Preliminary Investigation

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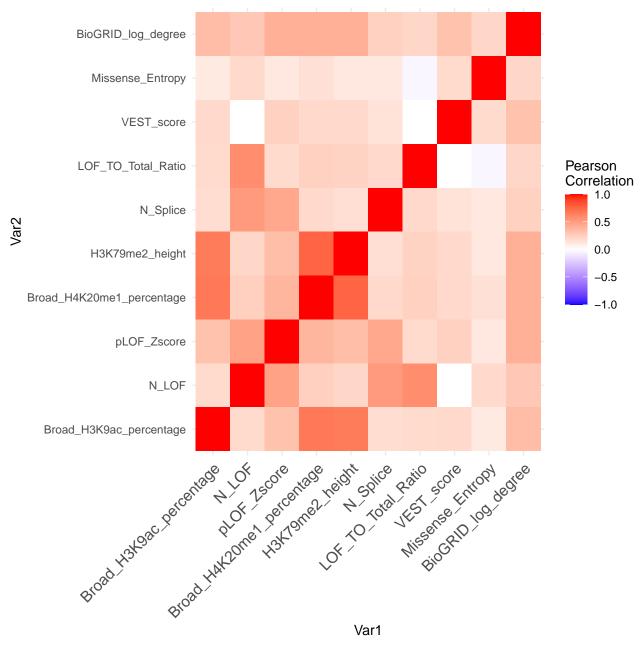
Setup

```
set.seed(110920)
library(tidyverse)
-- Attaching packages
v ggplot2 3.3.2
                    v purrr
                               0.3.4
v tibble 3.0.3
                    v dplyr
                               1.0.2
v tidvr
          1.1.2
                    v stringr 1.4.0
v readr
          1.3.1
                    v forcats 0.5.0
-- Conflicts -----
x dplyr::filter() masks stats::filter()
                  masks stats::lag()
x dplyr::lag()
# sample <- read.csv("sample.csv", stringsAsFactors = TRUE)
# sample
training <- read.csv("training.csv", stringsAsFactors = TRUE)</pre>
sort(abs(cor(training)["class", ]), decreasing = TRUE)[2:19]
                             Broad_H3K9ac_percentage
Broad_H4K2Ome1_percentage
                                                                 H4K20me1_width
                0.5309561
                                           0.4810581
                                                                      0.4789003
                                     H3K79me2_height
       BioGRID_log_degree
                                                                 H3K79me2_width
                0.4764364
                                           0.4719210
                                                                      0.4709266
Broad_H3K4me2_percentage Broad_H3K27ac_percentage
                                                                H4K20me1_height
                0.4693101
                                           0.4641367
                                                                      0.4595187
Broad_H3K4me1_percentage Broad_H3K79me2_percentage
                                                      Broad_H3K4me3_percentage
                0.4580446
                                           0.4571273
                                                                      0.4315878
Broad_H3K36me3_percentage
                                       H3K4me1_width
                                                                 H3K36me3_width
                0.4306293
                                           0.4287817
                                                                      0.4275987
              pLOF_Zscore
                                               N_LOF
                                                                  H3K4me2_width
                0.4227907
                                           0.4212450
                                                                      0.4080466
training$class <- factor(training$class)</pre>
levels(training$class) <- c("NG", "OG", "TSG")</pre>
dim(training)
[1] 3177
names(training)[c(1, 99)]
[1] "id"
            "class"
```

```
barplot(table(training$class))
1000
                 NG
                                          OG
                                                                  TSG
table(training$class) / nrow(training)
        NG
                    OG
                               TSG
0.89392509 0.05288008 0.05319484
any(is.na(training))
[1] FALSE
# library(ggplot2)
# scatter <- function(var) {</pre>
    ggplot(training, aes_string(var, "class")) +
      geom_jitter(width = 0.05, height = 0.1, size = 0.1,
#
                   colour = rqb(0, 0, 0, alpha = 1 / 3))
# }
# scat_plot <- lapply(names(training)[-99], scatter)</pre>
# library(gridExtra)
# grid.arrange(grobs = scat_plot[1:20], ncol = 4)
# grid.arrange(grobs = scat_plot[21:40], ncol = 4)
# grid.arrange(grobs = scat_plot[41:60], ncol = 4)
# grid.arrange(grobs = scat_plot[61:80], ncol = 4)
# grid.arrange(grobs = scat_plot[81:98], ncol = 4)
sig <- logical(98)</pre>
names(sig) <- names(training)[-99]</pre>
k <- 1
diffs <- logical(98)
for (var in names(training)[-99]) {
  model <- aov(training[[var]] ~ factor(training$class))</pre>
  sig[k] <- summary(model)[[1]][1, 5]</pre>
  diffs[k] <- all(TukeyHSD(model)$`factor(training$class)`[, 4] < 0.05)</pre>
  k \leftarrow k + 1
}
```

```
head(sort(sig[diffs]), 15)
                             Broad_H3K9ac_percentage
Broad_H4K20me1_percentage
                                                                 H4K20me1 width
            2.605382e-232
                                       8.993961e-184
                                                                  3.222855e-181
          H3K79me2_height
                                      H3K79me2_width Broad_H3K4me2_percentage
            4.253756e-176
                                       2.449130e-175
                                                                  4.311054e-174
Broad_H3K79me2_percentage Broad_H3K27ac_percentage
                                                                H4K20me1_height
            1.188924e-168
                                       2.345807e-168
                                                                  1.791736e-164
Broad_H3K4me1_percentage Broad_H3K36me3_percentage
                                                                H3K36me3 width
            1.065370e-163
                                       4.471080e-154
                                                                  2.142472e-147
              pLOF_Zscore Broad_H3K4me3_percentage
                                                                  H3K4me1 width
            5.366468e-145
                                       2.017738e-143
                                                                  5.760242e-141
score <- function (conf mat) {</pre>
  print(sum(diag(conf_mat) * c(1, 20, 20)))
  print(sum(diag(conf_mat) * c(1, 20, 20)) / sum(apply(conf_mat, 2, sum) * c(1, 20, 20)))
} #weighted scoring algorithm
classify <- function(probs) {</pre>
  if (any(probs[2:3] > 0.05)) {
    subset <- probs[2:3]</pre>
    output <- which(subset == max(subset))</pre>
    if (length(output) > 1) {
     output <- sample(1:2, 1)</pre>
      # if OG and TSG both have equal probabilities, pick one randomly
    }
  } else {
    output <- 0
  }
  output
# Selection of Predictors
vars <- training %>% dplyr::select(
 Broad_H3K9ac_percentage, N_LOF, pLOF_Zscore, Broad_H4K2Ome1_percentage,
 H3K79me2_height, N_Splice, LOF_TO_Total_Ratio, VEST_score,
 Missense_Entropy, BioGRID_log_degree, class
vars$class <- factor(vars$class)</pre>
levels(vars$class) <- c("NG", "OG", "TSG")</pre>
cor_mtx = round(cor(vars[, names(vars) != "class"]), 2)
library(reshape2)
Attaching package: 'reshape2'
The following object is masked from 'package:tidyr':
    smiths
#reshape it
melted_cor_mtx <- melt(cor_mtx)</pre>
#draw the heatmap
```

```
cor_heatmap = ggplot(data = melted_cor_mtx, aes(x=Var1, y=Var2, fill=value)) +
    geom_tile()
cor_heatmap = cor_heatmap +
scale_fill_gradient2(
    low = "blue", high = "red", mid = "white", midpoint = 0, limit = c(-1,1),
    space = "Lab", name="Pearson\nCorrelation"
) +
theme_minimal() +
theme(axis.text.x = element_text(angle = 45, vjust = 1, size = 12, hjust = 1))
cor_heatmap
```

