

# Modeling Data - Breast Cancer Analysis

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## INSTALL AND LOAD PACKAGES

```
# Install pacman ("package manager") if needed
if (!require("pacman")) install.packages("pacman")

## Loading required package: pacman

library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.3.6      v purrr   0.3.5
## v tibble  3.1.8      v dplyr  1.0.10
## v tidyr   1.2.1      v stringr 1.4.1
## v readr   2.1.3      v forcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

library(dplyr)
library(MASS)

##
## Attaching package: 'MASS'
##
## The following object is masked from 'package:dplyr':
##
##     select

library(caTools)
library(caret)

## Loading required package: lattice
##
## Attaching package: 'caret'
##
## The following object is masked from 'package:purrr':
##
##     lift
```

```
library(readr, dplyr)
library(ggplot2)
library(corrplot)
```

```
## corrplot 0.92 loaded
```

```
library(gridExtra)
```

```
##
## Attaching package: 'gridExtra'
##
## The following object is masked from 'package:dplyr':
##
##   combine
```

```
library(pROC)
```

```
## Type 'citation("pROC")' for a citation.
##
## Attaching package: 'pROC'
##
## The following objects are masked from 'package:stats':
##
##   cov, smooth, var
```

```
library(psych) # for pc
```

```
##
## Attaching package: 'psych'
##
## The following objects are masked from 'package:ggplot2':
##
##   %+%, alpha
```

```
library(GPARotation)
library(magrittr)
```

```
##
## Attaching package: 'magrittr'
##
## The following object is masked from 'package:purrr':
##
##   set_names
##
## The following object is masked from 'package:tidyr':
##
##   extract
```

```
library(rio)
```

## LOAD AND PREPARE DATA

```
df <- read_csv("../Data/breast_cancer_data.csv")
```

```
## Rows: 1138 Columns: 32
## -- Column specification -----
## Delimiter: ","
## chr (1): diagnosis
## dbl (31): id, radius_mean, texture_mean, perimeter_mean, area_mean, smoothne...
##
## 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.
```

```
head(df)
```

```
## # A tibble: 6 x 32
##       id diagn~1 radiu~2 textu~3 perim~4 area_~5 smoot~6 compa~7 conca~8 conca~9
##   <dbl> <chr>      <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
## 1 8.42e5 M          18.0    10.4   123.    1001    0.118    0.278    0.300    0.147
## 2 8.43e5 M          20.6    17.8   133.    1326    0.0847    0.0786    0.0869    0.0702
## 3 8.43e7 M          19.7    21.2   130     1203    0.110    0.160    0.197    0.128
## 4 8.43e7 M          11.4    20.4    77.6     386.    0.142    0.284    0.241    0.105
## 5 8.44e7 M          20.3    14.3   135.    1297    0.100    0.133    0.198    0.104
## 6 8.44e5 M          12.4    15.7    82.6     477.    0.128    0.17     0.158    0.0809
## # ... with 22 more variables: symmetry_mean <dbl>,
## #   fractal_dimension_mean <dbl>, radius_se <dbl>, texture_se <dbl>,
## #   perimeter_se <dbl>, area_se <dbl>, smoothness_se <dbl>,
## #   compactness_se <dbl>, concavity_se <dbl>, 'concave points_se' <dbl>,
## #   symmetry_se <dbl>, fractal_dimension_se <dbl>, radius_worst <dbl>,
## #   texture_worst <dbl>, perimeter_worst <dbl>, area_worst <dbl>,
## #   smoothness_worst <dbl>, compactness_worst <dbl>, concavity_worst <dbl>, ...
```

```
tail(df)
```

```
## # A tibble: 6 x 32
##       id diagn~1 radiu~2 textu~3 perim~4 area_~5 smoot~6 compa~7 conca~8 conca~9
##   <dbl> <chr>      <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
## 1 926125 M          20.9    25.1   143     1347    0.110    0.224    0.317    0.147
## 2 926424 M          21.6    22.4   142     1479    0.111    0.116    0.244    0.139
## 3 926682 M          20.1    28.2   131.     1261    0.0978    0.103    0.144    0.0979
## 4 926954 M          16.6    28.1   108.     858.    0.0846    0.102    0.0925    0.0530
## 5 927241 M          20.6    29.3   140.     1265    0.118    0.277    0.351    0.152
## 6 92751 B           7.76    24.5    47.9     181    0.0526    0.0436    0         0
## # ... with 22 more variables: symmetry_mean <dbl>,
## #   fractal_dimension_mean <dbl>, radius_se <dbl>, texture_se <dbl>,
## #   perimeter_se <dbl>, area_se <dbl>, smoothness_se <dbl>,
## #   compactness_se <dbl>, concavity_se <dbl>, 'concave points_se' <dbl>,
```

```
## # symmetry_se <dbl>, fractal_dimension_se <dbl>, radius_worst <dbl>,
## # texture_worst <dbl>, perimeter_worst <dbl>, area_worst <dbl>,
## # smoothness_worst <dbl>, compactness_worst <dbl>, concavity_worst <dbl>, ...
```

```
# review the dataset
```

```
str(df)
```

```
## spec_tbl_df [1,138 x 32] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ id : num [1:1138] 842302 842517 84300903 84348301 84358402 ...
## $ diagnosis : chr [1:1138] "M" "M" "M" "M" ...
## $ radius_mean : num [1:1138] 18 20.6 19.7 11.4 20.3 ...
## $ texture_mean : num [1:1138] 10.4 17.8 21.2 20.4 14.3 ...
## $ perimeter_mean : num [1:1138] 122.8 132.9 130 77.6 135.1 ...
## $ area_mean : num [1:1138] 1001 1326 1203 386 1297 ...
## $ smoothness_mean : num [1:1138] 0.1184 0.0847 0.1096 0.1425 0.1003 ...
## $ compactness_mean : num [1:1138] 0.2776 0.0786 0.1599 0.2839 0.1328 ...
## $ concavity_mean : num [1:1138] 0.3001 0.0869 0.1974 0.2414 0.198 ...
## $ concave points_mean : num [1:1138] 0.1471 0.0702 0.1279 0.1052 0.1043 ...
## $ symmetry_mean : num [1:1138] 0.242 0.181 0.207 0.26 0.181 ...
## $ fractal_dimension_mean : num [1:1138] 0.0787 0.0567 0.06 0.0974 0.0588 ...
## $ radius_se : num [1:1138] 1.095 0.543 0.746 0.496 0.757 ...
## $ texture_se : num [1:1138] 0.905 0.734 0.787 1.156 0.781 ...
## $ perimeter_se : num [1:1138] 8.59 3.4 4.58 3.44 5.44 ...
## $ area_se : num [1:1138] 153.4 74.1 94 27.2 94.4 ...
## $ smoothness_se : num [1:1138] 0.0064 0.00522 0.00615 0.00911 0.01149 ...
## $ compactness_se : num [1:1138] 0.049 0.0131 0.0401 0.0746 0.0246 ...
## $ concavity_se : num [1:1138] 0.0537 0.0186 0.0383 0.0566 0.0569 ...
## $ concave points_se : num [1:1138] 0.0159 0.0134 0.0206 0.0187 0.0188 ...
## $ symmetry_se : num [1:1138] 0.03 0.0139 0.0225 0.0596 0.0176 ...
## $ fractal_dimension_se : num [1:1138] 0.00619 0.00353 0.00457 0.00921 0.00511 ...
## $ radius_worst : num [1:1138] 25.4 25 23.6 14.9 22.5 ...
## $ texture_worst : num [1:1138] 17.3 23.4 25.5 26.5 16.7 ...
## $ perimeter_worst : num [1:1138] 184.6 158.8 152.5 98.9 152.2 ...
## $ area_worst : num [1:1138] 2019 1956 1709 568 1575 ...
## $ smoothness_worst : num [1:1138] 0.162 0.124 0.144 0.21 0.137 ...
## $ compactness_worst : num [1:1138] 0.666 0.187 0.424 0.866 0.205 ...
## $ concavity_worst : num [1:1138] 0.712 0.242 0.45 0.687 0.4 ...
## $ concave points_worst : num [1:1138] 0.265 0.186 0.243 0.258 0.163 ...
## $ symmetry_worst : num [1:1138] 0.46 0.275 0.361 0.664 0.236 ...
## $ fractal_dimension_worst : num [1:1138] 0.1189 0.089 0.0876 0.173 0.0768 ...
## - attr(*, "spec")=
## .. cols(
## .. id = col_double(),
## .. diagnosis = col_character(),
## .. radius_mean = col_double(),
## .. texture_mean = col_double(),
## .. perimeter_mean = col_double(),
## .. area_mean = col_double(),
## .. smoothness_mean = col_double(),
## .. compactness_mean = col_double(),
## .. concavity_mean = col_double(),
## .. 'concave points_mean' = col_double(),
## .. symmetry_mean = col_double(),
## .. fractal_dimension_mean = col_double(),
```

```
## .. radius_se = col_double(),
## .. texture_se = col_double(),
## .. perimeter_se = col_double(),
## .. area_se = col_double(),
## .. smoothness_se = col_double(),
## .. compactness_se = col_double(),
## .. concavity_se = col_double(),
## .. 'concave points_se' = col_double(),
## .. symmetry_se = col_double(),
## .. fractal_dimension_se = col_double(),
## .. radius_worst = col_double(),
## .. texture_worst = col_double(),
## .. perimeter_worst = col_double(),
## .. area_worst = col_double(),
## .. smoothness_worst = col_double(),
## .. compactness_worst = col_double(),
## .. concavity_worst = col_double(),
## .. 'concave points_worst' = col_double(),
## .. symmetry_worst = col_double(),
## .. fractal_dimension_worst = col_double()
## .. )
## - attr(*, "problems")=<externalptr>
```

