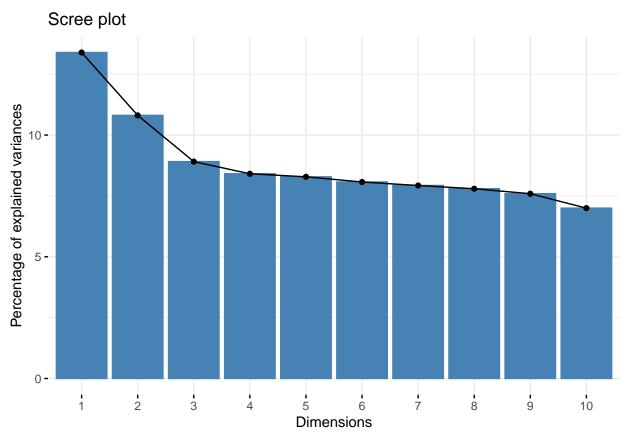
Quiz 9

Chaeeun Shin

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```
library(caret)
## Loading required package: ggplot2
## Loading required package: lattice
library(neuralnet)
library(randomForest)
## randomForest 4.7-1.1
## Type rfNews() to see new features/changes/bug fixes.
##
## Attaching package: 'randomForest'
## The following object is masked from 'package:ggplot2':
##
##
      margin
library(rpart)
library(rattle)
## Loading required package: tibble
## Loading required package: bitops
## Rattle: A free graphical interface for data science with R.
## Version 5.5.1 Copyright (c) 2006-2021 Togaware Pty Ltd.
## Type 'rattle()' to shake, rattle, and roll your data.
## Attaching package: 'rattle'
## The following object is masked from 'package:randomForest':
##
##
       importance
library(MASS)
library(tidyverse)
## Warning: package 'tidyr' was built under R version 4.2.3
## Warning: package 'readr' was built under R version 4.2.3
## Warning: package 'dplyr' was built under R version 4.2.3
## Warning: package 'stringr' was built under R version 4.2.3
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr
             1.1.4
                        v readr
                                     2.1.5
```

```
## v forcats 1.0.0 v stringr
                                      1.5.1
## v lubridate 1.9.3
                      v tidyr
                                      1.3.1
## v purrr
              1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::combine()
                          masks randomForest::combine()
## x dplyr::compute()
                          masks neuralnet::compute()
## x dplyr::filter()
                          masks stats::filter()
## x dplyr::lag()
                            masks stats::lag()
## x purrr::lift()
                            masks caret::lift()
## x randomForest::margin() masks ggplot2::margin()
## x dplyr::select()
                            masks MASS::select()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(glmnet)
## Loading required package: Matrix
## Attaching package: 'Matrix'
##
## The following objects are masked from 'package:tidyr':
##
##
       expand, pack, unpack
##
## The following object is masked from 'package:bitops':
##
##
       %&%
##
## Loaded glmnet 4.1-8
library(leaps)
library(ggplot2)
  1.
 (a)
GreatUnknown <- read.csv("GreatUnknown.csv")</pre>
GreatUnknown <- na.omit(GreatUnknown)</pre>
data <- scale(GreatUnknown[-13])</pre>
pc <- princomp(data,cor=TRUE)</pre>
 (b)
library(factoextra)
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
fviz_eig(pc)
```

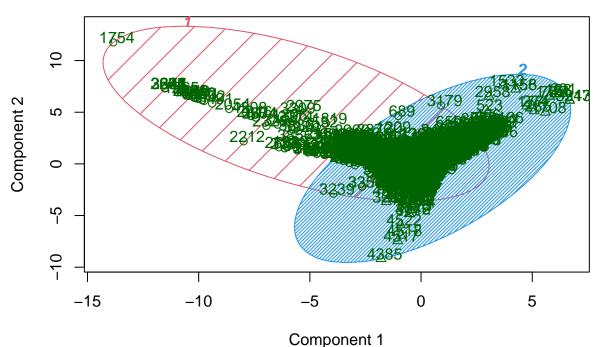


K-means fit

k.means.fit <- kmeans(data,2)
library(cluster)</pre>

clusplot(data,k.means.fit\$cluster,main="2D representation of the Cluster solution",color=TRUE,shade=TRUE

2D representation of the Cluster solution



These two components explain 24.2 % of the point variability.

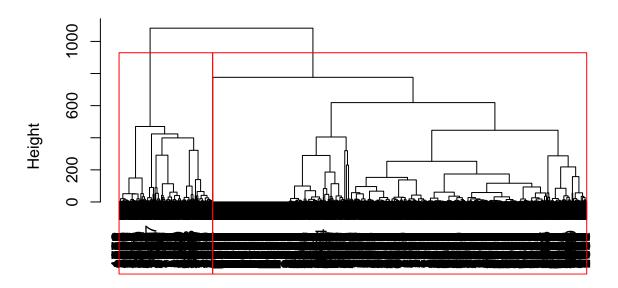
table(k.means.fit\$cluster, GreatUnknown\$y)

accuracy = (2643+913)/4601

H.ward

