Clustering and t-test

T-test for change in Clinical Outcome with p values

```
\% \ change \ in \ outcome = \frac{first \ read - second \ read}{first \ read}
```

```
data <- read_excel("finaldata.xlsx", sheet = NULL, col_types = c(rep("guess", 6),</pre>
                                                                  "date", rep("guess", 2),
                                                                  rep("numeric", 10)))
status <- factor(c("low", rep("intermediate", 4),</pre>
                    "low", "intermediate", "intermediate",
                    rep("low", 2), rep("intermediate", 3),
                    "low", rep("intermediate", 2), rep("low", 2)))
status <- forcats::fct relevel(status, c("low"))</pre>
#smoking <- as.factor(c("low", "high", "low", "high", "high", "low",
                         rep("high", 3), rep("low", 3), "high", "NA", "high", "low"))
\#smoking <- smoking[-c(8, 13, 18:20)]
smoking <-</pre>
patient <- 1:18
first_read <- data[seq(1, 54, 3), ]
second_read <- data[seq(2, 54, 3), ]</pre>
pval <- NULL
for(i in 10:19){
  change <- 1 - as.numeric(unlist(second_read[, i] / first_read[, i]))</pre>
  cat(names(second_read[, i]), "with p-value of", t.test(change ~ status)$p.value, "\n")
  pval <- c(pval, t.test(change ~ status)$p.value)</pre>
}
## Tumor Volume (mL) with p-value of 0.651912
## Largest Node Volume (mL) with p-value of 0.06978384
## SULmax Tumor with p-value of 0.2943616
## SULmedian with p-value of 0.3630501
## SULpeak with p-value of 0.3854198
## SULmax Largest Node with p-value of 0.0148899
## SULmedian node with p-value of 0.6197499
## SULpeak node with p-value of 0.03951802
## Diffusion Mean Tumor with p-value of 0.1825276
## Diffusion mean ADC Largest Node with p-value of 0.7309697
```

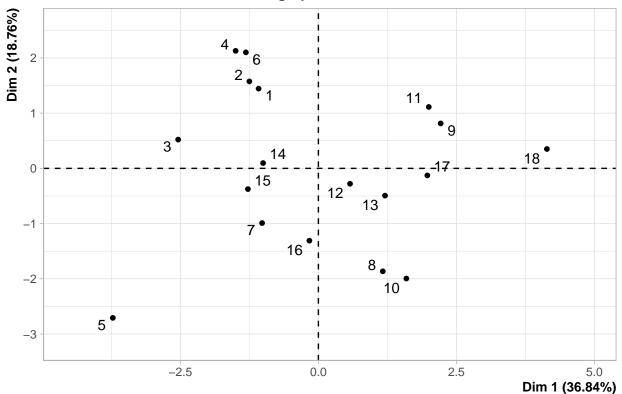
```
change_matrix <- array(NA, c(18, 10))

for(i in 1:10){
   change_matrix[, i] <- 1 - as.numeric(unlist(second_read[1:18, i + 9] / first_read[1:18, i + 9]))
}

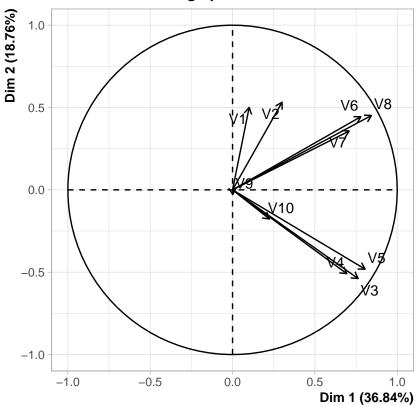
change_matrix[is.na(change_matrix)] <- 0

pca <- PCA(change_matrix)</pre>
```

PCA graph of individuals



PCA graph of variables



knitr::kable(cbind(names(data)[10:19], as.numeric(round(pca\$var\$coord[, 1], 3))))

