

# Model 3

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## Helper Packages And Modeling Packages

```
# Load Helper Packages
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(ggplot2)

## Warning: package 'ggplot2' was built under R version 4.2.2
library(stringr)
library(gridExtra)

##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##   combine
library(readr)
library(bestNormalize)

## Warning: package 'bestNormalize' was built under R version 4.2.2
# Load Modeling Packages
library(cluster)
library(factoextra)

## Warning: package 'factoextra' was built under R version 4.2.2
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.2 --
## v tibble  3.1.8     v purrr   0.3.5
```

```

## v tidyverse 1.2.1      vforcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x gridExtra::combine() masks dplyr::combine()
## x dplyr::filter()     masks stats::filter()
## x dplyr::lag()        masks stats::lag()
library(mclust)

## Warning: package 'mclust' was built under R version 4.2.2
## Package 'mclust' version 6.0.0
## Type 'citation("mclust")' for citing this R package in publications.
##
## Attaching package: 'mclust'
##
## The following object is masked from 'package:purrr':
##
##     map

```

### *CLUSTERING TECHNIQUES*

In this model, we will execute and differentiate the clustering techniques results from K-Means, Hierarchical and Model based clustering disregarding the binary output and categorical variables in the data set. We will start by normalizing the data “*radiomics\_completedata*”.

## Loading the Data

The data contains 197 rows and 431 columns with *Failure.binary* binary output.

```

Model3Arawdata <- read_csv("radiomics_completedata.csv")

## Rows: 197 Columns: 431
## -- Column specification --
## Delimiter: ","
## chr (1): Institution
## dbl (430): Failure.binary, Failure, Entropy_cooc.W.ADC, GLNU_align.H.PET, Mi...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.

Model3Arawdata

## # A tibble: 197 x 431
##   Institution Failure~1 Failure Entro~2 GLNU_~3 Min_h~4 Max_h~5 Mean_~6 Varia~7
##   <chr>          <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 A            0       49.3     12.9     46.3     6.25     17.8     9.78
## 2 A            1       12.6     12.2     27.5     11.0     26.5     15.4
## 3 A            0       79.8     12.8     90.2     2.78     6.88     4.30
## 4 A            1       17.9     13.5     326.     6.30     22.0     10.3
## 5 A            0       39.6     12.6     89.6     3.58     7.92     4.45
## 6 A            1       4.77    13.2     102.     2.60     6.21     3.77
## 7 A            0       25.0     12.2     36.8     8.65     28.2     14.9
## 8 A            0       35.8     12.3     51.0     5.71     13.0     7.73
## 9 A            1       35.3     13.4     27.2     5.88     14.8     9.12

```

```

## 10 A           1   17.8    12.6   20.2   5.70   17.1    8.55   6.12
## # ... with 187 more rows, 422 more variables:
## #   Standard_Deviation_hist.PET <dbl>, Skewness_hist.PET <dbl>,
## #   Kurtosis_hist.PET <dbl>, Energy_hist.PET <dbl>, Entropy_hist.PET <dbl>,
## #   AUC_hist.PET <dbl>, H_suv.PET <dbl>, Volume.PET <dbl>,
## #   X3D_surface.PET <dbl>, ratio_3ds_vol.PET <dbl>,
## #   ratio_3ds_vol_norm.PET <dbl>, irregularity.PET <dbl>,
## #   tumor_length.PET <dbl>, Compactness_v1.PET <dbl>, ...

```

## *Data Reprocessing*

### Checking for null and missing values

We are using *anyNA()* function to determine if there is any missing value in the data.

```

anyNA(Model3Arawdata)

## [1] FALSE
#The output will show either *True* or *False*. There are missing values If True, thus you have to omit them.

#[1] FALSE

# The result is False, hence, the data has no missing values.

```

Checking the Normality of the Data We are using *Shapiro-Wilk's Test* to check the normality of the data.

```

Model3Anumrawdata <- Model3Arawdata %>% select_if(is.numeric)

Model3Anumrawdata <- Model3Anumrawdata[ , -1]

Model3ASWtest <- apply(Model3Anumrawdata, 2, function(x){shapiro.test(x)})

```

Next we need to list only the p-value of the respective variables to proceed with the test. We are using the *unlist()* and *lapply()* functions to achieve this goal.

```

Model3ADRpvalue <- unlist(lapply(Model3ASWtest, function(x) x$p.value))

sum(Model3ADRpvalue<0.05) # not normally distributed

## [1] 428

sum(Model3ADRpvalue>0.05) # normally distributed

## [1] 1

Model3ASWtest$Entropy_cooc.W.ADC

##
##  Shapiro-Wilk normality test
##
## data: x
## W = 0.98903, p-value = 0.135

```

```

# [1] 428
# [1] 1

# Currently, there are 428 variables that are not normally distributed and only the Entropy_cooc.W.ADC

```

The goal is that all variables should be normally distributed. Next, we are using `orderNorm()` function. And we need to exclude the `Entropy_cooc.W.ADC` since it is already normally distributed.

```

Model3ADRtransrawdata <- Model3Arawdata[,c(3,5:length(names(Model3Arawdata)))]]

Model3ADRtransrawdata <- apply(Model3ADRtransrawdata,2,orderNorm)

Model3ADRtransrawdata <- lapply(Model3ADRtransrawdata, function(x) x$x.t)

Model3ADRtransrawdata <- Model3ADRtransrawdata%>%as.data.frame()

Model3ASWtest <- apply(Model3ADRtransrawdata,2,shapiro.test)

Model3ASWtest <- unlist(lapply(Model3ASWtest, function(x) x$p.value))

```

Next, we will be testing the data to check the normality or the transformed data.

```

sum(Model3ASWtest <0.05) # for not normally distributed

## [1] 0

sum(Model3ASWtest >0.05) # for normally distributed

## [1] 428

#[1] 0
#[1] 428

# Now, the 428 variables that were initially not normally distributed are now normally distributed.

Model3Arawdata[,c(3,5:length(names(Model3Arawdata)))] = Model3ADRtransrawdata

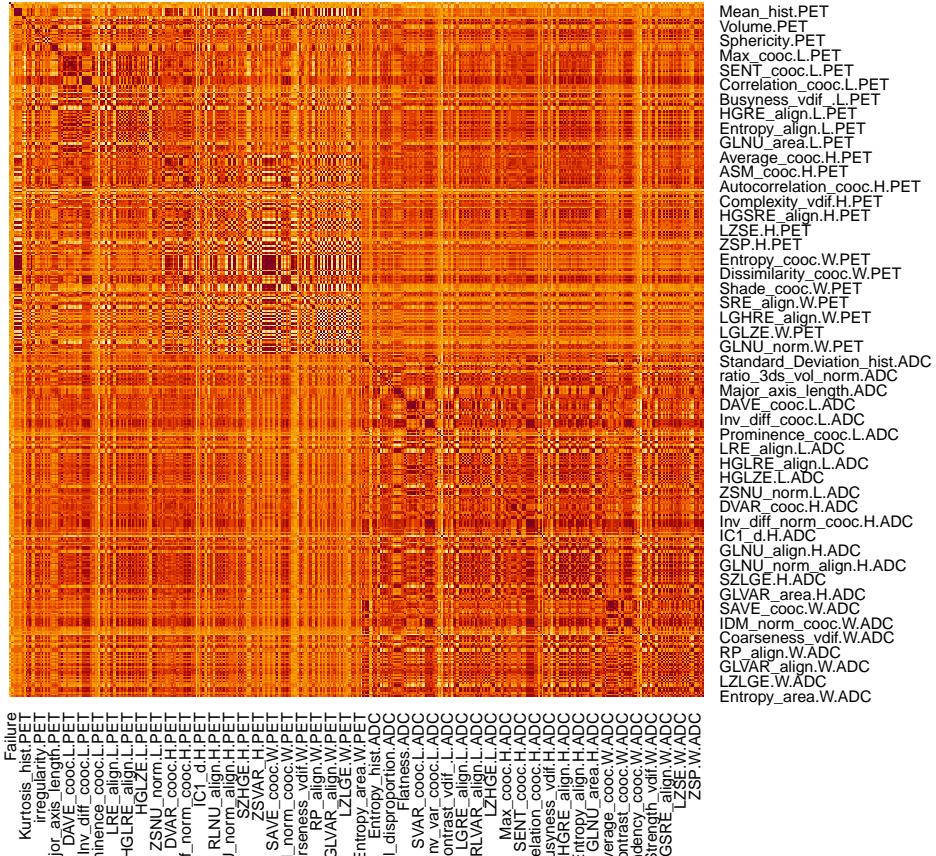
```

We are getting the correlation of the whole data except the categorical variables

```

Model3ACorrMat = cor(Model3Arawdata[,-c(1,2)])
heatmap(Model3ACorrMat,Rowv=NA,Colv=NA,scale="none",revC = T)

```



Finally, we will convert the data frame output of data reprocessing into “csv” file, which will we use for the entire Final Project.

```
library(data.table)

##
## Attaching package: 'data.table'

## The following object is masked from 'package:purrr':
##
##     transpose

## The following objects are masked from 'package:dplyr':
##
##     between, first, last

fwrite(Model3Arawdata, "Model3_Final_Project_Data.csv")
```

Lastly, let's check if the dataframe we have exported to CSV is really the normal data.

```
Model3Arawdata1 <- read_csv("Model3_Final_Project_Data.csv")

## Rows: 197 Columns: 431
## -- Column specification --
## Delimiter: ","
## chr  (1): Institution
## dbl (430): Failure.binary, Failure, Entropy_cooc.W.ADC, GLNU_align.H.PET, Mi...
##
## i Use `spec()` to retrieve the full column specification for this data.
```

```

## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
Model3Arawdata1

## # A tibble: 197 x 431
##   Institution Failure Entro~2 GLNU_~3 Min_h~4 Max_h~5 Mean_~6 Varia~7
##   <chr>          <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 A            0     1.15    12.9   -0.433   -0.270   -0.257   -0.192   0.0509
## 2 A            1    -0.533   12.2   -1.02    0.671    0.405    0.490    0.687
## 3 A            0     2.24    12.8    0.179   -1.41    -1.57    -1.53    -1.57
## 4 A            1    -0.140   13.5    2.00    -0.218   0.0764   -0.153   0.0127
## 5 A            0     0.787   12.6    0.153   -1.06    -1.15    -1.45    -1.91
## 6 A            1    -2.80    13.2    0.391   -1.57    -1.91    -1.72    -1.84
## 7 A            0     0.218   12.2   -0.687    0.284    0.519    0.405    0.915
## 8 A            0     0.623   12.3   -0.405   -0.461   -0.719   -0.639   -0.623
## 9 A            1     0.578   13.4   -1.04    -0.391   -0.593   -0.378   -0.244
## 10 A           1    -0.160   12.6   -1.38    -0.475   -0.310   -0.475   -0.0127
## # ... with 187 more rows, 422 more variables:
## #   Standard_Deviation_hist.PET <dbl>, Skewness_hist.PET <dbl>,
## #   Kurtosis_hist.PET <dbl>, Energy_hist.PET <dbl>, Entropy_hist.PET <dbl>,
## #   AUC_hist.PET <dbl>, H_suv.PET <dbl>, Volume.PET <dbl>,
## #   X3D_surface.PET <dbl>, ratio_3ds_vol.PET <dbl>,
## #   ratio_3ds_vol_norm.PET <dbl>, irregularity.PET <dbl>,
## #   tumor_length.PET <dbl>, Compactness_v1.PET <dbl>, ...
Model3Anumrawdata1 <- Model3Arawdata1 %>% select_if(is.numeric)

Model3Anumrawdata1 <- Model3Anumrawdata1[ , -1]

Model3ASWtest1 <- apply(Model3Anumrawdata1, 2, function(y){shapiro.test(y)})

Model3ADRpvalue1 <- unlist(lapply(Model3ASWtest1, function(y) y$p.value))

sum(Model3ADRpvalue1<0.05) # not normally distributed

## [1] 0
sum(Model3ADRpvalue1>0.05) # normally distributed

## [1] 429
#Yes! We were able to produce the correct CSV file and we are now ready to use it for the entire project

```

### MODEL 3

```

# Load the dataset
model3data <- read_csv("Model3_Final_Project_Data.csv")

## Rows: 197 Columns: 431
## -- Column specification -----
## Delimiter: ","
## chr  (1): Institution
## dbl (430): Failure.binary, Failure, Entropy_cooc.W.ADC, GLNU_align.H.PET, Mi...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.

```

```

model3data

## # A tibble: 197 x 431
##   Institution Failure~1 Failure Entro~2 GLNU_~3 Min_h~4 Max_h~5 Mean_~6 Varia~7
##   <chr>          <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 A              0     1.15     12.9   -0.433   -0.270   -0.257   -0.192   0.0509
## 2 A              1    -0.533    12.2   -1.02     0.671    0.405     0.490   0.687
## 3 A              0     2.24     12.8    0.179   -1.41    -1.57    -1.53   -1.57
## 4 A              1    -0.140    13.5    2.00    -0.218    0.0764   -0.153   0.0127
## 5 A              0     0.787    12.6    0.153   -1.06    -1.15    -1.45   -1.91
## 6 A              1    -2.80     13.2    0.391   -1.57    -1.91    -1.72   -1.84
## 7 A              0     0.218    12.2   -0.687    0.284    0.519     0.405   0.915
## 8 A              0     0.623    12.3   -0.405   -0.461   -0.719   -0.639   -0.623
## 9 A              1     0.578    13.4   -1.04    -0.391   -0.593   -0.378   -0.244
## 10 A             1    -0.160    12.6   -1.38    -0.475   -0.310   -0.475   -0.0127
## # ... with 187 more rows, 422 more variables:
## #   Standard_Deviation_hist.PET <dbl>, Skewness_hist.PET <dbl>,
## #   Kurtosis_hist.PET <dbl>, Energy_hist.PET <dbl>, Entropy_hist.PET <dbl>,
## #   AUC_hist.PET <dbl>, H_suv.PET <dbl>, Volume.PET <dbl>,
## #   X3D_surface.PET <dbl>, ratio_3ds_vol.PET <dbl>,
## #   ratio_3ds_vol_norm.PET <dbl>, irregularity.PET <dbl>,
## #   tumor_length.PET <dbl>, Compactness_v1.PET <dbl>, ...

```

## Scaling/Standardizing the Data

```

model3fdata <- scale(model3data[c(3:431)])
head(model3fdata)

## # A tibble: 6 x 431
##   Failure Entropy_cooc.W.ADC GLNU_align.H.PET Min_hist.PET Max_hist.PET
##   <dbl>    <dbl>        <dbl>        <dbl>        <dbl>
## 1 1.1542812 0.55290547   -0.4333655   -0.2706594   -0.25747509
## 2 -0.5335831 -0.06486729  -1.0166957    0.6709820    0.40556015
## 3 2.2372022  0.45990825   0.1792170   -1.4150181   -1.56968040
## 4 -0.1405364 1.14318298   1.9997211   -0.2181792   0.07647307
## 5 0.7872480  0.34499368   0.1533960   -1.0603294   -1.15426355
## 6 -2.8042412 0.84917904   0.3917753   -1.5696804   -1.91367122
## # ... with 422 more variables: Mean_hist.PET <dbl>,
## #   Variance_hist.PET <dbl>,
## #   Standard_Deviation_hist.PET <dbl>,
## #   Skewness_hist.PET <dbl>,
## #   Kurtosis_hist.PET <dbl>,
## #   Energy_hist.PET <dbl>,
## #   Entropy_hist.PET <dbl>,
## #   AUC_hist.PET <dbl>,
## #   H_suv.PET <dbl>,
## #   Volume.PET <dbl>,
## #   X3D_surface.PET <dbl>,
## #   ratio_3ds_vol.PET <dbl>,
## #   ratio_3ds_vol_norm.PET <dbl>,
## #   irregularity.PET <dbl>,
## #   tumor_length.PET <dbl>,
## #   Compactness_v1.PET <dbl>,
## #   ...
## #   Failure Entropy_cooc.W.ADC GLNU_align.H.PET Min_hist.PET Max_hist.PET
##   <dbl>    <dbl>        <dbl>        <dbl>        <dbl>
## 1 -0.1921714 0.05095448   -0.06370859
## 2 0.4900236  0.68701098   0.50442967
## 3 -1.5275205 -1.56968040  -1.48790984
## 4 -0.1533960 0.01273346   -0.08924999
## 5 -1.4505021 -1.91367122  -1.91367122
## 6 -1.7165060 -1.83980971  -1.83980971
## # ... with 422 more variables: Skewness_hist.PET <dbl>,
## #   Kurtosis_hist.PET <dbl>,
## #   Energy_hist.PET <dbl>,
## #   Entropy_hist.PET <dbl>,
## #   AUC_hist.PET <dbl>,
## #   H_suv.PET <dbl>,
## #   Volume.PET <dbl>,
## #   X3D_surface.PET <dbl>,
## #   ratio_3ds_vol.PET <dbl>,
## #   ...
## #   Failure Entropy_cooc.W.ADC GLNU_align.H.PET Min_hist.PET Max_hist.PET
##   <dbl>    <dbl>        <dbl>        <dbl>        <dbl>
## 1 -0.1533960 -0.3105060   -0.19217136  -0.1533960
## 2 0.1405248  -0.2443353   -0.08924999  -0.7700043
## 3 -1.3180073 -1.4505021   -0.20515797  -0.1405248
## 4 0.1792170  0.8586682   -0.46150957  0.2181792
## 5 1.1298434  1.4879098   0.01273346  -0.9546481
## 6 -0.2838909 0.1020415   -0.17921704  0.2312375
## # ... with 422 more variables: H_suv.PET <dbl>,
## #   Volume.PET <dbl>,
## #   X3D_surface.PET <dbl>,
## #   ratio_3ds_vol.PET <dbl>,
## #   ...
## #   Failure Entropy_cooc.W.ADC GLNU_align.H.PET Min_hist.PET Max_hist.PET
##   <dbl>    <dbl>        <dbl>        <dbl>        <dbl>
## 1 -0.5189411 0.05095448  -0.8586682   -0.8404111  -0.2443353
## 2 -0.3508513 1.03826584  -1.4150181   -0.5631547   0.6394275
## 3 -0.7032182 -1.28828023 -0.2181792   0.2838909  -0.2706594