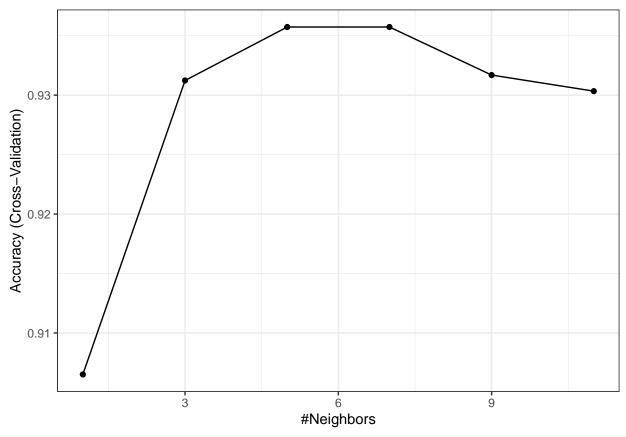


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Techniques

KNN

```
set.seed(nrow(training) +421314)
library(caret)
Loading required package: lattice
Attaching package: 'caret'
The following object is masked from 'package:purrr':
    lift
vars_test <- createDataPartition(vars$class, p = 0.7,</pre>
                                    list = FALSE)
vars_train <- vars[vars_test, ]</pre>
vars_test <- vars[-vars_test, ]</pre>
train_cont <- trainControl(method = "cv", number = 10, classProbs = TRUE,</pre>
                            savePredictions = TRUE
mod <- train(class ~ ., data = vars_train, method = "knn",</pre>
                preProc = c("center", "scale"),
                 trControl = train_cont,
                 tuneGrid = expand.grid(k = seq(from = 1, to = 11, by = 2))
ggplot(mod) + theme_bw()
```



```
for (k in seq(from = 1, to = 11, by = 2)) {
  preds <- predict(mod, newdata = vars_test, type = "prob")
  knn_mod <- table("pred" = unlist(apply(preds, 1, classify)), "obs" = vars_test$class)
  knn_mod
  score(knn_mod)
}</pre>
```

- [1] 2063
- [1] 0.723352
- [1] 2083
- [1] 0.7303647
- [1] 2103
- [1] 0.7373773
- [1] 2103
- [1] 0.7373773
- [1] 2083
- [1] 0.7303647
- [1] 2083
- [1] 0.7303647

QDA

```
train_cont <- trainControl(method = "cv", number = 10, classProbs = TRUE,</pre>
                            savePredictions = TRUE
mod <- train(class ~ ., data = vars_train, method = "qda",</pre>
                preProc = c("center", "scale"),
                trControl = train_cont)
preds <- predict(mod, newdata = vars_test, type = "prob")</pre>
qda_mod <- table("pred" = apply(preds, 1, classify), "obs" = vars_test$class)</pre>
{\tt qda\_mod}
    obs
pred NG OG TSG
  0 768 9 8
   1 16 28
              3
   2 68 13 39
score(qda_mod)
[1] 2108
[1] 0.7391304
```

LDA

```
train_cont <- trainControl(</pre>
method = "cv", number = 10, classProbs = TRUE, savePredictions = TRUE
mod <- train(class ~ ., data = vars_train, method = "lda",</pre>
                preProc = c("center", "scale"),
                trControl = train_cont
preds <- predict(mod, newdata = vars_test, type = "prob")</pre>
lda_mod <- table("pred"=apply(preds, 1, classify), "obs" = vars_test$class)</pre>
lda_mod
    obs
pred NG OG TSG
   0 804
          9
   1 19 30
               7
   2 29 11 37
score(lda_mod)
[1] 2144
[1] 0.7517532
```