Tumor Volume (mL)	0.101
Largest Node Volume (mL)	0.301
SULmax Tumor	0.763
SULmedian	0.694
SULpeak	0.805
SULmax Largest Node	0.779
SULmedian node	0.709
SULpeak node	0.844
Diffusion Mean Tumor	0.002
Diffusion mean ADC Largest Node	0.225

```
18
                                                                     11
                                9
                                10
                                                                     13
     0.5 -
Mean change in 10 outcomes
                                                                     12
45
                                                                     2
                                                                     16
             factor(cluster)
                                                                     3
      0.0 -
                                                                     7
                                                                     5
    -0.5 -
                                                               intermediate
                              low
                                           Risk Groups
# Patient with in cluster 1
```

knitr::kable(data.frame(Patient = dat\$patient, Cluster = dat\$cluster, value = round(change_matrix[, 6],

Patient	Cluster	value
1	2	0.427
2	2	0.289
3	1	0.029
4	2	0.363
5	1	-0.456
6	2	0.431
7	1	-0.232
8	2	0.287
9	2	0.605
10	2	0.548
11	2	0.636
12	2	0.422
13	2	0.529
14	2	0.413
15	2	0.376
16	2	0.228
17	2	0.500
18	2	0.789

Patient Characteristics

```
patient_data <- read_excel("patient.xlsx")</pre>
patient_data <- cbind(patient_data, change_matrix)</pre>
names(patient_data)[10:19] <- names(data)[9:18]</pre>
names(patient_data)[5] <- "Smoking"</pre>
patient_data <- data.frame(cbind(patient_data, cluster = dat$cluster))</pre>
## Tumor volume and SULpeak change and smoking are related
summary(lm(Tumor.Volume..mL. ~ Smoking, data = patient_data))
##
## Call:
## lm(formula = Tumor.Volume..mL. ~ Smoking, data = patient_data)
## Residuals:
##
                  1Q Median
                                    3Q
        Min
                                            Max
## -0.73028 -0.04801 0.07703 0.16304 0.35462
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.312043 0.080201
                                    3.891 0.0013 **
              -0.007487
                           0.003088 -2.425 0.0275 *
## Smoking
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2775 on 16 degrees of freedom
## Multiple R-squared: 0.2688, Adjusted R-squared: 0.223
## F-statistic: 5.88 on 1 and 16 DF, p-value: 0.02752
summary(lm(SULpeak ~ Smoking, data = patient_data))
```

```
##
## Call:
## lm(formula = SULpeak ~ Smoking, data = patient_data)
## Residuals:
##
       Min
                 1Q
                     Median
                                   3Q
                                           Max
## -0.79289 -0.06923 0.02890 0.14945 0.34876
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.440061
                          0.080297
                                   5.480 5.03e-05 ***
              -0.006420
                          0.003091 -2.077
## Smoking
                                             0.0543 .
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2779 on 16 degrees of freedom
## Multiple R-squared: 0.2123, Adjusted R-squared: 0.1631
## F-statistic: 4.313 on 1 and 16 DF, p-value: 0.05429
## Largest node volume change, SULmax tumor, and SULmedian and gender are related
summary(lm(Largest.Node.Volume..mL. ~ Sex, data = patient_data))
##
## Call:
## lm(formula = Largest.Node.Volume..mL. ~ Sex, data = patient_data)
## Residuals:
##
                 1Q
                     Median
                                   3Q
       Min
## -0.35757 -0.12107 0.01071 0.10741 0.36614
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
                           0.1403
                                   4.043 0.000942 ***
## (Intercept) 0.5673
                           0.1488 -2.187 0.043919 *
## SexM
               -0.3255
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.1984 on 16 degrees of freedom
## Multiple R-squared: 0.2302, Adjusted R-squared: 0.1821
## F-statistic: 4.784 on 1 and 16 DF, p-value: 0.04392
summary(lm(SULmax.Tumor ~ Sex, data = patient_data))
##
## Call:
## lm(formula = SULmax.Tumor ~ Sex, data = patient_data)
## Residuals:
       Min
                 1Q
                      Median
                                           Max
                                   30
## -0.52226 -0.17244 -0.01515 0.18506 0.37369
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                0.6055
                           0.1713
                                   3.535 0.00275 **
               -0.3742
                           0.1817 -2.060 0.05609 .
## SexM
```

```
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2422 on 16 degrees of freedom
## Multiple R-squared: 0.2096, Adjusted R-squared: 0.1602
## F-statistic: 4.242 on 1 and 16 DF, p-value: 0.05609
summary(lm(SULmedian ~ Sex, data = patient_data))
##
## Call:
## lm(formula = SULmedian ~ Sex, data = patient_data)
## Residuals:
##
       Min
                 1Q
                    Median
                                   ЗQ
## -0.34530 -0.06732 0.01716 0.10974 0.37577
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                           0.1395 4.134 0.00078 ***
## (Intercept) 0.5767
               -0.2932
                           0.1480 -1.981 0.06505 .
## SexM
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.1973 on 16 degrees of freedom
## Multiple R-squared: 0.197, Adjusted R-squared: 0.1468
## F-statistic: 3.924 on 1 and 16 DF, p-value: 0.06505
tb1 <- patient_data %>%
 dplyr::select(Age, cluster, Smoking, Sex, Tumor.site, Risk.grouping) %%
 tbl_summary(statistic = list(all_continuous() ~ "{mean} ({sd})",
                    all_categorical() ~ "{n} ({p}%)"))%>%
 bold_labels()
## Table printed with `knitr::kable()`, not {gt}. Learn why at
## http://www.danieldsjoberg.com/gtsummary/articles/rmarkdown.html
```

