## 133	0.7756071	0.7427276
## 134	0.7756071	0.7427276
## 135	0.7756071	0.7427276
## 136	0.7756071	0.7427276
## 137	0.7756071	0.7427276
## 138	0.7756071	0.7427276
## 139	0.7756071	0.7427276
## 140	0.7756071	0.7427276
## 141	0.7756071	0.7427276
## 142	0.7756071	0.7427276
## 143	0.7756071	0.7427276
## 144	0.7756071	0.7427276
## 145	0.8081557	0.8007580
## 146	0.8081557	0.8007580
## 147	0.8081557	0.8007580
## 148	0.8081557	0.8007580
## 149	0.8081557	0.8007580
## 150	0.8081557	0.8007580
## 151	0.8081557	0.8007580
## 152	0.8081557	0.8007580
## 153	0.8081557	0.8007580
## 154	0.8081557	0.8007580
## 155	0.8081557	0.8007580
## 156	0.8081557	0.8007580
## 157	0.8081557	0.8007580
## 158	0.8081557	0.8007580
## 159	0.8081557	0.8007580
## 160	0.8081557	0.8007580
## 161	0.6796223	0.6760241
## 162	0.6796223	0.6760241
## 163	0.6796223	0.6760241
## 164	0.6796223	0.6760241
## 165	0.6796223	0.6760241
## 166	0.6796223	0.6760241
## 167	0.6796223	0.6760241
## 168	0.6796223	0.6760241
## 169	0.6796223	0.6760241
## 170	0.6796223	0.6760241
## 171	0.6796223	0.6760241
## 172	0.6796223	0.6760241
## 173	0.6796223	0.6760241

## 174	0.6796223	0.6760241
## 175	0.6796223	0.6760241
## 176	0.6796223	0.6760241
## 177	0.4297822	0.3608378
## 178	0.4297822	0.3608378
## 179	0.4297822	0.3608378
## 180	0.4297822	0.3608378
## 181	0.4297822	0.3608378
## 182	0.4297822	0.3608378
## 183	0.4297822	0.3608378
## 184	0.4297822	0.3608378
## 185	0.4297822	0.3608378
## 186	0.4297822	0.3608378
## 187	0.4297822	0.3608378
## 188	0.4297822	0.3608378
## 189	0.4297822	0.3608378
## 190	0.4297822	0.3608378
## 191	0.4297822	0.3608378
## 192	0.4297822	0.3608378
## 193	0.5327732	0.5279986
## 194	0.5327732	0.5279986
## 195	0.5327732	0.5279986
## 196	0.5327732	0.5279986
## 197	0.5327732	0.5279986
## 198	0.5327732	0.5279986
## 199	0.5327732	0.5279986
## 200	0.5327732	0.5279986
## 201	0.5327732	0.5279986
## 202	0.5327732	0.5279986
## 203	0.5327732	0.5279986
## 204	0.5327732	0.5279986
## 205	0.5327732	0.5279986
## 206	0.5327732	0.5279986
## 207	0.5327732	0.5279986
## 208	0.5327732	0.5279986
## 209	0.7775782	0.7686443
## 210	0.7775782	0.7686443
## 211	0.7775782	0.7686443
## 212	0.7775782	0.7686443
## 213	0.7775782	0.7686443
## 214	0.7775782	0.7686443
## 215	0.7775782	0.7686443
## 216	0.7775782	0.7686443
## 217	0.7775782	0.7686443
## 218	0.7775782	0.7686443
## 219	0.7775782	0.7686443
## 220	0.7775782	0.7686443
## 221	0.7775782	0.7686443
## 222	0.7775782	0.7686443
## 223	0.7775782	0.7686443
## 224	0.7775782	0.7686443
## 225	0.2891650	0.1879559
## 226	0.2891650	0.1879559
## 227	0.2891650	0.1879559

```
## 228 0.2891650 0.1879559
## 229 0.2891650 0.1879559
## 230 0.2891650 0.1879559
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## 237 0.2891650 0.1879559
## 238 0.2891650 0.1879559
## 239 0.2891650 0.1879559
## 240 0.2891650 0.1879559
## 241 0.2030123 0.1903365
## 242 0.2030123 0.1903365
## 243 0.2030123 0.1903365
## 244 0.2030123 0.1903365
## 245 0.2030123 0.1903365
## 246 0.2030123 0.1903365
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## 253 0.2030123 0.1903365
## 254 0.2030123 0.1903365
## 255 0.2030123 0.1903365
## 256 0.2030123 0.1903365
```

```
result <- rc$result
vol <- as.data.frame(matrix(nrow = 16, ncol = 2))
colnames(vol) <- c("volume", "clade")
vol$clade <- sort(unique(Xrc$clade))
# product
vol[, 1] <- c(result[1:16, 6])
p1 <- ggplot(vol, aes(x = clade, y = volume)) + geom_bar(aes(fill = clade),
  stat = "identity", width = 0.8) + coord_flip() + scale_fill_manual(values = c("yellow",
  "gray", "aquamarine", "lightgreen", "cyan4", "wheat1", "tomato4",
  "tan1", "red", "violet", "steelblue1", "maroon", "forestgreen",
  "darkblue", "slateblue4", "darkred")) + ylab("Full niche hypervolume") +
  xlab("Genetic clade") + coord_flip() + labs(title = "Product") +
  guides(fill = F) + theme_classic()
```

Product

```
vol[, 1] <- c(result[1:16, 7])
p2 <- ggplot(vol, aes(x = clade, y = volume)) + geom_bar(aes(fill = clade),
```

```

stat = "identity", width = 0.8) + coord_flip() + scale_fill_manual(values = c("yellow",
"gray", "aquamarine", "lightgreen", "cyan4", "wheat1", "tomato4",
"tan1", "red", "violet", "steelblue1", "maroon", "forestgreen",
"darkblue", "slateblue4", "darkred")) + ylab("Full niche hypervolume") +
coord_flip() + labs(title = "Arithmetic mean") + guides(fill = F) +
theme_classic() + theme(axis.text.y = element_blank(), axis.title.y = element_blank())

```

Arithmetic mean

```

vol[, 1] <- c(result[1:16, 8])
p3 <- ggplot(vol, aes(x = clade, y = volume)) + geom_bar(aes(fill = clade),
stat = "identity", width = 0.8) + coord_flip() + scale_fill_manual(values = c("yellow",
"gray", "aquamarine", "lightgreen", "cyan4", "wheat1", "tomato4",
"tan1", "red", "violet", "steelblue1", "maroon", "forestgreen",
"darkblue", "slateblue4", "darkred")) + ylab("Full niche hypervolume") +
coord_flip() + labs(title = "Geometric mean") + guides(fill = F) +
theme_classic() + theme(axis.text.y = element_blank(), axis.title.y = element_blank())

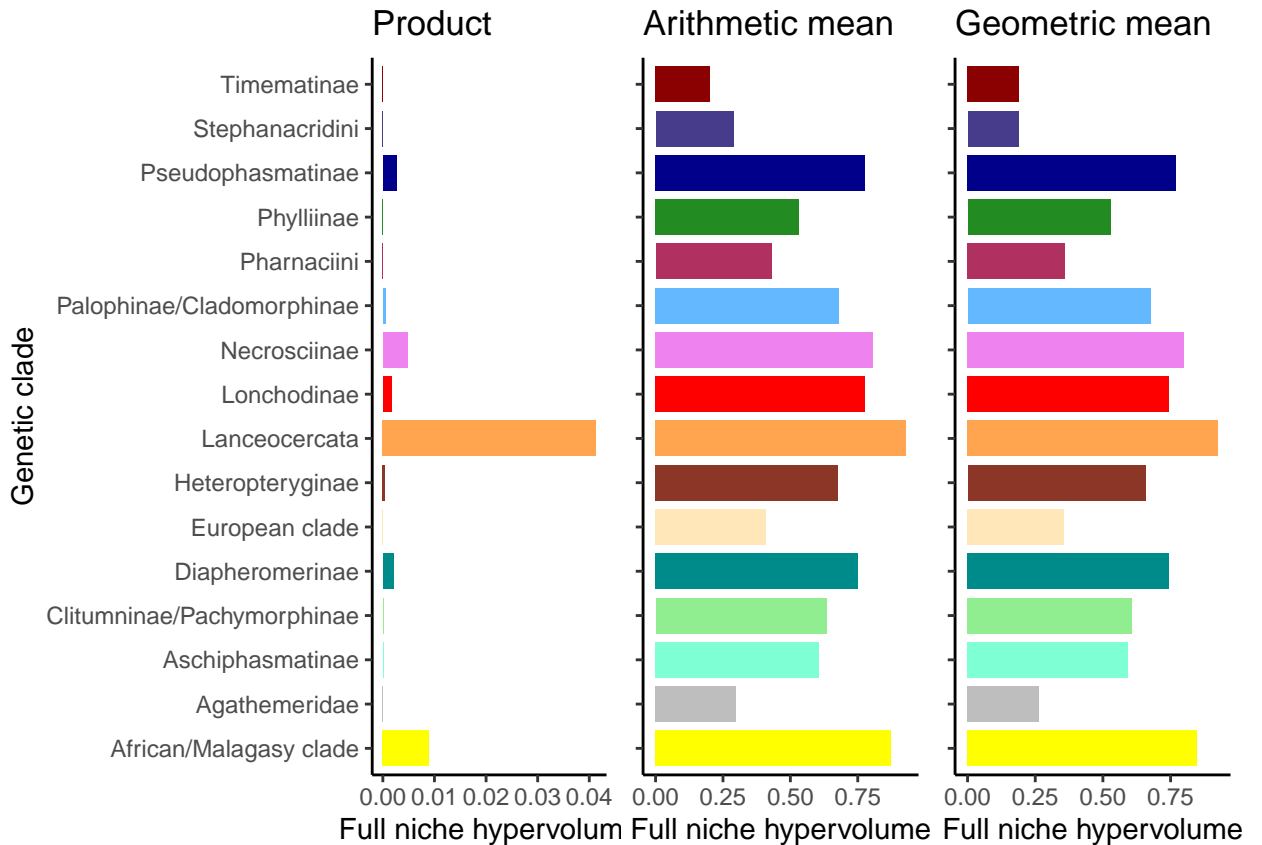
```

Geometric mean

```

plot_grid(p1, p2, p3, nrow = 1, rel_widths = (c(2, 1, 1)))

```



Plots

Hypervolume size - Different PCAs

Create datasets

```

# PCAmix - Size corrected
Xrb <- matrix(nrow = length(X$habitat2), ncol = 8)
colnames(Xrb) <- c("habitat", "PC1", "PC2", "PC3", "PC4", "PC5",
  "PC6", "PC7")
Xrb <- as.data.frame(Xrb)
Xrb$habitat <- as.character(X$habitat2)
Xrb$PC1 <- as.numeric(X$pca_mix_dim1)
Xrb$PC2 <- as.numeric(X$pca_mix_dim2)
Xrb$PC3 <- as.numeric(X$pca_mix_dim3)
Xrb$PC4 <- as.numeric(X$pca_mix_dim4)
Xrb$PC5 <- as.numeric(X$pca_mix_dim5)
Xrb$PC6 <- as.numeric(X$pca_mix_dim6)
Xrb$PC7 <- as.numeric(X$pca_mix_dim7)

# Phylogenetic PCA - Size corrected
Xrb2 <- matrix(nrow = length(X$habitat2), ncol = 9)
colnames(Xrb2) <- c("habitat", "PC1", "PC2", "PC3", "PC4", "PC5",
  "PC6", "PC7", "PC8")
Xrb2 <- as.data.frame(Xrb2)
Xrb2$habitat <- as.character(X$habitat2)

```

```

Xrb2$PC1 <- as.numeric(X$pPCA_sizecorr_dim1)
Xrb2$PC2 <- as.numeric(X$pPCA_sizecorr_dim2)
Xrb2$PC3 <- as.numeric(X$pPCA_sizecorr_dim3)
Xrb2$PC4 <- as.numeric(X$pPCA_sizecorr_dim4)
Xrb2$PC5 <- as.numeric(X$pPCA_sizecorr_dim5)
Xrb2$PC6 <- as.numeric(X$pPCA_sizecorr_dim6)
Xrb2$PC7 <- as.numeric(X$pPCA_sizecorr_dim7)
Xrb2$PC8 <- as.numeric(X$pPCA_sizecorr_dim8)

# Phylogenetic PCA - No size correction
Xrb3 <- matrix(nrow = length(X$habitat2), ncol = 7)
colnames(Xrb3) <- c("habitat", "PC1", "PC2", "PC3", "PC4", "PC5",
  "PC6")
Xrb3 <- as.data.frame(Xrb3)
Xrb3$habitat <- as.character(X$habitat2)
Xrb3$PC1 <- as.numeric(X$pPCA_nocorr_dim1)
Xrb3$PC2 <- as.numeric(X$pPCA_nocorr_dim2)
Xrb3$PC3 <- as.numeric(X$pPCA_nocorr_dim3)
Xrb3$PC4 <- as.numeric(X$pPCA_nocorr_dim4)
Xrb3$PC5 <- as.numeric(X$pPCA_nocorr_dim5)
Xrb3$PC6 <- as.numeric(X$pPCA_nocorr_dim6)

```

Using DynRB package

```
rc <- dynRB_VPa(Xrb)
```

```

## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
## [1] 6
## [1] 7
## [1] 8
## [1] 9
## [1] 10
## [1] 11
## [1] 12
## [1] 13
## [1] 14
## [1] 15
## [1] 16
## [1] 17
## [1] 18
## [1] 19
## [1] 20
## [1] 21
## [1] 22
## [1] 23
## [1] 24
## [1] 25

```

```

##                                     V1
## 1      hanging from branches and leaves
## 2              hanging from grass
## 3              laying on branches and leaves
## 4 laying on leaf litter, logs, bark, trunk
## 5              laying on palm leaves
## 6      hanging from branches and leaves
## 7              hanging from grass
## 8              laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10             laying on palm leaves
## 11     hanging from branches and leaves
## 12             hanging from grass
## 13     laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15             laying on palm leaves
## 16     hanging from branches and leaves
## 17             hanging from grass
## 18     laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20             laying on palm leaves
## 21     hanging from branches and leaves
## 22             hanging from grass
## 23     laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25             laying on palm leaves
##                                     V2      port_prod  vol_V1_prod
## 1      hanging from branches and leaves 1.000000e+00 8.026471e-02
## 2      hanging from branches and leaves 2.440322e-05 6.373489e-06
## 3      hanging from branches and leaves 9.552701e-02 2.708696e-02
## 4      hanging from branches and leaves 3.985398e-02 9.275388e-03
## 5      hanging from branches and leaves 0.000000e+00 6.429999e-05
## 6              hanging from grass 3.411457e-01 8.026471e-02
## 7              hanging from grass 1.000000e+00 6.373489e-06
## 8              hanging from grass 5.028213e-02 2.708696e-02
## 9              hanging from grass 0.000000e+00 9.275388e-03
## 10             hanging from grass 0.000000e+00 6.429999e-05
## 11     laying on branches and leaves 2.377482e-01 8.026471e-02
## 12     laying on branches and leaves 8.395322e-06 6.373489e-06
## 13     laying on branches and leaves 1.000000e+00 2.708696e-02
## 14     laying on branches and leaves 8.144047e-02 9.275388e-03
## 15     laying on branches and leaves 2.329100e-04 6.429999e-05
## 16 laying on leaf litter, logs, bark, trunk 3.149566e-01 8.026471e-02
## 17 laying on leaf litter, logs, bark, trunk 0.000000e+00 6.373489e-06
## 18 laying on leaf litter, logs, bark, trunk 2.559732e-01 2.708696e-02
## 19 laying on leaf litter, logs, bark, trunk 1.000000e+00 9.275388e-03
## 20 laying on leaf litter, logs, bark, trunk 2.242259e-05 6.429999e-05
## 21             laying on palm leaves 0.000000e+00 8.026471e-02
## 22             laying on palm leaves 0.000000e+00 6.373489e-06
## 23             laying on palm leaves 1.256811e-01 2.708696e-02
## 24             laying on palm leaves 6.106268e-03 9.275388e-03
## 25             laying on palm leaves 1.000000e+00 6.429999e-05
##      vol_V2_prod
## 1 8.026471e-02

```

```

## 2 8.026471e-02
## 3 8.026471e-02
## 4 8.026471e-02
## 5 8.026471e-02
## 6 6.373489e-06
## 7 6.373489e-06
## 8 6.373489e-06
## 9 6.373489e-06
## 10 6.373489e-06
## 11 2.708696e-02
## 12 2.708696e-02
## 13 2.708696e-02
## 14 2.708696e-02
## 15 2.708696e-02
## 16 9.275388e-03
## 17 9.275388e-03
## 18 9.275388e-03
## 19 9.275388e-03
## 20 9.275388e-03
## 21 6.429999e-05
## 22 6.429999e-05
## 23 6.429999e-05
## 24 6.429999e-05
## 25 6.429999e-05

```

```
rc$result
```

```

##                               V1
## 1      hanging from branches and leaves
## 2                hanging from grass
## 3      laying on branches and leaves
## 4 laying on leaf litter, logs, bark, trunk
## 5                laying on palm leaves
## 6      hanging from branches and leaves
## 7                hanging from grass
## 8      laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10               laying on palm leaves
## 11      hanging from branches and leaves
## 12               hanging from grass
## 13      laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15               laying on palm leaves
## 16      hanging from branches and leaves
## 17               hanging from grass
## 18      laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20               laying on palm leaves
## 21      hanging from branches and leaves
## 22               hanging from grass
## 23      laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25               laying on palm leaves
##                               V2      port_prod port_mean port_gmean

