My presentation

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1 How I can contribute to the SAPPHIRE project

1.1 Update the glottodist with metrics other than Gower's distance

We can implement other kinds of metrics like listed in [1], Take Eskin's distance as an example:

$$S_k(X_k, Y_k) = \begin{cases} 1 & \text{if } X_k = Y_k \\ \frac{n_k^2}{n_k^2 + 2} & \text{otherwise} \end{cases}$$

1.2 Add a function to compute the Hausdorff distance

def Hausdorff

1.3 Topological data analysis (TDA)

sdfs [2]

1.4 Spectral graph theory methods

```
library(glottospace)
library(dplyr)

##
## Attaching package: 'dplyr'

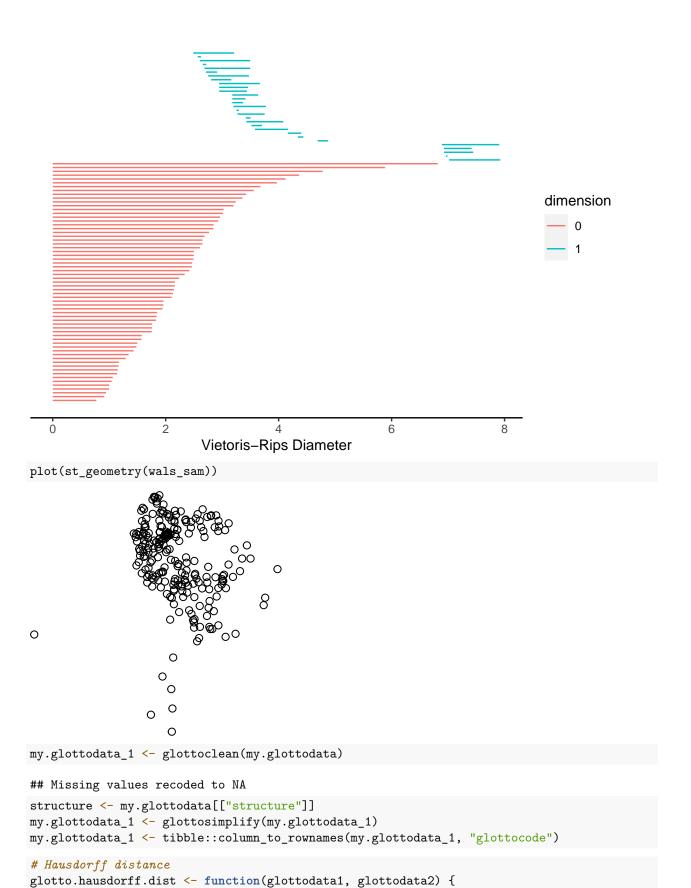
## The following objects are masked from 'package:stats':
##
## filter, lag

## The following objects are masked from 'package:base':
##
## intersect, setdiff, setequal, union
library(TDAstats)
library(sf)
```

```
## Linking to GEOS 3.10.2, GDAL 3.4.2, PROJ 8.2.1; sf_use_s2() is TRUE
# load the dataset wals
wals <- glottoget("wals")</pre>
#select the data wrt Aouth America
wals_sam <- subset(wals, continent == "South America")</pre>
# Define a function select.features(dataset, feature_names, a) to get all features of dataset that the
count.na <- function (dataset, feature.name) {</pre>
  num.features <- length(st_drop_geometry(dataset))</pre>
 return(sum(is.na(dataset[[feature.name]])) / nrow(dataset))
}
select.features <- function(dataset, feature_names, a) {</pre>
  counts <- c()
  for (x in feature names) {
    counts <- append(counts, count.na(wals_sam, x))</pre>
  result <- t(as.matrix(counts[counts<0.5]))</pre>
  colnames(result) <- feature_names[which(counts<0.5)]</pre>
  return(result)
}
feature_names <- colnames(wals_sam)[-c(1, 194:208)]
select.features(wals_sam, feature_names, 0.5)
##
           81A
                  82A
                         83A
                                    86A
                                          129A
## [1,] 0.4875 0.425 0.4125 0.4833333 0.4875
data <- select(wals_sam, 'glottocode', '81A', '82A', '83A', '86A', '129A')
data.obj <- st_drop_geometry(data)</pre>
data.obj$`81A` <- as.factor(data.obj$`81A`)</pre>
data.obj$`82A` <- as.factor(data.obj$`82A`)</pre>
data.obj$`83A` <- as.factor(data.obj$`83A`)</pre>
data.obj$`86A` <- as.factor(data.obj$`86A`)</pre>
data.obj$`129A` <- as.factor(data.obj$`129A`)</pre>
gtd.unique <- match(unique(data.obj$glottocode), data.obj$glottocode)</pre>
data.obj.unique <- data.obj[gtd.unique, ]</pre>
str(data.obj.unique)
## 'data.frame':
                     235 obs. of 6 variables:
## $ glottocode: chr "abip1241" "acha1250" "ache1246" "achu1248" ...
            : Factor w/ 7 levels "No dominant order",..: 5 5 NA NA NA NA NA NA NA 1 ...
## $ 81A
## $ 82A
                : Factor w/ 3 levels "No dominant order",...: 2 2 NA NA NA NA NA NA NA NA 1 ...
## $ 83A
                : Factor w/ 3 levels "No dominant order",...: 3 3 NA 2 NA NA NA NA NA 2 ...
                : Factor w/ 3 levels "Genitive-Noun",..: 2 1 NA 1 1 NA NA NA NA 1 ...
## $ 86A
                : Factor w/ 2 levels "Different", "Identical": 1 1 NA 1 NA 1 NA NA NA 1 ...
structure <- glottocreate_structuretable(varnames = c("81A", "82A", "83A", "86A", "129A"))
structure$type <- rep("factor", 5)</pre>
```