```

```

## [4,] 0.5932479 -0.25747509 0.4194224      0.9749140      -0.8224296
## [5,] 0.5781359 -1.61482620 -0.1405248     -0.6394275      0.6238874
## [6,] -0.3780645 -1.41501808 -0.7872387     -0.4055602      0.1533960
##      ratio_3ds_vol_norm.PET irregularity.PET tumor_length.PET
## [1,] -0.57813595      0.32389526      -0.33734333
## [2,] 0.39177534      0.49002356      -0.51894198
## [3,] 0.14052475     -0.02546898      -0.01273363
## [4,] 0.27065942     -1.34894519      0.68701181
## [5,] 0.37806448      0.33734270      -0.75299666
## [6,] 0.01273346      0.05095448      -0.21817968
##      Compactness_v1.PET Compactness_v2.PET Spherical_disproportion.PET
## [1,] -0.3038327     -0.4972144      -0.57813595
## [2,] -0.6709825     -0.9152478      0.39177534
## [3,] -0.5482991     -0.7700052      0.14052475
## [4,] -0.5932484     -0.8224305      0.27065942
## [5,] -0.6551238     -0.8960704      0.37806448
## [6,] -0.5044301     -0.7196129      0.01273346
##      Sphericity.PET Asphericity.PET Center_of_mass.PET Max_3D_diam.PET
## [1,] -0.1921714     -0.57813595      0.36442422      -0.6238874
## [2,] -0.8404111     0.39177534     -0.03820863      -0.7700043
## [3,] -0.6551233     0.14052475     -0.75299544      -0.2838909
## [4,] -0.7196121     0.27065942     0.47571828      0.3105060
## [5,] -0.8224296     0.37806448     -0.19217136      -0.8224296
## [6,] -0.5781359     0.01273346     -1.06032943      -0.5044297
##      Major_axis_length.PET Minor_axis_length.PET Least_axis_length.PET
## [1,] -0.8404111     -0.8047097      -0.3780645
## [2,] -0.7872387     -0.6394275      -0.7529954
## [3,] -0.2574751     -0.4900236      -0.2971722
## [4,] 0.2971722      0.7872387      0.9546481
## [5,] -0.7700043     -1.1298434      -0.7362013
## [6,] -0.4757183     -0.4473933      -1.3489452
##      Elongation.PET Flatness.PET Max_cooc.L.PET Average_cooc.L.PET
## [1,] -0.23123752    0.3238953     -0.35085134      -0.10204149
## [2,] -0.07647307    -0.1662927     0.03820863      -0.29717220
## [3,] -0.82242955    -0.3917753     -0.32389526      0.47571828
## [4,] 0.27065942    0.7196121     -0.15339601      -1.17939112
## [5,] -0.67098197    -0.2574751     -0.02546898      -1.31800728
## [6,] -0.29717220    -1.3180073     -0.27065942      -0.05095448
##      Variance_cooc.L.PET Entropy_cooc.L.PET DAVE_cooc.L.PET DVAR_cooc.L.PET
## [1,] 0.06370859      0.1020415     -0.11484968      -0.2971722
## [2,] 0.33734270      -0.7529954     0.46150957      0.6394275
## [3,] 0.11484968      0.3780645     -0.05095448      -0.2838909
## [4,] -1.83980971     -0.7700043     -1.34894519      -1.5275205
## [5,] -0.62388738     -1.2052851     -0.49002356      -0.3508513
## [6,] -0.20515797     0.1148497     -0.14052475      -0.2706594
##      DENT_cooc.L.PET SAVE_cooc.L.PET SVAR_cooc.L.PET SENT_cooc.L.PET
## [1,] -0.1662927     -0.10204149    0.2181792      0.23123752
## [2,] 0.4055602      -0.29717220    0.1662927      0.05095448
## [3,] -0.1405248     0.47571828     0.2706594      0.46150957
## [4,] -1.3489452     -1.17939112    -1.3489452      -0.82242955
## [5,] -0.5044297     -1.31800728    -0.8224296      -0.89606941
## [6,] -0.2051580     -0.06370859    -0.1792170      -0.06370859
##      ASM_cooc.L.PET Contrast_cooc.L.PET Dissimilarity_cooc.L.PET
## [1,] -2.838909e-01     -0.03820863    -0.11484968