```



```

## 1      hanging from branches and leaves 1.000000e+00 1.0000000 1.00000000
## 2      hanging from branches and leaves 2.440322e-05 0.2785816 0.13636491
## 3      hanging from branches and leaves 9.552701e-02 0.7055622 0.68074627
## 4      hanging from branches and leaves 3.985398e-02 0.6805866 0.56350439
## 5      hanging from branches and leaves 0.000000e+00 0.3491063 0.00000000
## 6              hanging from grass 3.411457e-01 0.8905466 0.52079172
## 7              hanging from grass 1.000000e+00 1.0000000 1.00000000
## 8              hanging from grass 5.028213e-02 0.6071049 0.38025652
## 9              hanging from grass 0.000000e+00 0.7916137 0.00000000
## 10             hanging from grass 0.000000e+00 0.4176534 0.00000000
## 11             laying on branches and leaves 2.377482e-01 0.7929074 0.76188259
## 12             laying on branches and leaves 8.395322e-06 0.2022186 0.11357210
## 13             laying on branches and leaves 1.000000e+00 1.0000000 1.00000000
## 14             laying on branches and leaves 8.144047e-02 0.6704278 0.66073281
## 15             laying on branches and leaves 2.329100e-04 0.3734930 0.18744212
## 16 laying on leaf litter, logs, bark, trunk 3.149566e-01 0.8730048 0.75080560
## 17 laying on leaf litter, logs, bark, trunk 0.000000e+00 0.3390452 0.00000000
## 18 laying on leaf litter, logs, bark, trunk 2.559732e-01 0.7884876 0.78182337
## 19 laying on leaf litter, logs, bark, trunk 1.000000e+00 1.0000000 1.00000000
## 20 laying on leaf litter, logs, bark, trunk 2.242259e-05 0.3216908 0.07940308
## 21             laying on palm leaves 0.000000e+00 0.8509387 0.00000000
## 22             laying on palm leaves 0.000000e+00 0.3071580 0.00000000
## 23             laying on palm leaves 1.256811e-01 0.8178359 0.44802457
## 24             laying on palm leaves 6.106268e-03 0.5905056 0.17040009
## 25             laying on palm leaves 1.000000e+00 1.0000000 1.00000000
##      vol_V1_prod vol_V1_mean vol_V1_gmean  vol_V2_prod vol_V2_mean vol_V2_gmean
## 1  8.026471e-02  0.9664312  0.9626571  8.026471e-02  0.9664312  0.9626571
## 2  6.373489e-06  0.4744041  0.4397935  8.026471e-02  0.9664312  0.9626571
## 3  2.708696e-02  0.9111737  0.9103507  8.026471e-02  0.9664312  0.9626571
## 4  9.275388e-03  0.8471250  0.8335036  8.026471e-02  0.9664312  0.9626571
## 5  6.429999e-05  0.5704378  0.5544605  8.026471e-02  0.9664312  0.9626571
## 6  8.026471e-02  0.9664312  0.9626571  6.373489e-06  0.4744041  0.4397935
## 7  6.373489e-06  0.4744041  0.4397935  6.373489e-06  0.4744041  0.4397935
## 8  2.708696e-02  0.9111737  0.9103507  6.373489e-06  0.4744041  0.4397935
## 9  9.275388e-03  0.8471250  0.8335036  6.373489e-06  0.4744041  0.4397935
## 10 6.429999e-05  0.5704378  0.5544605  6.373489e-06  0.4744041  0.4397935
## 11 8.026471e-02  0.9664312  0.9626571  2.708696e-02  0.9111737  0.9103507
## 12 6.373489e-06  0.4744041  0.4397935  2.708696e-02  0.9111737  0.9103507
## 13 2.708696e-02  0.9111737  0.9103507  2.708696e-02  0.9111737  0.9103507
## 14 9.275388e-03  0.8471250  0.8335036  2.708696e-02  0.9111737  0.9103507
## 15 6.429999e-05  0.5704378  0.5544605  2.708696e-02  0.9111737  0.9103507
## 16 8.026471e-02  0.9664312  0.9626571  9.275388e-03  0.8471250  0.8335036
## 17 6.373489e-06  0.4744041  0.4397935  9.275388e-03  0.8471250  0.8335036
## 18 2.708696e-02  0.9111737  0.9103507  9.275388e-03  0.8471250  0.8335036
## 19 9.275388e-03  0.8471250  0.8335036  9.275388e-03  0.8471250  0.8335036
## 20 6.429999e-05  0.5704378  0.5544605  9.275388e-03  0.8471250  0.8335036
## 21 8.026471e-02  0.9664312  0.9626571  6.429999e-05  0.5704378  0.5544605
## 22 6.373489e-06  0.4744041  0.4397935  6.429999e-05  0.5704378  0.5544605
## 23 2.708696e-02  0.9111737  0.9103507  6.429999e-05  0.5704378  0.5544605
## 24 9.275388e-03  0.8471250  0.8335036  6.429999e-05  0.5704378  0.5544605
## 25 6.429999e-05  0.5704378  0.5544605  6.429999e-05  0.5704378  0.5544605

```

```
rc2 <- dynRB_VPa(Xrb2)
```

```

## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
## [1] 6
## [1] 7
## [1] 8
## [1] 9
## [1] 10
## [1] 11
## [1] 12
## [1] 13
## [1] 14
## [1] 15
## [1] 16
## [1] 17
## [1] 18
## [1] 19
## [1] 20
## [1] 21
## [1] 22
## [1] 23
## [1] 24
## [1] 25
##
##                                     V1
## 1      hanging from branches and leaves
## 2              hanging from grass
## 3              laying on branches and leaves
## 4 laying on leaf litter, logs, bark, trunk
## 5              laying on palm leaves
## 6      hanging from branches and leaves
## 7              hanging from grass
## 8              laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10             laying on palm leaves
## 11      hanging from branches and leaves
## 12             hanging from grass
## 13              laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15             laying on palm leaves
## 16      hanging from branches and leaves
## 17             hanging from grass
## 18              laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20             laying on palm leaves
## 21      hanging from branches and leaves
## 22             hanging from grass
## 23              laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25             laying on palm leaves
##
##                                     V2      port_prod  vol_V1_prod
## 1      hanging from branches and leaves 1.000000e+00 8.877607e-02
## 2      hanging from branches and leaves 1.382038e-05 5.543901e-06

```

```

## 3      hanging from branches and leaves 1.923441e-02 6.701998e-03
## 4      hanging from branches and leaves 1.696952e-02 5.606689e-03
## 5      hanging from branches and leaves 2.304337e-04 4.538380e-05
## 6              hanging from grass 1.875653e-01 8.877607e-02
## 7              hanging from grass 1.000000e+00 5.543901e-06
## 8              hanging from grass 4.870423e-02 6.701998e-03
## 9              hanging from grass 0.000000e+00 5.606689e-03
## 10             hanging from grass 0.000000e+00 4.538380e-05
## 11      laying on branches and leaves 2.408277e-01 8.877607e-02
## 12      laying on branches and leaves 2.766688e-05 5.543901e-06
## 13      laying on branches and leaves 1.000000e+00 6.701998e-03
## 14      laying on branches and leaves 8.185961e-02 5.606689e-03
## 15      laying on branches and leaves 0.000000e+00 4.538380e-05
## 16 laying on leaf litter, logs, bark, trunk 3.121606e-01 8.877607e-02
## 17 laying on leaf litter, logs, bark, trunk 0.000000e+00 5.543901e-06
## 18 laying on leaf litter, logs, bark, trunk 1.170634e-01 6.701998e-03
## 19 laying on leaf litter, logs, bark, trunk 1.000000e+00 5.606689e-03
## 20 laying on leaf litter, logs, bark, trunk 2.701403e-04 4.538380e-05
## 21      laying on palm leaves 3.318853e-01 8.877607e-02
## 22      laying on palm leaves 0.000000e+00 5.543901e-06
## 23      laying on palm leaves 0.000000e+00 6.701998e-03
## 24      laying on palm leaves 3.877847e-02 5.606689e-03
## 25      laying on palm leaves 1.000000e+00 4.538380e-05
##      vol_V2_prod
## 1 8.877607e-02
## 2 8.877607e-02
## 3 8.877607e-02
## 4 8.877607e-02
## 5 8.877607e-02
## 6 5.543901e-06
## 7 5.543901e-06
## 8 5.543901e-06
## 9 5.543901e-06
## 10 5.543901e-06
## 11 6.701998e-03
## 12 6.701998e-03
## 13 6.701998e-03
## 14 6.701998e-03
## 15 6.701998e-03
## 16 5.606689e-03
## 17 5.606689e-03
## 18 5.606689e-03
## 19 5.606689e-03
## 20 5.606689e-03
## 21 4.538380e-05
## 22 4.538380e-05
## 23 4.538380e-05
## 24 4.538380e-05
## 25 4.538380e-05

```

```
rc2$result
```

```

##                                V1
## 1      hanging from branches and leaves

```

```

## 2             hanging from grass
## 3             laying on branches and leaves
## 4 laying on leaf litter, logs, bark, trunk
## 5             laying on palm leaves
## 6             hanging from branches and leaves
## 7             hanging from grass
## 8             laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10            laying on palm leaves
## 11            hanging from branches and leaves
## 12            hanging from grass
## 13            laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15            laying on palm leaves
## 16            hanging from branches and leaves
## 17            hanging from grass
## 18            laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20            laying on palm leaves
## 21            hanging from branches and leaves
## 22            hanging from grass
## 23            laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25            laying on palm leaves
##
##              V2      port_prod port_mean port_gmean
## 1             hanging from branches and leaves 1.000000e+00 1.0000000 1.0000000
## 2             hanging from branches and leaves 1.382038e-05 0.2564344 0.2272034
## 3             hanging from branches and leaves 1.923441e-02 0.6055165 0.5974303
## 4             hanging from branches and leaves 1.696952e-02 0.6202517 0.5540875
## 5             hanging from branches and leaves 2.304337e-04 0.3904977 0.3413156
## 6             hanging from grass 1.875653e-01 0.7827245 0.7212464
## 7             hanging from grass 1.000000e+00 1.0000000 1.0000000
## 8             hanging from grass 4.870423e-02 0.6666597 0.4102019
## 9             hanging from grass 0.000000e+00 0.5909686 0.0000000
## 10            hanging from grass 0.000000e+00 0.3781358 0.0000000
## 11            laying on branches and leaves 2.408277e-01 0.8020534 0.8055643
## 12            laying on branches and leaves 2.766688e-05 0.2937828 0.1675191
## 13            laying on branches and leaves 1.000000e+00 1.0000000 1.0000000
## 14            laying on branches and leaves 8.185961e-02 0.7000857 0.7069830
## 15            laying on branches and leaves 0.000000e+00 0.4302223 0.0000000
## 16 laying on leaf litter, logs, bark, trunk 3.121606e-01 0.8620425 0.7963066
## 17 laying on leaf litter, logs, bark, trunk 0.000000e+00 0.2665596 0.0000000
## 18 laying on leaf litter, logs, bark, trunk 1.170634e-01 0.7402571 0.7487617
## 19 laying on leaf litter, logs, bark, trunk 1.000000e+00 1.0000000 1.0000000
## 20 laying on leaf litter, logs, bark, trunk 2.701403e-04 0.4792061 0.2312780
## 21            laying on palm leaves 3.318853e-01 0.8460962 0.8289590
## 22            laying on palm leaves 0.000000e+00 0.2772102 0.0000000
## 23            laying on palm leaves 0.000000e+00 0.6999329 0.0000000
## 24            laying on palm leaves 3.877847e-02 0.7564975 0.4149414
## 25            laying on palm leaves 1.000000e+00 1.0000000 1.0000000
##
##      vol_V1_prod vol_V1_mean vol_V1_gmean  vol_V2_prod vol_V2_mean vol_V2_gmean
## 1  8.877607e-02  0.9791447  0.9778563 8.877607e-02  0.9791447  0.9778563
## 2  5.543901e-06  0.5211603  0.5146444 8.877607e-02  0.9791447  0.9778563
## 3  6.701998e-03  0.8751410  0.8631182 8.877607e-02  0.9791447  0.9778563

```

```

## 4 5.606689e-03 0.8616857 0.8449887 8.877607e-02 0.9791447 0.9778563
## 5 4.538380e-05 0.6545858 0.6121191 8.877607e-02 0.9791447 0.9778563
## 6 8.877607e-02 0.9791447 0.9778563 5.543901e-06 0.5211603 0.5146444
## 7 5.543901e-06 0.5211603 0.5146444 5.543901e-06 0.5211603 0.5146444
## 8 6.701998e-03 0.8751410 0.8631182 5.543901e-06 0.5211603 0.5146444
## 9 5.606689e-03 0.8616857 0.8449887 5.543901e-06 0.5211603 0.5146444
## 10 4.538380e-05 0.6545858 0.6121191 5.543901e-06 0.5211603 0.5146444
## 11 8.877607e-02 0.9791447 0.9778563 6.701998e-03 0.8751410 0.8631182
## 12 5.543901e-06 0.5211603 0.5146444 6.701998e-03 0.8751410 0.8631182
## 13 6.701998e-03 0.8751410 0.8631182 6.701998e-03 0.8751410 0.8631182
## 14 5.606689e-03 0.8616857 0.8449887 6.701998e-03 0.8751410 0.8631182
## 15 4.538380e-05 0.6545858 0.6121191 6.701998e-03 0.8751410 0.8631182
## 16 8.877607e-02 0.9791447 0.9778563 5.606689e-03 0.8616857 0.8449887
## 17 5.543901e-06 0.5211603 0.5146444 5.606689e-03 0.8616857 0.8449887
## 18 6.701998e-03 0.8751410 0.8631182 5.606689e-03 0.8616857 0.8449887
## 19 5.606689e-03 0.8616857 0.8449887 5.606689e-03 0.8616857 0.8449887
## 20 4.538380e-05 0.6545858 0.6121191 5.606689e-03 0.8616857 0.8449887
## 21 8.877607e-02 0.9791447 0.9778563 4.538380e-05 0.6545858 0.6121191
## 22 5.543901e-06 0.5211603 0.5146444 4.538380e-05 0.6545858 0.6121191
## 23 6.701998e-03 0.8751410 0.8631182 4.538380e-05 0.6545858 0.6121191
## 24 5.606689e-03 0.8616857 0.8449887 4.538380e-05 0.6545858 0.6121191
## 25 4.538380e-05 0.6545858 0.6121191 4.538380e-05 0.6545858 0.6121191

```

```
rc3 <- dynRB_VPa(Xrb3)
```

```

## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
## [1] 6
## [1] 7
## [1] 8
## [1] 9
## [1] 10
## [1] 11
## [1] 12
## [1] 13
## [1] 14
## [1] 15
## [1] 16
## [1] 17
## [1] 18
## [1] 19
## [1] 20
## [1] 21
## [1] 22
## [1] 23
## [1] 24
## [1] 25
##
## 1 hanging from branches and leaves
## 2 hanging from grass
## 3 laying on branches and leaves

```

```

## 4 laying on leaf litter, logs, bark, trunk
## 5         laying on palm leaves
## 6         hanging from branches and leaves
## 7         hanging from grass
## 8         laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10        laying on palm leaves
## 11        hanging from branches and leaves
## 12        hanging from grass
## 13        laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15        laying on palm leaves
## 16        hanging from branches and leaves
## 17        hanging from grass
## 18        laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20        laying on palm leaves
## 21        hanging from branches and leaves
## 22        hanging from grass
## 23        laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25        laying on palm leaves
##
##          V2      port_prod  vol_V1_prod
## 1         hanging from branches and leaves 1.000000e+00 0.1598451684
## 2         hanging from branches and leaves 1.681771e-04 0.0001222903
## 3         hanging from branches and leaves 3.748854e-02 0.0240751823
## 4         hanging from branches and leaves 1.658536e-02 0.0092887460
## 5         hanging from branches and leaves 2.585288e-03 0.0005412556
## 6         hanging from grass 2.455953e-01 0.1598451684
## 7         hanging from grass 1.000000e+00 0.0001222903
## 8         hanging from grass 7.442070e-02 0.0240751823
## 9         hanging from grass 7.164497e-04 0.0092887460
## 10        hanging from grass 0.000000e+00 0.0005412556
## 11        laying on branches and leaves 2.251331e-01 0.1598451684
## 12        laying on branches and leaves 3.220762e-04 0.0001222903
## 13        laying on branches and leaves 1.000000e+00 0.0240751823
## 14        laying on branches and leaves 1.396384e-01 0.0092887460
## 15        laying on branches and leaves 3.227660e-03 0.0005412556
## 16 laying on leaf litter, logs, bark, trunk 2.712031e-01 0.1598451684
## 17 laying on leaf litter, logs, bark, trunk 6.065730e-06 0.0001222903
## 18 laying on leaf litter, logs, bark, trunk 3.639053e-01 0.0240751823
## 19 laying on leaf litter, logs, bark, trunk 1.000000e+00 0.0092887460
## 20 laying on leaf litter, logs, bark, trunk 2.761694e-04 0.0005412556
## 21        laying on palm leaves 5.878160e-01 0.1598451684
## 22        laying on palm leaves 0.000000e+00 0.0001222903
## 23        laying on palm leaves 1.751566e-01 0.0240751823
## 24        laying on palm leaves 8.466414e-03 0.0092887460
## 25        laying on palm leaves 1.000000e+00 0.0005412556
##
##      vol_V2_prod
## 1 0.1598451684
## 2 0.1598451684
## 3 0.1598451684
## 4 0.1598451684
## 5 0.1598451684

```

```

## 6 0.0001222903
## 7 0.0001222903
## 8 0.0001222903
## 9 0.0001222903
## 10 0.0001222903
## 11 0.0240751823
## 12 0.0240751823
## 13 0.0240751823
## 14 0.0240751823
## 15 0.0240751823
## 16 0.0092887460
## 17 0.0092887460
## 18 0.0092887460
## 19 0.0092887460
## 20 0.0092887460
## 21 0.0005412556
## 22 0.0005412556
## 23 0.0005412556
## 24 0.0005412556
## 25 0.0005412556

```

```
rc3$result
```

```

##                               V1
## 1      hanging from branches and leaves
## 2                hanging from grass
## 3                laying on branches and leaves
## 4 laying on leaf litter, logs, bark, trunk
## 5                laying on palm leaves
## 6      hanging from branches and leaves
## 7                hanging from grass
## 8                laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10               laying on palm leaves
## 11      hanging from branches and leaves
## 12               hanging from grass
## 13               laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15               laying on palm leaves
## 16      hanging from branches and leaves
## 17               hanging from grass
## 18               laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20               laying on palm leaves
## 21      hanging from branches and leaves
## 22               hanging from grass
## 23               laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25               laying on palm leaves
##                               V2      port_prod port_mean port_gmean
## 1      hanging from branches and leaves 1.000000e+00 1.0000000 1.00000000
## 2      hanging from branches and leaves 1.681771e-04 0.2644282 0.17585220
## 3      hanging from branches and leaves 3.748854e-02 0.5902280 0.55012460
## 4      hanging from branches and leaves 1.658536e-02 0.5395237 0.44322494