## SUMMARIZE DATAFRAME

```
# Convert diagnosis field to factor
df$diagnosis <- as.factor(df$diagnosis)
# Summary of all variables.
summary(df)
```

```
##      id      diagnosis radius_mean texture_mean
## Min.   :    8670   B:714   Min.   : 6.981   Min.   : 9.71
## 1st Qu.:  869218   M:424   1st Qu.:11.700   1st Qu.:16.17
## Median :  906024           Median :13.370   Median :18.84
## Mean   : 30371831           Mean   :14.127   Mean   :19.29
## 3rd Qu.:  8813129           3rd Qu.:15.780   3rd Qu.:21.80
## Max.   :911320502           Max.   :28.110   Max.   :39.28
## perimeter_mean area_mean smoothness_mean compactness_mean
## Min.   : 43.79   Min.   : 143.5   Min.   :0.05263   Min.   :0.01938
## 1st Qu.: 75.17   1st Qu.: 420.3   1st Qu.:0.08637   1st Qu.:0.06492
## Median : 86.24   Median : 551.1   Median :0.09587   Median :0.09263
## Mean   : 91.97   Mean   : 654.9   Mean   :0.09636   Mean   :0.10434
## 3rd Qu.:104.10   3rd Qu.: 782.7   3rd Qu.:0.10530   3rd Qu.:0.13040
## Max.   :188.50   Max.   :2501.0   Max.   :0.16340   Max.   :0.34540
## concavity_mean concave points_mean symmetry_mean fractal_dimension_mean
## Min.   :0.00000   Min.   :0.00000   Min.   :0.1060   Min.   :0.04996
## 1st Qu.:0.02956   1st Qu.:0.02031   1st Qu.:0.1619   1st Qu.:0.05770
## Median :0.06154   Median :0.03350   Median :0.1792   Median :0.06154
## Mean   :0.08880   Mean   :0.04892   Mean   :0.1812   Mean   :0.06280
## 3rd Qu.:0.13070   3rd Qu.:0.07400   3rd Qu.:0.1957   3rd Qu.:0.06612
## Max.   :0.42680   Max.   :0.20120   Max.   :0.3040   Max.   :0.09744
## radius_se texture_se perimeter_se area_se
```

```
## Min. :0.1115 Min. :0.3602 Min. : 0.757 Min. : 6.802
## 1st Qu.:0.2324 1st Qu.:0.8339 1st Qu.: 1.606 1st Qu.: 17.850
## Median :0.3242 Median :1.1080 Median : 2.287 Median : 24.530
## Mean :0.4052 Mean :1.2169 Mean : 2.866 Mean : 40.337
## 3rd Qu.:0.4789 3rd Qu.:1.4740 3rd Qu.: 3.357 3rd Qu.: 45.190
## Max. :2.8730 Max. :4.8850 Max. :21.980 Max. :542.200
## smoothness_se compactness_se concavity_se concave points_se
## Min. :0.001713 Min. :0.002252 Min. :0.000000 Min. :0.000000
## 1st Qu.:0.005169 1st Qu.:0.013080 1st Qu.:0.01509 1st Qu.:0.007638
## Median :0.006380 Median :0.020450 Median :0.02589 Median :0.010930
## Mean :0.007041 Mean :0.025478 Mean :0.03189 Mean :0.011796
## 3rd Qu.:0.008146 3rd Qu.:0.032450 3rd Qu.:0.04205 3rd Qu.:0.014710
## Max. :0.031130 Max. :0.135400 Max. :0.39600 Max. :0.052790
## symmetry_se fractal_dimension_se radius_worst texture_worst
## Min. :0.007882 Min. :0.0008948 Min. : 7.93 Min. :12.02
## 1st Qu.:0.015160 1st Qu.:0.0022480 1st Qu.:13.01 1st Qu.:21.08
## Median :0.018730 Median :0.0031870 Median :14.97 Median :25.41
## Mean :0.020542 Mean :0.0037949 Mean :16.27 Mean :25.68
## 3rd Qu.:0.023480 3rd Qu.:0.0045580 3rd Qu.:18.79 3rd Qu.:29.72
## Max. :0.078950 Max. :0.0298400 Max. :36.04 Max. :49.54
## perimeter_worst area_worst smoothness_worst compactness_worst
## Min. : 50.41 Min. : 185.2 Min. :0.07117 Min. :0.02729
## 1st Qu.: 84.11 1st Qu.: 515.3 1st Qu.:0.11660 1st Qu.:0.14720
## Median : 97.66 Median : 686.5 Median :0.13130 Median :0.21190
## Mean :107.26 Mean : 880.6 Mean :0.13237 Mean :0.25427
## 3rd Qu.:125.40 3rd Qu.:1084.0 3rd Qu.:0.14600 3rd Qu.:0.33910
## Max. :251.20 Max. :4254.0 Max. :0.22260 Max. :1.05800
## concavity_worst concave points_worst symmetry_worst fractal_dimension_worst
## Min. :0.0000 Min. :0.000000 Min. :0.1565 Min. :0.05504
## 1st Qu.:0.1145 1st Qu.:0.06493 1st Qu.:0.2504 1st Qu.:0.07146
## Median :0.2267 Median :0.09993 Median :0.2822 Median :0.08004
## Mean :0.2722 Mean :0.11461 Mean :0.2901 Mean :0.08395
## 3rd Qu.:0.3829 3rd Qu.:0.16140 3rd Qu.:0.3179 3rd Qu.:0.09208
## Max. :1.2520 Max. :0.29100 Max. :0.6638 Max. :0.20750
```

```
prop.table(table(df$diagnosis))
```

```
##
## B M
## 0.6274165 0.3725835
```

```
dfii <- df[,3:ncol(df)]
```

```
# evaluate correlation all attributes in the dataset
dfii%>%
  cor(dfii) %>%
  round(2) %>%
  print()
```

```
## radius_mean texture_mean perimeter_mean area_mean
## radius_mean 1.00 0.32 1.00 0.99
## texture_mean 0.32 1.00 0.33 0.32
```