```

```

## [2,] -1.020415e-01      0.63942746      0.46150957
## [3,] -3.780645e-01     -0.02546898     -0.05095448
## [4,] -7.647307e-02     -1.28828023     -1.34894519
## [5,] -5.421149e-17     -0.35085134     -0.49002356
## [6,] -3.508513e-01     -0.07647307     -0.14052475
##   Inv_diff_cooc.L.PET Inv_diff_norm_cooc.L.PET IDM_cooc.L.PET
## [1,] -0.37806448      -0.33734270     -0.44739330
## [2,] -0.62388738      -0.75299544     -0.59324787
## [3,] -0.65512329      -0.44739330     -0.70321822
## [4,] 0.40556015       0.40556015      0.41942235
## [5,] 0.05095448       0.02546898      0.06370859
## [6,] -0.36442422      -0.28389090     -0.39177534
##   IDM_norm_cooc.L.PET Inv_var_cooc.L.PET Correlation_cooc.L.PET
## [1,] -0.32389526      -0.4194224        0.02546898
## [2,] -0.78723867      -0.5482987      -1.08292088
## [3,] -0.37806448      -0.6394275      0.06370859
## [4,] 0.36442422       0.4757183       0.39177534
## [5,] 0.05095448       0.1148497      -0.54829867
## [6,] -0.24433531      -0.3780645      -0.24433531
##   Autocorrelation_cooc.L.PET Tendency_cooc.L.PET Shade_cooc.L.PET
## [1,] -1.273346e-02     0.2181792      0.35085134
## [2,] -1.405248e-01     0.1662927     -0.05095448
## [3,] 7.529954e-01      0.2706594     -1.45050212
## [4,] -1.179391e+00     -1.3489452     -0.21817922
## [5,] -1.415018e+00     -0.8224296      0.75299544
## [6,] -5.421149e-17     -0.1792170     -0.08924999
##   Prominence_cooc.L.PET IC1_.L.PET IC2_.L.PET Coarseness_vdif_.L.PET
## [1,] 0.29717220  0.1148497 -0.03820863     -1.276768e-01
## [2,] 0.11484968 -0.1276767  0.07647307     -1.405249e-01
## [3,] 0.10204149  0.3373427 -0.24433531     -6.370871e-02
## [4,] -1.15426355  0.9546481 -0.77000435     -5.335627e-01
## [5,] -0.08924999  0.3780645 -0.36442422     -9.599724e-08
## [6,] -0.14052475  0.3105060 -0.23123752     -7.647318e-02
##   Contrast_vdif_.L.PET Busyness_vdif_.L.PET Complexity_vdif_.L.PET
## [1,] 0.10204149      -0.4900236     -0.03820863
## [2,] 0.70321822      -0.5631547      0.56315472
## [3,] 0.07647307      -0.8772156     -0.35085134
## [4,] -1.52752054      0.8960694      -0.99558794
## [5,] -0.11484968      -0.5482987     -0.08924999
## [6,] -0.08924999      -0.6394275     -0.36442422
##   Strength_vdif_.L.PET SRE_align.L.PET LRE_align.L.PET GLNU_align.L.PET
## [1,] 0.23123752 -1.533960e-01 -0.3780645     -0.5044297
## [2,] 0.37806448  6.370859e-02 -0.6709820     -0.6084963
## [3,] 0.05095448 -5.421149e-17 -0.6870110     -0.5482987
## [4,] -1.10607816 -6.238874e-01  0.3917753      1.1542636
## [5,] 0.80470972 -1.792170e-01 -0.4333655     -0.4615096
## [6,] 0.11484968 -2.312375e-01 -0.3644242     -0.5189411
##   RLNU_align.L.PET RP_align.L.PET LGRE_align.L.PET HGRE_align.L.PET
## [1,] -0.4473933 -0.12767671  0.05095448     -0.03820863
## [2,] -0.7529954  0.03820863  0.10204149     -0.10204149
## [3,] -0.3917753  0.02546898 -0.54829867     0.56315472
## [4,] 1.2596485 -0.65512329 -0.35085134     -1.12984341
## [5,] -0.7700043 -0.11484968  0.51894110     -1.38122903
## [6,] -0.3780645 -0.17921704 -0.33734270     0.08924999

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##      LGSRE_align.L.PET HGSRE_align.L.PET LGHRE_align.L.PET HGLRE_align.L.PET
## [1,] 0.06370859 -0.01273346 0.03820863 -0.07647307
## [2,] 0.10204149 -0.10204149 0.19217136 -0.23123752
## [3,] -0.54829867 0.56315472 -0.60849632 0.57813595
## [4,] -0.35085134 -1.12984341 -0.28389090 -1.10607816
## [5,] 0.54829867 -1.31800728 0.50442967 -1.41501808
## [6,] -0.32389526 0.08924999 -0.35085134 0.01273346
##      GLNU_norm_align.L.PET RLNU_norm_align.L.PET GLVAR_align.L.PET
## [1,] -2.838909e-01 -0.1148497 0.06370859
## [2,] -5.421149e-17 0.2051580 0.31050605
## [3,] -4.900236e-01 0.1533960 0.40556015
## [4,] -1.020415e-01 -0.7362013 -1.45050212
## [5,] 2.443353e-01 -0.1405248 -0.91524677
## [6,] -3.508513e-01 -0.1921714 -0.11484968
##      RLVAR_align.L.PET Entropy_align.L.PET SZSE.L.PET LZSE.L.PET LGLZE.L.PET
## [1,] -0.3238953 0.01273346 -0.5189411 0.1921714 0.02546898
## [2,] -0.5482987 -0.87721561 0.2574751 -0.7032182 0.10204149
## [3,] -0.5781359 0.27065942 0.5335625 -2.2371600 -0.49002356
## [4,] 0.4900236 -0.49002356 -0.9152468 0.4473933 -0.36442422
## [5,] -0.4055602 -1.52752054 0.3508513 -1.5696804 0.54829867
## [6,] -0.3373427 0.12767671 -0.2971722 -0.1405248 -0.37806448
##      HGLZE.L.PET SZLGE.L.PET SZHGE.L.PET LZLGE.L.PET LZHGE.L.PET
## [1,] -0.01273346 -0.02546898 -0.08924999 0.16629268 0.1148497
## [2,] -0.12767671 0.10204149 -0.10204149 0.12767671 -0.5044297
## [3,] 0.54829867 -0.40556015 0.59324787 -0.75299544 0.2574751
## [4,] -1.12984341 -0.36442422 -1.06032943 -0.03820863 -0.8224296
## [5,] -1.38122903 0.62388738 -1.23201233 0.27065942 -1.7747932
## [6,] 0.08924999 -0.39177534 0.02546898 -0.20515797 0.1405248
##      GLNU_area.L.PET ZSNU.L.PET ZSP.L.PET GLNU_norm.L.PET ZSNU_norm.L.PET
## [1,] -0.5189411 -0.5189411 -0.5044297 -0.29717220 -0.6084963
## [2,] -0.6551233 -0.7872387 0.2051580 -0.01273346 0.3238953
## [3,] -0.5335625 -0.3508513 0.5932479 -0.50442967 0.6084963
## [4,] 1.1542636 1.2052851 -1.0382658 -0.07647307 -1.1298434
## [5,] -0.4333655 -0.7529954 0.4194224 0.25747509 0.4615096
## [6,] -0.5044297 -0.4473933 -0.2706594 -0.35085134 -0.3238953
##      GLVAR_area.L.PET ZSVAR.L.PET Entropy_area.L.PET Max_cooc.H.PET
## [1,] 0.06370859 0.2312375 0.16629268 -0.6238874
## [2,] 0.27065942 -0.2312375 -0.93476628 -0.3373427
## [3,] 0.40556015 -1.6148262 -0.32389526 0.8404111
## [4,] -1.45050212 0.9152468 0.23123752 -0.3780645
## [5,] -0.91524677 -1.3180073 -1.71650601 1.9997211
## [6,] -0.24433531 -0.1921714 -0.05095448 1.2596485
##      Average_cooc.H.PET Variance_cooc.H.PET Entropy_cooc.H.PET DAVE_cooc.H.PET
## [1,] -0.2971722 -4.333655e-01 -0.47571828 -0.3105060
## [2,] -0.4900236 -3.917753e-01 0.08924999 0.2838909
## [3,] 0.3373427 -6.709820e-01 -1.45050212 -0.7700043
## [4,] -0.8772156 -5.421149e-17 -0.50442967 -0.4473933
## [5,] 0.6394275 -2.428639e+00 -1.99972110 -2.1037376
## [6,] 0.4194224 -1.205285e+00 -1.91367122 -1.2596485
##      DVAR_cooc.H.PET DENT_cooc.H.PET SAVE_cooc.H.PET SVAR_cooc.H.PET
## [1,] -0.4900236 0.28389090 -0.05095448 0.1148497
## [2,] -0.1020415 -0.95464809 -1.15426355 -0.5482987
## [3,] -0.1792170 0.16629268 0.49002356 0.4757183
## [4,] -0.5189411 0.03820863 -0.73620133 0.3238953

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## [5,] -2.4286392 -0.08924999 0.65512329 0.1020415
## [6,] -0.3780645 -1.77479325 0.56315472 -1.2596485
##   SENT_cooc.H.PET ASM_cooc.H.PET Contrast_cooc.H.PET
## [1,] 0.2574751 -0.4900236 -0.2838909
## [2,] 0.5044297 -0.6551233 0.2838909
## [3,] -0.6394275 0.9152468 -0.5631547
## [4,] 0.1662927 -0.4333655 -0.5335625
## [5,] -1.0603294 2.1037376 -2.4286392
## [6,] -0.9347663 1.3489452 -0.9749140
##   Dissimilarity_cooc.H.PET Inv_diff_cooc.H.PET Inv_diff_norm_cooc.H.PET
## [1,] -0.3105060 -0.4757183 -0.28389090
## [2,] 0.2838909 -1.0166957 -0.63942746
## [3,] -0.7700043 0.5335625 0.10204149
## [4,] -0.4473933 -0.2574751 -0.08924999
## [5,] -2.1037376 1.0382658 0.62388738
## [6,] -1.2596485 0.9347663 0.40556015
##   IDM_cooc.H.PET IDM_norm_cooc.H.PET Inv_var_cooc_.H.PET
## [1,] -0.4900236 -0.23123752 0.1792170
## [2,] -1.0603294 -0.60849632 0.2706594
## [3,] 0.7700043 -0.10204149 -0.5781359
## [4,] -0.2574751 -0.06370859 0.2838909
## [5,] 1.1793911 0.60849632 -0.2838909
## [6,] 1.0166957 0.16629268 -0.6238874
##   Correlation_cooc.H.PET Autocorrelation_cooc.H.PET Tendency_cooc.H.PET
## [1,] 0.01273346 -0.3508513 -0.2574751
## [2,] -0.84041109 -0.5932479 -0.4473933
## [3,] -0.12767671 0.3508513 -0.6870110
## [4,] 0.40556015 -0.7032182 0.2971722
## [5,] -0.99558794 0.6394275 -2.4286392
## [6,] -0.54829867 0.4333655 -1.1298434
##   Shade_cooc.H.PET Prominence_cooc.H.PET IC1_d.H.PET IC2_d.H.PET
## [1,] 0.56315472 -0.2706594 0.36442422 -0.08924999
## [2,] -0.03820863 -0.3917753 0.89606941 -0.73620133
## [3,] -0.24433531 -0.6709820 -0.19217136 -0.31050605
## [4,] -0.59324787 0.5044297 -0.27065942 0.33734270
## [5,] 1.83980971 -2.4286392 0.32389526 -1.23201233
## [6,] 0.53356248 -1.2882802 -0.03820863 -0.75299544
##   Coarseness_vdif.H.PET Contrast_vdif.H.PET Busyness_vdif.H.PET
## [1,] -0.11484971 -0.3780645 -0.6084963
## [2,] -0.03820861 -0.9749140 -0.8960694
## [3,] -0.29717235 1.2320123 -0.2706594
## [4,] -0.49002384 -0.5781359 0.6709820
## [5,] -0.19217145 -0.8772156 -0.6870110
## [6,] -0.36442443 1.4879098 -0.1020415
##   Complexity_vdif.H.PET Strength_vdif.H.PET SRE_align.H.PET LRE_align.H.PET
## [1,] -5.421149e-17 0.1533960 0.03820863 -0.5335625
## [2,] 2.971722e-01 0.4473933 0.49002356 -1.4505021
## [3,] -1.276767e-01 0.3238953 -1.20528514 0.6238874
## [4,] -2.443353e-01 -0.8224296 -0.35085134 -0.1020415
## [5,] -8.224296e-01 1.0166957 -1.56968040 0.8047097
## [6,] -2.706594e-01 0.3105060 -1.71650601 1.1542636
##   RLNU_align.H.PET RP_align.H.PET LGRE_align.H.PET HGRE_align.H.PET
## [1,] -0.3238953 0.01273346 -0.24433531 -0.4757183
## [2,] -0.5044297 0.49002356 -0.21817922 -0.7872387

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## [3,] -0.7032182 -1.34894519 -0.32389526 0.1276767
## [4,] 0.9347663 -0.44739330 -0.02546898 -0.4194224
## [5,] -1.0603294 -1.61482620 -0.49002356 0.6394275
## [6,] -0.8960694 -1.83980971 -0.43336550 0.5482987
## LGSRE_align.H.PET HGSRE_align.H.PET LGHRE_align.H.PET HGLRE_align.H.PET
## [1,] -0.21817922 -0.32389526 -0.23123748 -0.5932479
## [2,] -0.19217136 -0.05095448 -0.28389089 -1.7747932
## [3,] -0.33734270 -1.41501808 -0.33734273 0.9955879
## [4,] -0.06370859 -0.67098197 0.03820884 -0.1921714
## [5,] -0.49002356 0.63942746 -0.44739340 1.2596485
## [6,] -0.46150957 0.10204149 -0.32389528 1.3812290
## GLNU_norm_align.H.PET RLNU_norm_align.H.PET GLVAR_align.H.PET
## [1,] -0.3917753 -5.421149e-17 -0.4194224
## [2,] -0.6394275 5.335625e-01 -0.5335625
## [3,] 0.7362013 -1.318007e+00 -0.7032182
## [4,] -0.4900236 -4.473933e-01 0.1405248
## [5,] 1.4505021 -1.569680e+00 -2.4286392
## [6,] 1.1060782 -1.839810e+00 -1.2052851
## RLVAR_align.H.PET Entropy_align.H.PET SZSE.H.PET LZSE.H.PET LGLZE.H.PET
## [1,] -0.4900236 -0.2706594 -0.2706594 -0.1405248 -0.23123752
## [2,] -1.2596485 -0.1020415 0.7196121 -2.8041692 -0.21817922
## [3,] 1.0603294 -1.1298434 -0.9955879 0.9152468 -0.32389526
## [4,] 0.1020415 0.1662927 -0.4900236 0.8224296 -0.07647307
## [5,] 1.1793911 -1.7165060 -1.2052851 1.1298434 -0.49002356
## [6,] 1.4505021 -1.1542636 -1.2320123 1.4879098 -0.40556015
## HGLZE.H.PET SZLGE.H.PET SZHGE.H.PET LZLGE.H.PET LZHGE.H.PET
## [1,] 1.276767e-01 -0.2181792 -0.02546898 -0.37806448 -0.2838909
## [2,] -6.551233e-01 -0.1662927 0.36442422 -0.70321822 -1.9997211
## [3,] -5.421149e-17 -0.3373427 -1.31800728 0.01273346 0.9347663
## [4,] 5.932479e-01 -0.1276767 -0.62388738 0.70321822 0.5932479
## [5,] 6.870110e-01 -0.4900236 0.46150957 0.54829867 1.0166957
## [6,] 3.917753e-01 -0.4333655 -0.95464809 1.17939112 1.4150181
## GLNU_area.H.PET ZSNU.H.PET ZSP.H.PET GLNU_norm.H.PET ZSNU_norm.H.PET
## [1,] -0.4615096 -0.32389526 -0.1662927 -0.3917753 -0.2181792
## [2,] -0.5932479 -0.01273346 0.8960694 -0.6551233 0.8586682
## [3,] -0.2312375 -0.82242955 -0.9749140 0.8960694 -1.0603294
## [4,] 1.1298434 0.77000435 -0.6084963 -0.4615096 -0.4757183
## [5,] -0.6084963 -1.17939112 -1.1298434 0.9955879 -1.2320123
## [6,] -0.4757183 -1.12984341 -1.2882802 1.0382658 -1.3812290
## GLVAR_area.H.PET ZSVAR_H.PET Entropy_area.H.PET Max_cooc.W.PET
## [1,] -0.4333655 -0.2181792 -0.2838909 -0.4757183
## [2,] -0.4900236 -1.8398097 -0.8224296 -0.3780645
## [3,] -0.8224296 0.8224296 -0.9546481 0.4473933
## [4,] 0.1792170 0.7529954 0.6084963 -0.4333655
## [5,] -2.4286392 0.9749140 -1.1060782 1.5275205
## [6,] -0.9546481 1.3489452 -1.0382658 0.9152468
## Average_cooc.W.PET Variance_cooc.W.PET Entropy_cooc.W.PET DAVE_cooc.W.PET
## [1,] -0.11484968 0.02546898 -0.1020415 -0.1020415
## [2,] 0.27065942 0.70321822 0.1662927 0.5631547
## [3,] -1.15426355 -1.56968040 -1.1793911 -1.4879098
## [4,] -0.02546898 -0.01273346 -0.1148497 -0.2838909
## [5,] -2.42863924 -1.83980971 -1.9136712 -1.7165060
## [6,] -1.83980971 -1.91367122 -1.8398097 -1.7747932
## DVAR_cooc.W.PET DENT_cooc.W.PET SAVE_cooc.W.PET SVAR_cooc.W.PET

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## [1,] 0.01273346 -0.1921714 -0.11484968 0.05095448
## [2,] 0.78723867 0.2971722 0.27065942 0.67098197
## [3,] -1.48790984 -1.4150181 -1.15426355 -1.45050212
## [4,] -0.10204149 -0.3508513 -0.02546898 0.02546898
## [5,] -1.31800728 -1.6148262 -2.42863924 -1.83980971
## [6,] -1.77479325 -1.7747932 -1.83980971 -1.91367122
##      SENT_cooc.W.PET ASM_cooc.W.PET Contrast_cooc.W.PET
## [1,] 0.01273346 -0.3508516 0.01273346
## [2,] 0.37806448 -0.4403688 0.80470972
## [3,] -1.28828023 0.4757183 -1.45050212
## [4,] 0.05095448 -0.3238955 -0.15339601
## [5,] -1.83980971 1.4505024 -1.56968040
## [6,] -1.71650601 1.0166959 -1.77479325
##      Dissimilarity_cooc.W.PET Inv_diff_cooc.W.PET Inv_diff_norm_cooc.W.PET
## [1,] -0.1020415 -0.4757183 -0.28389090
## [2,] 0.5631547 -1.0382658 -0.85866816
## [3,] -1.4879098 0.4900236 -0.23123752
## [4,] -0.2838909 -0.2706594 0.37806448
## [5,] -1.7165060 0.7529954 0.11484968
## [6,] -1.7747932 0.7032182 -0.07647307
##      IDM_cooc.W.PET IDM_norm_cooc.W.PET Inv_var_cooc.W.PET
## [1,] -0.4757183 -0.27726923 -0.4333655
## [2,] -1.0382658 -0.78723892 -0.9749140
## [3,] 0.7196121 -0.20515808 0.7700043
## [4,] -0.2706594 0.36442425 -0.2181792
## [5,] 0.9749140 0.05095443 0.8047097
## [6,] 0.9546481 -0.03820870 0.9955879
##      Correlation_cooc.W.PET Autocorrelation_cooc.W.PET Tendency_cooc.W.PET
## [1,] 0.02546898 0.05095448 0.05095448
## [2,] -1.06032943 0.35085134 0.67098197
## [3,] 0.06370859 -1.12984341 -1.45050212
## [4,] 0.37806448 0.11484968 0.02546898
## [5,] -0.73620133 -2.23715996 -1.83980971
## [6,] -0.2838909 -1.83980971 -1.91367122
##      Shade_cooc.W.PET Prominence_cooc.W.PET IC1_d.W.PET IC2_d.W.PET
## [1,] 0.2838909 0.03820863 3.780645e-01 -0.11484968
## [2,] 0.5335625 0.67098197 2.574751e-01 0.02546898
## [3,] -1.5696804 -1.45050212 -5.421149e-17 -0.41942235
## [4,] 0.3917753 0.23123752 -1.020415e-01 0.17921704
## [5,] -0.6551233 -1.41501808 6.709820e-01 -1.45050212
## [6,] -1.2052851 -1.83980971 1.921714e-01 -0.95464809
##      Coarseness_vdif.W.PET Contrast_vdif.W.PET Busyness_vdif.W.PET
## [1,] -1.533960e-01 0.08924999 -0.43336550
## [2,] -1.405248e-01 1.06032943 -1.06032943
## [3,] -5.421149e-17 -1.03826584 0.54829867
## [4,] -5.482987e-01 -0.80470972 0.07647307
## [5,] 2.546898e-02 -1.41501808 0.80470972
## [6,] -7.647307e-02 -1.38122903 1.01669574
##      Complexity_vdif.W.PET Strength_vdif.W.PET SRE_align.W.PET LRE_align.W.PET
## [1,] -0.1020415 0.3917753 0.1148497 -0.5189411
## [2,] 0.5631547 1.0166957 0.4055602 -1.2882802
## [3,] -1.2320123 -1.4505021 -0.9749140 0.3644242
## [4,] 0.1792170 -0.4194224 -0.2838909 -0.1533960
## [5,] -1.2052851 -0.6551233 -1.2596485 0.4194224

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## [6,] -1.8398097 -1.6148262 -1.5275205 0.5482987
## GLNU_align.W.PET RLUU_align.W.PET RP_align.W.PET LGRE_align.W.PET
## [1,] -0.77000435 -0.3238953 0.07647307 -0.1405248
## [2,] -1.23201233 -0.6394275 0.41942235 -0.2971722
## [3,] -0.11484968 -0.5932479 -1.12984341 0.5482987
## [4,] 1.08292088 1.2052851 -0.32389526 -0.7529954
## [5,] -0.03820863 -0.9749140 -1.34894519 1.3812290
## [6,] 0.10204149 -0.6551233 -1.56968040 0.8960694
## HGRE_align.W.PET LGSRE_align.W.PET HGSRE_align.W.PET LGHRE_align.W.PET
## [1,] -5.421149e-17 -0.1276767 -5.421149e-17 -0.1405248
## [2,] 3.780645e-01 -0.2971722 3.780645e-01 -0.3644242
## [3,] -1.232012e+00 0.4900236 -1.205285e+00 0.6238874
## [4,] 1.662927e-01 -0.7529954 1.148497e-01 -0.6709820
## [5,] -2.428639e+00 1.2596485 -1.913671e+00 1.6148262
## [6,] -1.839810e+00 0.8586682 -1.839810e+00 1.0829209
## HGLRE_align.W.PET GLNU_norm_align.W.PET RLUU_norm_align.W.PET
## [1,] -0.1276767 -0.3780645 0.05095448
## [2,] 0.3238953 -0.6084963 0.49002356
## [3,] -1.0603294 0.5932479 -1.15426355
## [4,] 0.1792170 -0.5044297 -0.37806448
## [5,] -2.4286392 1.4879098 -1.52752054
## [6,] -1.8398097 1.1060782 -1.66352051
## GLVAR_align.W.PET RLVAR_align.W.PET Entropy_align.W.PET SZSE.W.PET
## [1,] 0.03820863 -0.43336555 -0.24433531 -0.1020415
## [2,] 0.68701098 -0.99558812 -0.03820863 0.6084963
## [3,] -1.56968040 0.93476655 -1.12984341 -0.9347663
## [4,] 0.02546898 -0.02546894 0.06370859 -0.4900236
## [5,] -1.91367122 0.99558823 -2.23715996 -1.2052851
## [6,] -1.83980971 1.25964883 -1.25964848 -1.3180073
## LZSE.W.PET LGLZE.W.PET HGLZE.W.PET SZLGE.W.PET SZHGE.W.PET LZLGE.W.PET
## [1,] -0.4473933 -0.2443353 -5.421149e-17 -0.2574751 0.01273346 -0.01273346
## [2,] -1.4879098 -0.3373427 3.917753e-01 -0.2051580 0.46150957 -0.53356248
## [3,] 0.6709820 0.7700043 -1.232012e+00 0.8047097 -1.31800728 0.54829867
## [4,] 0.2051580 -0.7032182 1.148497e-01 -0.7872387 0.11484968 -0.27065942
## [5,] 0.7700043 1.3489452 -1.913671e+00 1.2052851 -1.83980971 1.20528514
## [6,] 1.2882802 0.9152468 -1.839810e+00 0.9546481 -1.91367122 1.61482620
## LZHGE.W.PET GLNU_area.W.PET ZSNU.W.PET ZSP.W.PET GLNU_norm.W.PET
## [1,] -0.4194224 -0.5781359 -0.2312375 -0.2181792 -0.4055602
## [2,] -0.2181792 -1.0382658 -0.3238953 0.5932479 -0.5932479
## [3,] -0.6394275 -0.1148497 -0.6870110 -1.0382658 0.7362013
## [4,] 0.4333655 1.1060782 0.8960694 -0.5932479 -0.5189411
## [5,] -2.1037376 -0.2574751 -1.1793911 -1.1542636 1.3489452
## [6,] -0.8960694 -0.1533960 -0.9749140 -1.4879098 1.1793911
## ZSNU_norm.W.PET GLVAR_area.W.PET ZSVAR.W.PET Entropy_area.W.PET
## [1,] -0.2051580 0.03820863 -0.3373427 -0.3508513
## [2,] 0.6870110 0.62388738 -1.0166957 -0.4473933
## [3,] -0.9749140 -1.28828023 0.8224296 -1.1542636
## [4,] -0.5335625 0.06370859 0.4055602 0.3780645
## [5,] -1.3180073 -1.77479325 0.8047097 -1.5696804
## [6,] -1.4505021 -1.83980971 1.2882802 -1.0166957
## Min_hist.ADC Max_hist.ADC Mean_hist.ADC Variance_hist.ADC
## [1,] 0.4475228 -0.31070344 -0.05095448 0.4333655
## [2,] -1.1111899 -0.42661942 -0.31050605 -0.2181792
## [3,] 0.6402049 0.50450261 0.06370859 1.1060782

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## [4,] -1.1111899 0.51901885 -0.11484968 0.7032182
## [5,] -1.1111899 -0.06382412 -1.01669574 0.3644242
## [6,] -1.1111899 0.08918521 0.46150957 1.7747932
## Standard_Deviation_hist.ADC Skewness_hist.ADC Kurtosis_hist.ADC
## [1,] 0.2443353 0.9546481 -0.2181792
## [2,] -0.3917753 -1.5696804 0.7362013
## [3,] 0.7362013 1.4150181 1.2320123
## [4,] 0.4333655 -0.4333655 -0.4757183
## [5,] 0.1533960 -0.3105060 -0.1792170
## [6,] 1.1542636 -1.0829209 -2.8041692
## Energy_hist.ADC Entropy_hist.ADC AUC_hist.ADC Volume.ADC X3D_surface.ADC
## [1,] 0.15339594 -1.61482620 -0.2181796 -0.8404111 -1.3180073
## [2,] -0.07647335 -0.82242955 -0.9546489 -1.2320123 -0.9546481
## [3,] -0.25090024 -0.01273346 0.3780644 -0.2706594 -0.4757183
## [4,] -0.41942296 0.41942235 -0.2443357 0.3917753 0.2574751
## [5,] -0.18569070 -0.20515797 -0.5482992 -0.2181792 -0.4615096
## [6,] -0.28389138 -0.02546898 -1.0166966 -0.7362013 -0.3373427
## ratio_3ds_vol.ADC ratio_3ds_vol_norm.ADC irregularity.ADC
## [1,] 0.7700047 -0.02546898 0.43336550
## [2,] 0.1533962 -1.10607816 -0.03820863
## [3,] -0.4757183 -1.52752054 -1.34894519
## [4,] -0.5044297 0.29717220 -0.84041109
## [5,] -0.3780645 -1.15426355 -0.99558794
## [6,] 0.3105063 0.41942235 -0.27065942
## Compactness_v1.ADC Compactness_v2.ADC Spherical_disproportion.ADC
## [1,] -0.24433531 -0.5044297 -0.02546898
## [2,] 0.03820863 0.4194224 -1.10607816
## [3,] 0.12767671 0.6084963 -1.52752054
## [4,] -0.44739330 -1.0829209 0.29717220
## [5,] 0.05095448 0.4757183 -1.15426355
## [6,] -0.53356248 -1.3812290 0.41942235
## Sphericity.ADC Asphericity.ADC Center_of_mass.ADC Max_3D_diam.ADC
## [1,] -0.4900236 -0.02546898 0.2574751 -1.38122903
## [2,] 0.3780645 -1.10607816 0.2971722 -0.77000435
## [3,] 0.5044297 -1.52752054 0.7032182 -0.59324787
## [4,] -1.1298434 0.31050605 0.5781359 0.07647307
## [5,] 0.4194224 -1.15426355 -0.4055602 -0.68701098
## [6,] -1.4150181 0.50442967 0.7362013 -0.50442967
## Major_axis_length.ADC Minor_axis_length.ADC Least_axis_length.ADC
## [1,] -0.50442967 -1.52752054 -1.48790984
## [2,] -1.38122903 -0.75299544 -0.39177534
## [3,] -0.97491405 -0.31050605 -0.21817922
## [4,] -0.02546898 -0.01273346 0.19217136
## [5,] -1.06032943 -0.36442422 0.01273346
## [6,] -0.24433531 -0.37806448 -0.53356248
## Elongation.ADC Flatness.ADC Max_cooc.L.ADC Average_cooc.L.ADC
## [1,] -2.42863924 -2.4286397 0.10204149 -0.7196121
## [2,] 0.01273346 0.4194225 -0.29717220 0.4055602
## [3,] 0.28389090 0.1792171 -0.01273346 -2.4286392
## [4,] -0.39177534 0.1084434 -0.08924999 -0.4055602
## [5,] 0.46150957 0.7032184 -0.12767671 -0.3105060
## [6,] -0.67098197 -0.9749142 -0.49002356 0.3373427
## Variance_cooc.L.ADC Entropy_cooc.L.ADC DAVE_cooc.L.ADC DVAR_cooc.L.ADC
## [1,] 0.7872387 -0.87721561 0.36442422 1.3180073

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## [2,]      -0.5932479      -0.57813595     -0.51894110     -0.6238874
## [3,]       1.0166957      -0.08924999      0.07647307     1.1298434
## [4,]      -0.7700043      -0.60849632     -1.08292088    -1.1542636
## [5,]      -0.4194224      -0.25747509     -0.31050605    -0.4900236
## [6,]       1.2052851      0.60849632      0.29717220     0.8586682
##   DENT_cooc.L.ADC SAVE_cooc.L.ADC SVAR_cooc.L.ADC SENT_cooc.L.ADC
## [1,]      0.43336550     -0.7196121      0.6709820      -0.2181792
## [2,]     -0.51894110      0.4055602     -0.7529954     -1.3812290
## [3,]      0.07647307     -2.4286392      1.1793911      0.6394275
## [4,]     -0.97491405     -0.4055602     -0.6394275     -0.1405248
## [5,]     -0.33734270     -0.3105060     -0.6084963     -0.2312375
## [6,]      0.37806448      0.3373427      1.3812290     -0.9749140
##   ASM_cooc.L.ADC Contrast_cooc.L.ADC Dissimilarity_cooc.L.ADC
## [1,]  2.730047e-06      1.1060782      0.36442422
## [2,]  -1.662910e-01     -0.5044297     -0.51894110
## [3,]  -1.020394e-01      0.6870110      0.07647307
## [4,]  -1.276748e-01     -1.1060782     -1.08292088
## [5,]  -2.443342e-01     -0.3105060     -0.31050605
## [6,]  -4.686025e-01      0.7872387      0.29717220
##   Inv_diff_cooc.L.ADC Inv_diff_norm_cooc.L.ADC IDM_cooc.L.ADC
## [1,]      -0.2838909     -0.67098275    -1.921714e-01
## [2,]      -0.2312375     -0.05095454    -2.574751e-01
## [3,]      -0.1276767     -0.35085175    -5.421149e-17
## [4,]      0.2706594      0.31050641    2.838909e-01
## [5,]      -0.3373427     -0.17921725   -3.373427e-01
## [6,]      -0.7032182     -0.68701178   -6.709820e-01
##   IDM_norm_cooc.L.ADC Inv_var_cooc.L.ADC Correlation_cooc.L.ADC
## [1,]      -0.87721561    -2.443353e-01    -0.8586682
## [2,]      -0.07647307    -2.051580e-01    -0.5781359
## [3,]      -0.49002356    -5.421149e-17    0.4900236
## [4,]      0.19217136     2.443353e-01    0.4473933
## [5,]      -0.16629268    -3.373427e-01   -0.6084963
## [6,]      -0.62388738    -7.032182e-01    0.5932479
##   Autocorrelation_.L.ADC Tendency_cooc.L.ADC Shade_.L.ADC
## [1,]      -0.7196121      0.6709820     1.4150181
## [2,]       0.5781359     -0.7529954     -1.1793911
## [3,]      -1.9136712      1.1793911     2.2371600
## [4,]      -0.3917753     -0.6394275     -0.3780645
## [5,]      -0.2971722     -0.6084963     -0.4473933
## [6,]       0.6238874      1.3812290     -1.5696804
##   Prominence_cooc.L.ADC IC1_.L.ADC IC2_.L.ADC Coarseness_vdif_.L.ADC
## [1,]      1.1298434     -0.93476628    0.50442967    0.47571994
## [2,]     -0.6238874      0.60849632    -0.28389090    0.06370943
## [3,]      2.2371600     -0.07647307    0.20515797    -0.17921668
## [4,]     -0.9152468      0.16629268   -0.06370859    -0.31050595
## [5,]     -0.6084963      0.70321822   -0.33734270   -0.05095386
## [6,]      1.1793911     -0.37806448    0.44739330   -0.07008883
##   Contrast_vdif_.L.ADC Busyness_vdif_.L.ADC Complexity_vdif_.L.ADC
## [1,]      1.1060782      -1.0603294      0.5631547
## [2,]     -0.3105060      -0.9955879     -0.8586682
## [3,]      0.3373427      0.2443353      0.8586682
## [4,]     -0.9546481      0.1662927     -1.0603294
## [5,]     -0.1533960      -0.5044297     -0.6551233
## [6,]      0.8404111     -0.7700043      0.6870110

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##      Strength_vdif_.L.ADC SRE_align.L.ADC LRE_align.L.ADC GLNU_align.L.ADC
## [1,] 1.3812290 0.01910130 -0.51894110 -1.3180073
## [2,] 0.6238874 -0.02546860 -0.43336550 -0.5932479
## [3,] 0.8404111 -0.31050614 -0.06370859 -0.2838909
## [4,] -0.7032182 -0.57813649 0.15339601 0.2706594
## [5,] 0.2838909 0.05095499 -0.49002356 -0.5189411
## [6,] 0.5932479 0.20515874 -0.87721561 -0.7872387
##      RLNU_align.L.ADC RP_align.L.ADC LGRE_align.L.ADC HGRE_align.L.ADC
## [1,] -1.5275205 0.05095448 -0.06370859 -0.3644242
## [2,] -0.7529954 -0.02546898 -0.39177534 0.4194224
## [3,] -0.2838909 -0.28389090 0.10204149 -1.9997211
## [4,] 0.2971722 -0.59324787 -0.12767671 -0.4757183
## [5,] -0.5781359 0.03820863 -0.23123752 -0.3508513
## [6,] -0.5189411 0.24433531 -0.40556015 0.6551233
##      LGSRE_align.L.ADC HGSRE_align.L.ADC LGHRE_align.L.ADC HGLRE_align.L.ADC
## [1,] -0.06370884 -0.2574751 -0.10204149 -0.4900236
## [2,] -0.39177578 0.4194224 -0.40556015 0.4194224
## [3,] 0.07647290 -1.9997211 0.08924999 -1.9997211
## [4,] -0.14052504 -0.4757183 -0.08924999 -0.3780645
## [5,] -0.23123786 -0.3105060 -0.24433531 -0.3508513
## [6,] -0.40556060 0.6709820 -0.41942235 0.6238874
##      GLNU_norm_align.L.ADC RLNU_norm_align.L.ADC GLVAR_align.L.ADC
## [1,] -0.11484968 0.05095457 0.8586682
## [2,] -0.08924999 -0.05095441 -0.5482987
## [3,] -0.21817922 -0.29717219 0.8960694
## [4,] -0.12767671 -0.67098206 -0.8047097
## [5,] -0.24433531 0.03820873 -0.3644242
## [6,] -0.75299544 0.29717236 1.0603294
##      RLVAR_align.L.ADC Entropy_align.L.ADC SZSE.L.ADC LZSE.L.ADC LGLZE.L.ADC
## [1,] -0.43336634 -0.4194224 0.24433538 -0.87721561 -0.07647307
## [2,] -0.41942318 -0.8586682 -0.03820862 -0.59324787 -0.39177534
## [3,] -0.07647358 0.1921714 -0.89606958 0.23123752 0.11484968
## [4,] 0.21817897 -0.3644242 -0.51894119 -0.02546898 -0.19217136
## [5,] -0.36442500 -0.3780645 -0.29717224 -0.10204149 -0.20515797
## [6,] -0.78723983 0.6238874 0.20515804 -1.20528514 -0.40556015
##      HGLZE.L.ADC SZLGE.L.ADC SZHGE.L.ADC LZLGE.L.ADC LZHGE.L.ADC
## [1,] -0.2838909 -0.06370838 -0.1662927 -0.14052487 -0.9955879
## [2,] 0.4194224 -0.37806447 0.3373427 -0.47571848 0.2971722
## [3,] -1.9997211 0.08925030 -1.9997211 0.50442973 -1.6635205
## [4,] -0.4900236 -0.25747501 -0.6551233 -0.06370868 -0.4333655
## [5,] -0.4473933 -0.19217124 -0.4473933 -0.27065957 -0.3917753
## [6,] 0.6238874 -0.40556016 0.6238874 -0.50442988 0.3780645
##      GLNU_area.L.ADC ZSNU.L.ADC ZSP.L.ADC GLNU_norm.L.ADC ZSNU_norm.L.ADC
## [1,] -1.3180073 -1.4879098 0.25747509 -0.11484983 0.2971730
## [2,] -0.5482987 -0.9546481 0.03820863 -0.08925014 -0.0764728
## [3,] -0.3917753 -0.4615096 -0.97491405 -0.32389548 -1.1793926
## [4,] 0.2971722 0.3508513 -0.51894110 -0.12767687 -0.6870117
## [5,] -0.5335625 -0.6394275 -0.33734270 -0.25747529 -0.3105061
## [6,] -0.7529954 -0.4900236 0.33734270 -0.71961240 0.2312383
##      GLVAR_area.L.ADC ZSVAR.L.ADC Entropy_area.L.ADC Max_cooc.H.ADC
## [1,] 0.8772156 -0.80470972 -0.7529954 -0.11484998
## [2,] -0.5631547 -0.63942746 -1.2882802 -0.27065984
## [3,] 0.8224296 0.44739330 0.5781359 0.07647291
## [4,] -0.8586682 0.06370859 -0.2051580 -0.12767702

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## [5,] -0.3373427 0.01273346 -0.1276767 -0.39865881
## [6,] 1.0603294 -1.25964848 0.5932479 -0.07647334
## Average_cooc.H.ADC Variance_cooc.H.ADC Entropy_cooc.H.ADC DAVE_cooc.H.ADC
## [1,] -1.2052851 -0.9546481 0.49002356 0.10204149
## [2,] 0.5781359 -0.8047097 -0.70321822 0.01273346
## [3,] -0.8047097 0.6394275 -0.23123752 -0.67098197
## [4,] -0.5631547 -0.9955879 0.12767671 -1.23201233
## [5,] -0.1792170 -1.7165060 -0.06370859 -0.03820863
## [6,] -0.8404111 0.5631547 -0.75299544 -1.20528514
## DVAR_cooc.H.ADC DENT_cooc.H.ADC SAVE_cooc.H.ADC SVAR_cooc.H.ADC
## [1,] 0.07647307 0.01273556 -1.2052851 -0.8404111
## [2,] -0.36442422 -0.03820676 0.5781359 -0.4900236
## [3,] -0.35085134 -0.57813651 -0.7872387 0.3508513
## [4,] -1.28828023 -1.25965212 -0.5932479 0.1020415
## [5,] -0.23123752 -0.02546706 -0.1792170 -0.8224296
## [6,] -0.80470972 -1.12984646 -0.8404111 0.4615096
## SENT_cooc.H.ADC ASM_cooc.H.ADC Contrast_cooc.H.ADC
## [1,] 0.2838909 0.02548452 0.06370859
## [2,] -1.7165060 -0.20515786 -0.05095448
## [3,] 0.1405248 -0.08285155 -0.62388738
## [4,] -0.1662927 -0.29717824 -1.31800728
## [5,] -0.5631547 -0.25090274 -0.06370859
## [6,] -0.6870110 -0.05094405 -1.10607816
## Dissimilarity_cooc.H.ADC Inv_diff_cooc.H.ADC Inv_diff_norm_cooc.H.ADC
## [1,] 0.10204149 -0.3780651 -0.50442967
## [2,] 0.01273346 -0.6238881 -0.40556015
## [3,] -0.67098197 0.1792168 0.01273346
## [4,] -1.23201233 0.2574749 0.28389090
## [5,] -0.03820863 -0.2838914 -0.32389526
## [6,] -1.20528514 0.3917753 0.31050605
## IDM_cooc.H.ADC IDM_norm_cooc.H.ADC Inv_var_cooc.H.ADC
## [1,] -0.3373427 -0.41942235 -0.2312376
## [2,] -0.6870110 -0.29717220 -0.7032184
## [3,] 0.2051580 -0.06370859 0.2051580
## [4,] 0.1921714 0.27065942 0.1792170
## [5,] -0.2971722 -0.28389090 -0.3644243
## [6,] 0.3508513 0.23123752 0.3644243
## Correlation_cooc.H.ADC Autocorrelation_cooc.H.ADC Tendency_cooc.H.ADC
## [1,] -0.7529954 -1.487910e+00 -0.8404111
## [2,] -0.4473933 5.932479e-01 -0.4900236
## [3,] 0.2051580 -2.574751e-01 0.3508513
## [4,] 0.4900236 -5.421149e-17 0.1020415
## [5,] -0.5482987 -4.055602e-01 -0.8224296
## [6,] 0.5044297 -1.273346e-02 0.4615096
## Shade_cooc.H.ADC Prominence_cooc.H.ADC IC1_d.H.ADC IC2_d.H.ADC
## [1,] 0.5482987 -0.93476628 -1.1793914 0.50442967
## [2,] -1.6635205 -0.78723867 0.2971724 -0.01273346
## [3,] 1.2596485 0.51894110 0.1020416 0.14052475
## [4,] -0.1276767 -0.01273346 0.4757185 -0.15339601
## [5,] -0.8960694 -0.91524677 0.3644245 -0.03820863
## [6,] -0.7872387 0.28389090 -0.4757183 0.37806448
## Coarseness_vdif.H.ADC Contrast_vdif.H.ADC Busyness_vdif.H.ADC
## [1,] 6.709821e-01 0.03820863 -1.6635205
## [2,] 3.820853e-02 -0.08924999 -0.7362013

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## [3,]      -1.276769e-01      -1.06032943      -0.3105060
## [4,]      -3.238955e-01      -1.34894519       0.2181792
## [5,]      -7.647320e-02      -0.01273346      -0.5044297
## [6,]      -1.147554e-07      -1.28828023      -0.6394275
## Complexity_vdif.H.ADC Strength_vdif.H.ADC SRE_align.H.ADC LRE_align.H.ADC
## [1,]      5.095448e-02       1.3180073       0.05095439      -0.6709820
## [2,]     -1.405248e-01       0.7362013       0.02546889      -0.6551233
## [3,]     -1.038266e+00       0.1662927      -0.41248158       0.2181792
## [4,]     -1.348945e+00      -0.5335625      -0.53356273       0.2574751
## [5,]     -5.421149e-17       0.2971722      -0.14052490      -0.3373427
## [6,]     -1.259648e+00       0.6394275      -0.35085154      -0.1405248
## GLNU_align.H.ADC RLNU_align.H.ADC RP_align.H.ADC LGRE_align.H.ADC
## [1,]     -1.4505021      -1.4505021       0.14052483      0.01273851
## [2,]     -0.7529954      -0.7529954       0.08925006      -0.12767341
## [3,]     -0.2971722      -0.3105060      -0.46150963      0.05096001
## [4,]      0.3373427       0.3373427      -0.56315481      -0.36442388
## [5,]     -0.5781359      -0.5781359      -0.08924997      -0.21817705
## [6,]     -0.5189411      -0.5189411      -0.28389092      -0.23123551
## HGRE_align.H.ADC LGSRE_align.H.ADC HGSRE_align.H.ADC LGHRE_align.H.ADC
## [1,]      0.3105060       0.03821309      0.41942235      -3.917756e-01
## [2,]     -0.6551233      -0.07646978      -0.07647307      -3.780647e-01
## [3,]     -2.4286392       0.06371331      -2.42863924      -2.706596e-01
## [4,]     -0.2181792      -0.37806428      -0.50442967      5.095454e-02
## [5,]     -0.4194224      -0.25747366      -0.12767671      -2.574752e-01
## [6,]      0.2312375      -0.31050515      -0.01273346      2.756749e-08
## HGLRE_align.H.ADC GLNU_norm_align.H.ADC RLNU_norm_align.H.ADC
## [1,]     -0.8047097      -0.03821078      2.838911e-01
## [2,]     -0.8224296      -0.14696808      2.051581e-01
## [3,]      0.6084963      -0.22472047      -4.473934e-01
## [4,]     -0.1020415      -0.47575501      -6.551235e-01
## [5,]     -0.5482987      -0.14696808      6.658463e-08
## [6,]     -0.5335625      -0.30385552      -2.838910e-01
## GLVAR_align.H.ADC RLVAR_align.H.ADC Entropy_align.H.ADC SZSE.H.ADC
## [1,]      0.3508513      -0.59324787      -1.20528537      0.24433561
## [2,]      0.2051580      -0.54829867      -0.82242969      0.08925023
## [3,]     -1.4879098       0.21817922      -0.05095444      -0.85866834
## [4,]     -0.3644242       0.05095448      0.31050617      -0.39177532
## [5,]      0.1792170      -0.35085134      -0.54829874      -0.16629256
## [6,]     -0.4473933      -0.15339601      -0.23123752      -0.65512338
## LZSE.H.ADC LGLZE.H.ADC HGLZE.H.ADC SZLGE.H.ADC SZHGE.H.ADC LZLGE.H.ADC
## [1,]    -0.85875773      0.19217133      -0.2971722      2.377814e-01      0.1405248      -0.75299544
## [2,]    -0.80479542      0.03820853      -0.2051580      6.370858e-02      0.1148497      -0.60849632
## [3,]      0.62390401     -0.12126089      -2.8041692     -2.181794e-01     -2.1037376      0.06370859
## [4,]     -0.05098619     -0.21817945      -0.2838909     -3.105063e-01     -0.4055602      -0.12767671
## [5,]     -0.17925794     -0.08925016      -1.3812290     -4.875143e-08     -1.2052851      -0.14052475
## [6,]      0.14050676     -0.41248169      0.3105060     -3.917756e-01     -0.5044297      0.54829867
## LZHGE.H.ADC GLNU_area.H.ADC ZSNU.H.ADC ZSP.H.ADC GLNU_norm.H.ADC
## [1,]     -0.78723867      -1.4505021     -1.4879695      0.3105064      0.0001161525
## [2,]     -0.95464809      -0.7529954     -0.7530320      0.2181795      -0.2838687582
## [3,]      0.70321822      -0.3105060     -0.3239184     -1.2320127      -0.0126215246
## [4,]     -0.02546898      0.3373427      0.3238925     -0.4333655      -0.4264086264
## [5,]      0.11484968      -0.5781359     -0.5932795     -0.1921713      -0.1533306504
## [6,]     -0.10204149      -0.5335625     -0.5335922     -0.6238875      -0.2116180558
## ZSNU_norm.H.ADC GLVAR_area.H.ADC ZSVAR.H.ADC Entropy_area.H.ADC