```

```

## 5      hanging from branches and leaves 2.585288e-03 0.4302638 0.35592400
## 6              hanging from grass 2.455953e-01 0.7770595 0.57382187
## 7              hanging from grass 1.000000e+00 1.0000000 1.00000000
## 8              hanging from grass 7.442070e-02 0.5338428 0.45632362
## 9              hanging from grass 7.164497e-04 0.3707744 0.09240878
## 10             hanging from grass 0.000000e+00 0.5122723 0.00000000
## 11      laying on branches and leaves 2.251331e-01 0.7623472 0.72675461
## 12      laying on branches and leaves 3.220762e-04 0.2323698 0.18875065
## 13      laying on branches and leaves 1.000000e+00 1.0000000 1.00000000
## 14      laying on branches and leaves 1.396384e-01 0.7018413 0.70004036
## 15      laying on branches and leaves 3.227660e-03 0.4348383 0.34078969
## 16 laying on leaf litter, logs, bark, trunk 2.712031e-01 0.8020086 0.69366503
## 17 laying on leaf litter, logs, bark, trunk 6.065730e-06 0.1883817 0.04213101
## 18 laying on leaf litter, logs, bark, trunk 3.639053e-01 0.8256921 0.82582207
## 19 laying on leaf litter, logs, bark, trunk 1.000000e+00 1.0000000 1.00000000
## 20 laying on leaf litter, logs, bark, trunk 2.761694e-04 0.4269285 0.09729556
## 21      laying on palm leaves 5.878160e-01 0.9281819 0.85773836
## 22      laying on palm leaves 0.000000e+00 0.4123846 0.00000000
## 23      laying on palm leaves 1.751566e-01 0.7537908 0.63674571
## 24      laying on palm leaves 8.466414e-03 0.6269661 0.16974714
## 25      laying on palm leaves 1.000000e+00 1.0000000 1.00000000
##      vol_V1_prod vol_V1_mean vol_V1_gmean  vol_V2_prod vol_V2_mean vol_V2_gmean
## 1 0.1598451684 0.9825744 0.9814757 0.1598451684 0.9825744 0.9814757
## 2 0.0001222903 0.5218075 0.5039255 0.1598451684 0.9825744 0.9814757
## 3 0.0240751823 0.8735418 0.8677168 0.1598451684 0.9825744 0.9814757
## 4 0.0092887460 0.8183300 0.7917287 0.1598451684 0.9825744 0.9814757
## 5 0.0005412556 0.6320294 0.5979652 0.1598451684 0.9825744 0.9814757
## 6 0.1598451684 0.9825744 0.9814757 0.0001222903 0.5218075 0.5039255
## 7 0.0001222903 0.5218075 0.5039255 0.0001222903 0.5218075 0.5039255
## 8 0.0240751823 0.8735418 0.8677168 0.0001222903 0.5218075 0.5039255
## 9 0.0092887460 0.8183300 0.7917287 0.0001222903 0.5218075 0.5039255
## 10 0.0005412556 0.6320294 0.5979652 0.0001222903 0.5218075 0.5039255
## 11 0.1598451684 0.9825744 0.9814757 0.0240751823 0.8735418 0.8677168
## 12 0.0001222903 0.5218075 0.5039255 0.0240751823 0.8735418 0.8677168
## 13 0.0240751823 0.8735418 0.8677168 0.0240751823 0.8735418 0.8677168
## 14 0.0092887460 0.8183300 0.7917287 0.0240751823 0.8735418 0.8677168
## 15 0.0005412556 0.6320294 0.5979652 0.0240751823 0.8735418 0.8677168
## 16 0.1598451684 0.9825744 0.9814757 0.0092887460 0.8183300 0.7917287
## 17 0.0001222903 0.5218075 0.5039255 0.0092887460 0.8183300 0.7917287
## 18 0.0240751823 0.8735418 0.8677168 0.0092887460 0.8183300 0.7917287
## 19 0.0092887460 0.8183300 0.7917287 0.0092887460 0.8183300 0.7917287
## 20 0.0005412556 0.6320294 0.5979652 0.0092887460 0.8183300 0.7917287
## 21 0.1598451684 0.9825744 0.9814757 0.0005412556 0.6320294 0.5979652
## 22 0.0001222903 0.5218075 0.5039255 0.0005412556 0.6320294 0.5979652
## 23 0.0240751823 0.8735418 0.8677168 0.0005412556 0.6320294 0.5979652
## 24 0.0092887460 0.8183300 0.7917287 0.0005412556 0.6320294 0.5979652
## 25 0.0005412556 0.6320294 0.5979652 0.0005412556 0.6320294 0.5979652

```

```

result1 <- rc$result
result2 <- rc2$result
result3 <- rc3$result

```

```

vol <- as.data.frame(matrix(nrow = 5, ncol = 2))
colnames(vol) <- c("volume", "habitat")

```



```

vol$habitat <- c("hanging from branches and leaves", "hanging from grass",
  "laying on branches and leaves", "laying on leaf litter, logs, bark, trunk",
  "laying on palm leaves")

vol1 <- vol
vol2 <- vol
vol3 <- vol

# product
vol1[, 1] <- c(result1[1:5], 6)
vol2[, 1] <- c(result2[1:5], 6)
vol3[, 1] <- c(result3[1:5], 6)

p1 <- ggplot(vol1, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Niche hypervolume size") +
  # xlab('')+
coord_cartesian(ylim = c(-1e-04, 0.1)) + labs(title = "Mixed PCA - Size correction",
  subtitle = "Dynamic range boxes - Product") + scale_x_discrete(labels = NULL,
  breaks = NULL) + guides(fill = F) + theme_classic()

p2 <- ggplot(vol2, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Niche hypervolume size") +
  # xlab('')+
coord_cartesian(ylim = c(-1e-04, 0.1)) + labs(title = "Phylogenetic PCA - Size correction",
  subtitle = "Dynamic range boxes - Product") + scale_x_discrete(labels = NULL,
  breaks = NULL) + guides(fill = F) + theme_classic()

p3 <- ggplot(vol3, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Niche hypervolume size") +
  # xlab('')+
coord_cartesian(ylim = c(-1e-04, 0.1)) + labs(title = "Phylogenetic PCA - No size correction",
  subtitle = "Dynamic range boxes - Product") + scale_x_discrete(labels = NULL,
  breaks = NULL) + guides(fill = F) + theme_classic()

```

Using hypervolume package

```

# PCAmix - Size corrected

Xrb_litter = Xrb[Xrb$habitat == "laying on leaf litter, logs, bark, trunk",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]
Xrb_hang = Xrb[Xrb$habitat == "hanging from branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]
Xrb_branch = Xrb[Xrb$habitat == "laying on branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]
Xrb_grass = Xrb[Xrb$habitat == "hanging from grass", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]
Xrb_palm = Xrb[Xrb$habitat == "laying on palm leaves", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6", "PC7")]

```

```
hv1 = hypervolume_box(Xrb_litter, name = "litter")
```

```
##  
## Building tree...  
## done.  
## Ball query...  
##  
## done.  
## Binding random points... done.  
## Beginning volume calculation... done.
```

```
hv2 = hypervolume_box(Xrb_hang, name = "hang")
```

```
##  
## Building tree...  
## done.  
## Ball query...  
##  
## done.  
## Binding random points... done.  
## Beginning volume calculation... done.
```

```
hv3 = hypervolume_box(Xrb_branch, name = "branch")
```

```
##  
## Building tree...  
## done.  
## Ball query...  
##  
## done.  
## Binding random points... done.  
## Beginning volume calculation... done.
```

```
hv4 = hypervolume_box(Xrb_grass, name = "grass")
```

```
##  
## Building tree...  
## done.  
## Ball query...  
##  
## done.  
## Binding random points... done.  
## Beginning volume calculation... done.
```

```
hv5 = hypervolume_box(Xrb_palm, name = "palm")
```

```
##  
## Building tree...  
## done.
```

```

## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

values <- c(get_volume(hv2), get_volume(hv4), get_volume(hv3),
  get_volume(hv1), get_volume(hv5))
hyper <- data.frame(values, names(values))
colnames(hyper) <- c("volume", "habitat")
hyper$habitat <- factor(hyper$habitat, levels = hyper$habitat)

pl4 <- ggplot(hyper, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Niche hypervolume size") +
  # xlab('')+
labs(title = "Mixed PCA - Size correction", subtitle = "Kernel density estimates") +
  scale_x_discrete(labels = NULL, breaks = NULL) + guides(fill = F) +
  theme_classic()

# Phylogenetic PCA - Size corrected

Xrb2_litter = Xrb2[Xrb2$habitat == "laying on leaf litter, logs, bark, trunk",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6", "PC7", "PC8")]
Xrb2_hang = Xrb2[Xrb2$habitat == "hanging from branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6", "PC7", "PC8")]
Xrb2_branch = Xrb2[Xrb2$habitat == "laying on branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6", "PC7", "PC8")]
Xrb2_grass = Xrb2[Xrb2$habitat == "hanging from grass", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6", "PC7", "PC8")]
Xrb2_palm = Xrb2[Xrb2$habitat == "laying on palm leaves", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6", "PC7", "PC8")]

hv12 = hypervolume_box(Xrb2_litter, name = "litter")

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

hv22 = hypervolume_box(Xrb2_hang, name = "hang")

##
## Building tree...
## done.
## Ball query...
##
## done.

```

```
## Binding random points... done.  
## Beginning volume calculation... done.
```

```
hv32 = hypervolume_box(Xrb2_branch, name = "branch")
```

```
##  
## Building tree...  
## done.  
## Ball query...  
##  
## done.  
## Binding random points... done.  
## Beginning volume calculation... done.
```

```
hv42 = hypervolume_box(Xrb2_grass, name = "grass")
```

```
##  
## Building tree...  
## done.  
## Ball query...  
##  
## done.  
## Binding random points... done.  
## Beginning volume calculation... done.
```

```
hv52 = hypervolume_box(Xrb2_palm, name = "palm")
```

```
##  
## Building tree...  
## done.  
## Ball query...  
##  
## done.  
## Binding random points... done.  
## Beginning volume calculation... done.
```

```
values2 <- c(get_volume(hv22), get_volume(hv42), get_volume(hv32),  
            get_volume(hv12), get_volume(hv52))  
hyper2 <- data.frame(values2, names(values2))  
colnames(hyper2) <- c("volume", "habitat")  
hyper2$habitat <- factor(hyper2$habitat, levels = hyper2$habitat)
```

```
pl5 <- ggplot(hyper2, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),  
                    stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",  
                    "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Niche hypervolume size") +  
                    # xlab('')+
```

```
labs(title = "Phylogenetic PCA - Size correction", subtitle = "Kernel density estimates") +  
scale_x_discrete(labels = NULL, breaks = NULL) + guides(fill = F) +  
theme_classic()
```

```
# Phylogenetic PCA - No size correction
```

```

Xrb3_litter = Xrb3[Xrb3$habitat == "laying on leaf litter, logs, bark, trunk",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6")]
Xrb3_hang = Xrb3[Xrb3$habitat == "hanging from branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6")]
Xrb3_branch = Xrb3[Xrb3$habitat == "laying on branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6")]
Xrb3_grass = Xrb3[Xrb3$habitat == "hanging from grass", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6")]
Xrb3_palm = Xrb3[Xrb3$habitat == "laying on palm leaves", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6")]

hv13 = hypervolume_box(Xrb3_litter, name = "litter")

```

```

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

```

```

hv23 = hypervolume_box(Xrb3_hang, name = "hang")

```

```

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

```

```

hv33 = hypervolume_box(Xrb3_branch, name = "branch")

```

```

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

```

```

hv43 = hypervolume_box(Xrb3_grass, name = "grass")

```

```

##
## Building tree...
## done.
## Ball query...

```

```

##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

hv53 = hypervolume_box(Xrb3_palm, name = "palm")

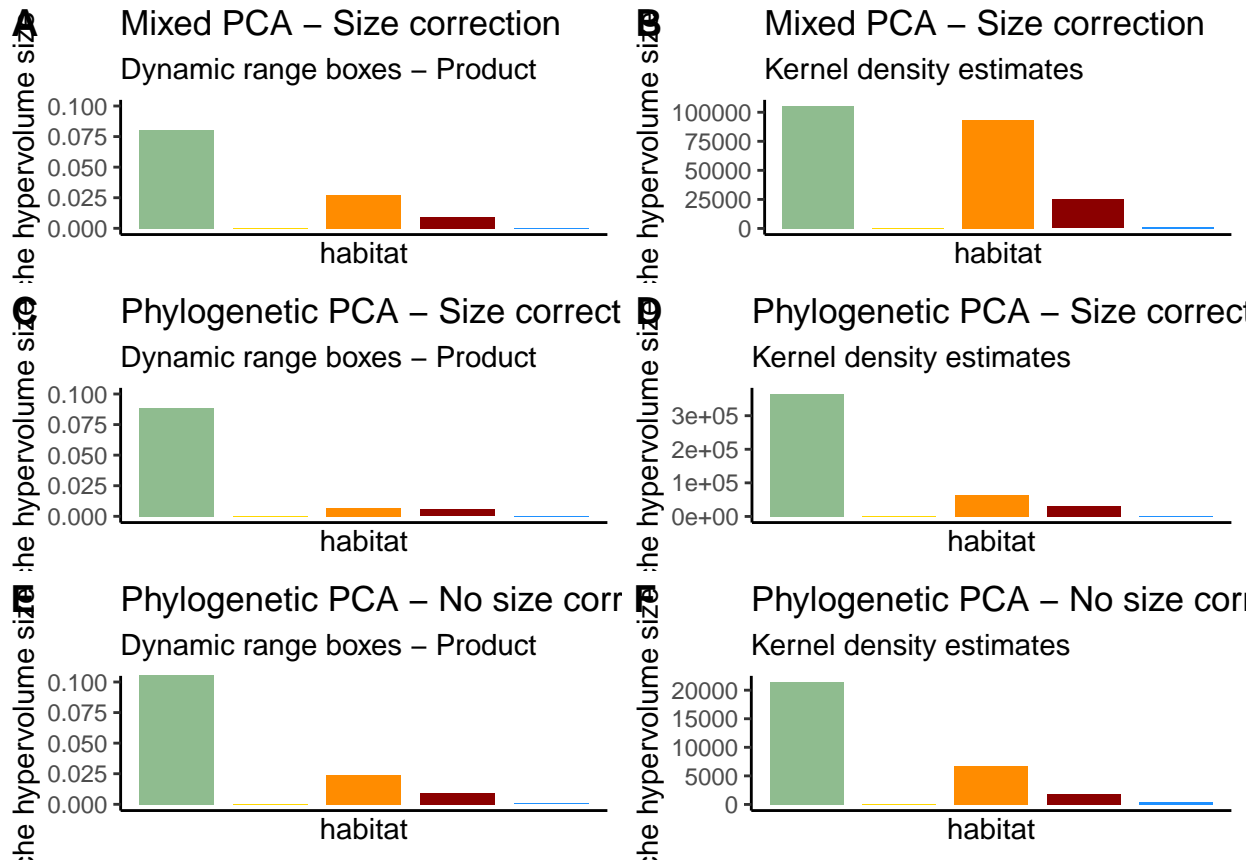
##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

values3 <- c(get_volume(hv23), get_volume(hv43), get_volume(hv33),
             get_volume(hv13), get_volume(hv53))
hyper3 <- data.frame(values3, names(values3))
colnames(hyper3) <- c("volume", "habitat")
hyper3$habitat <- factor(hyper3$habitat, levels = hyper3$habitat)

pl6 <- ggplot(hyper3, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
                    stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
                    "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Niche hypervolume size") +
                    # xlab('')+
labs(title = "Phylogenetic PCA - No size correction", subtitle = "Kernel density estimates") +
    scale_x_discrete(labels = NULL, breaks = NULL) + guides(fill = F) +
    theme_classic()

plot_grid(pl1, pl4, pl2, pl5, pl3, pl6, ncol = 2, labels = "AUTO")

```



Predicting habitat (Random forests - PCA mix + residuals)

Can habitat be predicted by morphology? tutorial: <https://towardsdatascience.com/random-forest-in-r-f66adf80ec9>

Prepare dataset

```
XX <- X[, -1]
XX <- XX[, -c(16:52)] #only keep morpho data and habitat use
XX <- subset(XX, is.na(XX$habitat2) == F)
XX$habitat2 <- as.factor(XX$habitat2)
```

```
head(XX)
```

```
##                pca_mix_dim1 pca_mix_dim2 pca_mix_dim3 pca_mix_dim4
## Acanthoxyla_geisovii_AXG1 -1.3759671  0.085671079  -1.0824179   1.6197692
## Clitarchus_hookeri_CLH1  -2.8740381  1.369276215   0.4360864  -0.4345218
## Argosarchus_horridus_ARG  -3.9956566  0.005241395  -1.2947459   0.5195014
## Micrararchus_hystriculeus_MIH  0.9150644 -0.267091036  -1.7022573   0.4727771
## Canachus_alligator_CAN1    4.1915667  1.030433904  -0.6096281   1.5343849
## Canachus_sp._CAN2         4.6060583  1.649304906  -0.3693754   1.6917260
##                pca_mix_dim5 pca_mix_dim6 pca_mix_dim7 pca_mix_dim8
```

```

## Acanthoxyla_geisovii_AXG1      -0.7840636   -0.7307814   -0.5069875   -0.7024527
## Clitarchus_hookeri_CLH1         0.7347754    0.1038844    0.2192024    0.4229190
## Argosarchus_horridus_ARG        -3.0610214   -0.6666210    0.8030136   -0.7137576
## Micrarchus_hystriculeus_MIH     -1.8435060   -0.2895571   -0.1021009   -0.2832688
## Canachus_alligator_CAN1         0.4317944    1.0413946   -0.2469166    0.1280258
## Canachus_sp._CAN2               1.2429044    1.2574744    0.1902168    0.1216815
##                                pca_mix_dim9  pca_mix_dim10  pca_mix_dim11
## Acanthoxyla_geisovii_AXG1     -0.9431956   -0.18055995    0.3234469
## Clitarchus_hookeri_CLH1       -0.3051649   -0.43393416   -0.2500117
## Argosarchus_horridus_ARG       -0.1101320   -0.52138089    0.3056898
## Micrarchus_hystriculeus_MIH   -0.5014256    0.24502244    0.4049826
## Canachus_alligator_CAN1        0.3904848    0.03806992   -0.1937711
## Canachus_sp._CAN2             0.9825112    0.41346075   -0.2189815
##                                pca_mix_dim12  pca_mix_dim13  pca_mix_dim14
## Acanthoxyla_geisovii_AXG1     -0.3701815   -0.1146489   -0.245664320
## Clitarchus_hookeri_CLH1       -0.4065280    0.7680544   -0.284338664
## Argosarchus_horridus_ARG       -0.6794614   -0.1720579    0.177054919
## Micrarchus_hystriculeus_MIH    0.0448744   -0.1112182   -0.001549396
## Canachus_alligator_CAN1       -0.7539665   -0.2374013   -0.247930234
## Canachus_sp._CAN2             -0.4920160    0.0377261   -0.136942420
##                                pca_mix_dim15
## Acanthoxyla_geisovii_AXG1     -0.2092843
## Clitarchus_hookeri_CLH1       -0.3350510
## Argosarchus_horridus_ARG       -0.4757940
## Micrarchus_hystriculeus_MIH    0.3703019
## Canachus_alligator_CAN1       -0.1688248
## Canachus_sp._CAN2             -0.2384600
##                                habitat2
## Acanthoxyla_geisovii_AXG1      hanging from branches and leaves
## Clitarchus_hookeri_CLH1         hanging from branches and leaves
## Argosarchus_horridus_ARG        hanging from branches and leaves
## Micrarchus_hystriculeus_MIH     laying on leaf litter, logs, bark, trunk
## Canachus_alligator_CAN1         laying on leaf litter, logs, bark, trunk
## Canachus_sp._CAN2               laying on leaf litter, logs, bark, trunk