## perimeter_mean	1.00	0.33	1.00	0.99
## area_mean	0.99	0.32	0.99	1.00
## smoothness_mean	0.17	-0.02	0.21	0.18
## compactness_mean	0.51	0.24	0.56	0.50
## concavity_mean	0.68	0.30	0.72	0.69
## concave points_mean	0.82	0.29	0.85	0.82
## symmetry_mean	0.15	0.07	0.18	0.15
## fractal_dimension_mean	-0.31	-0.08	-0.26	-0.28
## radius_se	0.68	0.28	0.69	0.73
## texture_se	-0.10	0.39	-0.09	-0.07
## perimeter_se	0.67	0.28	0.69	0.73
## area_se	0.74	0.26	0.74	0.80
## smoothness_se	-0.22	0.01	-0.20	-0.17
## compactness_se	0.21	0.19	0.25	0.21
## concavity_se	0.19	0.14	0.23	0.21
## concave points_se	0.38	0.16	0.41	0.37
## symmetry_se	-0.10	0.01	-0.08	-0.07
## fractal_dimension_se	-0.04	0.05	-0.01	-0.02
## radius_worst	0.97	0.35	0.97	0.96
## texture_worst	0.30	0.91	0.30	0.29
## perimeter_worst	0.97	0.36	0.97	0.96
## area_worst	0.94	0.34	0.94	0.96
## smoothness_worst	0.12	0.08	0.15	0.12
## compactness_worst	0.41	0.28	0.46	0.39
## concavity_worst	0.53	0.30	0.56	0.51
## concave points_worst	0.74	0.30	0.77	0.72
## symmetry_worst	0.16	0.11	0.19	0.14
## fractal_dimension_worst	0.01	0.12	0.05	0.00
##				
##	smoothness_mean	compactness_mean	concavity_mean	
## radius_mean	0.17	0.51	0.68	
## texture_mean	-0.02	0.24	0.30	
## perimeter_mean	0.21	0.56	0.72	
## area_mean	0.18	0.50	0.69	
## smoothness_mean	1.00	0.66	0.52	
## compactness_mean	0.66	1.00	0.88	
## concavity_mean	0.52	0.88	1.00	
## concave points_mean	0.55	0.83	0.92	
## symmetry_mean	0.56	0.60	0.50	
## fractal_dimension_mean	0.58	0.57	0.34	
## radius_se	0.30	0.50	0.63	
## texture_se	0.07	0.05	0.08	
## perimeter_se	0.30	0.55	0.66	
## area_se	0.25	0.46	0.62	
## smoothness_se	0.33	0.14	0.10	
## compactness_se	0.32	0.74	0.67	
## concavity_se	0.25	0.57	0.69	
## concave points_se	0.38	0.64	0.68	
## symmetry_se	0.20	0.23	0.18	
## fractal_dimension_se	0.28	0.51	0.45	
## radius_worst	0.21	0.54	0.69	
## texture_worst	0.04	0.25	0.30	
## perimeter_worst	0.24	0.59	0.73	
## area_worst	0.21	0.51	0.68	
## smoothness_worst	0.81	0.57	0.45	

## compactness_worst	0.47	0.87	0.75
## concavity_worst	0.43	0.82	0.88
## concave points_worst	0.50	0.82	0.86
## symmetry_worst	0.39	0.51	0.41
## fractal_dimension_worst	0.50	0.69	0.51
##	concave points_mean	symmetry_mean	
## radius_mean	0.82	0.15	
## texture_mean	0.29	0.07	
## perimeter_mean	0.85	0.18	
## area_mean	0.82	0.15	
## smoothness_mean	0.55	0.56	
## compactness_mean	0.83	0.60	
## concavity_mean	0.92	0.50	
## concave points_mean	1.00	0.46	
## symmetry_mean	0.46	1.00	
## fractal_dimension_mean	0.17	0.48	
## radius_se	0.70	0.30	
## texture_se	0.02	0.13	
## perimeter_se	0.71	0.31	
## area_se	0.69	0.22	
## smoothness_se	0.03	0.19	
## compactness_se	0.49	0.42	
## concavity_se	0.44	0.34	
## concave points_se	0.62	0.39	
## symmetry_se	0.10	0.45	
## fractal_dimension_se	0.26	0.33	
## radius_worst	0.83	0.19	
## texture_worst	0.29	0.09	
## perimeter_worst	0.86	0.22	
## area_worst	0.81	0.18	
## smoothness_worst	0.45	0.43	
## compactness_worst	0.67	0.47	
## concavity_worst	0.75	0.43	
## concave points_worst	0.91	0.43	
## symmetry_worst	0.38	0.70	
## fractal_dimension_worst	0.37	0.44	
##	fractal_dimension_mean	radius_se	texture_se
## radius_mean	-0.31	0.68	-0.10
## texture_mean	-0.08	0.28	0.39
## perimeter_mean	-0.26	0.69	-0.09
## area_mean	-0.28	0.73	-0.07
## smoothness_mean	0.58	0.30	0.07
## compactness_mean	0.57	0.50	0.05
## concavity_mean	0.34	0.63	0.08
## concave points_mean	0.17	0.70	0.02
## symmetry_mean	0.48	0.30	0.13
## fractal_dimension_mean	1.00	0.00	0.16
## radius_se	0.00	1.00	0.21
## texture_se	0.16	0.21	1.00
## perimeter_se	0.04	0.97	0.22
## area_se	-0.09	0.95	0.11
## smoothness_se	0.40	0.16	0.40
## compactness_se	0.56	0.36	0.23
## concavity_se	0.45	0.33	0.19



## concave points_se	0.34	0.51	0.23
## symmetry_se	0.35	0.24	0.41
## fractal_dimension_se	0.69	0.23	0.28
## radius_worst	-0.25	0.72	-0.11
## texture_worst	-0.05	0.19	0.41
## perimeter_worst	-0.21	0.72	-0.10
## area_worst	-0.23	0.75	-0.08
## smoothness_worst	0.50	0.14	-0.07
## compactness_worst	0.46	0.29	-0.09
## concavity_worst	0.35	0.38	-0.07
## concave points_worst	0.18	0.53	-0.12
## symmetry_worst	0.33	0.09	-0.13
## fractal_dimension_worst	0.77	0.05	-0.05
##			
## radius_mean	perimeter_se	area_se	smoothness_se
## texture_mean	0.67	0.74	-0.22
## perimeter_mean	0.28	0.26	0.01
## area_mean	0.69	0.74	-0.20
## smoothness_mean	0.73	0.80	-0.17
## compactness_mean	0.30	0.25	0.33
## concavity_mean	0.55	0.46	0.14
## concave points_mean	0.66	0.62	0.10
## symmetry_mean	0.71	0.69	0.03
## fractal_dimension_mean	0.31	0.22	0.19
## radius_se	0.04	-0.09	0.40
## texture_se	0.97	0.95	0.16
## perimeter_se	0.22	0.11	0.40
## area_se	1.00	0.94	0.15
## smoothness_se	0.94	1.00	0.08
## compactness_se	0.15	0.08	1.00
## concavity_se	0.42	0.28	0.34
## concave points_se	0.36	0.27	0.27
## symmetry_se	0.56	0.42	0.33
## fractal_dimension_se	0.27	0.13	0.41
## radius_worst	0.24	0.13	0.43
## texture_worst	0.70	0.76	-0.23
## perimeter_worst	0.20	0.20	-0.07
## area_worst	0.72	0.76	-0.22
## smoothness_worst	0.73	0.81	-0.18
## compactness_worst	0.13	0.13	0.31
## concavity_worst	0.34	0.28	-0.06
## concave points_worst	0.42	0.39	-0.06
## symmetry_worst	0.55	0.54	-0.10
## fractal_dimension_worst	0.11	0.07	-0.11
##	0.09	0.02	0.10
##	concavity_se	concave points_se	symmetry_se
## radius_mean	0.19	0.38	-0.10
## texture_mean	0.14	0.16	0.01
## perimeter_mean	0.23	0.41	-0.08
## area_mean	0.21	0.37	-0.07
## smoothness_mean	0.25	0.38	0.20
## compactness_mean	0.57	0.64	0.23
## concavity_mean	0.69	0.68	0.18
## concave points_mean	0.44	0.62	0.10
## symmetry_mean	0.34	0.39	0.45