Characteristic	N = 18
$\overline{\mathbf{Age}}$	61 (7)
cluster	
1	3(17%)
2	15 (83%)
Smoking	15 (22)
Sex	, ,
F	2 (11%)
M	16 (89%)
Tumor.site	, ,
BOT	9 (50%)
Tonsil	9 (50%)
Risk.grouping	,
Intermediate	11 (61%)
Low	7 (39%)

To suppress this message, include `message = FALSE` in code chunk header.

- ## Table printed with `knitr::kable()`, not {gt}. Learn why at
 ## http://www.danieldsjoberg.com/gtsummary/articles/rmarkdown.html
 ## To suppress this message include `message = FALSE` in code shunk header
- ## To suppress this message, include `message = FALSE` in code chunk header.

Characteristic	1, N = 3	2, N = 15	p-value
Age	68 (5)	59 (7)	0.057
Smoking	30 (40)	12 (17)	0.2
Sex			> 0.9
F	0 (0%)	2(13%)	
M	3 (100%)	13 (87%)	
Tumor.site	, ,	, ,	> 0.9
BOT	2(67%)	7 (47%)	
Tonsil	1 (33%)	8 (53%)	
Risk.grouping	, ,	,	0.2
Intermediate	3 (100%)	8 (53%)	
Low	0 (0%)	7 (47%)	

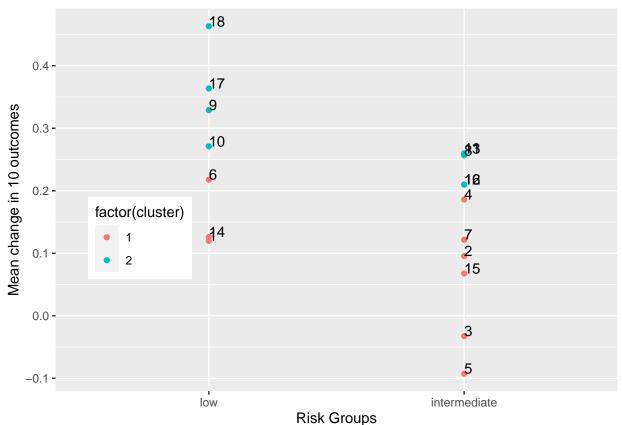
```
tbl_merge(list(tb1, tb2), tab_spanner = c(NA_character_, "**Cluster**"))
```

- ## Table printed with `knitr::kable()`, not {gt}. Learn why at
- ## http://www.danieldsjoberg.com/gtsummary/articles/rmarkdown.html
- ## To suppress this message, include `message = FALSE` in code chunk header.