```

```
##Run randomforest
```

Null accuracy

Predict the most frequent habitat in the training set for the test set.

```

a <- NULL
acc <- NULL

# Define training dataset with always the same proportions
# of species from each habitat
trainIndex1 <- createDataPartition(XX$habitat2, p = 3/4, list = FALSE,
  times = 1)
train <- XX[trainIndex1, ]
# Test dataset
test <- subset(XX, !(rownames(XX) %in% rownames(train)))
# Most frequent habitat in test dataset
h <- names(which.max(table(train$habitat2)))

```



```

# Prediction = always the most frequent habitat
pred <- rep(h, length(test$habitat2))
pred <- as.factor(pred)
test2 <- test[, dim(test)[2]]
cm <- table(pred, test2)
cm2 <- prop.table(cm, 1) * 100
null <- cm2[2]
null

```

```
## [1] 53.06122
```

Training and prediction with morphological data

Adding one PC at a time (repeat each step 1000 times)

```

r <- vector(mode = "list", length = 15)
aa <- vector(mode = "list", length = 15)
racc <- NULL
a <- NULL
for (i in c(1:15)) {
  dat <- XX[, c(1:i, 16)]
  dat$habitat2 <- as.factor(dat$habitat2)
  acc <- NULL
  if (i == 1) {
    mtry = 1
  }
  if (i != 1) {
    mtry = 2
  }
  for (j in c(1:1000)) {
    trainIndex1 <- createDataPartition(XX$habitat2, p = 3/4,
      list = FALSE, times = 1)
    train <- dat[trainIndex1, ]
    test <- subset(dat, !(rownames(dat) %in% rownames(train)))
    test2 <- test[, dim(test)[2]]
    rf <- randomForest(habitat2 ~ ., data = train, mtry = mtry,
      maxnodes = 50)
    pred = as.factor(predict(rf, newdata = test))
    cm = table(pred, test2)
    a <- rbind(a, prop.table(cm, 1) * 100)
    acc <- cbind(acc, mean(pred == test2))
  }
  aa[[i]] <- a
  r[[i]] <- acc #store all accuracy results
  racc <- cbind(racc, mean(acc)) #mean accuracy results
  a <- NULL
}

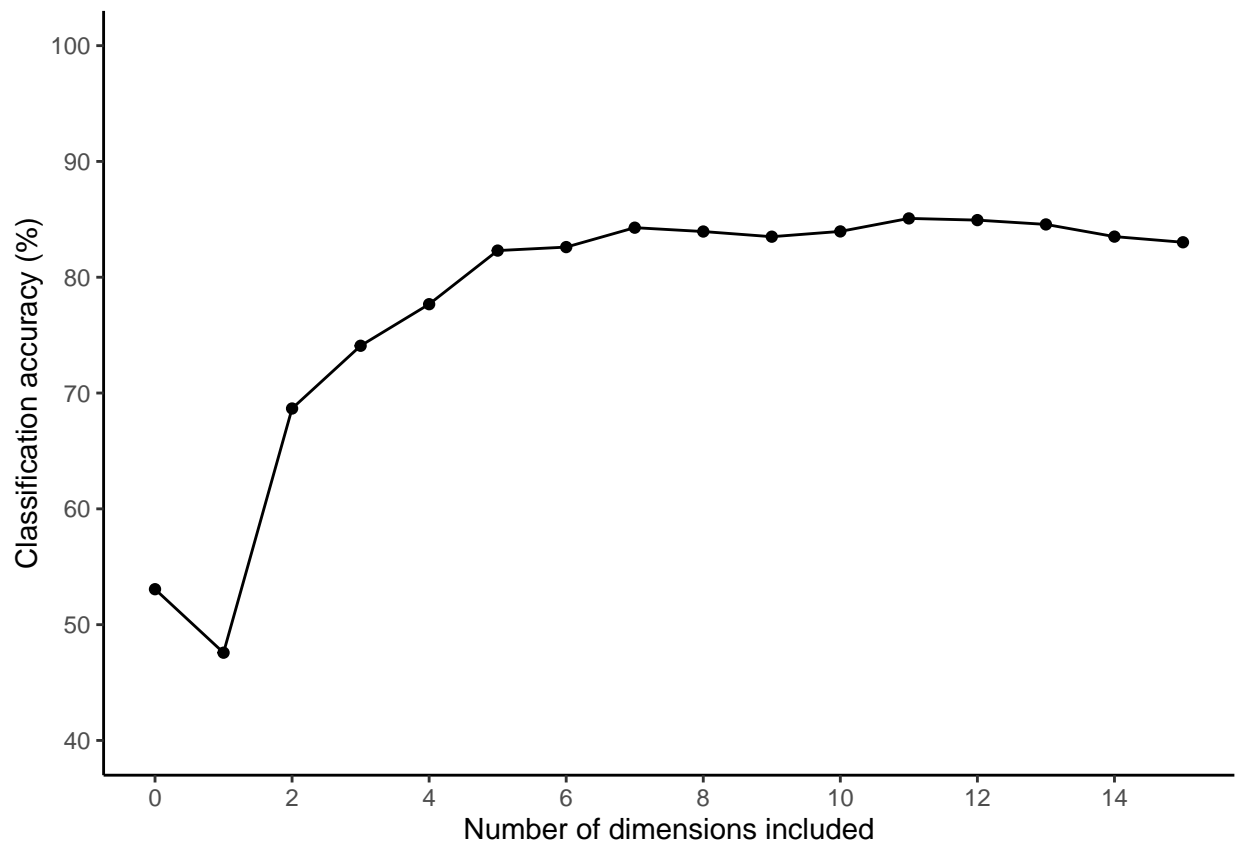
```

Plot overall prediction accuracy

```

accu <- cbind(c(1:15), as.vector(racc * 100))
colnames(accu) <- c("var", "accur")
accu <- as.data.frame(accu)
accu <- rbind(accu, c(0, null))
ggplot(accu, aes(x = var, y = accur)) + geom_point() + geom_line() +
  scale_x_continuous(name = "Number of dimensions included",
    limits = c(0, 15), breaks = seq(0, 15, 2)) + scale_y_continuous(name = "Classification accuracy",
    limits = c(40, 100), breaks = seq(40, 100, 10)) + theme_classic()

```



Prediction accuracy for 2 dimensions

```

d <- aa[[2]]
mean2 <- function(x) {
  mean(x, na.rm = T)
}
b <- aggregate(d, list(row.names(d)), mean2)
bb <- melt(b)
bb$variable <- factor(bb$variable, levels = c("laying on leaf litter, logs, bark, trunk",
  "hanging from grass", "laying on branches and leaves", "hanging from branches and leaves",
  "laying on palm leaves"))
bb$Group.1 <- factor(bb$Group.1, levels = c("laying on leaf litter, logs, bark, trunk",
  "hanging from grass", "laying on branches and leaves", "hanging from branches and leaves",
  "laying on palm leaves"))

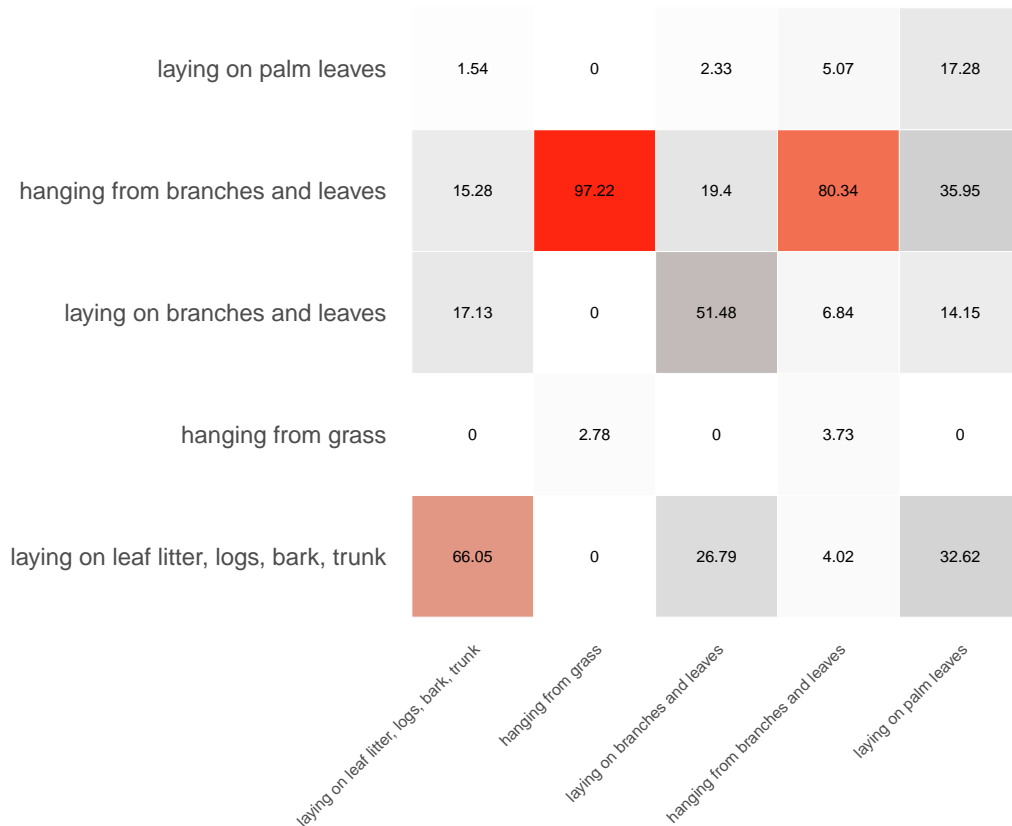
```

```

bb <- bb %>%
  arrange(Group.1, variable)

p2 <- ggplot(data = bb, aes(x = Group.1, y = variable, fill = value)) +
  geom_tile(color = "white") + scale_fill_gradient2(low = "white",
  high = "red", mid = "grey", midpoint = 50, limit = c(0, 100),
  space = "Lab", name = "% of classified species") + theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, vjust = 1, size = 6,
  hjust = 1)) + coord_fixed() + geom_text(aes(x = Group.1,
  y = variable, label = round(value, digits = 2)), color = "black",
  size = 2) + theme(axis.title.x = element_blank(), axis.title.y = element_blank(),
  panel.grid.major = element_blank(), panel.border = element_blank(),
  panel.background = element_blank(), axis.ticks = element_blank(),
  legend.position = "none")
p2

```



```

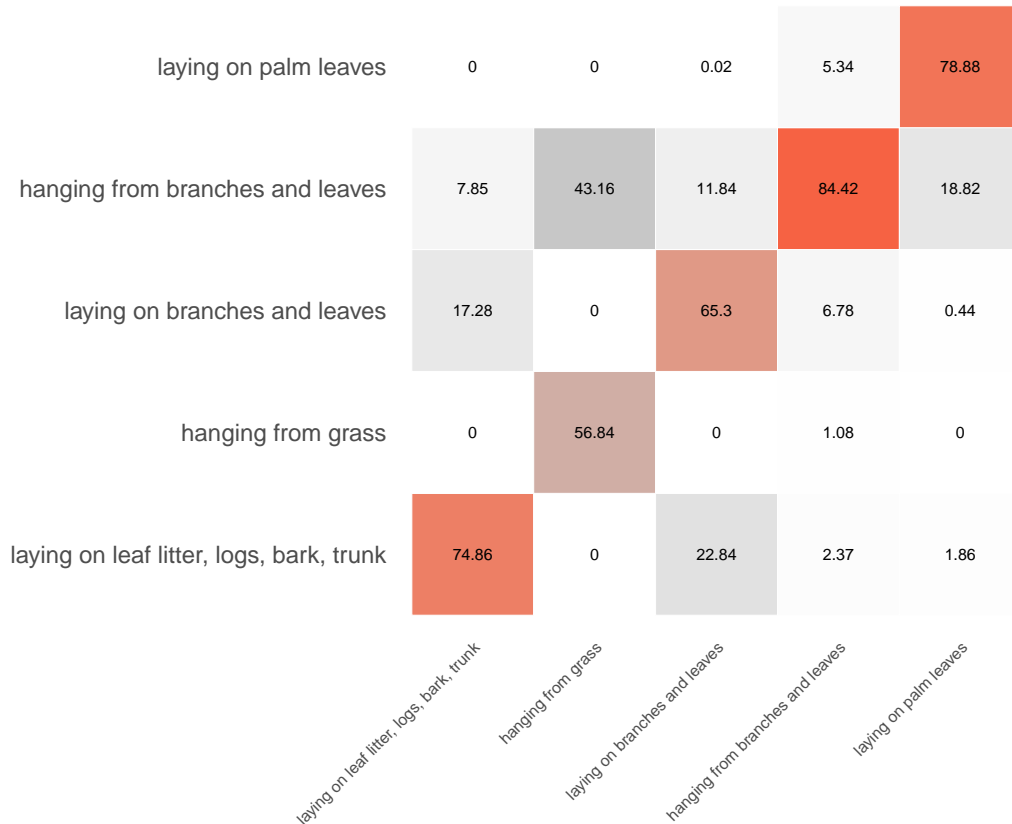
p22 <- ggplot(data = bb, aes(x = Group.1, y = variable, fill = value)) +
  geom_tile(color = "white") + scale_fill_gradient2(low = "white",
  high = "red", mid = "grey", midpoint = 50, limit = c(0, 100),
  space = "Lab", name = "% of classified species") + theme_minimal() +
  labs(title = "2 dimensions") + theme(axis.text.x = element_text(angle = 45,
  vjust = 1, size = 6, hjust = 1)) + coord_fixed() + theme(axis.title.x = element_blank(),
  axis.title.y = element_blank(), panel.grid.major = element_blank(),
  panel.border = element_blank(), panel.background = element_blank(),
  axis.ticks = element_blank(), legend.position = "none")

```

Prediction accuracy for 3 dimensions

```
d <- aa[[4]]
b <- aggregate(d, list(row.names(d)), mean2)
bb <- melt(b)
bb$variable <- factor(bb$variable, levels = c("laying on leaf litter, logs, bark, trunk",
      "hanging from grass", "laying on branches and leaves", "hanging from branches and leaves",
      "laying on palm leaves"))
bb$Group.1 <- factor(bb$Group.1, levels = c("laying on leaf litter, logs, bark, trunk",
      "hanging from grass", "laying on branches and leaves", "hanging from branches and leaves",
      "laying on palm leaves"))
bb <- bb %>%
  arrange(Group.1, variable)

p4 <- ggplot(data = bb, aes(x = Group.1, y = variable, fill = value)) +
  geom_tile(color = "white") + scale_fill_gradient2(low = "white",
    high = "red", mid = "grey", midpoint = 50, limit = c(0, 100),
    space = "Lab", name = "% of classified species") + theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, vjust = 1, size = 6,
    hjust = 1)) + coord_fixed() + geom_text(aes(x = Group.1,
    y = variable, label = round(value, digits = 2)), color = "black",
    size = 2) + theme(axis.title.x = element_blank(), axis.title.y = element_blank(),
    panel.grid.major = element_blank(), panel.border = element_blank(),
    panel.background = element_blank(), axis.ticks = element_blank(),
    legend.position = "none")
p4
```