## fractal_dimension_mean	0.45	0.34	0.35
## radius_se	0.33	0.51	0.24
## texture_se	0.19	0.23	0.41
## perimeter_se	0.36	0.56	0.27
## area_se	0.27	0.42	0.13
## smoothness_se	0.27	0.33	0.41
## compactness_se	0.80	0.74	0.39
## concavity_se	1.00	0.77	0.31
## concave points_se	0.77	1.00	0.31
## symmetry_se	0.31	0.31	1.00
## fractal_dimension_se	0.73	0.61	0.37
## radius_worst	0.19	0.36	-0.13
## texture_worst	0.10	0.09	-0.08
## perimeter_worst	0.23	0.39	-0.10
## area_worst	0.19	0.34	-0.11
## smoothness_worst	0.17	0.22	-0.01
## compactness_worst	0.48	0.45	0.06
## concavity_worst	0.66	0.55	0.04
## concave points_worst	0.44	0.60	-0.03
## symmetry_worst	0.20	0.14	0.39
## fractal_dimension_worst	0.44	0.31	0.08
##	fractal_dimension_se	radius_worst	texture_worst
## radius_mean	-0.04	0.97	0.30
## texture_mean	0.05	0.35	0.91
## perimeter_mean	-0.01	0.97	0.30
## area_mean	-0.02	0.96	0.29
## smoothness_mean	0.28	0.21	0.04
## compactness_mean	0.51	0.54	0.25
## concavity_mean	0.45	0.69	0.30
## concave points_mean	0.26	0.83	0.29
## symmetry_mean	0.33	0.19	0.09
## fractal_dimension_mean	0.69	-0.25	-0.05
## radius_se	0.23	0.72	0.19
## texture_se	0.28	-0.11	0.41
## perimeter_se	0.24	0.70	0.20
## area_se	0.13	0.76	0.20
## smoothness_se	0.43	-0.23	-0.07
## compactness_se	0.80	0.20	0.14
## concavity_se	0.73	0.19	0.10
## concave points_se	0.61	0.36	0.09
## symmetry_se	0.37	-0.13	-0.08
## fractal_dimension_se	1.00	-0.04	0.00
## radius_worst	-0.04	1.00	0.36
## texture_worst	0.00	0.36	1.00
## perimeter_worst	0.00	0.99	0.37
## area_worst	-0.02	0.98	0.35
## smoothness_worst	0.17	0.22	0.23
## compactness_worst	0.39	0.48	0.36
## concavity_worst	0.38	0.57	0.37
## concave points_worst	0.22	0.79	0.36
## symmetry_worst	0.11	0.24	0.23
## fractal_dimension_worst	0.59	0.09	0.22
##	perimeter_worst	area_worst	smoothness_worst
## radius_mean	0.97	0.94	0.12

## texture_mean	0.36	0.34	0.08
## perimeter_mean	0.97	0.94	0.15
## area_mean	0.96	0.96	0.12
## smoothness_mean	0.24	0.21	0.81
## compactness_mean	0.59	0.51	0.57
## concavity_mean	0.73	0.68	0.45
## concave points_mean	0.86	0.81	0.45
## symmetry_mean	0.22	0.18	0.43
## fractal_dimension_mean	-0.21	-0.23	0.50
## radius_se	0.72	0.75	0.14
## texture_se	-0.10	-0.08	-0.07
## perimeter_se	0.72	0.73	0.13
## area_se	0.76	0.81	0.13
## smoothness_se	-0.22	-0.18	0.31
## compactness_se	0.26	0.20	0.23
## concavity_se	0.23	0.19	0.17
## concave points_se	0.39	0.34	0.22
## symmetry_se	-0.10	-0.11	-0.01
## fractal_dimension_se	0.00	-0.02	0.17
## radius_worst	0.99	0.98	0.22
## texture_worst	0.37	0.35	0.23
## perimeter_worst	1.00	0.98	0.24
## area_worst	0.98	1.00	0.21
## smoothness_worst	0.24	0.21	1.00
## compactness_worst	0.53	0.44	0.57
## concavity_worst	0.62	0.54	0.52
## concave points_worst	0.82	0.75	0.55
## symmetry_worst	0.27	0.21	0.49
## fractal_dimension_worst	0.14	0.08	0.62
##	compactness_worst	concavity_worst	concave points_worst
## radius_mean	0.41	0.53	0.74
## texture_mean	0.28	0.30	0.30
## perimeter_mean	0.46	0.56	0.77
## area_mean	0.39	0.51	0.72
## smoothness_mean	0.47	0.43	0.50
## compactness_mean	0.87	0.82	0.82
## concavity_mean	0.75	0.88	0.86
## concave points_mean	0.67	0.75	0.91
## symmetry_mean	0.47	0.43	0.43
## fractal_dimension_mean	0.46	0.35	0.18
## radius_se	0.29	0.38	0.53
## texture_se	-0.09	-0.07	-0.12
## perimeter_se	0.34	0.42	0.55
## area_se	0.28	0.39	0.54
## smoothness_se	-0.06	-0.06	-0.10
## compactness_se	0.68	0.64	0.48
## concavity_se	0.48	0.66	0.44
## concave points_se	0.45	0.55	0.60
## symmetry_se	0.06	0.04	-0.03
## fractal_dimension_se	0.39	0.38	0.22
## radius_worst	0.48	0.57	0.79
## texture_worst	0.36	0.37	0.36
## perimeter_worst	0.53	0.62	0.82
## area_worst	0.44	0.54	0.75

## smoothness_worst	0.57	0.52	0.55
## compactness_worst	1.00	0.89	0.80
## concavity_worst	0.89	1.00	0.86
## concave points_worst	0.80	0.86	1.00
## symmetry_worst	0.61	0.53	0.50
## fractal_dimension_worst	0.81	0.69	0.51
##	symmetry_worst	fractal_dimension_worst	
## radius_mean	0.16	0.01	
## texture_mean	0.11	0.12	
## perimeter_mean	0.19	0.05	
## area_mean	0.14	0.00	
## smoothness_mean	0.39	0.50	
## compactness_mean	0.51	0.69	
## concavity_mean	0.41	0.51	
## concave points_mean	0.38	0.37	
## symmetry_mean	0.70	0.44	
## fractal_dimension_mean	0.33	0.77	
## radius_se	0.09	0.05	
## texture_se	-0.13	-0.05	
## perimeter_se	0.11	0.09	
## area_se	0.07	0.02	
## smoothness_se	-0.11	0.10	
## compactness_se	0.28	0.59	
## concavity_se	0.20	0.44	
## concave points_se	0.14	0.31	
## symmetry_se	0.39	0.08	
## fractal_dimension_se	0.11	0.59	
## radius_worst	0.24	0.09	
## texture_worst	0.23	0.22	
## perimeter_worst	0.27	0.14	
## area_worst	0.21	0.08	
## smoothness_worst	0.49	0.62	
## compactness_worst	0.61	0.81	
## concavity_worst	0.53	0.69	
## concave points_worst	0.50	0.51	
## symmetry_worst	1.00	0.54	
## fractal_dimension_worst	0.54	1.00	

```
# view correlation visually
```

```
dfii%>%
```

```
  cor(dfii) %>%
```

```
  corrplot(
```

```
    type = "upper",      # matrix: full, upper, or lower
```

```
    diag = F,           # remove diagonal
```

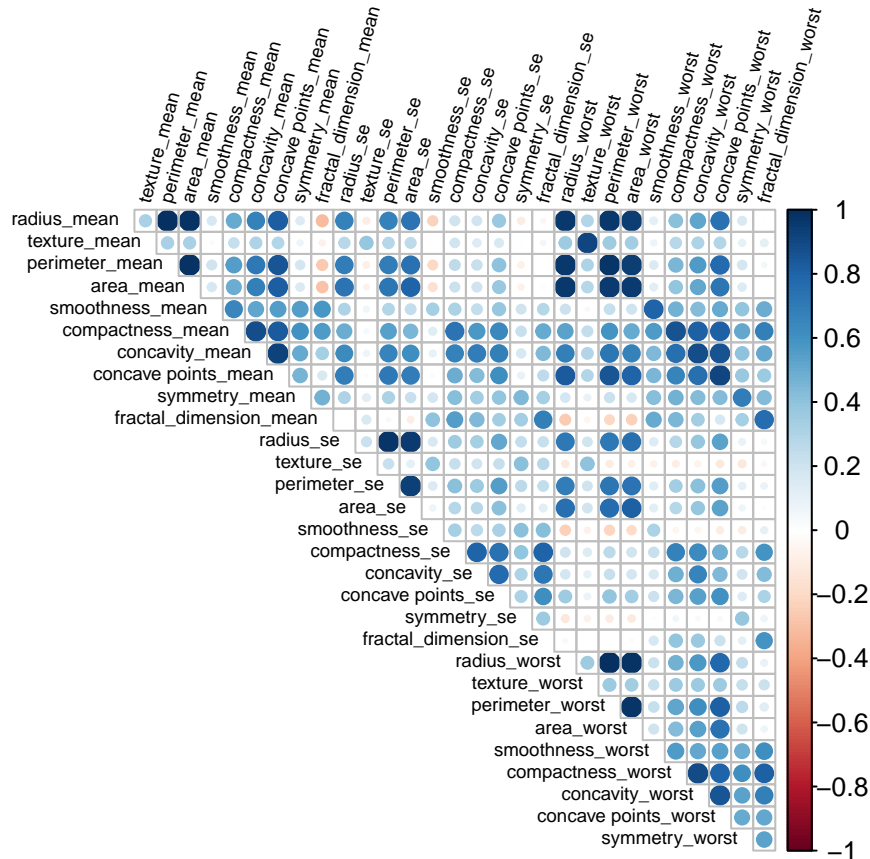
```
    order = "original", # order for labels
```

```
    tl.col = "black",    # font color
```

```
    tl.srt = 75,         # label angel
```

```
    tl.cex = 0.6         # font size
```

```
)
```



```
dfii %%% cor.test(radius_mean, perimeter_mean)
```

```
##
## Pearson's product-moment correlation
##
## data: radius_mean and perimeter_mean
## t = 513.79, df = 1136, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.9975910 0.9980906
## sample estimates:
## cor
## 0.9978553
```

```
dfii %%% cor.test(radius_mean, fractal_dimension_worst)
```

```
##
## Pearson's product-moment correlation
##
## data: radius_mean and fractal_dimension_worst
## t = 0.23816, df = 1136, p-value = 0.8118
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## -0.05106638 0.06515043
## sample estimates:
```

