# PCA: Genomic

## Import data and library

```
library(ggplot2)
data <- read.table("p4dataset2023.txt", header = FALSE, stringsAsFactors = FALSE)</pre>
```

#### Define data matrix X

```
meta <- data[,c(1:3)] # The first three columns are metadata
raw_data <- data[,-c(1:3)]

# Define a function to find the mode of each column
find_modes <- function(col){
    count <- table(col)
    max_count <- max(count)
    mode <- names(count[count ==max_count])
    return(mode)
}

modes <- sapply(raw_data,find_modes)

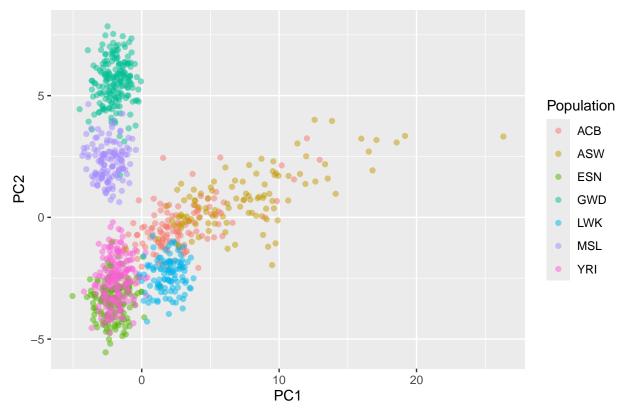
# Create a binary matrix X
X <- matrix(0, nrow = nrow(raw_data), ncol = ncol(raw_data))
for(i in 1:ncol(raw_data)){
    X[,i] <- ifelse(raw_data[,i] == modes[i], 0, 1)
}
X <- as.data.frame(X)</pre>
```

## Perform PCA on sample covariance matrix of X

```
pca <- prcomp(X, center = TRUE, scale = FALSE)</pre>
```

### Plot of PC1 and PC2

### Scatter Plot of PC1 and PC2

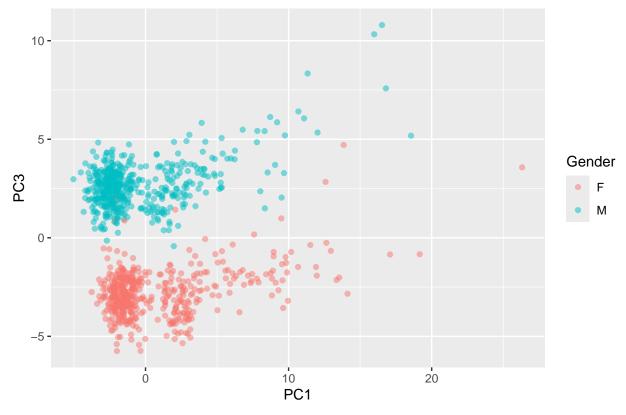


### Interpretation

• From the above scatter plot, the first two principal components PC1 and PC2 preserve the information of geographical location and the proximity of the populations.

## Plot of PC1 and PC3

## Scatter Plot of PC1 and PC3

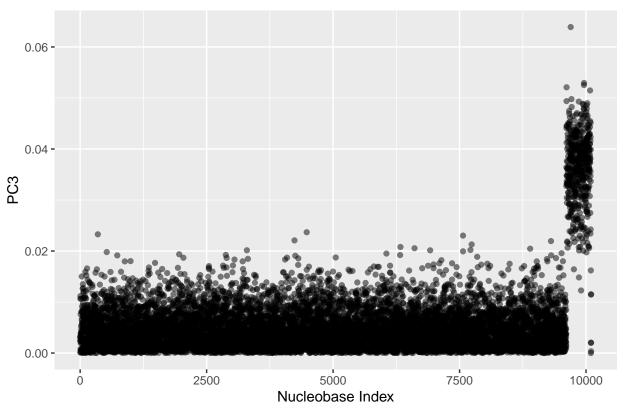


## Interpretation

• From the above scatter plot, the third principal component PC3 preserve the information related to gender.

### Plot of nucleobase index and PC3

### Scatter Plot of Nucleobase Index and PC3



### Interpretation

- The absolute value of the third principal component PC3 is significantly larger in the latter part of the nucleobase index
- It is possibly because of differences in the number and type of genes on the X and Y chromosomes, and PC3 captures this difference