```

```

## [1,] 0.28389119 -0.03820863 -6.870142e-01 -1.41501808
## [2,] 0.17921729 0.44739330 -7.362047e-01 -0.85866816
## [3,] -1.25964882 -2.42863924 1.716510e+00 0.41942235
## [4,] -0.41942234 -0.51894110 -1.336426e-06 0.07647307
## [5,] -0.08924985 -0.01273346 -1.273483e-02 -0.53356248
## [6,] -0.80470987 -1.20528514 2.706588e-01 0.01273346
##      Max_cooc.W.ADC Average_cooc.W.ADC Variance_cooc.W.ADC DAVE_cooc.W.ADC
## [1,] 0.1662977 -8.772156e-01 0.3644242 0.28389090
## [2,] -0.1986587 4.473933e-01 -0.2181792 -0.05095448
## [3,] -0.2312360 -9.152468e-01 1.4150181 0.44739330
## [4,] -0.4900243 3.508513e-01 0.6551233 0.23123752
## [5,] -0.2971712 -5.421149e-17 0.1533960 0.33734270
## [6,] -0.3508508 7.529954e-01 1.8398097 1.01669574
##      DVAR_cooc.W.ADC DENT_cooc.W.ADC SAVE_cooc.W.ADC SVAR_cooc.W.ADC
## [1,] 0.85866816 0.1405248 -0.87721561 0.15339601
## [2,] -0.05095448 -0.1020415 0.46150957 -0.23123752
## [3,] 1.45050212 0.3780645 -0.91524677 1.41501808
## [4,] 0.32389526 0.1792170 0.39177534 0.73620133
## [5,] 0.29717220 0.2574751 0.01273346 0.06370859
## [6,] 1.52752054 0.6394275 0.70321822 1.83980971
##      SENT_cooc.W.ADC ASM_cooc.W.ADC Contrast_cooc.W.ADC
## [1,] -0.2312375 0.03821171 0.62388738
## [2,] -1.3489452 -0.13409891 0.02546898
## [3,] 0.5932479 -0.17921903 1.25964848
## [4,] 0.1792170 -0.41248879 0.32389526
## [5,] 0.1148497 -0.24433882 0.46150957
## [6,] -0.8772156 -0.36443054 1.61482620
##      Dissimilarity_cooc.W.ADC Inv_diff_cooc.W.ADC Inv_diff_norm_cooc.W.ADC
## [1,] 0.28389090 0.2706595 -0.67098197
## [2,] -0.05095448 -0.3508514 -0.05095448
## [3,] 0.44739330 -0.2312375 -0.36442422
## [4,] 0.23123752 -0.3373427 0.29717220
## [5,] 0.33734270 -0.6551234 -0.16629268
## [6,] 1.01669574 -0.9955881 -0.68701098
##      IDM_cooc.W.ADC IDM_norm_cooc.W.ADC Inv_var_cooc.W.ADC
## [1,] 0.2706595 -0.87721561 0.2443353
## [2,] -0.5482988 -0.06370859 -0.5631547
## [3,] -0.2838910 -0.51894110 -0.2312375
## [4,] -0.3917754 0.19217136 -0.4757183
## [5,] -0.5932480 -0.16629268 -0.7362013
## [6,] -0.9152469 -0.65512329 -0.8960694
##      Correlation_cooc.W.ADC Autocorrelation_cooc.W.ADC Tendency_cooc.W.ADC
## [1,] -0.8586682 -0.8224296 0.15339601
## [2,] -0.5781359 0.6084963 -0.23123752
## [3,] 0.4900236 -0.8047097 1.41501808
## [4,] 0.4615096 0.6551233 0.73620133
## [5,] -0.6084963 0.1148497 0.06370859
## [6,] 0.5932479 1.2320123 1.83980971
##      Shade_cooc.W.ADC Prominence_cooc.W.ADC IC1_d.W.ADC IC2_d.W.ADC
## [1,] 1.15426355 0.4055621 -1.0829209 0.4900236
## [2,] -1.31800728 -0.1020407 -0.2181792 0.1921714
## [3,] 2.80416916 2.1037436 -0.4055602 0.2838909
## [4,] 0.47571828 0.7362041 0.4055602 -0.1148497
## [5,] 0.06370859 0.2706611 -0.3917753 0.2574751