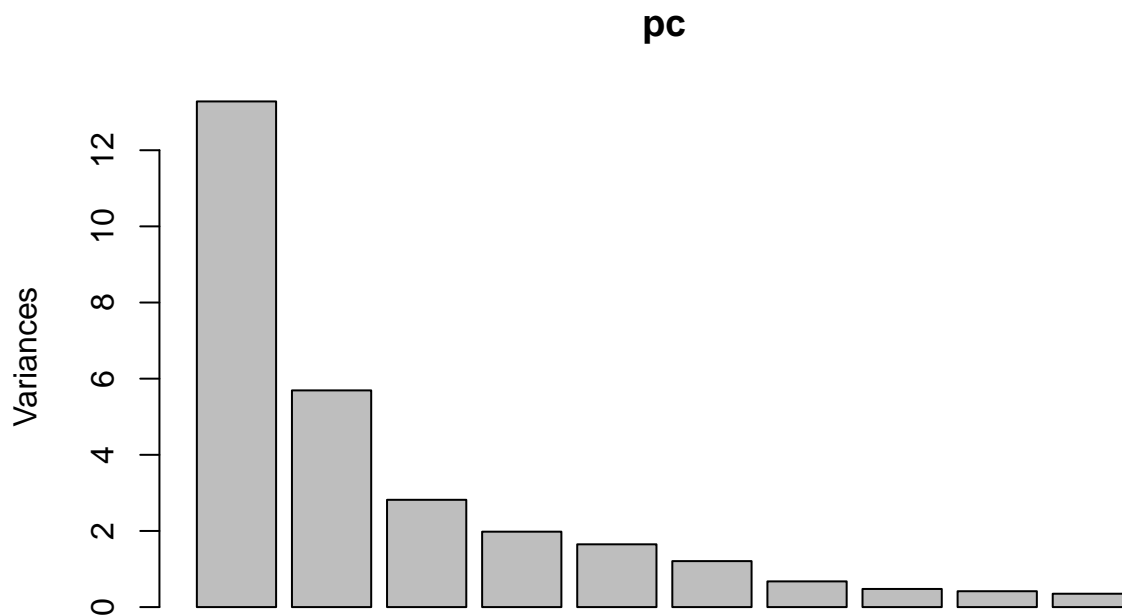
```
##          cor
## 0.007065886
```

## PRINCIPAL COMPONENT MODEL

```
pc <- dfii %>%
  prcomp(
    center = TRUE,
    scale = TRUE
  )
summary(pc)
```

```
## Importance of components:
##          PC1      PC2      PC3      PC4      PC5      PC6      PC7
## Standard deviation  3.6444 2.3857 1.67867 1.40735 1.28403 1.09880 0.82172
## Proportion of Variance 0.4427 0.1897 0.09393 0.06602 0.05496 0.04025 0.02251
## Cumulative Proportion 0.4427 0.6324 0.72636 0.79239 0.84734 0.88759 0.91010
##          PC8      PC9      PC10     PC11     PC12     PC13     PC14
## Standard deviation  0.69037 0.6457 0.59219 0.5421 0.51104 0.49128 0.39624
## Proportion of Variance 0.01589 0.0139 0.01169 0.0098 0.00871 0.00805 0.00523
## Cumulative Proportion 0.92598 0.9399 0.95157 0.9614 0.97007 0.97812 0.98335
##          PC15     PC16     PC17     PC18     PC19     PC20     PC21
## Standard deviation  0.30681 0.28260 0.24372 0.22939 0.22244 0.17652 0.1731
## Proportion of Variance 0.00314 0.00266 0.00198 0.00175 0.00165 0.00104 0.0010
## Cumulative Proportion 0.98649 0.98915 0.99113 0.99288 0.99453 0.99557 0.9966
##          PC22     PC23     PC24     PC25     PC26     PC27     PC28
## Standard deviation  0.16565 0.15602 0.1344 0.12442 0.09043 0.08307 0.03987
## Proportion of Variance 0.00091 0.00081 0.0006 0.00052 0.00027 0.00023 0.00005
## Cumulative Proportion 0.99749 0.99830 0.9989 0.99942 0.99969 0.99992 0.99997
##          PC29     PC30
## Standard deviation  0.02736 0.01153
## Proportion of Variance 0.00002 0.00000
## Cumulative Proportion 1.00000 1.00000
```

```
# plot the principal component
plot(pc)
```



## VERY SIMPLE STRUCTURE

Use “Very Simple Structure” to suggest number of factors Note: MAP = Minimum Absolute Partial correlation; n is the proposed maximum number of factors

```
dfii %>%
  dplyr::select(1:30) %>%
  vss(n = 10)
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```

## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully

## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.

## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully

## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.

## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully

## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.

## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully

## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.

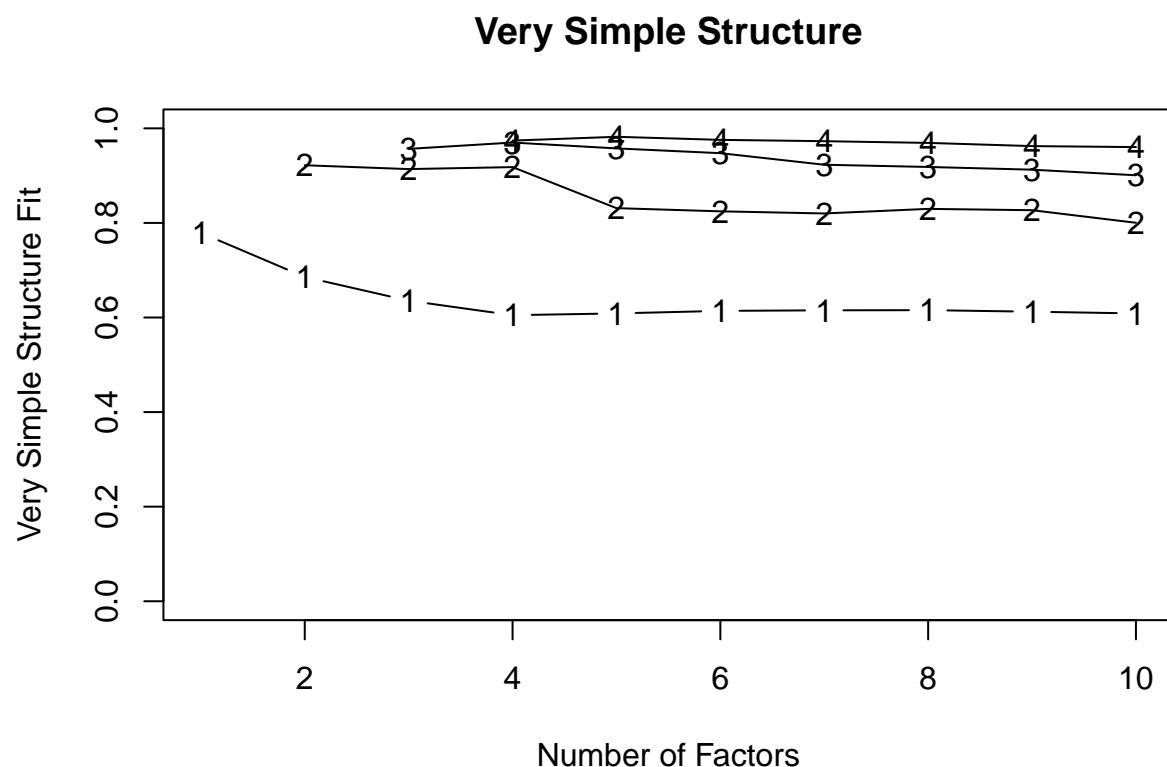
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully

## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.

## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully