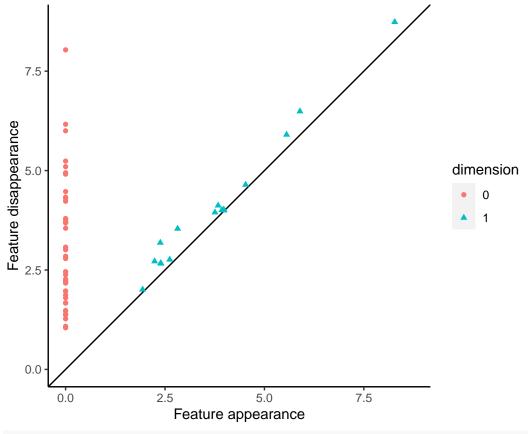
```
my.glottodata <- glottocreate_addtable(data.obj.unique, structure, name="structure")</pre>
my.glottodist <- glottodist(my.glottodata)</pre>
## Missing values recoded to NA
## All variables have two or more levels (excluding NA)
my.glottodist[is.na(my.glottodist)] <- 0</pre>
my.glottodata.phom <- calculate_homology(my.glottodist)</pre>
plot_persist(my.glottodata.phom)
   8
Feature disappearance
                                                                      dimension
                                                                          0
   2
                     2
                         Feature appearance
```

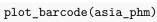
plot_barcode(my.glottodata.phom)