```

```

## [6,] -1.66352051 1.6635255 -1.2320123 0.6084963
## Coarseness_vdif.W.ADC Contrast_vdif.W.ADC Busyness_vdif.W.ADC
## [1,] 0.35085134 1.8398097 -0.41942260
## [2,] 0.06370850 0.2312375 -0.82242999
## [3,] -0.19217154 0.7362013 -0.05095456
## [4,] -0.31050626 -0.2574751 -0.39177558
## [5,] -0.03820876 0.4615096 -0.63942781
## [6,] -0.07647320 1.2320123 -0.84041153
## Complexity_vdif.W.ADC Strength_vdif.W.ADC SRE_align.W.ADC LRE_align.W.ADC
## [1,] -0.5482987 1.34894519 -0.40556015 -0.1662927
## [2,] -0.2051580 0.71961205 -0.02546898 -0.6084963
## [3,] 0.6870110 1.28828023 -0.17921704 -0.3373427
## [4,] 0.5189411 0.05095448 -0.29717220 -0.2971722
## [5,] 0.1662927 0.65512329 -0.03820863 -0.4900236
## [6,] 1.1793911 1.23201233 0.10204149 -0.8772156
## GLNU_align.W.ADC RLNU_align.W.ADC RP_align.W.ADC LGRE_align.W.ADC
## [1,] -1.4505021 -1.5275205 -0.39177534 0.01273328
## [2,] -0.8586682 -0.7529954 0.01273346 -0.34408956
## [3,] -0.4333655 -0.2971722 -0.16629268 -0.32389562
## [4,] 0.1020415 0.3373427 -0.25747509 -0.07647329
## [5,] -0.8772156 -0.5781359 -0.02546898 -0.39177575
## [6,] -1.2052851 -0.5189411 0.10204149 -0.42638400
## HGRE_align.W.ADC LGSRE_align.W.ADC HGSRE_align.W.ADC LGHRE_align.W.ADC
## [1,] -0.7362013 0.01273324 -0.7362013 3.295782e-06
## [2,] 0.4615096 -0.33734313 0.4615096 -3.508505e-01
## [3,] -0.7700043 -0.32389568 -0.7700043 -3.238942e-01
## [4,] 0.7529954 -0.08925027 0.7529954 -1.273025e-02
## [5,] 0.1405248 -0.39177581 0.1405248 -3.917748e-01
## [6,] 1.2320123 -0.42638406 1.2882802 -4.333653e-01
## HGLRE_align.W.ADC GLNU_norm_align.W.ADC RLNU_norm_align.W.ADC
## [1,] -0.7362013 0.08924981 -0.37806451
## [2,] 0.4473933 -0.07647338 0.05095455
## [3,] -0.7700043 -0.19217177 -0.08924996
## [4,] 0.7700043 -0.43336611 -0.23123751
## [5,] 0.1405248 -0.29717269 0.03820870
## [6,] 1.2052851 -0.53356317 0.21817933
## GLVAR_align.W.ADC RLVAR_align.W.ADC Entropy_align.W.ADC SZSE.W.ADC
## [1,] 0.4333655 -0.1405238 0.2312375 0.37806457
## [2,] -0.2181792 -0.4473935 -0.3238953 -0.47571838
## [3,] 1.1060782 -0.2181786 -0.1662927 0.41942244
## [4,] 0.7196121 -0.2051573 0.4473933 0.12767674
## [5,] 0.3644242 -0.3238950 0.2706594 -0.07647308
## [6,] 1.7747932 -0.5631553 0.6084963 0.29717226
## LZSE.W.ADC LGLZE.W.ADC HGLZE.W.ADC SZLGE.W.ADC SZHGE.W.ADC LZLGE.W.ADC
## [1,] -1.2052851 0.02546896 -0.7362013 0.05095473 -0.7362013 -0.03820863
## [2,] -0.4194224 -0.33734299 0.4615096 -0.33734264 0.4757183 -0.39177534
## [3,] 0.2443353 -0.32389553 -0.7700043 -0.32389519 -0.7700043 -0.31050605
## [4,] -0.6870110 -0.24433553 0.7529954 -0.25747499 0.7529954 -0.01273346
## [5,] -0.1533960 -0.39177567 0.1405248 -0.39177531 0.1405248 -0.41942235
## [6,] -1.0166957 -0.41942269 1.2320123 -0.41942233 1.2320123 -0.46150957
## LZHGE.W.ADC GLNU_area.W.ADC ZSNU.W.ADC ZSP.W.ADC GLNU_norm.W.ADC
## [1,] -0.7872387 -1.4505021 -1.3812290 0.49002356 0.03820875
## [2,] 0.4473933 -0.8404111 -0.7529954 -0.32389526 -0.11484963
## [3,] -0.7362013 -0.4615096 -0.2838909 0.29717220 0.29717244

```

```

## [4,] 0.6709820 0.1148497 0.3780645 0.25747509 -0.43336560
## [5,] 0.1662927 -0.9347663 -0.5781359 -0.06370859 0.33061178
## [6,] 1.2052851 -1.1793911 -0.5189411 0.44739330 -0.51894124
## ZSNU_norm.W.ADC GLVAR_area.W.ADC ZSVAR.W.ADC Entropy_area.W.ADC
## [1,] 0.43336550 0.4333655 -0.8047097 -1.66352051
## [2,] -0.16629268 -0.2181792 -0.4055602 -0.39177534
## [3,] -0.57813595 1.0603294 0.4194224 0.25747509
## [4,] 0.17921704 0.7196121 -0.4473933 0.39177534
## [5,] -0.02546898 0.3644242 -0.1533960 0.03820863
## [6,] 0.35085134 1.7747932 -0.7032182 0.56315472
sum(is.na(model3fdata))

## [1] 0

```

### 1. K-Means Clustering

K-Means Clustering is one of the most well-known and commonly used clustering algorithms for partitioning observations into a set of k groups.

The primary objective of k-means clustering is to make clusters within-cluster variation that is minimized. We will perform K-means clustering with 3 clusters, 100 maximum number of iterations, and 100 n start.

#### K-MEANS CLUSTERING

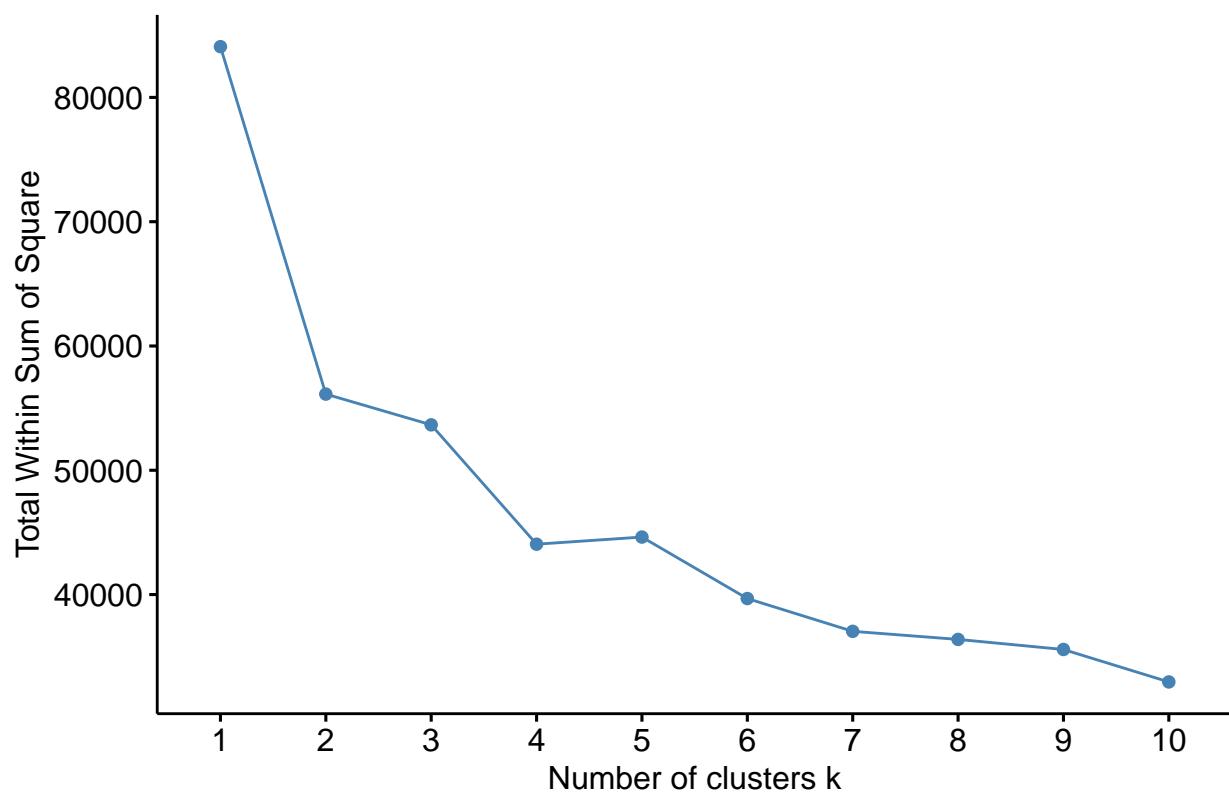
```
dataclust <- kmeans(model3fdata, centers = 3, iter.max = 100, nstart = 100)
```

Also, we need to determine and visualize optimal number of clusters.