```





```
##
## Very Simple Structure
## Call: vss(x = ., n = 10)
## VSS complexity 1 achieves a maximum of 0.78 with 1 factors
## VSS complexity 2 achieves a maximum of 0.92 with 2 factors
##
## The Velicer MAP achieves a minimum of 0.07 with 9 factors
## BIC achieves a minimum of 13154.76 with 10 factors
## Sample Size adjusted BIC achieves a minimum of 13726.5 with 10 factors
##
## Statistics by number of factors
##   vss1 vss2  map dof chisq prob sqresid fit RMSEA  BIC SABIC complex eChisq
## 1  0.78 0.00 0.171 405 58779 0 49.94 0.78 0.36 55929 57216 1.0 42948
## 2  0.69 0.92 0.087 376 39285 0 17.64 0.92 0.30 36640 37834 1.3 12052
## 3  0.64 0.91 0.073 348 33490 0 9.82 0.96 0.29 31041 32146 1.5 5812
## 4  0.60 0.92 0.079 321 30570 0 5.86 0.97 0.29 28311 29330 1.5 3345
## 5  0.61 0.83 0.071 295 26904 0 3.13 0.99 0.28 24828 25765 1.7 1439
## 6  0.61 0.82 0.072 270 25412 0 1.58 0.99 0.29 23512 24370 1.9 597
## 7  0.62 0.82 0.075 246 21868 0 1.11 1.00 0.28 20137 20918 2.0 290
## 8  0.62 0.83 0.080 223 19217 0 0.93 1.00 0.27 17648 18356 2.1 184
## 9  0.61 0.83 0.068 201 16330 0 0.74 1.00 0.27 14915 15554 2.1 109
## 10 0.61 0.80 0.077 180 14421 0 0.63 1.00 0.26 13155 13726 2.2 64
##   SRMR eCRMS eBIC
## 1 0.2083 0.216 40098
## 2 0.1103 0.119 9406
## 3 0.0766 0.086 3363
```

```
## 4  0.0581 0.068  1086
## 5  0.0381 0.046  -637
## 6  0.0246 0.031 -1303
## 7  0.0171 0.023 -1441
## 8  0.0136 0.019 -1385
## 9  0.0105 0.015 -1305
## 10 0.0081 0.013 -1202
```

```
# Or use "nfactors" to do the same; includes VSS
dfii %>%
  dplyr::select(1:30) %>%
  nfactors(n = 10)
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.

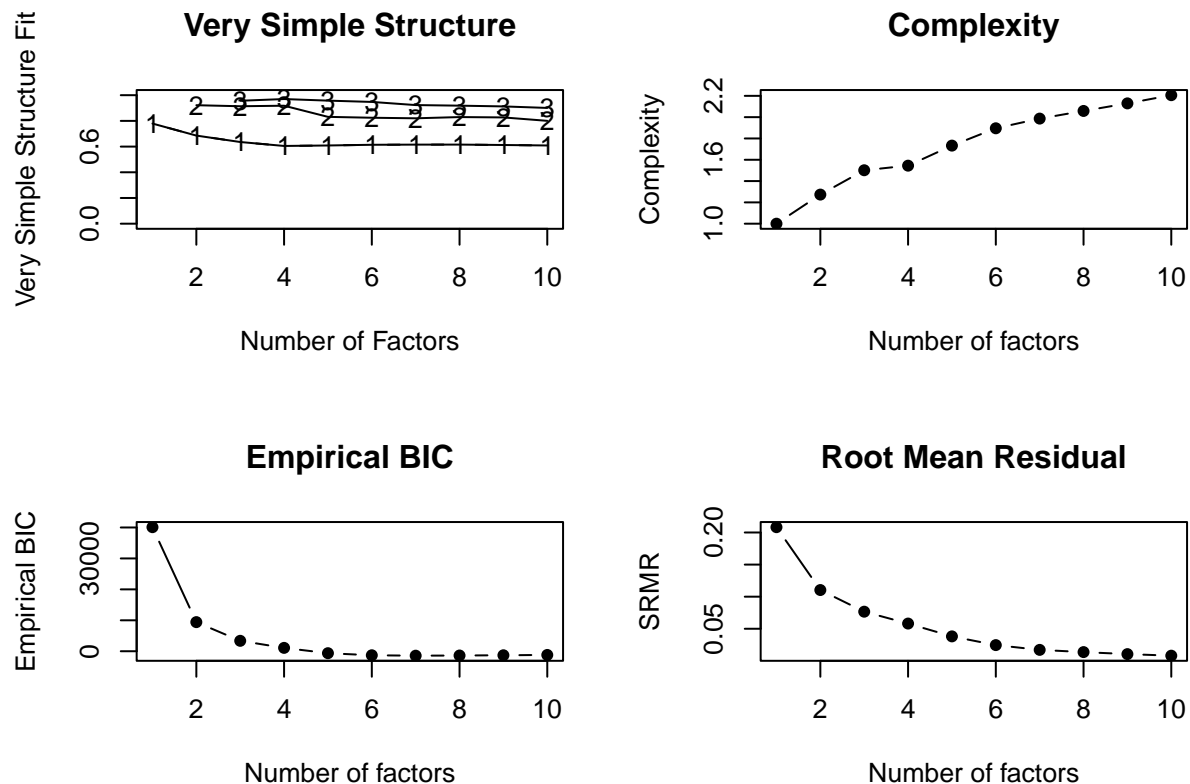
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully

## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.

## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully

## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.

## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully
```



```
##
## Number of factors
## Call: vss(x = x, n = n, rotate = rotate, diagonal = diagonal, fm = fm,
```

```
##      n.obs = n.obs, plot = FALSE, title = title, use = use, cor = cor)
## VSS complexity 1 achieves a maximum of 0.78 with 1 factors
## VSS complexity 2 achieves a maximum of 0.92 with 2 factors
## The Velicer MAP achieves a minimum of 0.07 with 9 factors
## Empirical BIC achieves a minimum of -1440.79 with 7 factors
## Sample Size adjusted BIC achieves a minimum of 13726.5 with 10 factors
##
## Statistics by number of factors
##      vss1 vss2  map dof chisq prob sqresid fit RMSEA  BIC SABIC complex eChisq
## 1  0.78 0.00 0.171 405 58779 0 49.94 0.78 0.36 55929 57216 1.0 42948
## 2  0.69 0.92 0.087 376 39285 0 17.64 0.92 0.30 36640 37834 1.3 12052
## 3  0.64 0.91 0.073 348 33490 0 9.82 0.96 0.29 31041 32146 1.5 5812
## 4  0.60 0.92 0.079 321 30570 0 5.86 0.97 0.29 28311 29330 1.5 3345
## 5  0.61 0.83 0.071 295 26904 0 3.13 0.99 0.28 24828 25765 1.7 1439
## 6  0.61 0.82 0.072 270 25412 0 1.58 0.99 0.29 23512 24370 1.9 597
## 7  0.62 0.82 0.075 246 21868 0 1.11 1.00 0.28 20137 20918 2.0 290
## 8  0.62 0.83 0.080 223 19217 0 0.93 1.00 0.27 17648 18356 2.1 184
## 9  0.61 0.83 0.068 201 16330 0 0.74 1.00 0.27 14915 15554 2.1 109
## 10 0.61 0.80 0.077 180 14421 0 0.63 1.00 0.26 13155 13726 2.2 64
##      SRMR eCRMS eBIC
## 1  0.2083 0.216 40098
## 2  0.1103 0.119 9406
## 3  0.0766 0.086 3363
## 4  0.0581 0.068 1086
## 5  0.0381 0.046 -637
## 6  0.0246 0.031 -1303
## 7  0.0171 0.023 -1441
## 8  0.0136 0.019 -1385
## 9  0.0105 0.015 -1305
## 10 0.0081 0.013 -1202
```