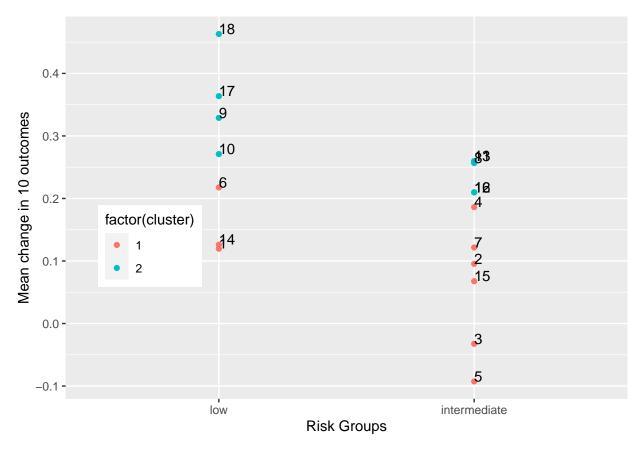
Characteristic	N = 18	1, N = 3	2, N = 15	p-value
Age	61 (7)	68 (5)	59 (7)	0.057
cluster				
1	3 (17%)			
2	15 (83%)			
Smoking	15 (22)	30(40)	12 (17)	0.2
Sex				> 0.9
F	2(11%)	0(0%)	2(13%)	
M	16 (89%)	3 (100%)	13~(87%)	
Tumor.site				> 0.9
BOT	9 (50%)	2(67%)	7 (47%)	
Tonsil	9 (50%)	1 (33%)	8 (53%)	
Risk.grouping				0.2
Intermediate	11 (61%)	3 (100%)	8 (53%)	
Low	7 (39%)	0 (0%)	7 (47%)	

Cross validation

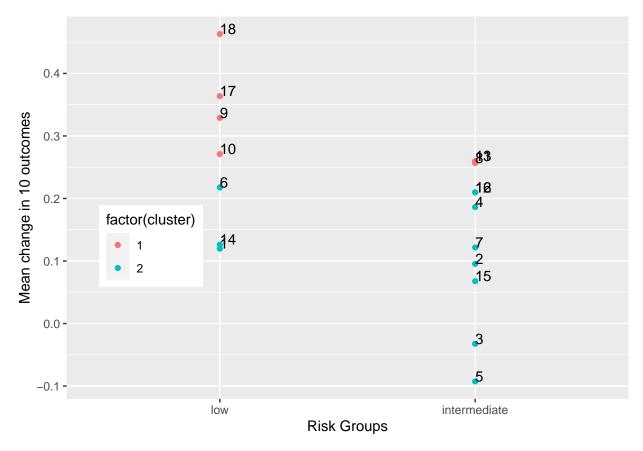
Removing Tumor Volume (mL)



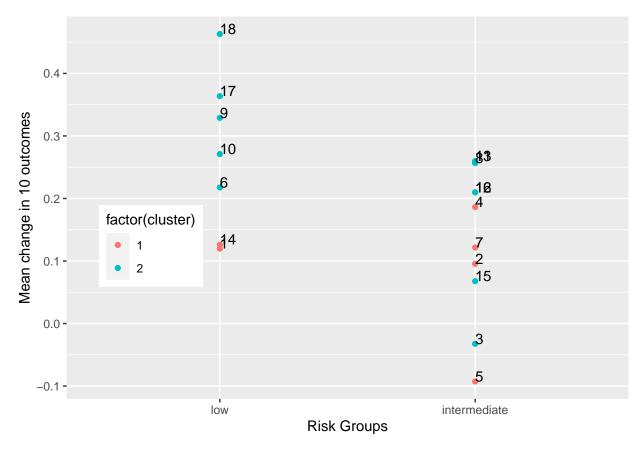
Removing Largest Node Volume (mL)