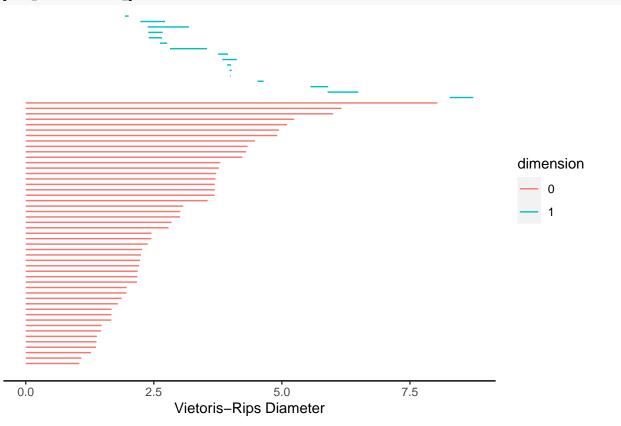


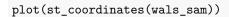
```
overlap.dist <- function (x, y){
  sum(x != y) / length(x)
overlap.x.Y.dist <- function(x, Y) {</pre>
  result <- c()
  for (i in 1:nrow(Y)) {
    result <- append(result, overlap.dist(x, Y[i,]))</pre>
 return(min(result))
overlap.supX.Y.dist <- function(X, Y) {</pre>
  result <- c()
  for (i in 1:nrow(X)) {
    result <- append(result, overlap.x.Y.dist(X[i, ], Y))</pre>
 }
 return(max(result))
}
overlap.hausdorff.dist <- function(X, Y) {</pre>
 result <- max(overlap.supX.Y.dist(X, Y), overlap.supX.Y.dist(Y,X))</pre>
 return(result)
}
wals_asia <- subset(wals, continent=="Asia")</pre>
data_asia <- select(wals_asia, 'glottocode', '81A', '82A', '83A', '86A', '129A')
data_asia <- st_drop_geometry(data_asia)</pre>
data_asia$`81A` <- as.factor(data_asia$`81A`)</pre>
data_asia$`82A` <- as.factor(data_asia$`82A`)</pre>
data_asia$`83A` <- as.factor(data_asia$`83A`)</pre>
data_asia$`86A` <- as.factor(data_asia$`86A`)</pre>
data_asia$`129A` <- as.factor(data_asia$`129A`)</pre>
str(data_asia)
## 'data.frame':
                    599 obs. of 6 variables:
## $ glottocode: chr "nort3139" "abkh1244" "abun1252" "abui1241" ...
## $ 81A
           : Factor w/ 6 levels "No dominant order",..: NA 3 4 3 1 NA 3 6 6 NA ...
## $ 82A
               : Factor w/ 3 levels "No dominant order",..: NA 2 2 2 1 2 2 3 3 NA ...
## $ 83A
               : Factor w/ 3 levels "No dominant order",..: NA 2 3 2 1 2 2 3 3 NA ...
## $ 86A
                : Factor w/ 3 levels "Genitive-Noun",..: NA 1 1 1 3 1 1 3 3 NA ...
                : Factor w/ 2 levels "Different", "Identical": NA 1 NA NA NA NA NA NA NA NA NA ...
data_asia.unique <- match(unique(data_asia$glottocode), data_asia$glottocode)
data_asia.unique <- data_asia[data_asia.unique, ]</pre>
str(data_asia.unique)
## 'data.frame':
                    561 obs. of 6 variables:
## $ glottocode: chr "nort3139" "abkh1244" "abun1252" "abui1241" ...
## $ 81A
           : Factor w/ 6 levels "No dominant order",..: NA 3 4 3 1 NA 3 6 6 NA ...
```