## FACTOR ANALYSIS

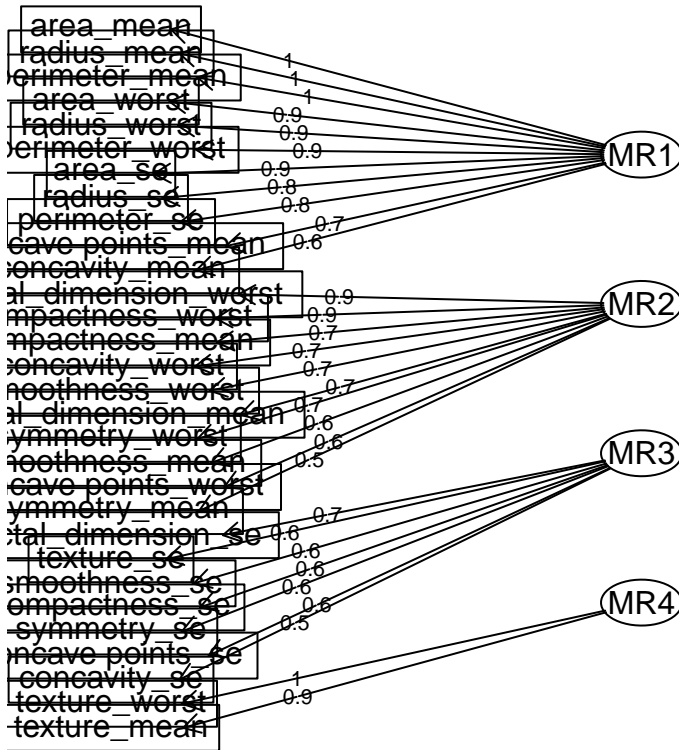
Factor analysis using minimum residual (minres) method and oblimin rotation, which is useful for simple structure. Need to enter desired number of factors (from VSS or nfactors above).

```
# Calculate and plot factors with fa()
dfii %>%
  dplyr::select(1:30) %>%      # Select variables
  fa(                          # Use fa() function
    nfactors = 4,              # Use four factors
    rotate = "oblimin"         # Oblimin oblique rotation
  ) %T>%                       # T-pipe
  fa.diagram() %>%             # Diagram of factors and variables
  print()                      # Print results
```

```
## Warning in fa.stats(r = r, f = f, phi = phi, n.obs = n.obs, np.obs = np.obs, :
## The estimated weights for the factor scores are probably incorrect. Try a
## different factor score estimation method.
```

```
## Warning in fac(r = r, nfactors = nfactors, n.obs = n.obs, rotate = rotate, : An
## ultra-Heywood case was detected. Examine the results carefully
```

## Factor Analysis



```
## Factor Analysis using method = minres
## Call: fa(r = ., nfactors = 4, rotate = "oblimin")
## Standardized loadings (pattern matrix) based upon correlation matrix
##
```

	MR1	MR2	MR3	MR4	h2	u2	com
## radius_mean	0.97	0.00	-0.13	0.04	0.95	0.047	1.0
## texture_mean	0.07	-0.05	0.10	0.89	0.83	0.171	1.0
## perimeter_mean	0.96	0.05	-0.10	0.04	0.96	0.043	1.0
## area_mean	0.98	-0.04	-0.06	0.03	0.96	0.036	1.0
## smoothness_mean	0.09	0.62	0.14	-0.15	0.47	0.531	1.3
## compactness_mean	0.33	0.74	0.21	0.01	0.92	0.082	1.6
## concavity_mean	0.55	0.53	0.24	0.05	0.91	0.090	2.4
## concave points_mean	0.74	0.41	0.09	0.00	0.92	0.082	1.6
## symmetry_mean	0.06	0.54	0.23	-0.05	0.42	0.585	1.4
## fractal_dimension_mean	-0.43	0.72	0.40	-0.10	0.81	0.191	2.3
## radius_se	0.81	-0.14	0.45	-0.01	0.84	0.160	1.6
## texture_se	-0.16	-0.25	0.62	0.46	0.55	0.452	2.4
## perimeter_se	0.80	-0.10	0.46	-0.01	0.85	0.152	1.6
## area_se	0.87	-0.15	0.31	-0.03	0.81	0.189	1.3
## smoothness_se	-0.19	0.02	0.62	-0.03	0.40	0.597	1.2
## compactness_se	0.07	0.48	0.58	0.06	0.74	0.257	2.0
## concavity_se	0.11	0.37	0.54	0.02	0.58	0.423	1.9
## concave points_se	0.34	0.27	0.57	-0.04	0.66	0.343	2.1
## symmetry_se	-0.09	0.06	0.58	-0.03	0.35	0.650	1.1
## fractal_dimension_se	-0.13	0.38	0.66	-0.02	0.67	0.326	1.7
## radius_worst	0.94	0.08	-0.17	0.08	0.98	0.020	1.1
## texture_worst	-0.02	0.09	-0.05	1.00	1.02	-0.017	1.0

```

## perimeter_worst      0.93  0.13 -0.14  0.08 0.99  0.014 1.1
## area_worst           0.94  0.03 -0.11  0.07 0.95  0.053 1.0
## smoothness_worst    -0.02  0.72 -0.09  0.00 0.49  0.510 1.0
## compactness_worst   0.16  0.85 -0.08  0.15 0.88  0.120 1.2
## concavity_worst      0.32  0.74 -0.02  0.14 0.84  0.163 1.4
## concave points_worst 0.59  0.61 -0.12  0.06 0.92  0.076 2.1
## symmetry_worst      -0.01  0.67 -0.14  0.05 0.43  0.565 1.1
## fractal_dimension_worst -0.24  0.92  0.02  0.08 0.81  0.192 1.2
##
##              MR1  MR2  MR3  MR4
## SS loadings    9.87 6.99 3.77 2.26
## Proportion Var 0.33 0.23 0.13 0.08
## Cumulative Var 0.33 0.56 0.69 0.76
## Proportion Explained 0.43 0.31 0.16 0.10
## Cumulative Proportion 0.43 0.74 0.90 1.00
##
## With factor correlations of
##      MR1  MR2  MR3  MR4
## MR1 1.00 0.28 0.08 0.27
## MR2 0.28 1.00 0.24 0.15
## MR3 0.08 0.24 1.00 0.02
## MR4 0.27 0.15 0.02 1.00
##
## Mean item complexity = 1.5
## Test of the hypothesis that 4 factors are sufficient.
##
## The degrees of freedom for the null model are 435 and the objective function was 70.65 with Chi S
## The degrees of freedom for the model are 321 and the objective function was 27.21
##
## The root mean square of the residuals (RMSR) is 0.06
## The df corrected root mean square of the residuals is 0.07
##
## The harmonic number of observations is 1138 with the empirical chi square 3344.58 with prob < 0
## The total number of observations was 1138 with Likelihood Chi Square = 30569.59 with prob < 0
##
## Tucker Lewis Index of factoring reliability = 0.481
## RMSEA index = 0.288 and the 90 % confidence intervals are 0.285 0.291
## BIC = 28310.7
## Fit based upon off diagonal values = 0.99