Prediction accuracy for 7 dimensions

```

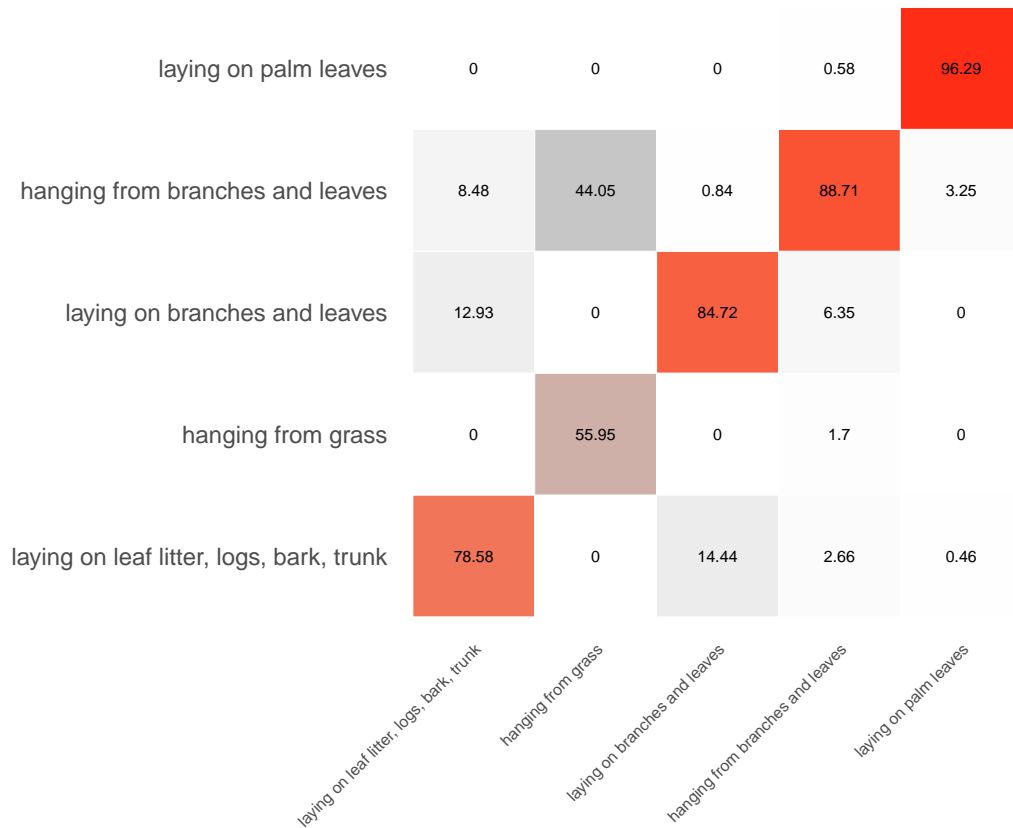
d <- aa[[7]]
b <- aggregate(d, list(row.names(d)), mean2)
bb <- melt(b)
bb$variable <- factor(bb$variable, levels = c("laying on leaf litter, logs, bark, trunk",
      "hanging from grass", "laying on branches and leaves", "hanging from branches and leaves",
      "laying on palm leaves"))
bb$Group.1 <- factor(bb$Group.1, levels = c("laying on leaf litter, logs, bark, trunk",
      "hanging from grass", "laying on branches and leaves", "hanging from branches and leaves",
      "laying on palm leaves"))
bb <- bb %>%
  arrange(Group.1, variable)

p7 <- ggplot(data = bb, aes(x = Group.1, y = variable, fill = value)) +
  geom_tile(color = "white") + scale_fill_gradient2(low = "white",
    high = "red", mid = "grey", midpoint = 50, limit = c(0, 100),
    space = "Lab", name = "% of classified species") + theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, vjust = 1, size = 6,
    hjust = 1)) + coord_fixed() + geom_text(aes(x = Group.1,
    y = variable, label = round(value, digits = 2)), color = "black",
    size = 2) + theme(axis.title.x = element_blank(), axis.title.y = element_blank(),
    panel.grid.major = element_blank(), panel.border = element_blank(),
    panel.background = element_blank(), axis.ticks = element_blank(),

```

```
legend.position = "none")
```

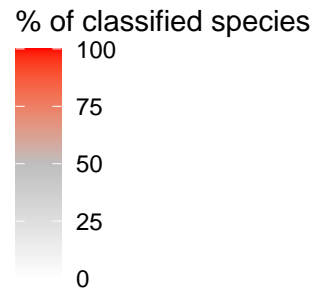
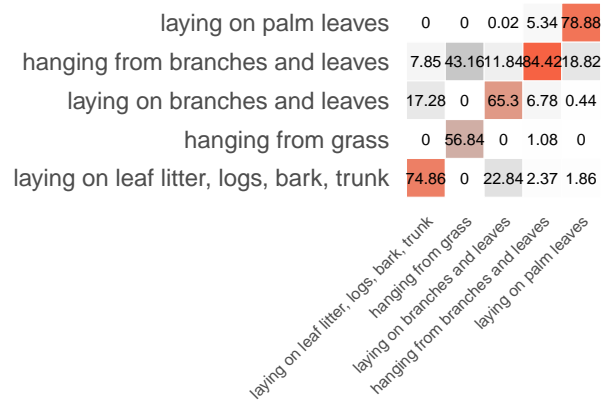
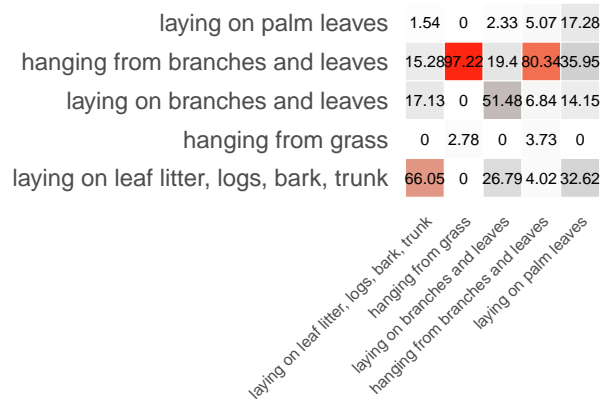
p7



```
p77 <- ggplot(data = bb, aes(x = Group.1, y = variable, fill = value)) +
  geom_tile(color = "white") + scale_fill_gradient2(low = "white",
  high = "red", mid = "grey", midpoint = 50, limit = c(0, 100),
  space = "Lab", name = "% of classified species") + theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, vjust = 1, size = 6,
  hjust = 1)) + labs(title = "7 dimensions") + coord_fixed() +
  theme(axis.title.x = element_blank(), axis.title.y = element_blank(),
  panel.grid.major = element_blank(), panel.border = element_blank(),
  panel.background = element_blank(), axis.ticks = element_blank(),
  legend.position = "none")
```

```
ppp <- ggplot(data = bb, aes(x = Group.1, y = variable, fill = value)) +
  geom_tile(color = "white") + scale_fill_gradient2(low = "white",
  high = "red", mid = "grey", midpoint = 50, limit = c(0, 100),
  space = "Lab", name = "% of classified species") + theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, vjust = 1, size = 6,
  hjust = 1)) + coord_fixed() + theme(axis.title.x = element_blank(),
  axis.title.y = element_blank(), panel.grid.major = element_blank(),
  panel.border = element_blank(), panel.background = element_blank(),
  axis.ticks = element_blank())
legend <- cowplot::get_legend(ppp)
```

```
l <- as_ggplot(legend)
plot_grid(p2, p4, p7, 1, ncol = 2)
```



```
# plot for figure
plot_grid(p22, p77, 1, nrow = 1)
```



Predicting habitat (Random forests - phylogenetic PCA + residuals)

Prepare dataset

```
XX <- X[, -1]
XX <- XX[, -c(1:15, 31:52)] #only keep morpho data and habitat use
XX <- subset(XX, is.na(XX$habitat2) == F)
XX$habitat2 <- as.factor(XX$habitat2)
```

```
head(XX)
```

```
##                pPCA_sizecorr_dim1 pPCA_sizecorr_dim2
## Acanthoxyla_geisovii_AXG1         4.5965581        -0.9060956
## Clitarchus_hookeri_CLH1           5.6814708        -1.6982495
## Argosarchus_horridus_ARG          7.3635322         0.8738964
## Micrarchus_hystriculeus_MIH        2.3296265         0.9535926
## Canachus_alligator_CAN1          -0.9045619        -0.9779065
## Canachus_sp._CAN2                -1.3483295        -1.4074393
##                pPCA_sizecorr_dim3 pPCA_sizecorr_dim4
## Acanthoxyla_geisovii_AXG1         0.8293949        -1.4687452
## Clitarchus_hookeri_CLH1           1.5277214         0.7039254
```


## Argosarchus_horridus_ARG	1.2545787	-0.9043370
## Micrarchus_hystriculeus_MIH	0.6438192	-1.0963965
## Canachus_alligator_CAN1	0.5615505	0.6987421
## Canachus_sp._CAN2	0.8543314	1.5526521
##	pPCA_sizecorr_dim5	pPCA_sizecorr_dim6
## Acanthoxyla_geisovii_AXG1	0.3568420	0.6291487
## Clitarchus_hookeri_CLH1	0.2780851	0.7978928
## Argosarchus_horridus_ARG	1.9344974	1.2439369
## Micrarchus_hystriculeus_MIH	0.9093774	0.9242074
## Canachus_alligator_CAN1	-1.2264691	1.1462574
## Canachus_sp._CAN2	-1.8703784	0.8670012
##	pPCA_sizecorr_dim7	pPCA_sizecorr_dim8
## Acanthoxyla_geisovii_AXG1	1.1675522	-0.1531384
## Clitarchus_hookeri_CLH1	0.6326144	-0.5988136
## Argosarchus_horridus_ARG	0.3868782	-0.2810615
## Micrarchus_hystriculeus_MIH	0.8438302	-0.2567734
## Canachus_alligator_CAN1	1.0873844	-1.1727578
## Canachus_sp._CAN2	0.6937075	-1.2800386
##	pPCA_sizecorr_dim9	pPCA_sizecorr_dim10
## Acanthoxyla_geisovii_AXG1	0.18208679	-0.001994741
## Clitarchus_hookeri_CLH1	-0.11390874	-0.240364319
## Argosarchus_horridus_ARG	0.32395784	-0.927072097
## Micrarchus_hystriculeus_MIH	-0.09634414	0.398199140
## Canachus_alligator_CAN1	-0.08718907	-0.448264921
## Canachus_sp._CAN2	-0.42310522	-0.335930235
##	pPCA_sizecorr_dim11	pPCA_sizecorr_dim12
## Acanthoxyla_geisovii_AXG1	-0.04498002	0.03350272
## Clitarchus_hookeri_CLH1	0.45783251	0.59178603
## Argosarchus_horridus_ARG	-0.18201816	-0.01935252
## Micrarchus_hystriculeus_MIH	-0.25486309	-0.01741955
## Canachus_alligator_CAN1	-0.03944657	-0.50324673
## Canachus_sp._CAN2	0.18174984	-0.47864899
##	pPCA_sizecorr_dim13	pPCA_sizecorr_dim14
## Acanthoxyla_geisovii_AXG1	0.11396619	-0.0276691484
## Clitarchus_hookeri_CLH1	0.03104352	0.1022408958
## Argosarchus_horridus_ARG	0.32195802	-0.0003249141
## Micrarchus_hystriculeus_MIH	0.55626523	0.0099649626
## Canachus_alligator_CAN1	-0.22006626	-0.1982651632
## Canachus_sp._CAN2	-0.12493447	0.0708690988
##	pPCA_sizecorr_dim15	
## Acanthoxyla_geisovii_AXG1	-0.06812651	
## Clitarchus_hookeri_CLH1	-0.33770374	
## Argosarchus_horridus_ARG	-0.57951647	
## Micrarchus_hystriculeus_MIH	-0.18791475	
## Canachus_alligator_CAN1	-0.18591290	
## Canachus_sp._CAN2	-0.26687417	
##		habitat2
## Acanthoxyla_geisovii_AXG1	hanging from branches and leaves	
## Clitarchus_hookeri_CLH1	hanging from branches and leaves	
## Argosarchus_horridus_ARG	hanging from branches and leaves	
## Micrarchus_hystriculeus_MIH	laying on leaf litter, logs, bark, trunk	
## Canachus_alligator_CAN1	laying on leaf litter, logs, bark, trunk	
## Canachus_sp._CAN2	laying on leaf litter, logs, bark, trunk	

```
##Run randomforest ### Null accuracy Predict the most frequent habitat in the training set for the test set.
```

```
a <- NULL
acc <- NULL

# Define training dataset with always the same proportions
# of species from each habitat
trainIndex1 <- createDataPartition(XX$habitat2, p = 3/4, list = FALSE,
  times = 1)
train <- XX[trainIndex1, ]
# Test dataset
test <- subset(XX, !(rownames(XX) %in% rownames(train)))
# Most frequent habitat in test dataset
h <- names(which.max(table(train$habitat2)))
# Prediction = always the most frequent habitat
pred <- rep(h, length(test$habitat2))
pred <- as.factor(pred)
test2 <- test[, dim(test)[2]]
cm <- table(pred, test2)
cm2 <- prop.table(cm, 1) * 100
null <- cm2[2]
null
```

```
## [1] 53.06122
```

Training and prediction with morphological data

Adding one PC at a time (repeat each step 1000 times)

```
r <- vector(mode = "list", length = 15)
aa <- vector(mode = "list", length = 15)
racc <- NULL
a <- NULL
for (i in c(1:15)) {
  dat <- XX[, c(1:i, 16)]
  dat$habitat2 <- as.factor(dat$habitat2)
  acc <- NULL
  if (i == 1) {
    mtry = 1
  }
  if (i != 1) {
    mtry = 2
  }
  for (j in c(1:1000)) {
    trainIndex1 <- createDataPartition(XX$habitat2, p = 3/4,
      list = FALSE, times = 1)
    train <- dat[trainIndex1, ]
    test <- subset(dat, !(rownames(dat) %in% rownames(train)))
    test2 <- test[, dim(test)[2]]
    rf <- randomForest(habitat2 ~ ., data = train, mtry = mtry,
      maxnodes = 50)
    pred = as.factor(predict(rf, newdata = test))
```

```

    cm = table(pred, test2)
    a <- rbind(a, prop.table(cm, 1) * 100)
    acc <- cbind(acc, mean(pred == test2))
  }
  aa[[i]] <- a
  r[[i]] <- acc #store all accuracy results
  racc <- cbind(racc, mean(acc)) #mean accuracy results
  a <- NULL
}

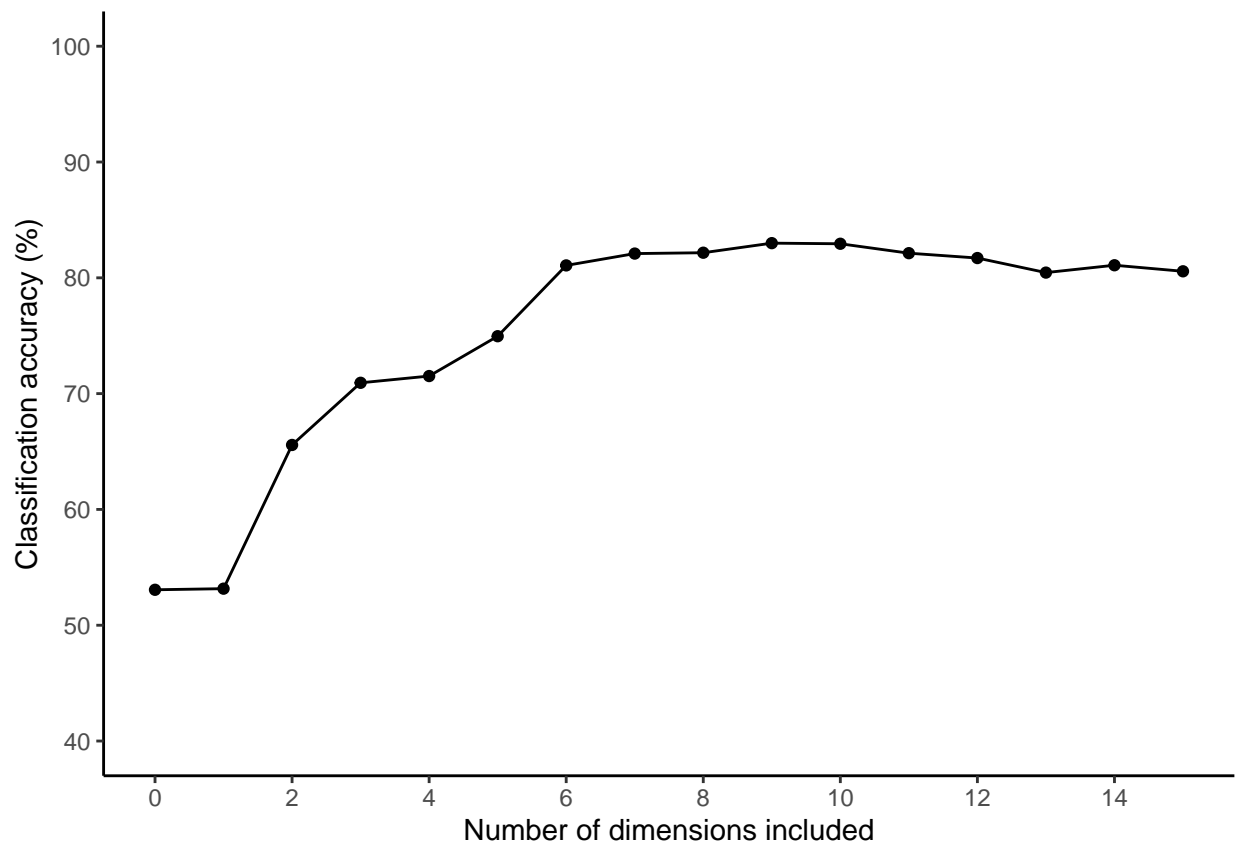
```

Plot overall prediction accuracy

```

accu <- cbind(c(1:15), as.vector(racc * 100))
colnames(accu) <- c("var", "accur")
accu <- as.data.frame(accu)
accu <- rbind(accu, c(0, null))
ggplot(accu, aes(x = var, y = accur)) + geom_point() + geom_line() +
  scale_x_continuous(name = "Number of dimensions included",
    limits = c(0, 15), breaks = seq(0, 15, 2)) + scale_y_continuous(name = "Classification accuracy",
    limits = c(40, 100), breaks = seq(40, 100, 10)) + theme_classic()

```



Predicting habitat (Random forests - PCA mix + residuals)