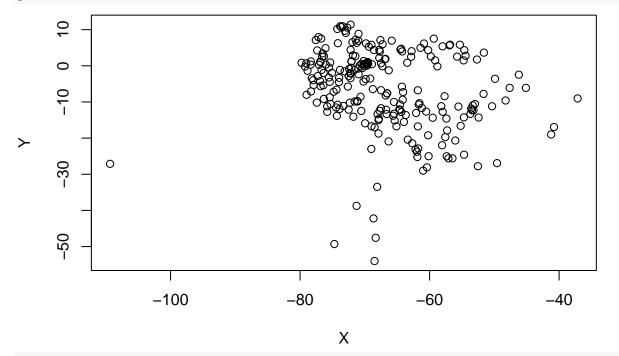
```
## $ 82A
                 : Factor w/ 3 levels "No dominant order",..: NA 2 2 2 1 2 2 3 3 NA ...
## $ 83A
                 : Factor w/ 3 levels "No dominant order",...: NA 2 3 2 1 2 2 3 3 NA ...
                 : Factor w/ 3 levels "Genitive-Noun",..: NA 1 1 1 3 1 1 3 3 NA ...
## $ 86A
                 : Factor w/ 2 levels "Different", "Identical": NA 1 NA ...
## $ 129A
glottodata.asia <- glottocreate_addtable(data_asia.unique, structure, name="structure")</pre>
glottodata.asia_1 <- glottoclean(glottodata.asia)</pre>
## Missing values recoded to NA
structure <- glottodata.asia_1[["structure"]]</pre>
glottodata.asia_1 <- glottosimplify(glottodata.asia_1)</pre>
glottodata.asia_1 <- tibble::column_to_rownames(glottodata.asia_1, "glottocode")</pre>
my.glottodata_1 <- data.frame(lapply(my.glottodata_1, as.character), stringsAsFactors = FALSE)
my.glottodata_1[is.na(my.glottodata_1)] <- "unknown"</pre>
glottodata.asia_1 <- data.frame(lapply(glottodata.asia_1, as.character), stringsAsFactors = FALSE)</pre>
glottodata.asia_1[is.na(glottodata.asia_1)] <- "unknown"</pre>
# overlap.hausdorff.dist(glottodata.asia_1, my.glottodata_1)
asia.dist <- glottodist(glottodata.asia)</pre>
## Missing values recoded to NA
## All variables have two or more levels (excluding NA)
asia.dist[is.na(asia.dist)] <- 0
asia_phm <- calculate_homology(asia.dist)</pre>
plot_persist(asia_phm)
```



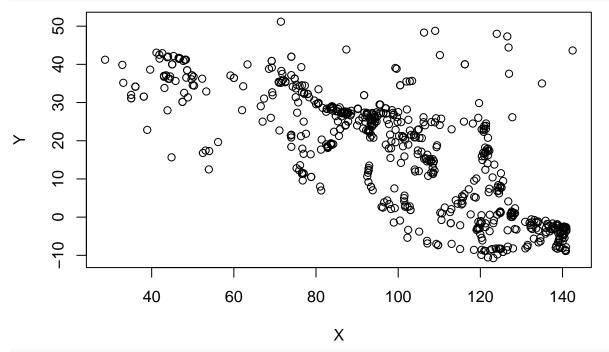




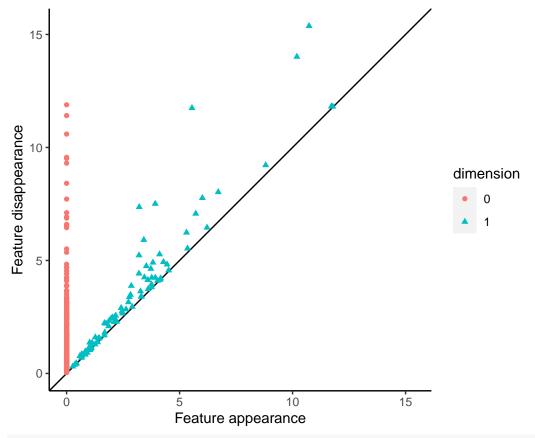




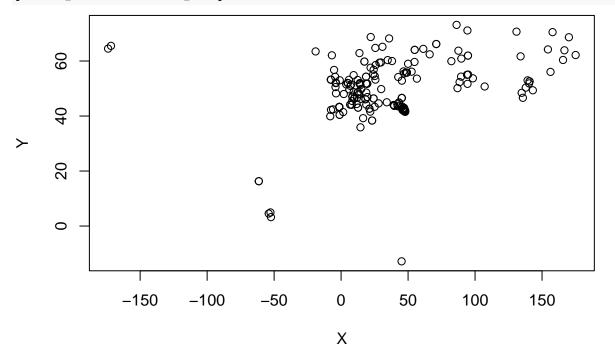
plot(st_coordinates(wals_asia))