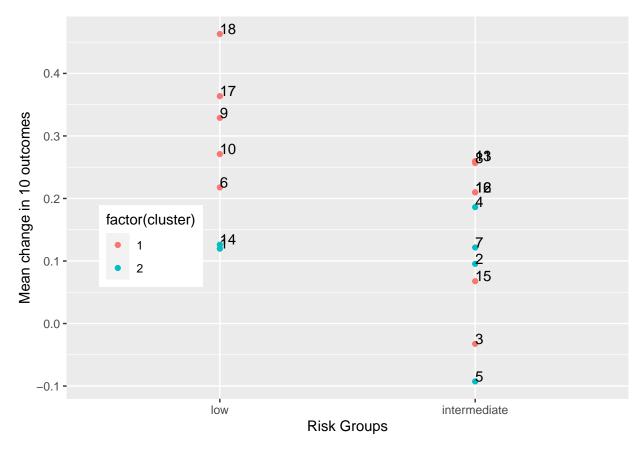
Removing SULmax Tumor



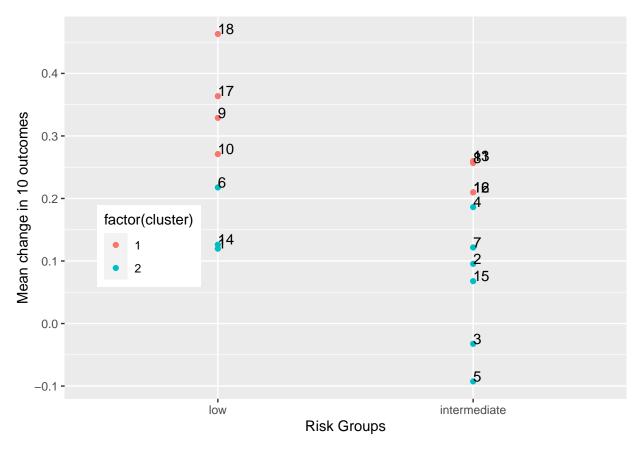
Removing SULmedian



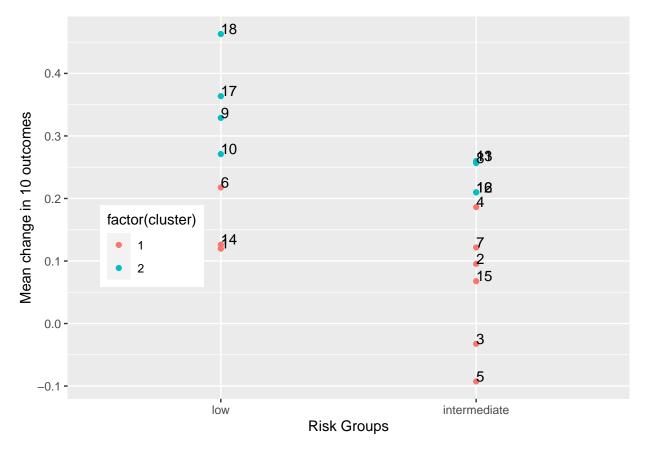
Removing SULpeak



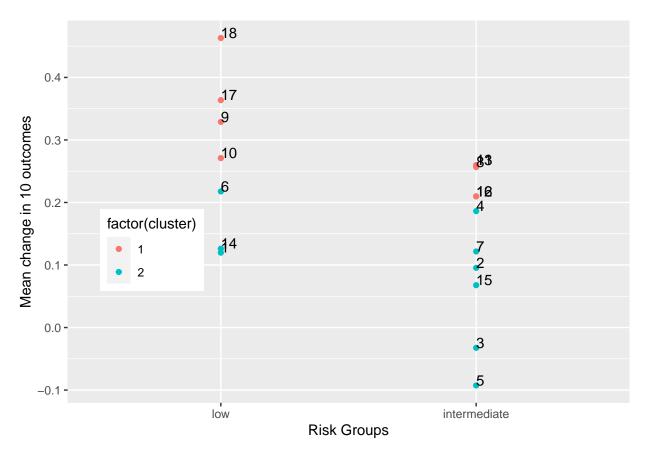
Removing SULmax Largest Node



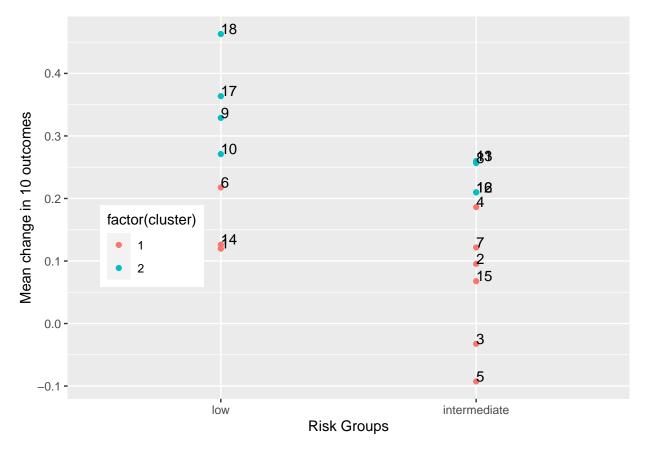
Removing SULmedian node



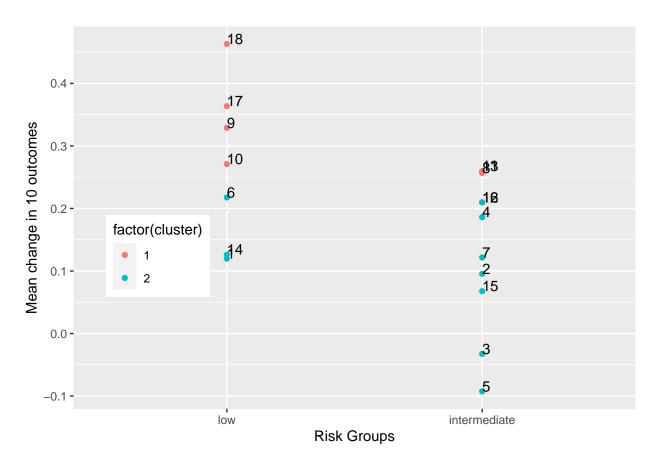
Removing SULpeak node



Removing Diffusion Mean Tumor



Removing Diffusion mean ADC Largest Node



Clustering table after removing each column

colnames(grouping) <- names(second_read[10:19])
knitr::kable(grouping)</pre>

Tumor				,	SULmax			Diffusion	Diffusion mean