Prepare dataset

```
XX <- X[, -1]
XX <- XX[, -c(1:30, 46:52)] #only keep morpho data and habitat use
XX <- subset(XX, is.na(XX$habitat2) == F)
XX$habitat2 <- as.factor(XX$habitat2)
```

```
head(XX)
```

```
##                pPCA_nocorr_dim1 pPCA_nocorr_dim2 pPCA_nocorr_dim3
## Acanthoxyla_geisovii_AXG1      -1.5414094      3.20842617      0.94624934
## Clitarchus_hookeri_CLH1         0.2617767      4.13404914      1.00254647
## Argosarchus_horridus_ARG       -1.9151226      5.20233387      1.68709400
## Micrarchus_hystriculeus_MIH     1.1451876      1.43071581      0.76552178
## Canachus_alligator_CAN1        -5.3362217      0.06605149      0.01476928
## Canachus_sp._CAN2              -4.5640696     -0.01438301     -0.22327286
##                pPCA_nocorr_dim4 pPCA_nocorr_dim5 pPCA_nocorr_dim6
## Acanthoxyla_geisovii_AXG1      1.2365590     -0.07485868      0.06737058
## Clitarchus_hookeri_CLH1         1.3409043     -0.22156799      0.33548288
## Argosarchus_horridus_ARG         0.7398764     -0.44336709      0.93946703
## Micrarchus_hystriculeus_MIH     0.1684776     -0.05225778      0.46512019
## Canachus_alligator_CAN1         0.4822347      0.13929531     -0.58727026
## Canachus_sp._CAN2               0.4178866      0.38068404     -0.73466778
##                pPCA_nocorr_dim7 pPCA_nocorr_dim8 pPCA_nocorr_dim9
## Acanthoxyla_geisovii_AXG1      0.8176705     -0.1255503     -0.41020756
## Clitarchus_hookeri_CLH1         0.4052168      0.3392119      0.27337845
## Argosarchus_horridus_ARG         0.4008504     -0.1396918     -0.05718353
## Micrarchus_hystriculeus_MIH     0.5729816      0.2410431     -0.68546292
## Canachus_alligator_CAN1         0.7339198      0.8667493      0.21085913
## Canachus_sp._CAN2               0.3562099      1.0768221      0.25460004
##                pPCA_nocorr_dim10 pPCA_nocorr_dim11
## Acanthoxyla_geisovii_AXG1      -0.37561675     -0.009681851
## Clitarchus_hookeri_CLH1        -0.29714039      0.205537860
## Argosarchus_horridus_ARG       -0.65896537     -0.399095170
## Micrarchus_hystriculeus_MIH    -0.16648291     -0.420959563
## Canachus_alligator_CAN1        -0.20142038      0.219005111
## Canachus_sp._CAN2              0.01443039      0.373028722
##                pPCA_nocorr_dim12 pPCA_nocorr_dim13
## Acanthoxyla_geisovii_AXG1      0.0580434945     -0.004448451
## Clitarchus_hookeri_CLH1        -0.1880016376     -0.187155253
## Argosarchus_horridus_ARG         0.3756427604     -0.489982351
## Micrarchus_hystriculeus_MIH     0.0007353492     -0.387099785
## Canachus_alligator_CAN1         0.2122434917     -0.247554869
## Canachus_sp._CAN2              0.2239732486     -0.383121375
##                pPCA_nocorr_dim14 pPCA_nocorr_dim15
## Acanthoxyla_geisovii_AXG1      0.06315413      0.097492751
## Clitarchus_hookeri_CLH1         0.21265022     -0.224353459
## Argosarchus_horridus_ARG         0.22518677     -0.123065972
## Micrarchus_hystriculeus_MIH     0.23001976      0.002747999
## Canachus_alligator_CAN1        -0.05136293     -0.046606515
```

```
## Canachus_sp._CAN2          0.09542549      -0.125398675
##                               habitat2
## Acanthoxyla_geisovii_AXG1    hanging from branches and leaves
## Clitarchus_hookeri_CLH1      hanging from branches and leaves
## Argosarchus_horridus_ARG     hanging from branches and leaves
## Micrarchus_hystriculeus_MIH  laying on leaf litter, logs, bark, trunk
## Canachus_alligator_CAN1      laying on leaf litter, logs, bark, trunk
## Canachus_sp._CAN2           laying on leaf litter, logs, bark, trunk
```

##Run randomforest ### Null accuracy Predict the most frequent habitat in the training set for the test set.

```
a <- NULL
acc <- NULL

# Define training dataset with always the same proportions
# of species from each habitat
trainIndex1 <- createDataPartition(XX$habitat2, p = 3/4, list = FALSE,
  times = 1)
train <- XX[trainIndex1, ]
# Test dataset
test <- subset(XX, !(rownames(XX) %in% rownames(train)))
# Most frequent habitat in test dataset
h <- names(which.max(table(train$habitat2)))
# Prediction = always the most frequent habitat
pred <- rep(h, length(test$habitat2))
pred <- as.factor(pred)
test2 <- test[, dim(test)[2]]
cm <- table(pred, test2)
cm2 <- prop.table(cm, 1) * 100
null <- cm2[2]
null
```

```
## [1] 53.06122
```

Training and prediction with morphological data

Adding one PC at a time (repeat each step 1000 times)

```
r <- vector(mode = "list", length = 15)
aa <- vector(mode = "list", length = 15)
racc <- NULL
a <- NULL
for (i in c(1:15)) {
  dat <- XX[, c(1:i, 16)]
  dat$habitat2 <- as.factor(dat$habitat2)
  acc <- NULL
  if (i == 1) {
    mtry = 1
  }
  if (i != 1) {
    mtry = 2
  }
}
```

```

for (j in c(1:1000)) {
  trainIndex1 <- createDataPartition(XX$habitat2, p = 3/4,
    list = FALSE, times = 1)
  train <- dat[trainIndex1, ]
  test <- subset(dat, !(rownames(dat) %in% rownames(train)))
  test2 <- test[, dim(test)[2]]
  rf <- randomForest(habitat2 ~ ., data = train, mtry = mtry,
    maxnodes = 50)
  pred = as.factor(predict(rf, newdata = test))
  cm = table(pred, test2)
  a <- rbind(a, prop.table(cm, 1) * 100)
  acc <- cbind(acc, mean(pred == test2))
}
aa[[i]] <- a
r[[i]] <- acc #store all accuracy results
racc <- cbind(racc, mean(acc)) #mean accuracy results
a <- NULL
}

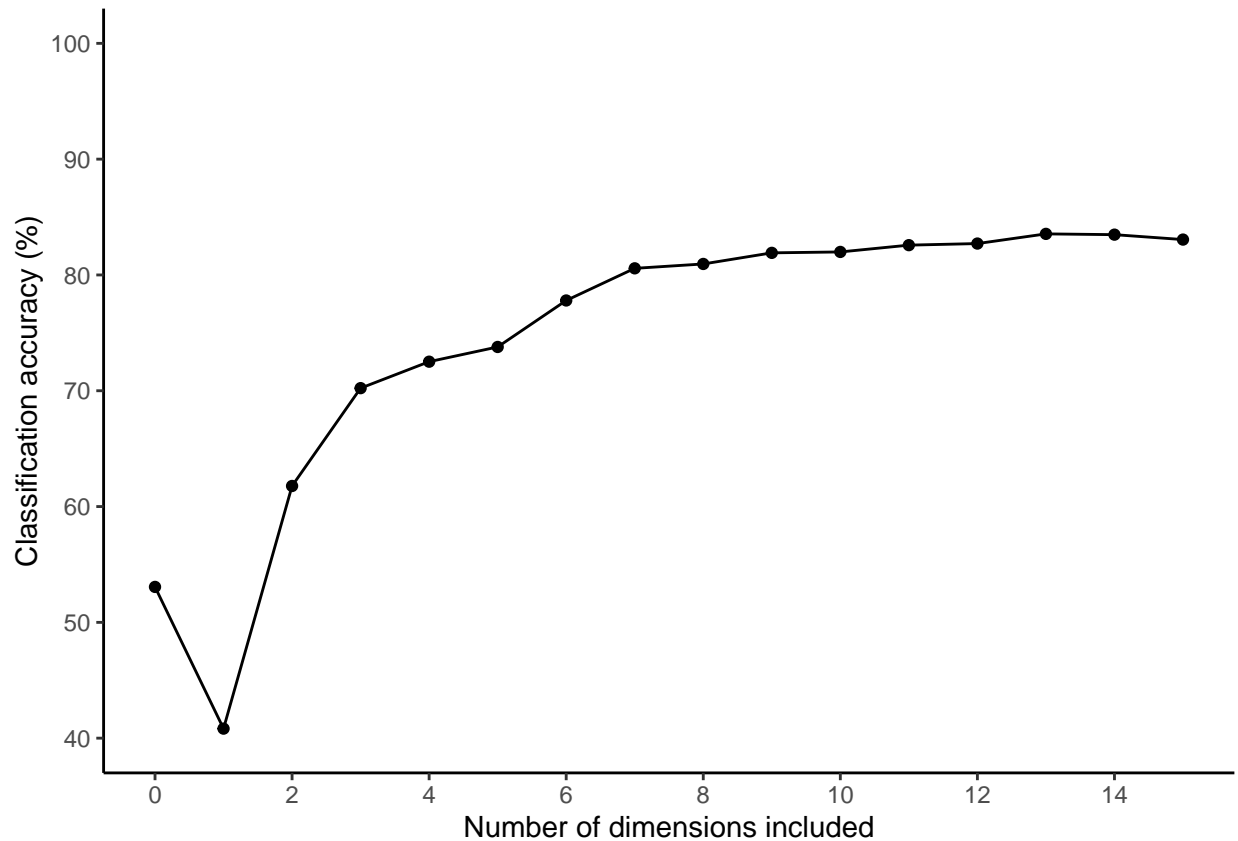
```

Plot overall prediction accuracy

```

accu <- cbind(c(1:15), as.vector(racc * 100))
colnames(accu) <- c("var", "accur")
accu <- as.data.frame(accu)
accu <- rbind(accu, c(0, null))
ggplot(accu, aes(x = var, y = accur)) + geom_point() + geom_line() +
  scale_x_continuous(name = "Number of dimensions included",
    limits = c(0, 15), breaks = seq(0, 15, 2)) + scale_y_continuous(name = "Classification accuracy",
    limits = c(40, 100), breaks = seq(40, 100, 10)) + theme_classic()

```



Environment hypervolumes and overlaps

```

dd <- read_excel("DatasetS1.xlsx", sheet = "Habitat")
dd <- as.data.frame(dd)
row.names(dd) <- dd$ID_in_tree

d2 <- dd[, c(6, 23:38)] #select environmental variables
row.names(d2) <- dd$ID_in_tree
d2 <- na.omit(d2)

# Replace vegetation layer by numerical variable (1:
# ground, 2: shrub, 3: canopy)
for (i in 1:length(d2$vegetation_layer)) {
  if (d2$vegetation_layer[i] == "ground/shrub") {
    d2$vegetation_layer[i] <- 1
  }
  if (d2$vegetation_layer[i] == "shrub/understory") {
    d2$vegetation_layer[i] <- 2
  }
  if (d2$vegetation_layer[i] == "understory/canopy") {
    d2$vegetation_layer[i] <- 3
  } else {
  }
}

```

```

}
d2$vegetation_layer <- as.numeric(d2$vegetation_layer)
d2 <- as.data.frame(scale(d2))

### Make phylogenetic tree match Cleaning both the data and
### the tree
cleaned_data <- clean.data(d2, ttree)
## Extracting the cleaned dataset and the cleaned tree
df <- cleaned_data$data
ttree <- cleaned_data$tree
name.check(ttree, df)

## [1] "OK"

# reorder data to match the order of tips in tree
d2 <- ReorderData(ttree, df, taxa.names = "row names")

```

Climate PCA

```

dpca <- d2
dpca <- na.omit(dpca)

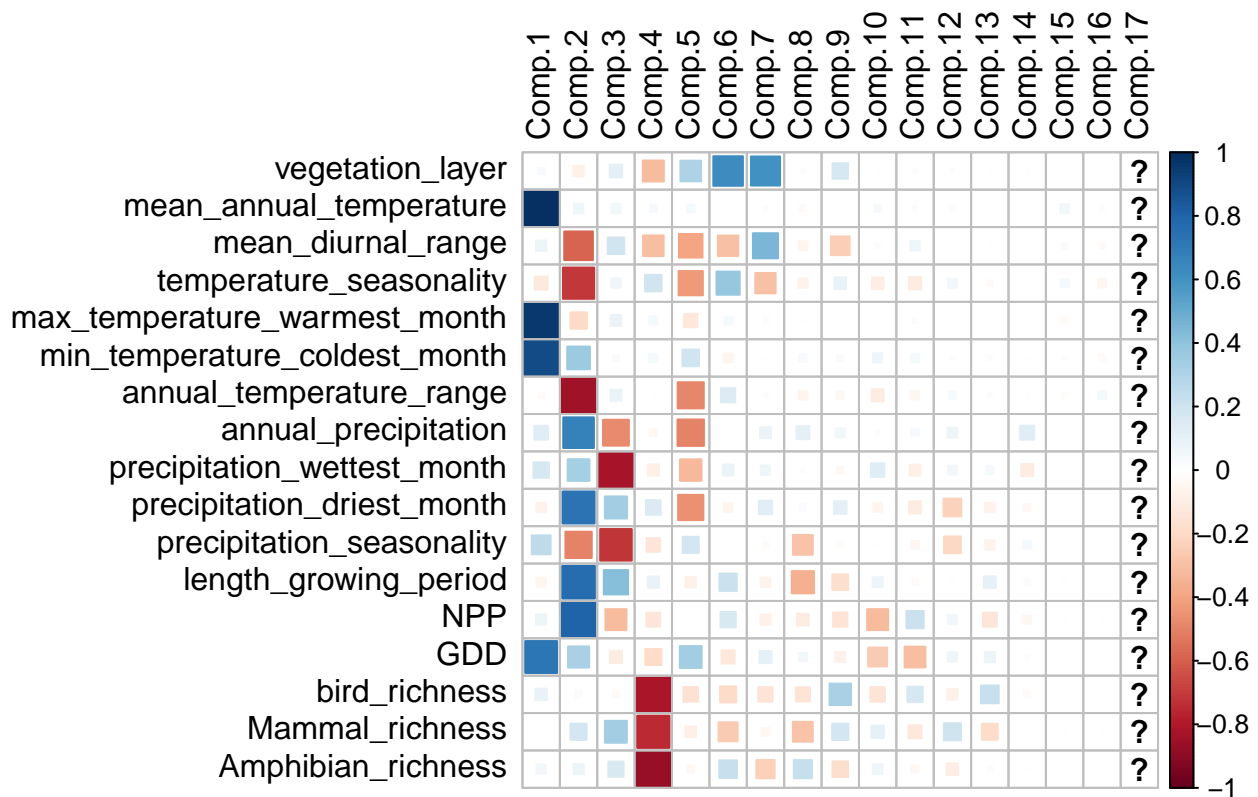
pPCA <- phyl.pca(ttree, dpca, method = "BM")
summary(pPCA)

## Importance of components:
##
##          PC1          PC2          PC3          PC4          PC5
## Standard deviation  0.5455323 0.3882250 0.2771509 0.25576341 0.22392173
## Proportion of Variance 0.4036433 0.2044202 0.1041812 0.08872247 0.06800632
## Cumulative Proportion 0.4036433 0.6080635 0.7122447 0.80096721 0.86897353
##
##          PC6          PC7          PC8          PC9          PC10
## Standard deviation  0.15941376 0.15142607 0.11674505 0.09962367 0.08492155
## Proportion of Variance 0.03446739 0.03109984 0.01848561 0.01346114 0.00978121
## Cumulative Proportion 0.90344091 0.93454075 0.95302636 0.96648750 0.97626871
##
##          PC11          PC12          PC13          PC14
## Standard deviation  0.080500684 0.075419421 0.057272677 0.036814134
## Proportion of Variance 0.008789333 0.007714773 0.004448891 0.001838171
## Cumulative Proportion 0.985058041 0.992772815 0.997221705 0.999059877
##
##          PC15          PC16          PC17
## Standard deviation  0.0222658568 0.0140493045          NaN
## Proportion of Variance 0.0006724122 0.0002677111 -6.531736e-17
## Cumulative Proportion 0.9997322889 1.0000000000 1.000000e+00

pPCA <- as.princomp(pPCA)

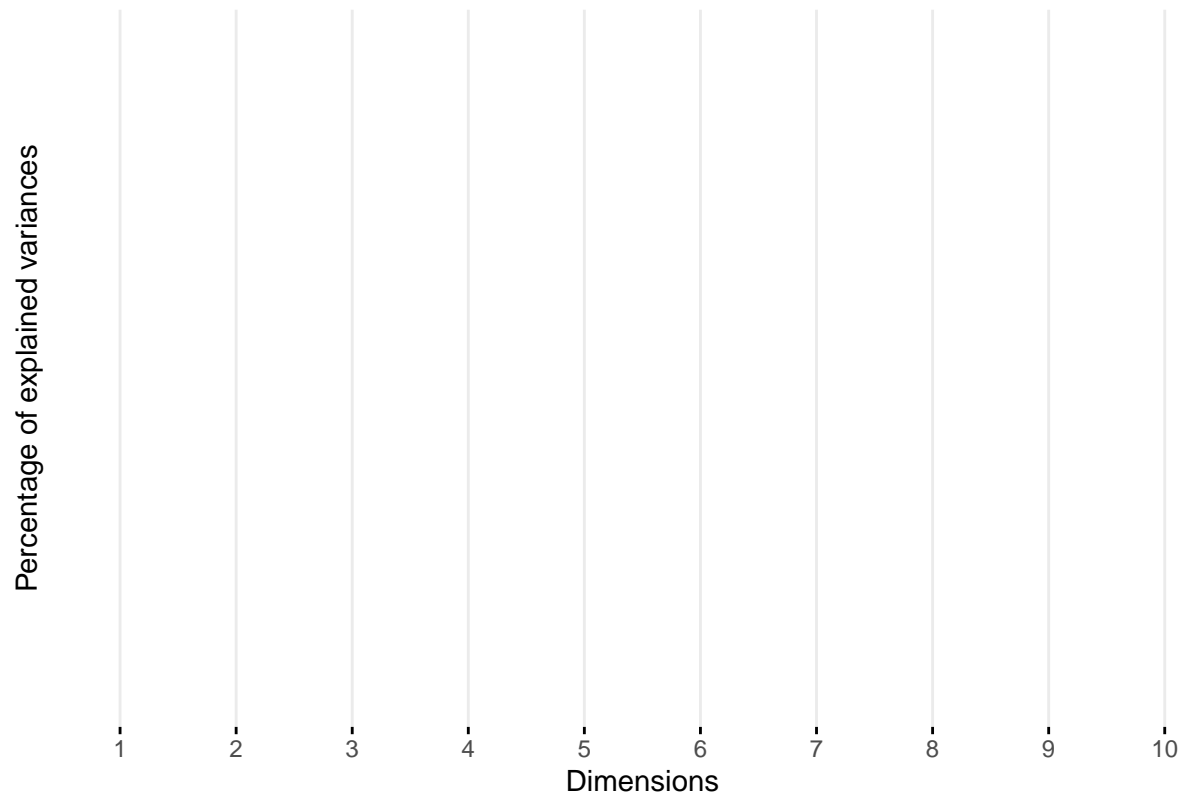
# Loadings correlation plot
corrplot(pPCA$loadings, method = "square", tl.col = "black")