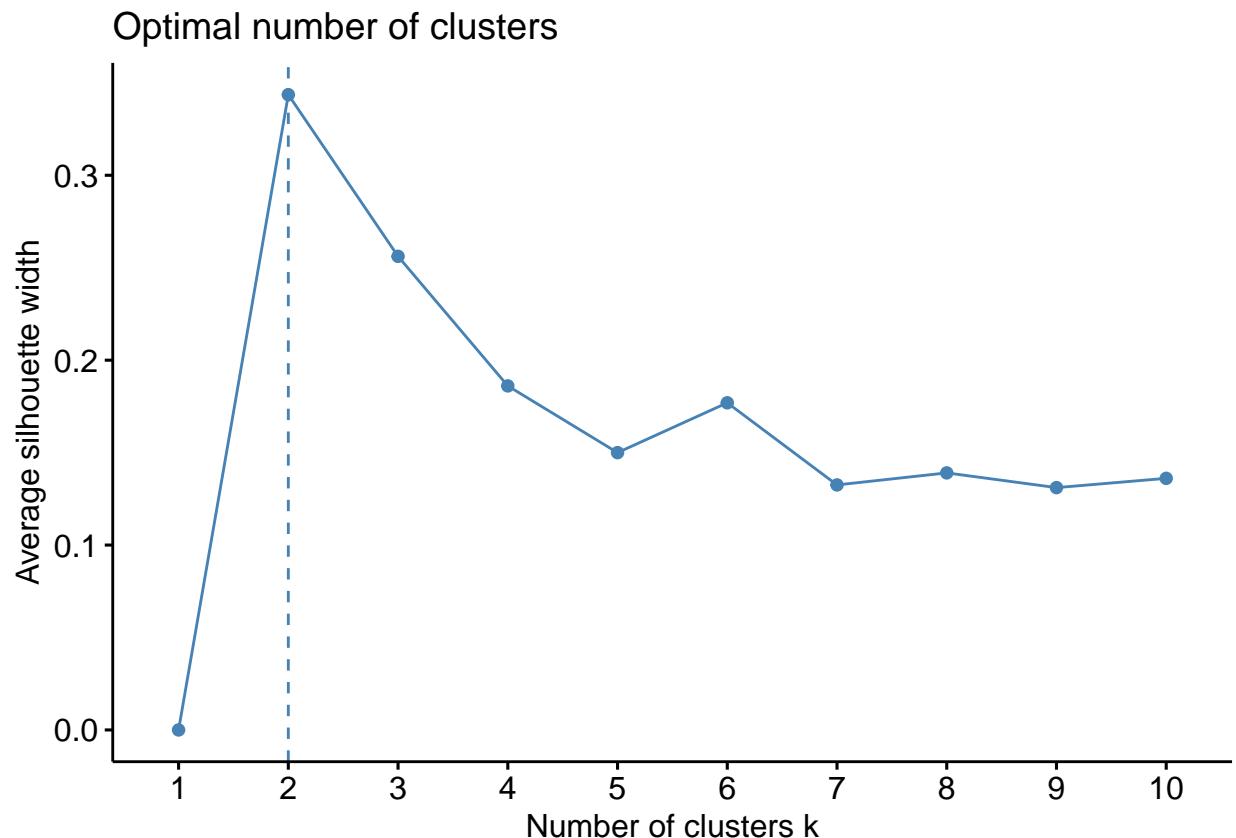
Another method to ascertain the most favorable value of the K number of clusters is by using *Within Sum of Squares*, *Silhouette* and *gap\_stat* plots. This advise us with 2 clusters. To plot the 2 clusters, we can use *fviz\_cluster()* function.

```
fviz_nbclust(model3fdata, kmeans, method = "wss")
```

### Optimal number of clusters

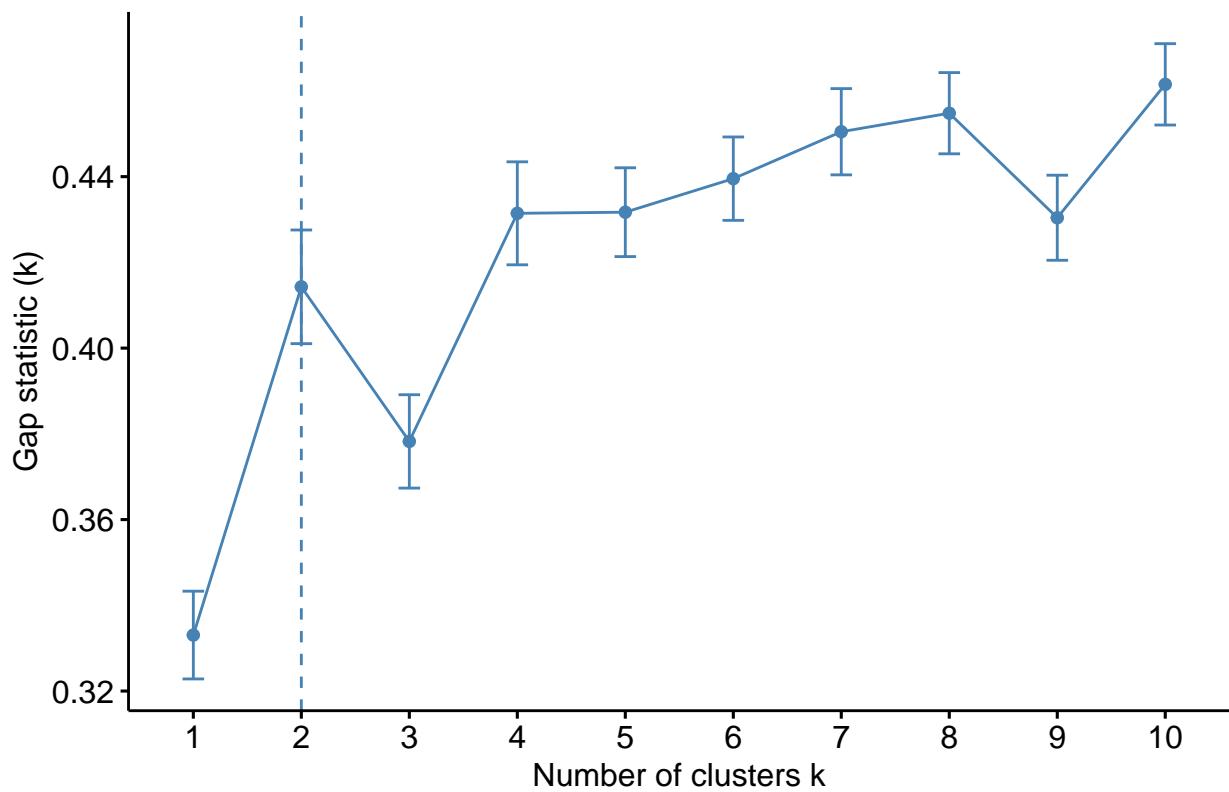


```
fviz_nbclust(model3fdata, kmeans, method = "silhouette")
```



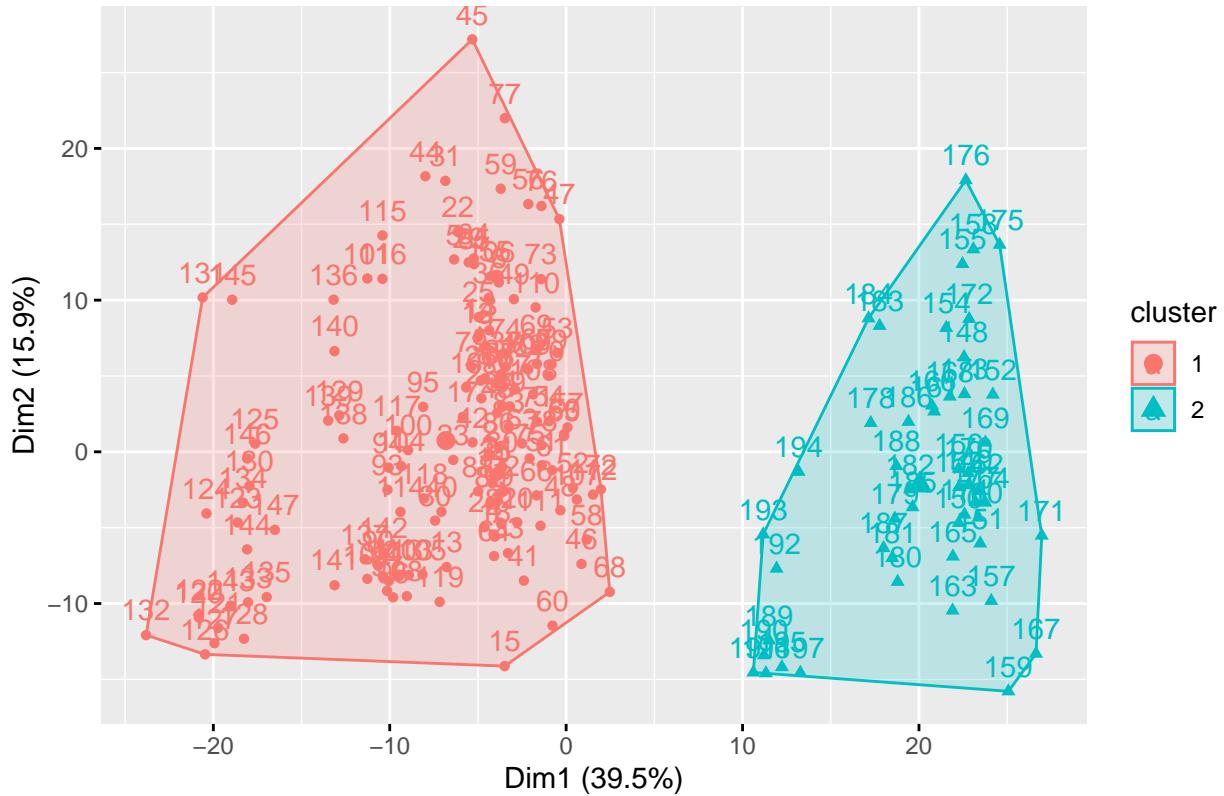
```
fviz_nbclust(model3fdata, kmeans, method = "gap_stat")
```

### Optimal number of clusters



```
dataclust <- kmeans(model3fdata, centers = 2, iter.max = 100, nstart = 100)
fviz_cluster(kmeans(model3fdata, centers = 2, iter.max = 100, nstart = 100), data = model3fdata)
```

Cluster plot



The quality of the k-means partition is gauged by the **SSwithin**, and we want it to be as little as possible. Hence, we have 33.2%.

```
dataclust$betweenss / dataclust$totss
```

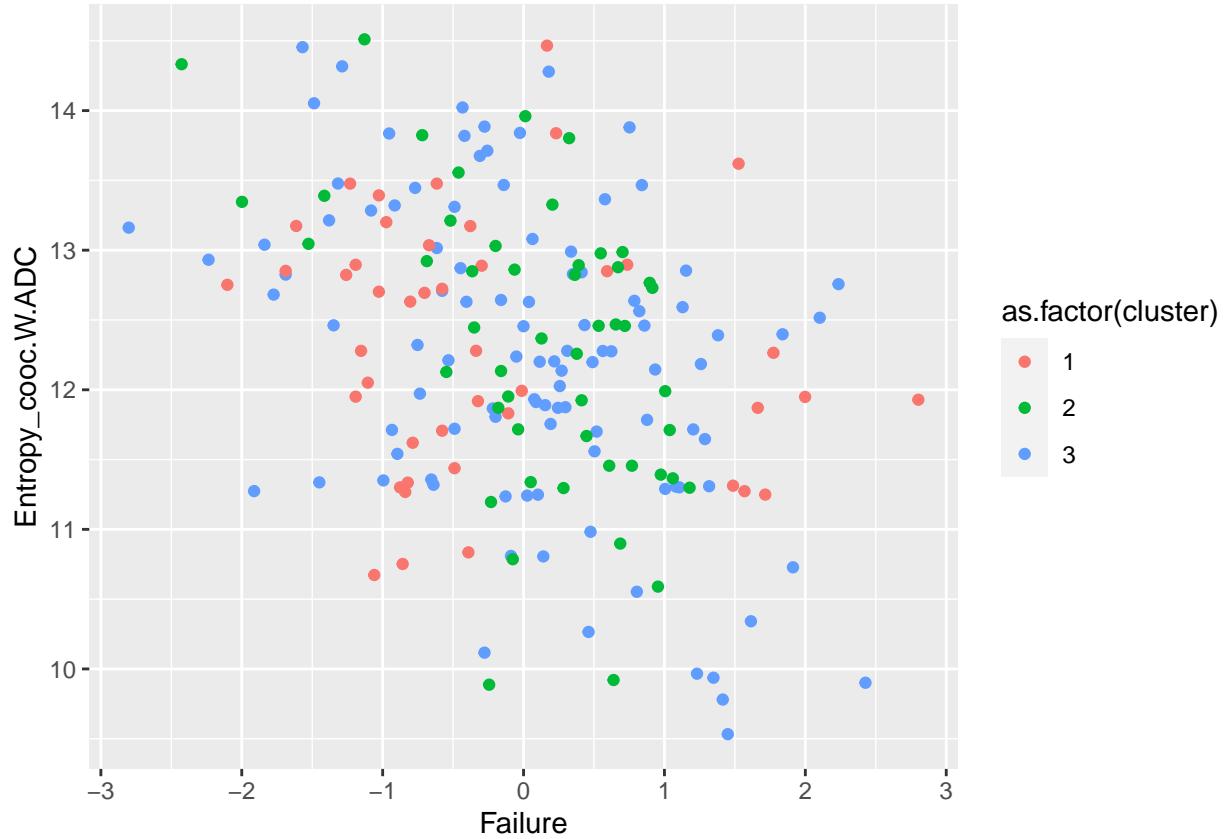
```
## [1] 0.3324795
```

Lastly, we can visualize clusters using the original variables where **x** is **Failure** and **y** is **Entropy\_cooc.W.ADC**

```
dataclust <- kmeans(model3fdata, centers = 3, iter.max = 100, nstart = 100)
```

```
model3data <- model3data |> mutate(cluster = dataclust$cluster)
```

```
model3data |> ggplot(aes(x = Failure, y = Entropy_cooc.W.ADC, col = as.factor(cluster))) + geom_point()
```



## 2.Hierarchical Clustering

In identifying the groupings in a data set, hierarchical clustering is a backup alternative to k-means clustering. Unlike k-means, in hierarchical clustering the number of clusters does not need to be preset, since this method can build hierarchy of clusters.