Volume	Largest Node	SULmax			Largest	SULmedia	nSULpeak	Mean	ADC Largest
(mL)	Volume (mL)	Tumor	SULmed	d S hiLpeak		node	node	Tumor	Node
1	1	2	1	2	2	1	2	1	2
1	1	2	1	2	2	1	2	1	2
1	1	2	2	1	2	1	2	1	2
1	1	2	1	2	2	1	2	1	2
1	1	2	1	2	2	1	2	1	2
1	1	2	2	1	2	1	2	1	2
1	1	2	1	2	2	1	2	1	2
2	2	1	2	1	1	2	1	2	1
2	2	1	2	1	1	2	1	2	1
2	2	1	2	1	1	2	1	2	1
2	2	1	2	1	1	2	1	2	1
2	2	1	2	1	1	2	1	2	1
2	2	1	2	1	1	2	1	2	1
1	1	2	1	2	2	1	2	1	2
1	1	2	2	1	2	1	2	1	2
2	2	2	2	1	1	2	1	2	2
2	2	1	2	1	1	2	1	2	1

Tumor					SULmax			Diffusion	Diffusion mean
Volume	Largest Node	SULmax			Largest	SULmedia	nSULpeak	Mean	ADC Largest
(mL)	Volume (mL)	Tumor	SULmed	l S aldLpeak	Node	node	node	Tumor	Node
2	2	1	2	1	1	2	1	2	1