```

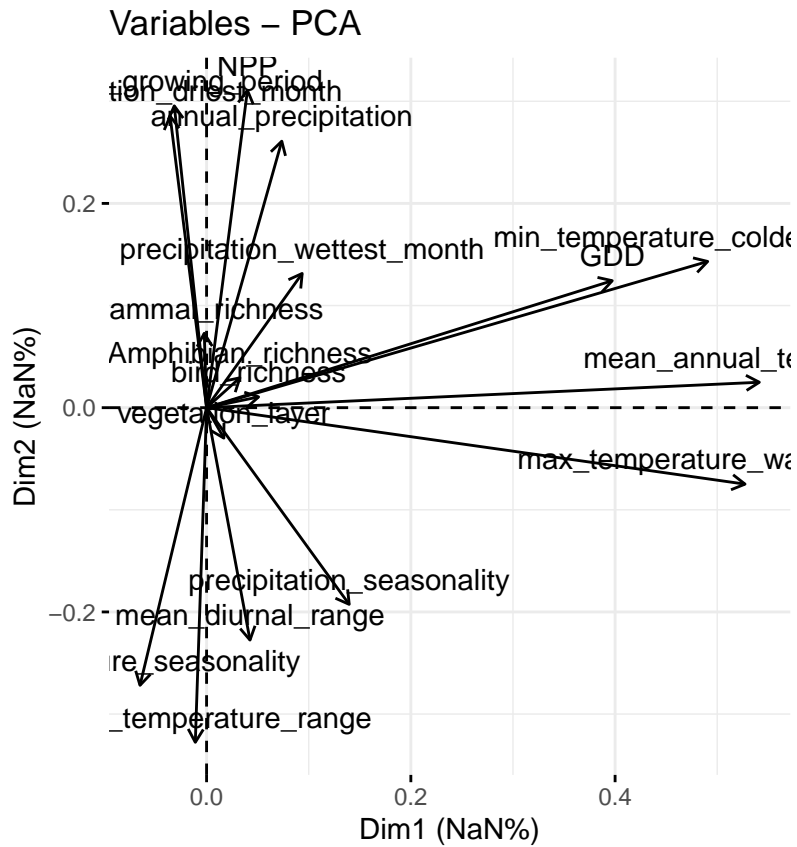



```
# Screeplot
fviz_screepplot(pPCA, addlabels = TRUE)
```

Scree plot



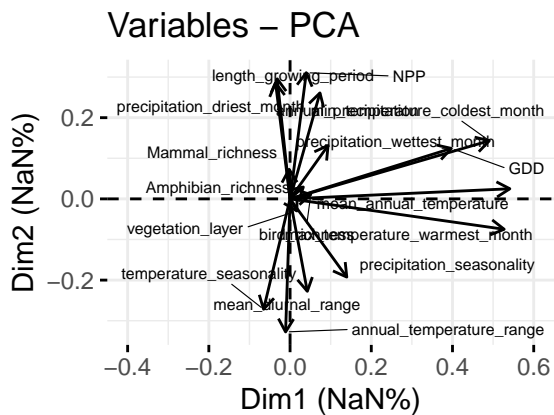
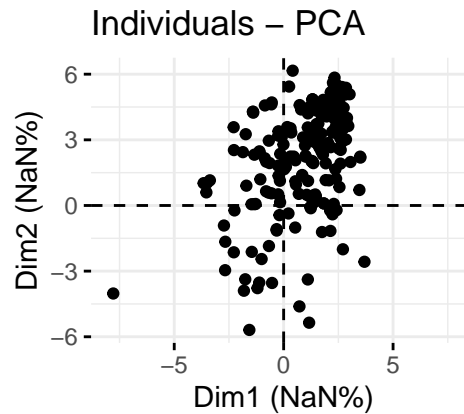
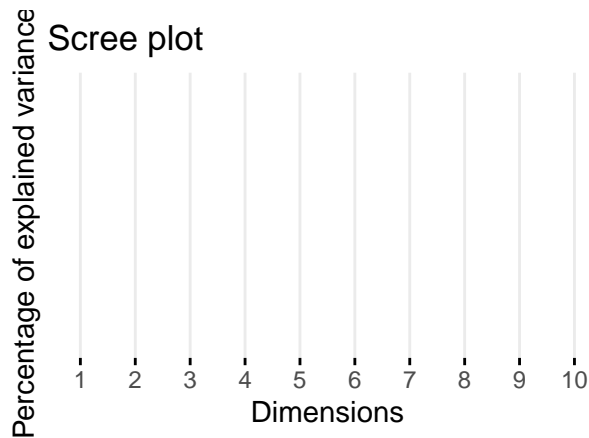
```
# Correlation circle  
fviz_pca_var(pPCA, col.var = "black") + coord_fixed()
```



```

pp <- fviz_eig(pPCA)
pp2 <- fviz_pca_ind(pPCA, geom = c("point"), repel = TRUE)
pp2 <- pp2 + coord_fixed() + xlim(-8, 8)
pp3 <- fviz_pca_var(pPCA, repel = TRUE, labelsize = 2)
pp3 <- pp3 + coord_fixed() + xlim(-0.4, 0.6)
plot_grid(pp, pp2, pp3, nrow = 2)

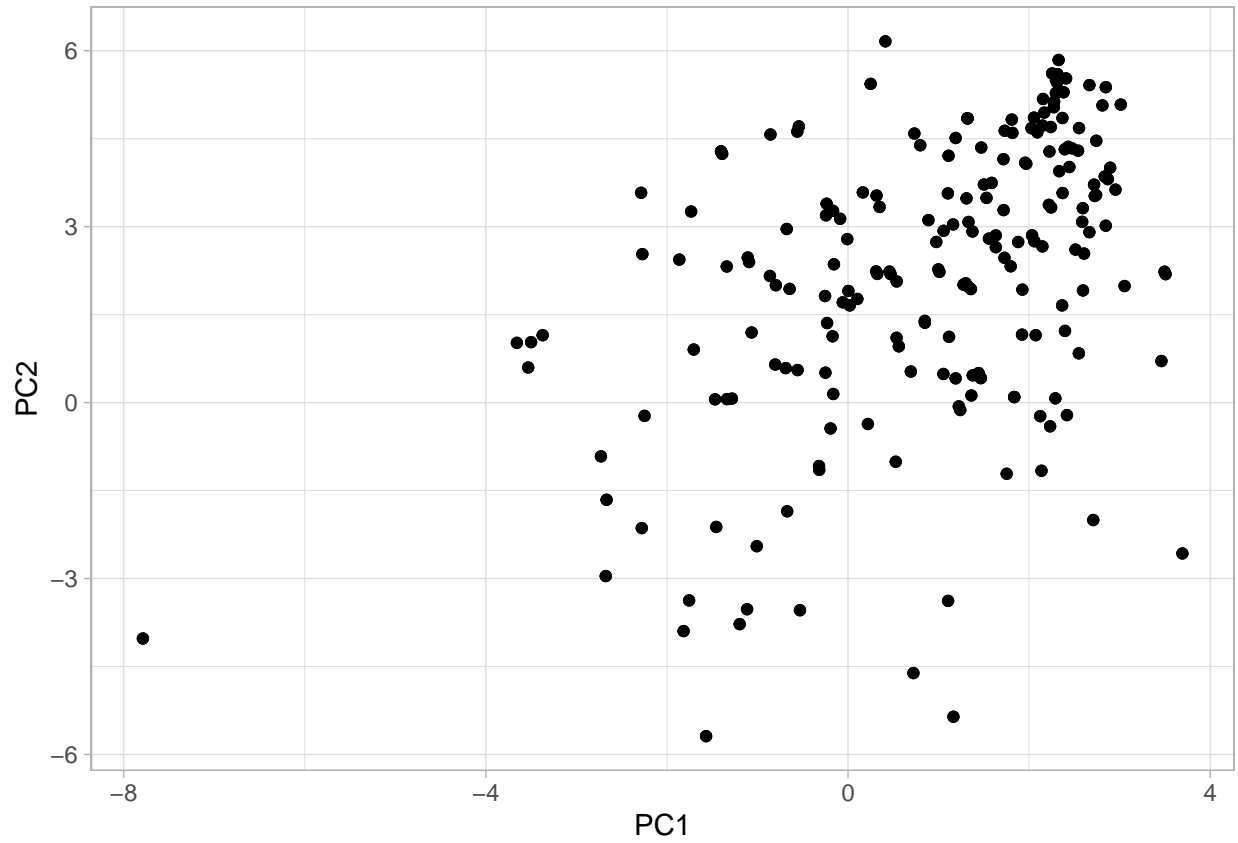
```



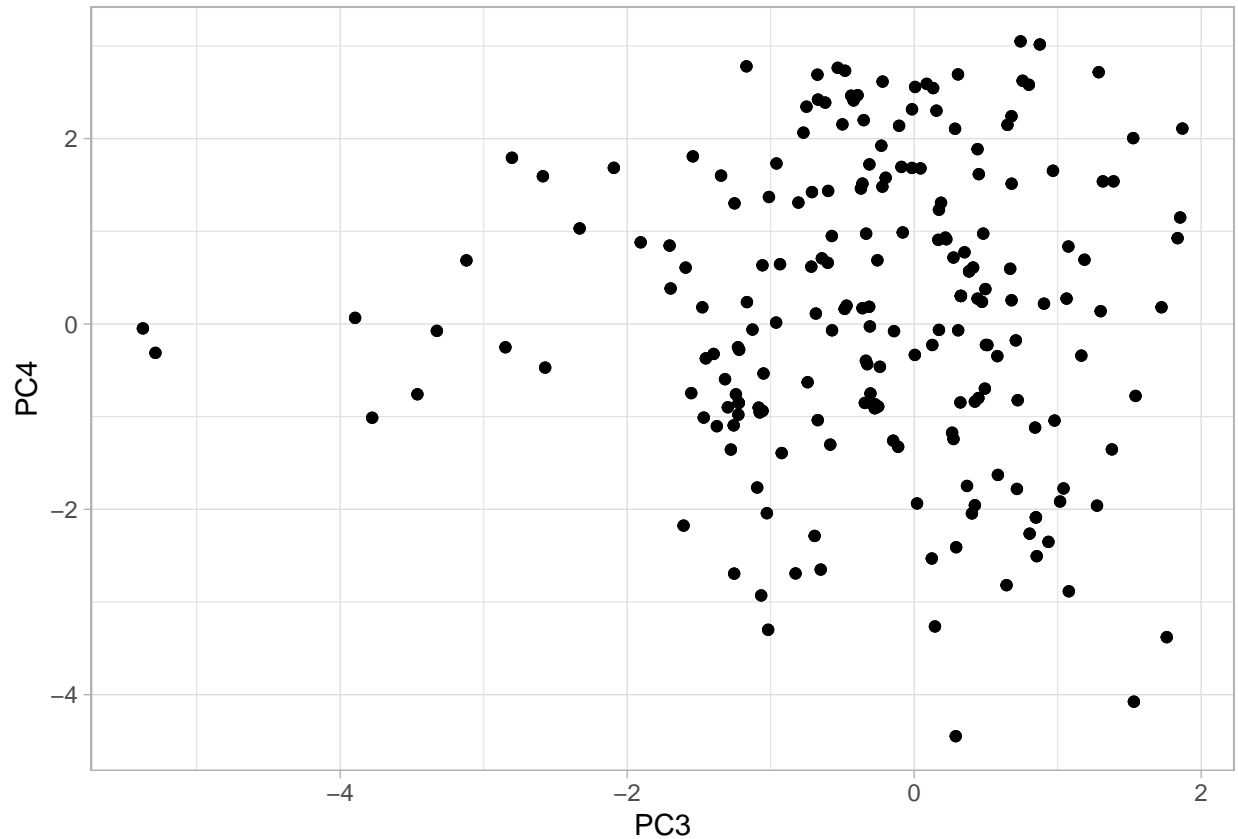
```

pcar <- as.data.frame(pPCA$scores)
ggplot(pcar, aes(x = Comp.1, y = Comp.2, label = rownames(pcar))) +
  geom_point() + #geom_text() + geom_point() +
  geom_point() + #geom_text() + #geom_text() +
  xlab("PC1") + ylab("PC2") + theme_light()

```



```
ggplot(pcar, aes(x = Comp.3, y = Comp.4, label = rownames(pcar))) +
  geom_point() + #geom_text() + geom_point() +
  geom_point() + #geom_text() + #geom_text() +
  xlab("PC3") + ylab("PC4") + theme_light()
```



```

# Take the first 6 dimensions (accounting for 90.3% of the
# variation)
denv <- as.data.frame(pcar[, 1:6])
colnames(denv) <- c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6")
denv$species <- row.names(denv)

# Add habitat column
hab <- as.data.frame(dd$habitat)
hab$species <- row.names(dd)
colnames(hab) <- c("habitat", "species")
env_space <- merge(denv, hab, by = "species")

```

Hypervolume size

Kernel density estimates

```

# Create appropriate dataset with 6 first dimensions of
# morphospace
Xrb <- matrix(nrow = length(env_space$habitat), ncol = 7)
colnames(Xrb) <- c("habitat", "PC1", "PC2", "PC3", "PC4", "PC5",
"PC6")
Xrb <- as.data.frame(Xrb)
Xrb$habitat <- as.character(env_space$habitat)

```

```

Xrb$PC1 <- as.numeric(env_space$PC1)
Xrb$PC2 <- as.numeric(env_space$PC2)
Xrb$PC3 <- as.numeric(env_space$PC3)
Xrb$PC4 <- as.numeric(env_space$PC4)
Xrb$PC5 <- as.numeric(env_space$PC5)
Xrb$PC6 <- as.numeric(env_space$PC6)

## Using hypervolume package
Xrb_litter = Xrb[Xrb$habitat == "laying on leaf litter, logs, bark, trunk",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6")]
Xrb_hang = Xrb[Xrb$habitat == "hanging from branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6")]
Xrb_branch = Xrb[Xrb$habitat == "laying on branches and leaves",
  c("PC1", "PC2", "PC3", "PC4", "PC5", "PC6")]
Xrb_grass = Xrb[Xrb$habitat == "hanging from grass", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6")]
Xrb_palm = Xrb[Xrb$habitat == "laying on palm leaves", c("PC1",
  "PC2", "PC3", "PC4", "PC5", "PC6")]

vol <- as.data.frame(matrix(nrow = 5, ncol = 2))
colnames(vol) <- c("volume", "habitat")
vol$habitat <- c("laying on leaf litter, logs, bark, trunk",
  "hanging from branches and leaves", "laying on branches and leaves",
  "hanging from grass", "laying on palm leaves")

hv1 = hypervolume_box(Xrb_litter, name = "litter")

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

hv2 = hypervolume_box(Xrb_hang, name = "hang")

##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.

hv3 = hypervolume_box(Xrb_branch, name = "branch")

##

```

```
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.
```

```
hv4 = hypervolume_box(Xrb_grass, name = "grass")
```

```
##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.
```

```
hv5 = hypervolume_box(Xrb_palm, name = "palm")
```

```
##
## Building tree...
## done.
## Ball query...
##
## done.
## Binding random points... done.
## Beginning volume calculation... done.
```

```
vol[1, 1] <- get_volume(hv1)
vol[2, 1] <- get_volume(hv2)
vol[3, 1] <- get_volume(hv3)
vol[4, 1] <- get_volume(hv4)
vol[5, 1] <- get_volume(hv5)

p1 <- ggplot(vol, aes(x = habitat, y = volume, fill = habitat)) +
  geom_bar(position = "dodge", stat = "identity") + xlab("Habitat") +
  ylab("Hypervolume size") + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + scale_x_discrete(labels = NULL,
  breaks = NULL) + labs(title = "Kernel density estimates") +
  theme_classic()
```

Dynamic range boxes

```
## Using dynRB Size and pairwise overlap
r <- dynRB_VPa(Xrb)
```

```
## [1] 1
## [1] 2
```



```

## 5      hanging from branches and leaves 0.0101317587 0.000720669
## 6              hanging from grass 0.3018722223 0.097187554
## 7              hanging from grass 1.0000000000 0.006420454
## 8              hanging from grass 0.2190941799 0.035481683
## 9              hanging from grass 0.4015063722 0.076003751
## 10             hanging from grass 0.0002453835 0.000720669
## 11             laying on branches and leaves 0.7080513951 0.097187554
## 12             laying on branches and leaves 0.0382763788 0.006420454
## 13             laying on branches and leaves 1.0000000000 0.035481683
## 14             laying on branches and leaves 0.4618436884 0.076003751
## 15             laying on branches and leaves 0.0133699268 0.000720669
## 16 laying on leaf litter, logs, bark, trunk 0.4144429486 0.097187554
## 17 laying on leaf litter, logs, bark, trunk 0.0418683796 0.006420454
## 18 laying on leaf litter, logs, bark, trunk 0.2852640434 0.035481683
## 19 laying on leaf litter, logs, bark, trunk 1.0000000000 0.076003751
## 20 laying on leaf litter, logs, bark, trunk 0.0082480308 0.000720669
## 21             laying on palm leaves 0.8145171129 0.097187554
## 22             laying on palm leaves 0.0020736525 0.006420454
## 23             laying on palm leaves 0.5699490799 0.035481683
## 24             laying on palm leaves 0.5795862151 0.076003751
## 25             laying on palm leaves 1.0000000000 0.000720669

```

```
## vol_V2_prod
```

```

## 1 0.097187554
## 2 0.097187554
## 3 0.097187554
## 4 0.097187554
## 5 0.097187554
## 6 0.006420454
## 7 0.006420454
## 8 0.006420454
## 9 0.006420454
## 10 0.006420454
## 11 0.035481683
## 12 0.035481683
## 13 0.035481683
## 14 0.035481683
## 15 0.035481683
## 16 0.076003751
## 17 0.076003751
## 18 0.076003751
## 19 0.076003751
## 20 0.076003751
## 21 0.000720669
## 22 0.000720669
## 23 0.000720669
## 24 0.000720669
## 25 0.000720669

```

```
r$result
```

```

## V1
## 1 hanging from branches and leaves
## 2 hanging from grass
## 3 laying on branches and leaves