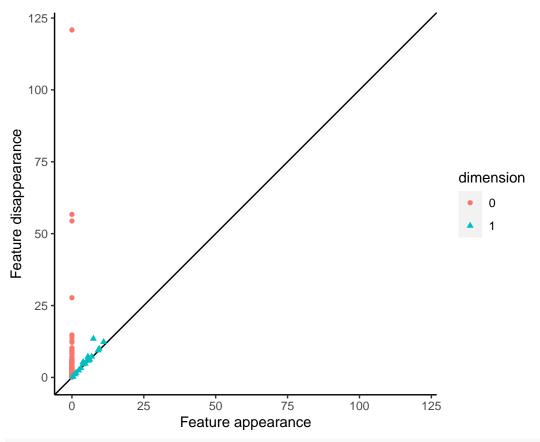
plot_persist(calculate_homology(st_coordinates(wals_asia)))



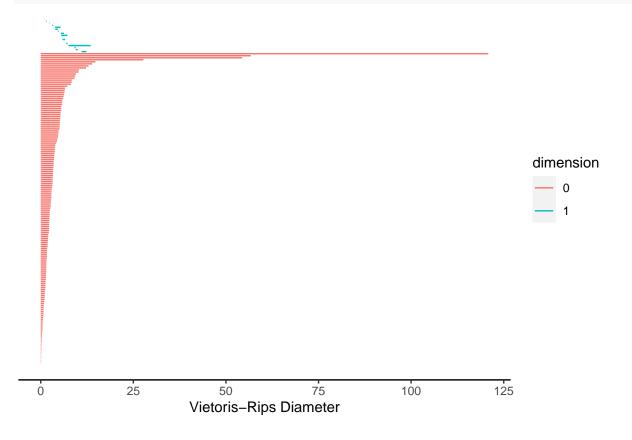
wals_europe <- subset(wals, continent=="Europe")
plot(st_coordinates(wals_europe))</pre>



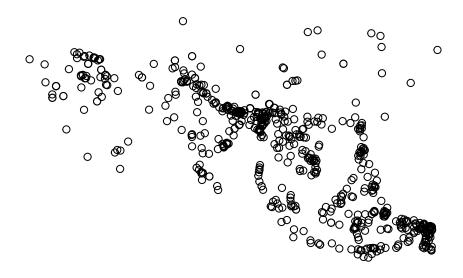
europe_phm <- calculate_homology(st_coordinates(wals_europe))
plot_persist(europe_phm)</pre>



plot_barcode(europe_phm)



```
wals_asia_geo <- st_geometry(wals_asia)</pre>
wals_asia_geo
## Geometry set for 599 features
## Geometry type: POINT
## Dimension:
## Bounding box: xmin: 28.6632 ymin: -10.7326 xmax: 142.4617 ymax: 51.17
## Geodetic CRS: WGS 84
## First 5 geometries:
## POINT (36.0468 34.1709)
## POINT (41.15911 43.05622)
## POINT (132.416 -0.57073)
## POINT (124.588 -8.31058)
## POINT (96.6032 3.90757)
plot_persist(calculate_homology(st_coordinates(wals_asia_geo)))
   15
Feature disappearance
                                                                   dimension
                         5
                                          10
                                                           15
                        Feature appearance
plot(st_geometry(wals_asia))
```



References

- [1] Shyam Boriah, Varun Chandola, and Vipin Kumar. Similarity measures for categorical data: A comparative evaluation. In SDM, 2008.
- [2] Nieves Atienza, Rocio Gonzalez-Díaz, and Manuel Soriano-Trigueros. On the stability of persistent entropy and new summary functions for topological data analysis. *Pattern Recognition*, 107:107509, nov 2020.