Also, before building a clustering model, a standardization of the data is prerequisite.

```
fdata <- model3data%>%
  select_if(is.numeric) %>% # select numeric columns
  select(-Failure.binary) %>%    # remove target column
  mutate_all(as.double) %>% # coerce to double type
  scale()
```

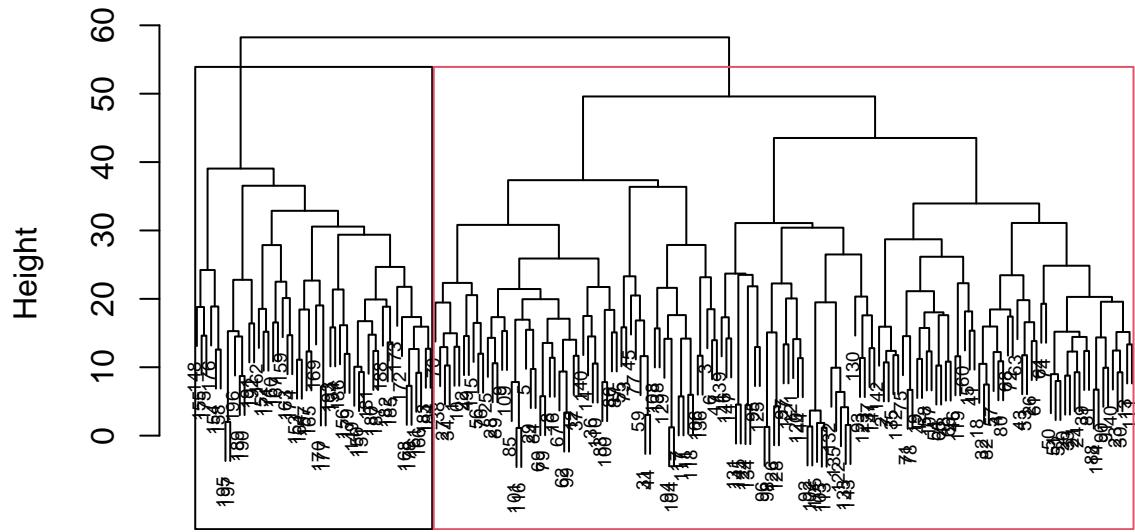
## Hierarchical clustering using Complete Linkage

Resembling to k-means, we evaluate first the dissimilarity of observations using distance measures to get the agglomeration coefficient (AC). For us to sustain these values and specify the agglomeration method to be used either “complete”, “average”, “single”, or “ward.D2”, is by using `hclust()` function.

```
cdata <- dist(fdata, method = "euclidean") #dissimilarity matrix

heirclustA <- hclust(cdata, method = "complete")
plot(heirclustA, cex = 0.6)
rect.hclust(heirclustA, k = 2, border = 1:4)
```

## Cluster Dendrogram



```
cdata
hclust (*, "complete")
```

### AGNES

We can also use agnes() function.

```
set.seed(123)
heirclustB <- agnes(fdata, method = "complete")
heirclustB$ac

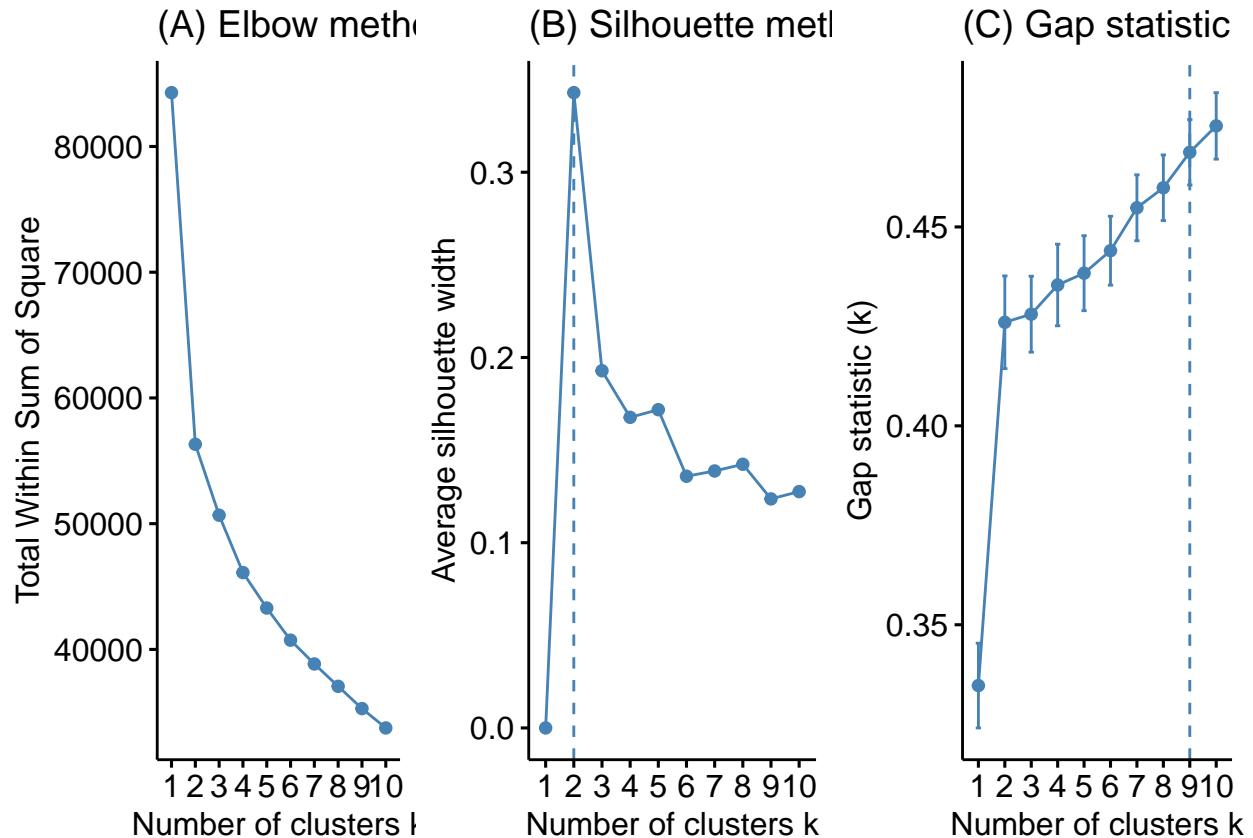
## [1] 0.8072963
```

### DIANA

```
heirclustC <- diana(fdata)
heirclustC$dc

## [1] 0.7915983

dianaA <- fviz_nbclust(fdata, FUN = hcut, method = "wss",
                        k.max = 10) +
  ggtitle("(A) Elbow method")
dianaB <- fviz_nbclust(fdata, FUN = hcut, method = "silhouette",
                        k.max = 10) +
  ggtitle("(B) Silhouette method")
dianaC <- fviz_nbclust(fdata, FUN = hcut, method = "gap_stat",
                        k.max = 10) +
  ggtitle("(C) Gap statistic")
gridExtra::grid.arrange(dianaA, dianaB, dianaC, nrow = 1)
```



#### Ward's method

```
heirclustD <- hclust(cdata, method = "ward.D2" )
resgroup <- cutree/heirclustD, k = 8)
table(resgroup)
```

```
## resgroup
## 1 2 3 4 5 6 7 8
## 71 33 12 21 10 19 22 9
```

#### 3. Model Based

Next is model-based clustering. Unlike the previous clustering methods, model-based clustering automatically ascertains the favorable number of clusters. Also, the Gaussian mixture models is applied.

We may use the `Mclust()` function, *which will leave G = NULL\** to necessitate the `Mclust()` to evaluate 1–9 clusters and select the optimal number of components based on BIC.

```
modelB1 <- Mclust(model3fdata[,1:10], G=3)
summary(modelB1)
```

```
## -----
## Gaussian finite mixture model fitted by EM algorithm
## -----
## 
## Mclust VEE (ellipsoidal, equal shape and orientation) model with 3 components:
## 
##   log-likelihood    n  df      BIC      ICL
##             -1074.583 197 89 -2619.371 -2638.94
```

```

## 
## Clustering table:
##   1   2   3
## 111 50 36
modelB2 = Mclust(model3fdata, 1:9) #from sir lecture

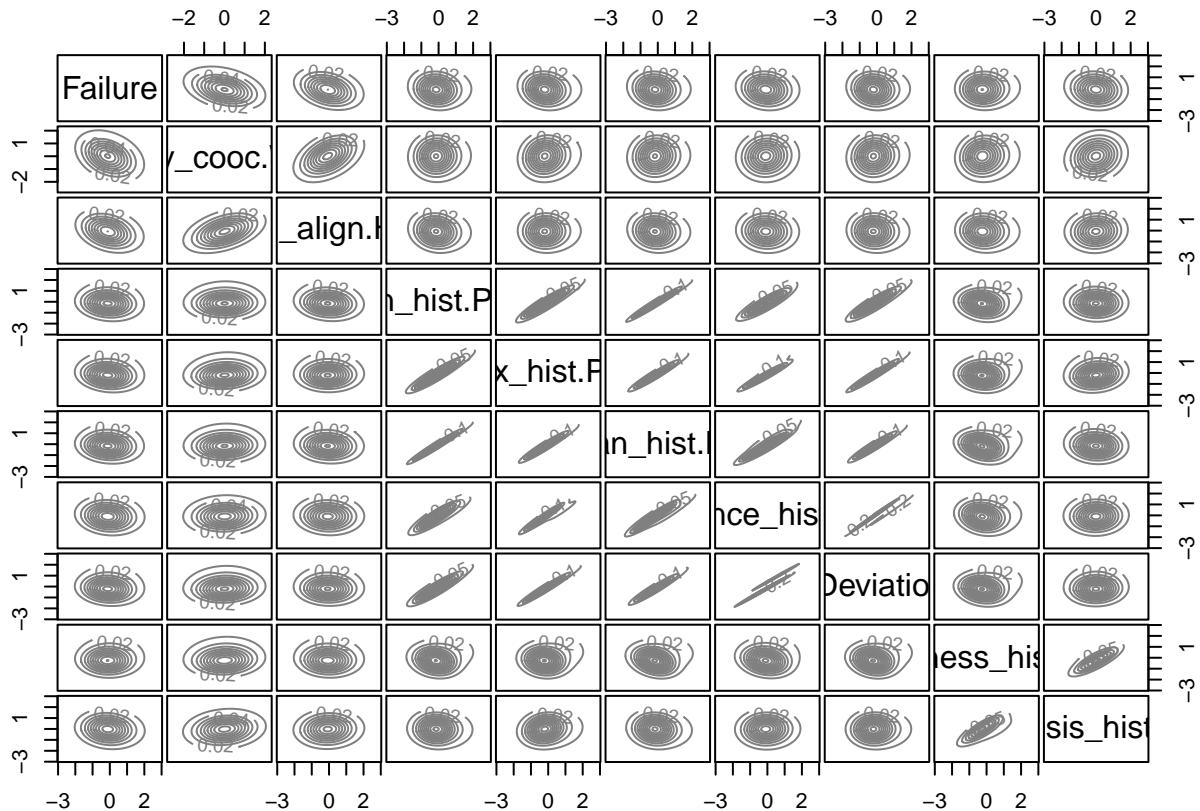
#Error in plot.new() : figure margins too large
summary(modelB2)

## -----
## Gaussian finite mixture model fitted by EM algorithm
## -----
## 
## Mclust VEI (diagonal, equal shape) model with 9 components:
## 
## log-likelihood   n    df      BIC      ICL
##          -77829.42 197 4306 -178408.3 -178408.3
## 
## Clustering table:
##   1   2   3   4   5   6   7   8   9
## 32 29 25 11 21 41 19 10  9

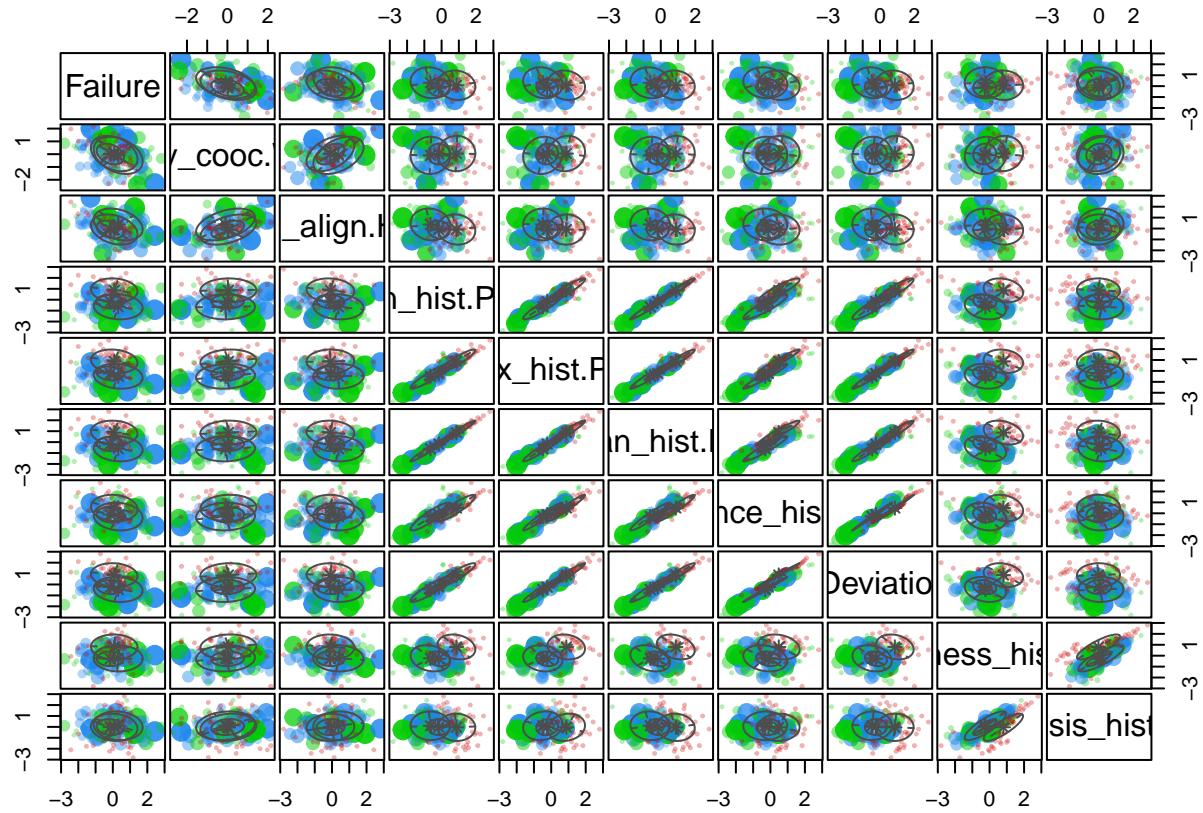
```

## Plot results

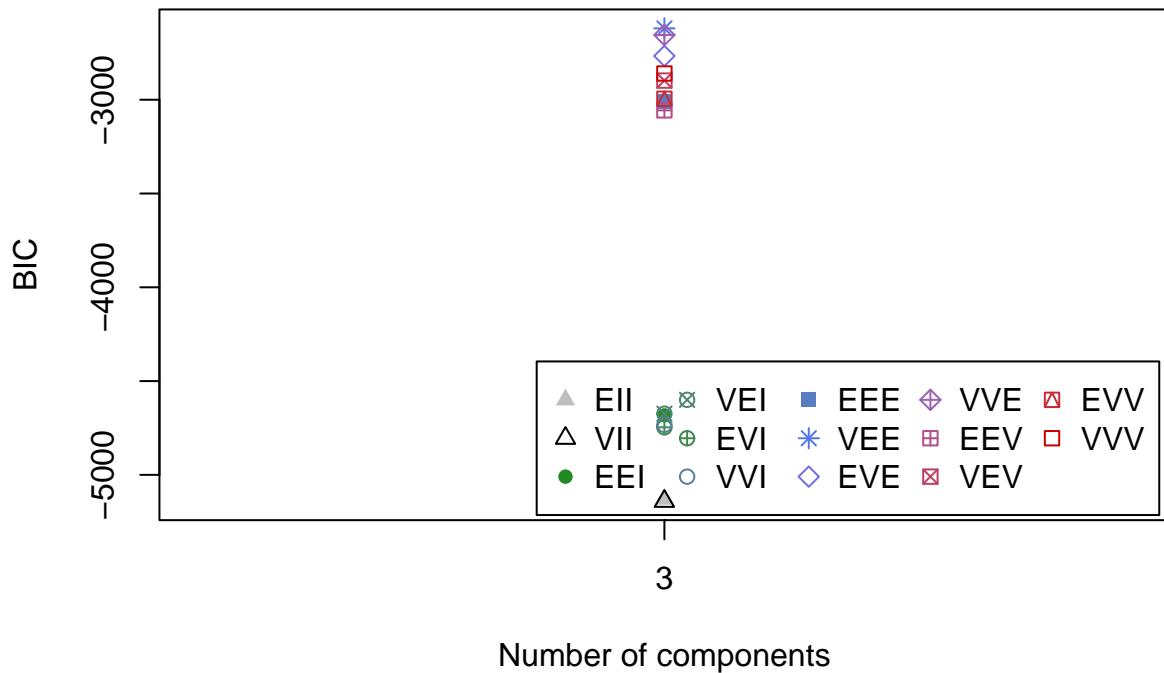
```
plot(modelB1, what = "density") #cannot plot 1:428 #dugay pag 1:10
```