```

```

## 4 laying on leaf litter, logs, bark, trunk
## 5         laying on palm leaves
## 6         hanging from branches and leaves
## 7         hanging from grass
## 8         laying on branches and leaves
## 9 laying on leaf litter, logs, bark, trunk
## 10        laying on palm leaves
## 11        hanging from branches and leaves
## 12        hanging from grass
## 13        laying on branches and leaves
## 14 laying on leaf litter, logs, bark, trunk
## 15        laying on palm leaves
## 16        hanging from branches and leaves
## 17        hanging from grass
## 18        laying on branches and leaves
## 19 laying on leaf litter, logs, bark, trunk
## 20        laying on palm leaves
## 21        hanging from branches and leaves
## 22        hanging from grass
## 23        laying on branches and leaves
## 24 laying on leaf litter, logs, bark, trunk
## 25        laying on palm leaves
##
##          V2      port_prod port_mean port_gmean
## 1         hanging from branches and leaves 1.000000000 1.0000000 1.0000000
## 2         hanging from branches and leaves 0.0214109091 0.4995906 0.43277356
## 3         hanging from branches and leaves 0.3554436485 0.8393246 0.83565413
## 4         hanging from branches and leaves 0.3279836889 0.8326677 0.82049163
## 5         hanging from branches and leaves 0.0101317587 0.4601335 0.44319774
## 6         hanging from grass 0.3018722223 0.7210752 0.64644967
## 7         hanging from grass 1.0000000000 1.0000000 1.00000000
## 8         hanging from grass 0.2190941799 0.6873573 0.59145444
## 9         hanging from grass 0.4015063722 0.8037271 0.67871730
## 10        hanging from grass 0.0002453835 0.3676887 0.06811329
## 11        laying on branches and leaves 0.7080513951 0.9352841 0.93707532
## 12        laying on branches and leaves 0.0382763788 0.5344836 0.45329544
## 13        laying on branches and leaves 1.0000000000 1.0000000 1.00000000
## 14        laying on branches and leaves 0.4618436884 0.8655839 0.86686431
## 15        laying on branches and leaves 0.0133699268 0.4790873 0.45321769
## 16 laying on leaf litter, logs, bark, trunk 0.4144429486 0.8574291 0.85156034
## 17 laying on leaf litter, logs, bark, trunk 0.0418683796 0.5626040 0.47487859
## 18 laying on leaf litter, logs, bark, trunk 0.2852640434 0.8055650 0.80159596
## 19 laying on leaf litter, logs, bark, trunk 1.0000000000 1.0000000 1.00000000
## 20 laying on leaf litter, logs, bark, trunk 0.0082480308 0.4318278 0.38420358
## 21        laying on palm leaves 0.8145171129 0.9561271 0.93438447
## 22        laying on palm leaves 0.0020736525 0.5203652 0.09723164
## 23        laying on palm leaves 0.5699490799 0.8924510 0.85201334
## 24        laying on palm leaves 0.5795862151 0.8844649 0.78094260
## 25        laying on palm leaves 1.0000000000 1.0000000 1.00000000
##
##  vol_V1_prod vol_V1_mean vol_V1_gmean vol_V2_prod vol_V2_mean vol_V2_gmean
## 1 0.097187554 0.9592122 0.9550130 0.097187554 0.9592122 0.9550130
## 2 0.006420454 0.7795008 0.7737219 0.097187554 0.9592122 0.9550130
## 3 0.035481683 0.8938553 0.8856993 0.097187554 0.9592122 0.9550130
## 4 0.076003751 0.9453340 0.9382550 0.097187554 0.9592122 0.9550130
## 5 0.000720669 0.6204625 0.6183855 0.097187554 0.9592122 0.9550130

```

```

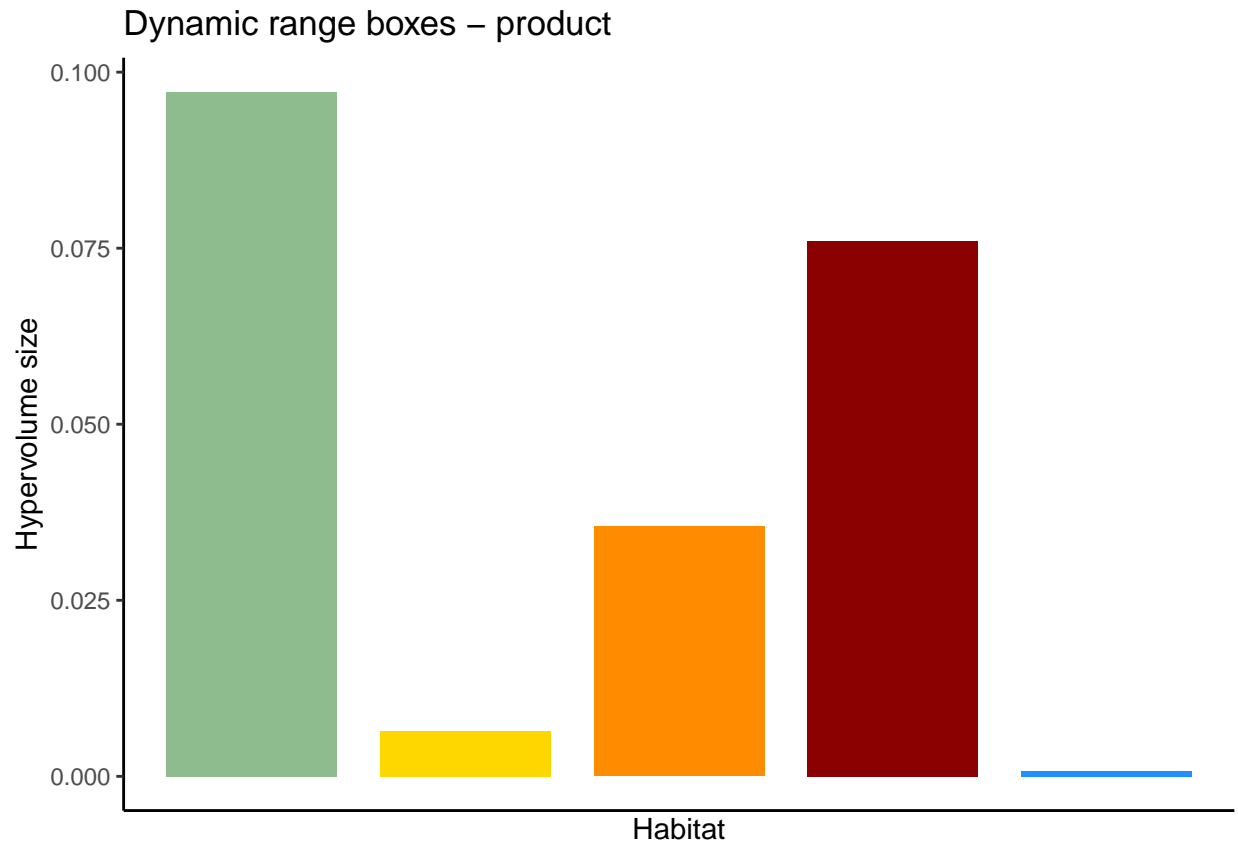
## 6 0.097187554 0.9592122 0.9550130 0.006420454 0.7795008 0.7737219
## 7 0.006420454 0.7795008 0.7737219 0.006420454 0.7795008 0.7737219
## 8 0.035481683 0.8938553 0.8856993 0.006420454 0.7795008 0.7737219
## 9 0.076003751 0.9453340 0.9382550 0.006420454 0.7795008 0.7737219
## 10 0.000720669 0.6204625 0.6183855 0.006420454 0.7795008 0.7737219
## 11 0.097187554 0.9592122 0.9550130 0.035481683 0.8938553 0.8856993
## 12 0.006420454 0.7795008 0.7737219 0.035481683 0.8938553 0.8856993
## 13 0.035481683 0.8938553 0.8856993 0.035481683 0.8938553 0.8856993
## 14 0.076003751 0.9453340 0.9382550 0.035481683 0.8938553 0.8856993
## 15 0.000720669 0.6204625 0.6183855 0.035481683 0.8938553 0.8856993
## 16 0.097187554 0.9592122 0.9550130 0.076003751 0.9453340 0.9382550
## 17 0.006420454 0.7795008 0.7737219 0.076003751 0.9453340 0.9382550
## 18 0.035481683 0.8938553 0.8856993 0.076003751 0.9453340 0.9382550
## 19 0.076003751 0.9453340 0.9382550 0.076003751 0.9453340 0.9382550
## 20 0.000720669 0.6204625 0.6183855 0.076003751 0.9453340 0.9382550
## 21 0.097187554 0.9592122 0.9550130 0.000720669 0.6204625 0.6183855
## 22 0.006420454 0.7795008 0.7737219 0.000720669 0.6204625 0.6183855
## 23 0.035481683 0.8938553 0.8856993 0.000720669 0.6204625 0.6183855
## 24 0.076003751 0.9453340 0.9382550 0.000720669 0.6204625 0.6183855
## 25 0.000720669 0.6204625 0.6183855 0.000720669 0.6204625 0.6183855

```

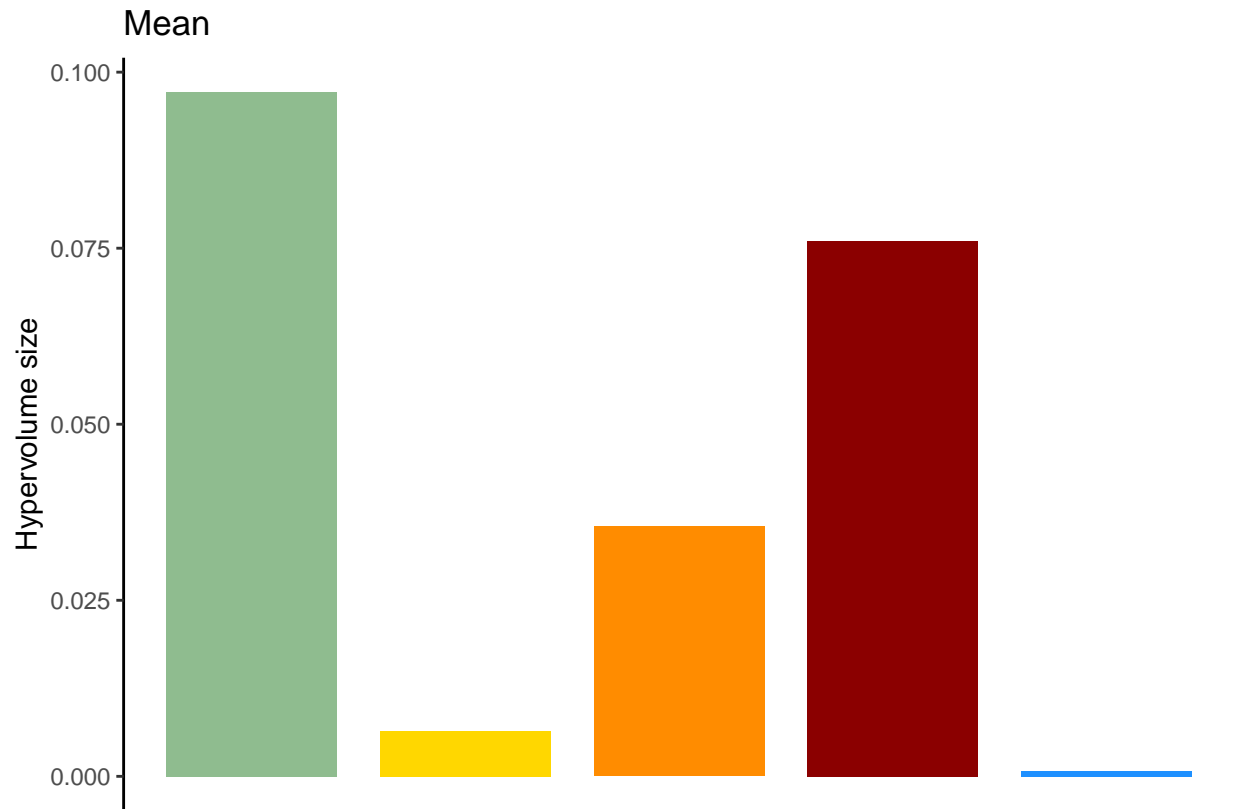
```

# Plot full hypervolume size
result <- r$result
voll <- as.data.frame(matrix(nrow = 5, ncol = 2))
colnames(voll) <- c("volume", "habitat")
voll$habitat <- c("hanging from branches and leaves", "hanging from grass",
  "laying on branches and leaves", "laying on leaf litter, logs, bark, trunk",
  "laying on palm leaves")
# product
voll[, 1] <- c(result[1:5, 6])
p2 <- ggplot(voll, aes(x = habitat, y = volume)) + xlab("Habitat") +
  ylab("Hypervolume size") + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Hypervolume size") +
  # xlab('')+
labs(title = "Dynamic range boxes - product") + scale_x_discrete(labels = NULL,
  breaks = NULL) + guides(fill = F) + theme_classic()
p2

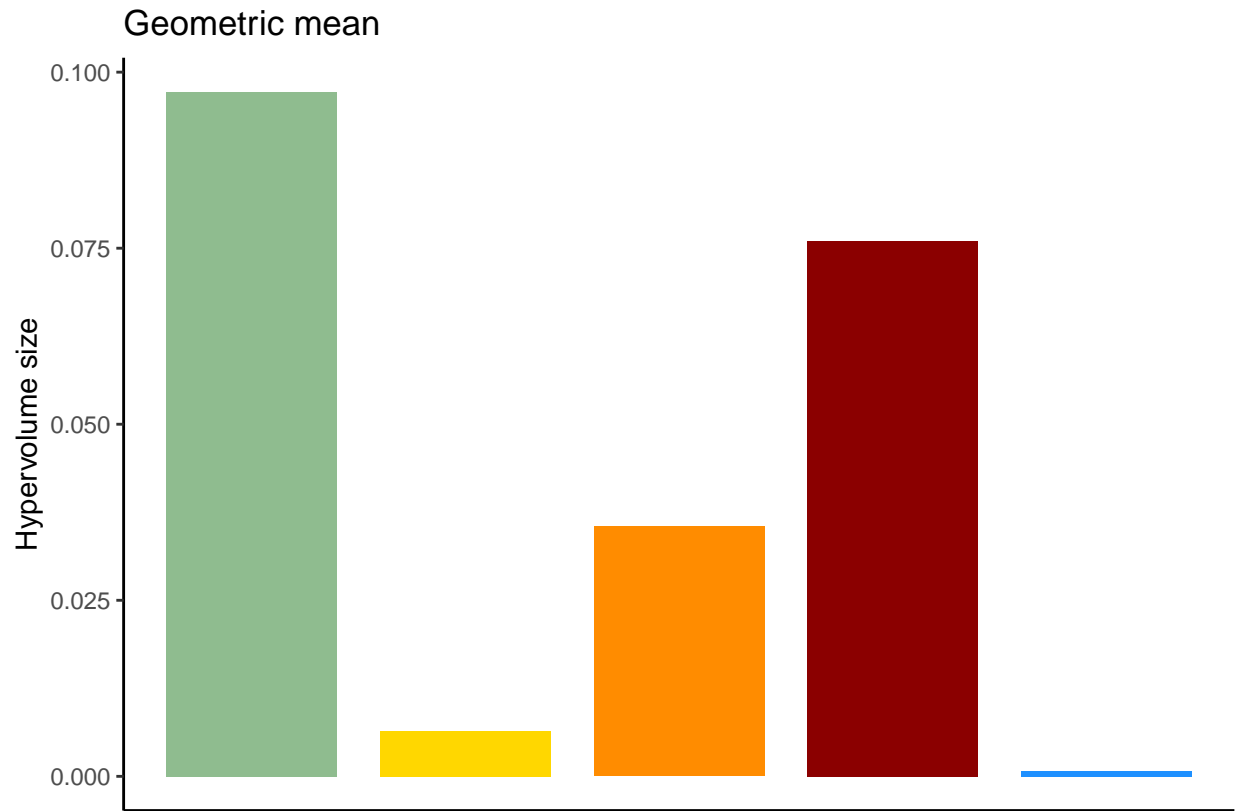
```



```
# mean
vol[, 1] <- c(result[1:5, 7])
ggplot(voll, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Full niche hypervolume") +
  ylab("Hypervolume size") + xlab("") + labs(title = "Mean") +
  scale_x_discrete(labels = NULL, breaks = NULL) + guides(fill = F) +
  theme_classic()
```

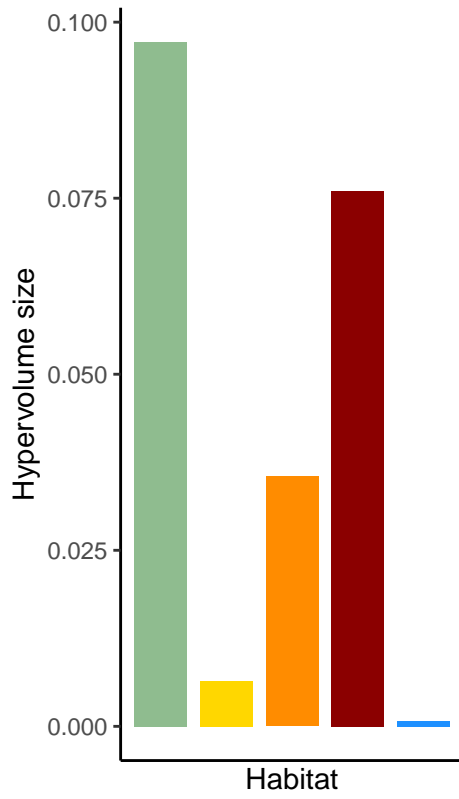


```
# gmean
vol[, 1] <- c(result[1:5, 8])
ggplot(voll, aes(x = habitat, y = volume)) + geom_bar(aes(fill = habitat),
  stat = "identity", width = 0.8) + scale_fill_manual(values = c("darkseagreen",
  "gold", "darkorange", "darkred", "dodgerblue")) + ylab("Full niche hypervolume") +
  ylab("Hypervolume size") + xlab("") + labs(title = "Geometric mean") +
  scale_x_discrete(labels = NULL, breaks = NULL) + guides(fill = F) +
  theme_classic()
```



```
# Final  
plot_grid(p2, p1, nrow = 1, rel_widths = c(1, 1.6), labels = "AUTO")
```

A Dynamic range boxes -



B Kernel density estimates