```
d <- dist(data,method="euclidean")
H.fit <- hclust(d,method="ward.D")
plot(H.fit)
rect.hclust(H.fit,k=2,border="red")</pre>
```

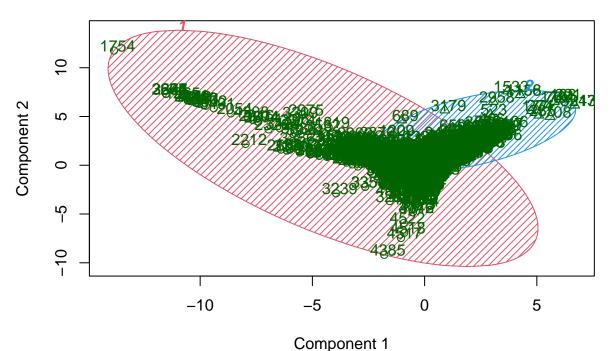
Cluster Dendrogram



d hclust (*, "ward.D")

```
groups <- cutree(H.fit,k=2)
clusters <- factor(groups,levels=1:2,labels=c(0,1))
clusplot(data,groups,main="2D representation of the Cluster solution",color=TRUE,shade=TRUE,labels=2,line</pre>
```

2D representation of the Cluster solution



These two components explain 24.2 % of the point variability.

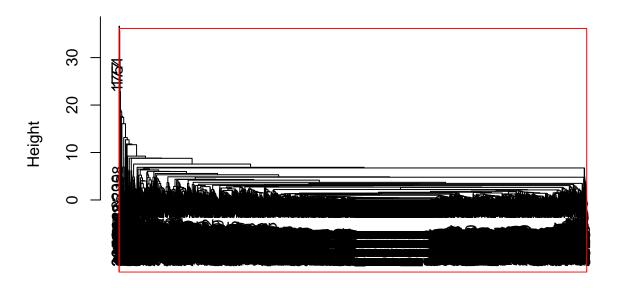
table(GreatUnknown\$y,clusters)

```
## clusters
## 0 1
## 0 2621 167
## 1 1059 754
# accuracy = (2621+754)/4601
```

H.Average

```
H.fit <- hclust(d,method="average")
plot(H.fit)
rect.hclust(H.fit,k=2,border="red")</pre>
```

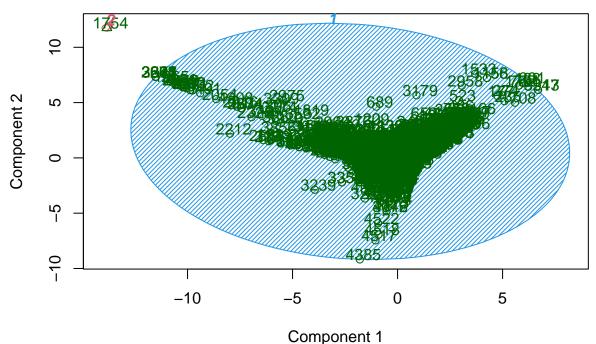
Cluster Dendrogram



d hclust (*, "average")