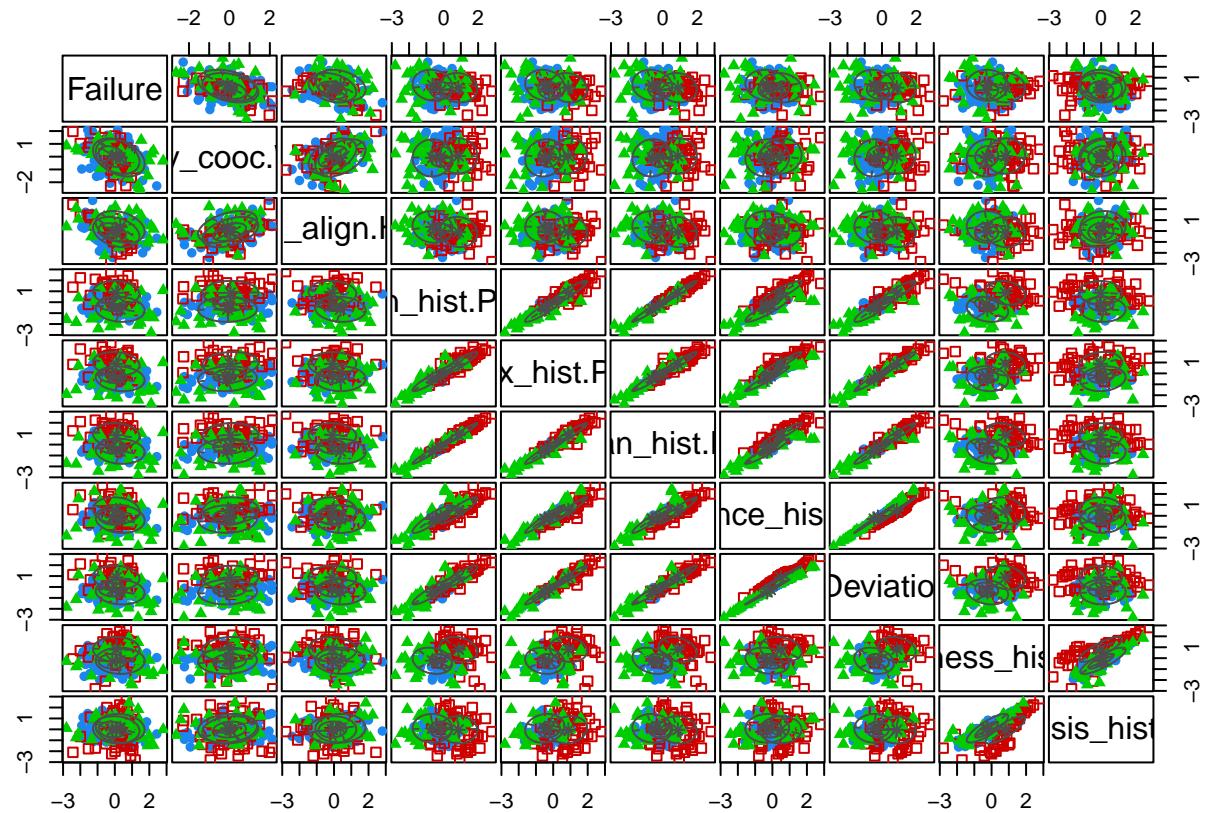
```
plot(modelB1, what = "uncertainty")
```



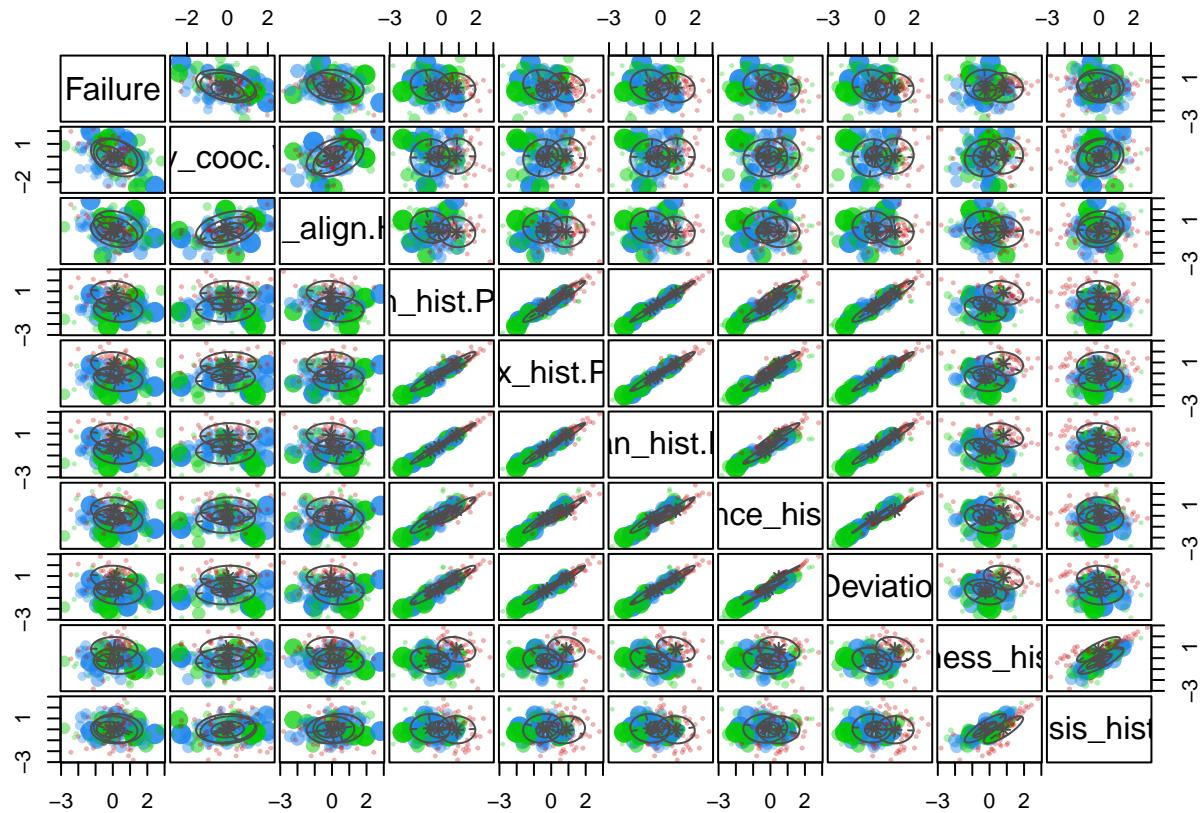
```
legargs <- list(x = "bottomright", ncol = 5)
plot(modelB1, what = 'BIC', legendArgs = legargs)
```



```
plot(modelB1, what = 'classification')
```



```
plot(modelB1, what = 'uncertainty')
```



```

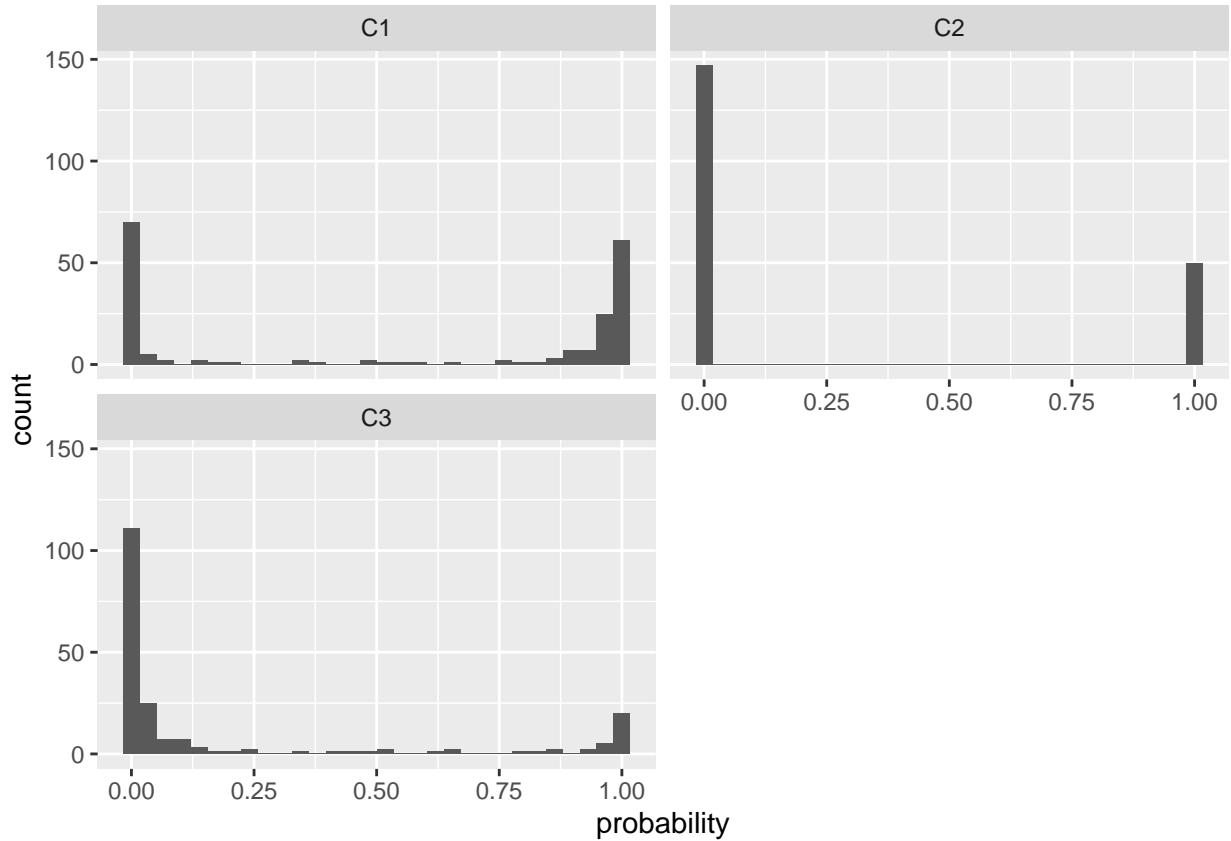
prbbelts <- modelB1$z
colnames(prbbelts) <- paste0('G', 1:3)

prbbelts <- prbbelts %>%
  as.data.frame() %>%
  mutate(id = row_number()) %>%
  tidyr::gather(cluster, probability, -id)

ggplot(prbbelts, aes(probability)) +
  geom_histogram() +
  facet_wrap(~ cluster, nrow = 2)

## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

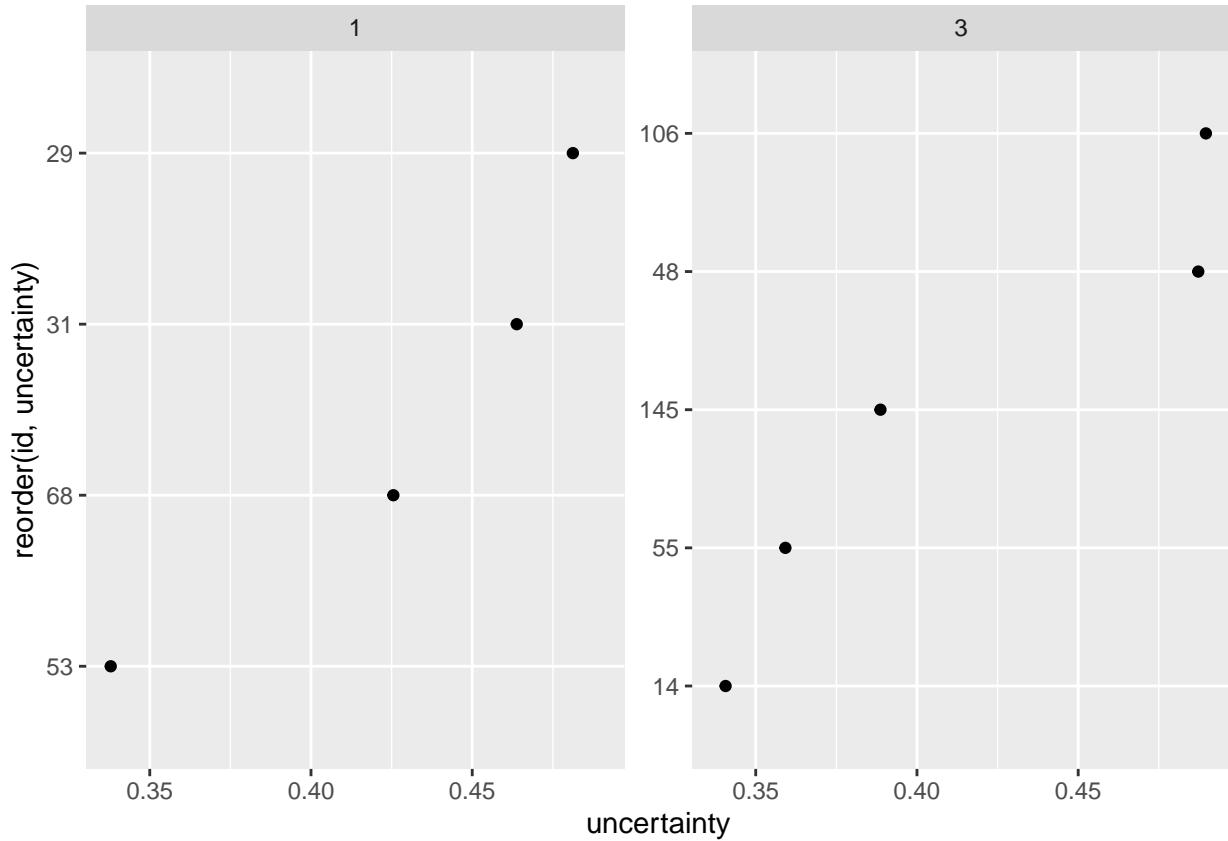
```



```

uncertainty <- data.frame(
  id = 1:nrow(model3fdata),
  datafclust = modelB1$classification,
  uncertainty = modelB1$uncertainty
)
uncertainty %>%
  group_by(datafclust) %>%
  filter(uncertainty > 0.25) %>%
  ggplot(aes(uncertainty, reorder(id, uncertainty))) +
  geom_point() +
  facet_wrap(~ datafclust, scales = 'free_y', nrow = 1)

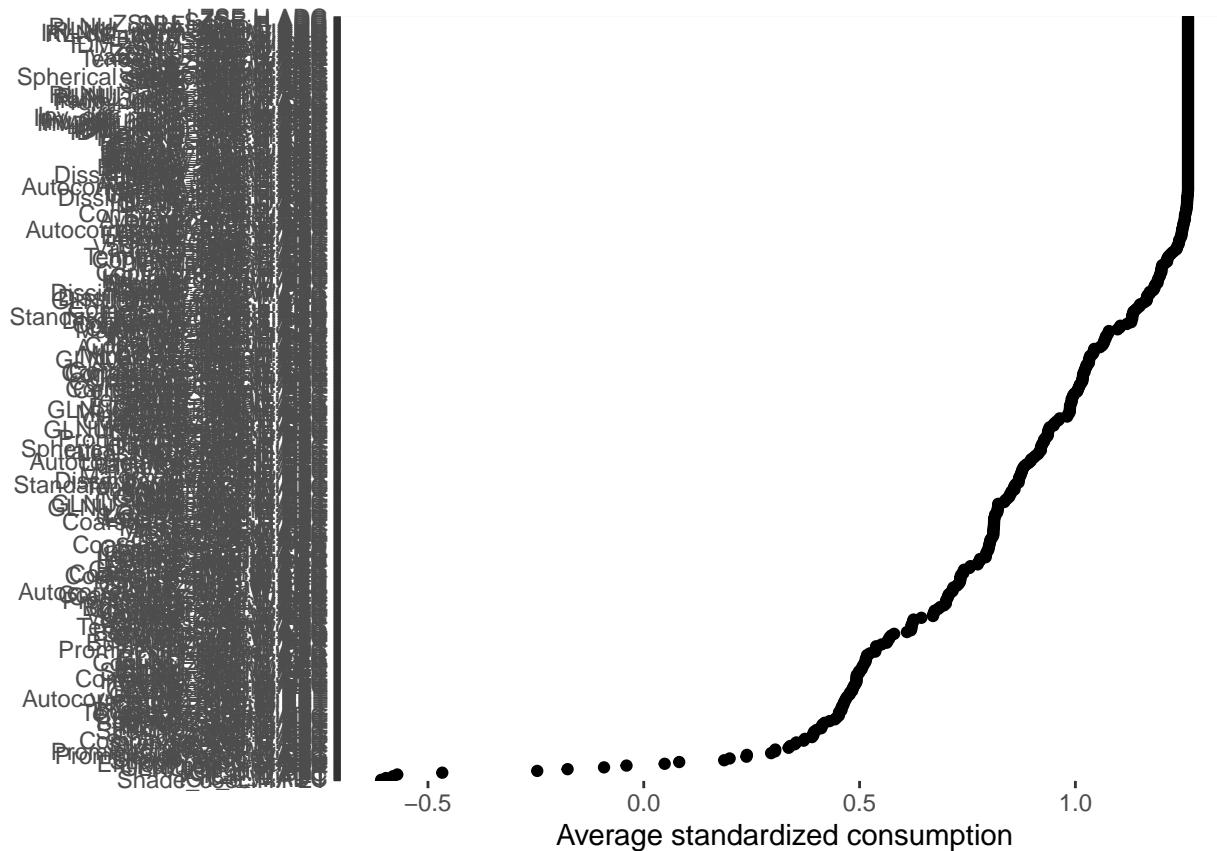
```



```

datafclustB <- model3fdata %>%
  scale() %>%
  as.data.frame() %>%
  mutate(cluster = modelB1$classification) %>%
  filter(cluster == 2) %>%
  select(-cluster)

datafclustB %>%
  tidyr::gather(product, std_count) %>%
  group_by(product) %>%
  summarize(avg = mean(std_count)) %>%
  ggplot(aes(avg, reorder(product, avg))) +
  geom_point() +
  labs(x = "Average standardized consumption", y = NULL)
  
```



## Conclusion

By the results we can, therefore say that, using k-means clustering the best number of clusters is 2 with SSwithin = 33.2%. Also Hierarchical, gap statistics suggest 9 clusters with 84.90% AC and 84.29%. Lastly, model-based suggest 3 optimal number of clusters with BIC -2632.206.