```

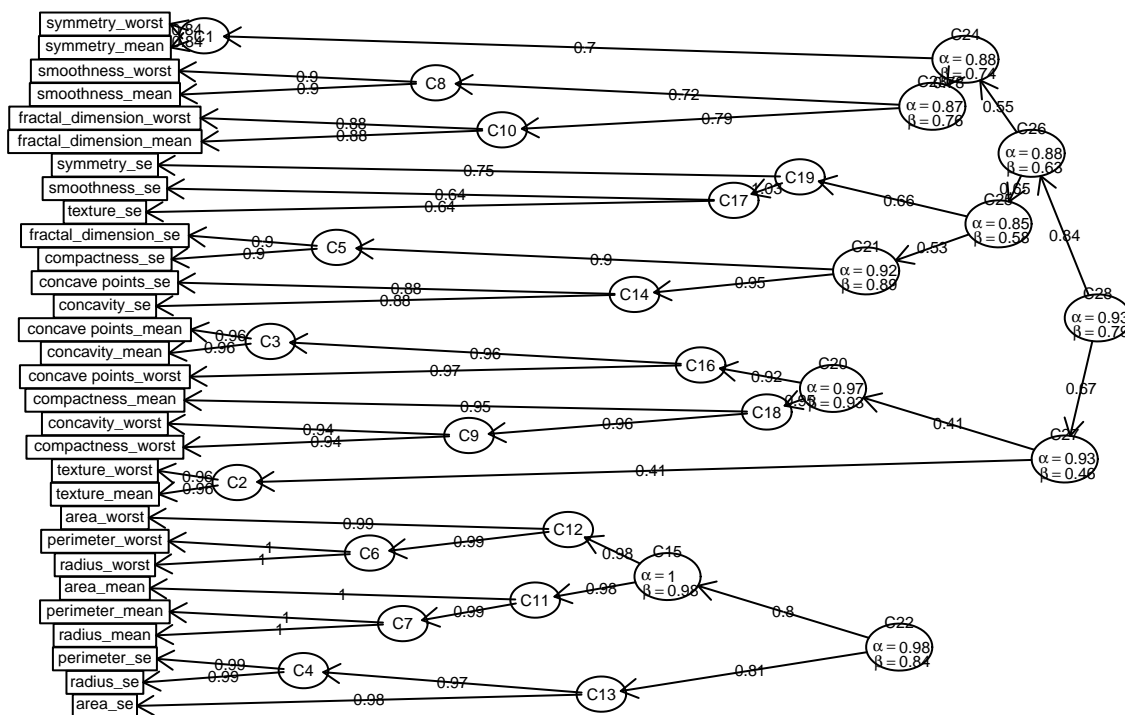
## HIERARCHICAL CLUSTERING

```

# Hierarchical clustering of items with iclust()
dfii %>%
  dplyr::select(1:30) %>%
  iclust()

```

# ICLUST



```
## ICLUST (Item Cluster Analysis)
## Call: iclust(r.mat = .)
##
## Purified Alpha:
## C22 C28
## 0.98 0.92
##
## G6* reliability:
## C22 C28
## 1 1
##
## Original Beta:
## C22 C28
## 0.84 0.79
##
## Cluster size:
## C22 C28
## 11 19
##
## Item by Cluster Structure matrix:
##      0 P C22 C28
## radius_mean C22 C22 0.95 0.33
## texture_mean C28 C28 0.34 0.36
## perimeter_mean C22 C22 0.96 0.38
## area_mean C22 C22 0.97 0.33
## smoothness_mean C28 C28 0.28 0.65
```

```

## compactness_mean      C28 C28  0.60  0.90
## concavity_mean        C28 C28  0.76  0.84
## concave points_mean   C28 C22  0.87  0.72
## symmetry_mean         C28 C28  0.26  0.65
## fractal_dimension_mean C28 C28 -0.15 0.66
## radius_se              C22 C22  0.86  0.46
## texture_se             C28 C28  0.00  0.26
## perimeter_se           C22 C22  0.85  0.49
## area_se                C22 C22  0.88  0.38
## smoothness_se          C28 C28 -0.09 0.33
## compactness_se         C28 C28  0.31  0.81
## concavity_se           C28 C28  0.29  0.71
## concave points_se      C28 C28  0.47  0.71
## symmetry_se            C28 C28  0.01  0.39
## fractal_dimension_se   C28 C28  0.08  0.66
## radius_worst           C22 C22  0.96  0.37
## texture_worst          C28 C28  0.31  0.38
## perimeter_worst        C22 C22  0.97  0.43
## area_worst             C22 C22  0.96  0.36
## smoothness_worst       C28 C28  0.21  0.61
## compactness_worst      C28 C28  0.46  0.81
## concavity_worst        C28 C28  0.57  0.81
## concave points_worst   C28 C22  0.77  0.73
## symmetry_worst         C28 C28  0.20  0.57
## fractal_dimension_worst C28 C28  0.10  0.75
##
## With eigenvalues of:
## C22 C28
## 9.9 8.6
##
## Purified scale intercorrelations
## reliabilities on diagonal
## correlations corrected for attenuation above diagonal:
##      C22  C28
## C22 0.98 0.48
## C28 0.45 0.92
##
## Cluster fit = 0.81   Pattern fit = 0.95   RMSR = 0.12

```

## PC WITH K FACTORS

```

# First PCA with no rotation, specify 5 factors
dfii %>% principal(nfactors = 4)

```

```

## Principal Components Analysis
## Call: principal(r = ., nfactors = 4)
## Standardized loadings (pattern matrix) based upon correlation matrix
##      RC1  RC2  RC3  RC4  h2  u2 com
## radius_mean      0.95  0.12 -0.13  0.10 0.95 0.049 1.1
## texture_mean      0.25  0.06  0.05  0.91 0.90 0.105 1.2
## perimeter_mean     0.95  0.17 -0.10  0.10 0.95 0.045 1.1
## area_mean          0.97  0.09 -0.08  0.09 0.96 0.039 1.0

```



```

## smoothness_mean      0.16  0.65  0.26 -0.19  0.55  0.452 1.6
## compactness_mean     0.46  0.76  0.33  0.05  0.91  0.091 2.1
## concavity_mean       0.66  0.60  0.32  0.11  0.91  0.092 2.5
## concave points_mean  0.80  0.50  0.16  0.05  0.92  0.079 1.8
## symmetry_mean        0.14  0.58  0.34 -0.05  0.47  0.527 1.8
## fractal_dimension_mean -0.31  0.66  0.54 -0.09  0.83  0.174 2.4
## radius_se            0.83 -0.01  0.40  0.03  0.85  0.151 1.4
## texture_se           -0.06 -0.22  0.60  0.54  0.70  0.298 2.3
## perimeter_se         0.82  0.03  0.42  0.04  0.85  0.145 1.5
## area_se              0.87 -0.02  0.26  0.01  0.83  0.167 1.2
## smoothness_se        -0.14  0.00  0.70 -0.05  0.51  0.487 1.1
## compactness_se       0.21  0.51  0.67  0.11  0.76  0.237 2.1
## concavity_se         0.22  0.40  0.64  0.07  0.62  0.378 2.0
## concave points_se    0.43  0.32  0.64  0.00  0.70  0.303 2.3
## symmetry_se          -0.04  0.05  0.67 -0.03  0.45  0.545 1.0
## fractal_dimension_se -0.03  0.38  0.75  0.02  0.71  0.287 1.5
## radius_worst         0.94  0.20 -0.16  0.13  0.97  0.028 1.2
## texture_worst        0.20  0.19 -0.07  0.94  0.95  0.045 1.2
## perimeter_worst      0.94  0.25 -0.13  0.14  0.98  0.022 1.2
## area_worst           0.95  0.16 -0.11  0.12  0.95  0.053 1.1
## smoothness_worst     0.06  0.76  0.01 -0.01  0.58  0.423 1.0
## compactness_worst    0.32  0.86  0.07  0.19  0.88  0.123 1.4
## concavity_worst      0.46  0.77  0.11  0.19  0.84  0.155 1.8
## concave points_worst 0.68  0.66 -0.01  0.11  0.92  0.082 2.1
## symmetry_worst       0.07  0.72 -0.05  0.06  0.53  0.475 1.0
## fractal_dimension_worst -0.08  0.88  0.19  0.11  0.83  0.173 1.1
##
##              RC1  RC2  RC3  RC4
## SS loadings    10.19 6.93 4.40 2.26
## Proportion Var  0.34 0.23 0.15 0.08
## Cumulative Var  0.34 0.57 0.72 0.79
## Proportion Explained 0.43 0.29 0.19 0.09
## Cumulative Proportion 0.43 0.72 0.91 1.00
##
## Mean item complexity = 1.5
## Test of the hypothesis that 4 components are sufficient.
##
## The root mean square of the residuals (RMSR) is 0.06
## with the empirical chi square 3773.14 with prob < 0
##
## Fit based upon off diagonal values = 0.98

```

```

# Second PCA with oblimin (oblique) rotation
dfii %>%
  principal(
    nfactors = 4,
    rotate = "oblimin"
  ) %>%
  plot() # Plot position of variables on components

```

## Principal Component Analysis