```
groups <- cutree(H.fit,k=2)
clusters <- factor(groups,levels=1:2,labels=c(0,1))
clusplot(data,groups,main="2D representation of the Cluster solution",color=TRUE,shade=TRUE,labels=2,line</pre>
```

2D representation of the Cluster solution



These two components explain 24.2 % of the point variability.

table(GreatUnknown\$y,clusters)

```
## clusters

## 0 1

## 0 2788 0

## 1 1812 1

# accuracy = (2788+1)/4601
```

Comparison: K-means>H.ward>H.average Based on their accuracy on confusion matrix.

2.

(a)

```
library(devtools)

## Loading required package: usethis

## Warning: package 'usethis' was built under R version 4.2.3

library(ggbiplot)

## Warning: package 'ggbiplot' was built under R version 4.2.3

##

## Attaching package: 'ggbiplot'

## The following object is masked from 'package:rattle':

##

## wine
```

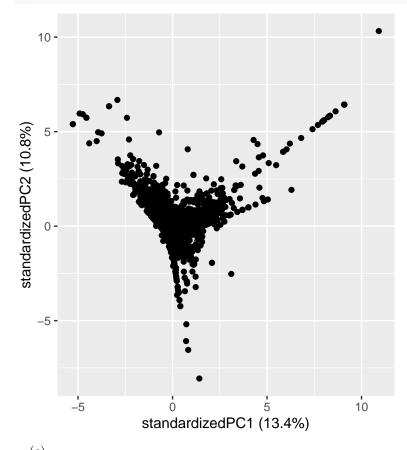
```
GreatUnknown.pca <- prcomp(data,center=TRUE,scale.=TRUE)
summary(GreatUnknown.pca)</pre>
```

```
## Importance of components:
                                           PC3
                                                    PC4
                                                            PC5
##
                             PC1
                                    PC2
                                                                    PC6
                                                                            PC7
## Standard deviation
                          1.2677 1.1389 1.0340 1.00462 0.99711 0.98418 0.97543
## Proportion of Variance 0.1339 0.1081 0.0891 0.08411 0.08285 0.08072 0.07929
## Cumulative Proportion 0.1339 0.2420 0.3311 0.41522 0.49807 0.57879 0.65807
##
                              PC8
                                      PC9
                                             PC10
                                                      PC11
                                                              PC12
## Standard deviation
                          0.96718 0.95444 0.91643 0.89253 0.78756
## Proportion of Variance 0.07795 0.07591 0.06999 0.06638 0.05169
## Cumulative Proportion 0.73603 0.81194 0.88193 0.94831 1.00000
```

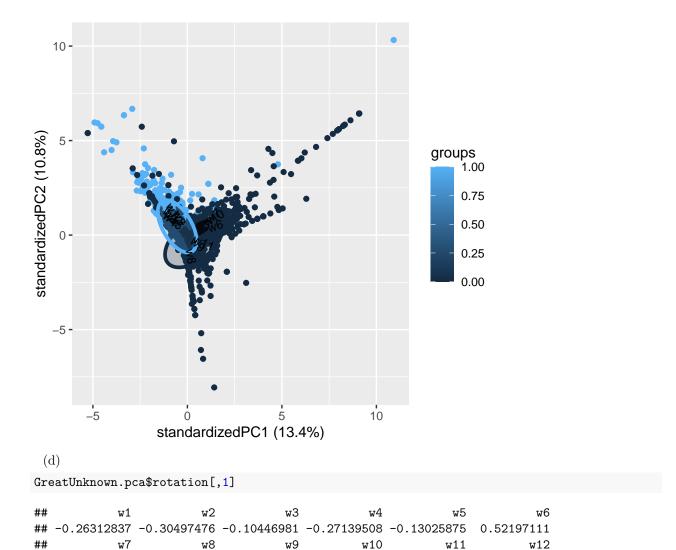
By the row "Proportion of Variance", we can find what percentage of the variations each PC would explain.

(b)

ggbiplot(GreatUnknown.pca)



ggbiplot(GreatUnknown.pca, ellipse=TRUE, groups=GreatUnknown\$y)



PC1 contains parts of each $w1\sim w12$ that accounts for the greatest possible variance.

0.49664611 0.05810733 0.08953016 0.40916977 0.10868236 -0.16273359