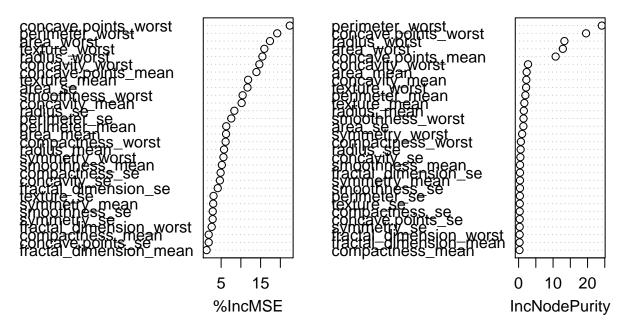
random Forest

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```
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.2.1 --
NA
NA
NA
NA
## -- Conflicts ------ tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
setwd("~/Downloads")
dat_raw = read.csv("data.csv")
dat_raw = dat_raw[, 1:32]
dat_covar = dat_raw %>% select(-id, -diagnosis)
dat_label = cbind(dat_raw$diagnosis) %>% as.data.frame
set.seed(123)
index = sample(1:nrow(dat_raw), size = trunc(.8 * nrow(dat_raw)))
dat_raw$diagnosis = as.integer(factor(dat_raw$diagnosis))-1
Train <- dat_raw[index,-1]</pre>
Test <- dat_raw[-index,-1]</pre>
#implement random forest
library(randomForest)
## randomForest 4.6-12
## Type rfNews() to see new features/changes/bug fixes.
## Attaching package: 'randomForest'
## The following object is masked from 'package:dplyr':
##
##
      combine
## The following object is masked from 'package:ggplot2':
##
##
      {\tt margin}
```

library(caret) ## Loading required package: lattice ## ## Attaching package: 'caret' ## The following object is masked from 'package:purrr': ## ## lift learn_rf <- randomForest(diagnosis~., data=Train, ntree=500, proximity=T, importance=T) ## Warning in randomForest.default(m, y, ...): The response has five or fewer ## unique values. Are you sure you want to do regression? ##variance importance plot varImpPlot(learn_rf)</pre>

learn_rf



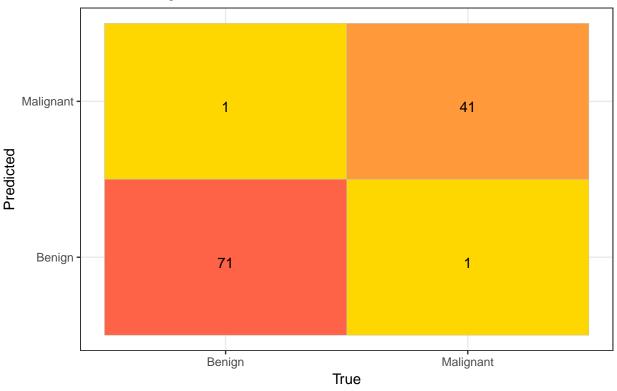
```
#append predicted values onto test set to plot confusion matrix
Test$predicted <- round(predict(learn_rf ,Test),0)

plotConfusionMatrix <- function(testset, sSubtitle) {
   tst <- data.frame(testset$predicted, testset$diagnosis)
   opts <- c("Predicted", "True")</pre>
```

```
names(tst) <- opts</pre>
    cf <- plyr::count(tst)</pre>
    cf[opts][cf[opts]==0] <- "Benign"</pre>
    cf[opts][cf[opts]==1] <- "Malignant"</pre>
    ggplot(data = cf, mapping = aes(x = True, y = Predicted)) +
      labs(title = "Confusion matrix", subtitle = sSubtitle) +
      geom_tile(aes(fill = freq), colour = "grey") +
      geom_text(aes(label = sprintf("%1.0f", freq)), vjust = 1) +
      scale_fill_gradient(low = "gold", high = "tomato") +
      theme_bw() + theme(legend.position = "none")
}
#calculate and plot AUC
library(pROC)
## Type 'citation("pROC")' for a citation.
##
## Attaching package: 'pROC'
## The following objects are masked from 'package:stats':
##
       cov, smooth, var
print(sprintf("Area under curve (AUC) : %.3f",auc(Test$diagnosis, Test$predicted)))
## [1] "Area under curve (AUC) : 0.981"
plotConfusionMatrix(Test, "Prediction using RandomForest with 500 trees")
```

Confusion matrix

Prediction using RandomForest with 500 trees



#check prediction accuracy

Refine covariates according to the result of the varimpplot

```
features_list <- c("perimeter_worst", "area_worst", "concave.points_worst", "radius_worst", "concavity
    "radius_mean", "radius_se", "perimeter_mean", "perimeter_se", "compactness_worst", "smoothness_worst",

#define train and validation set

Train_red = Train[,features_list]

#training
train_red_rf <- randomForest(diagnosis ~.,Train_red,ntree=500,importance=T)

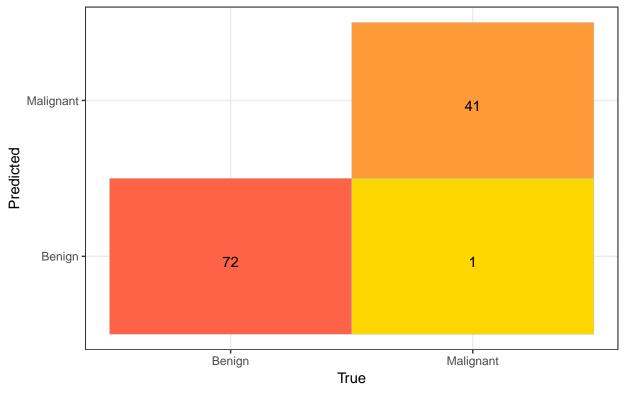
## Warning in randomForest.default(m, y, ...): The response has five or fewer
## unique values. Are you sure you want to do regression?

#get prediction
Test_red$predicted <- round(predict(train_red_rf ,Test_red),0)</pre>
```

plotConfusionMatrix(Test_red, "Prediction using RandomForest with reduced features set")

Confusion matrix

Prediction using RandomForest with reduced features set



```
#increase in true negative
#calculate AUC
print(sprintf("Area under curve (AUC) : %.3f",auc(Test_red$diagnosis, Test_red$predicted)))
```

[1] "Area under curve (AUC) : 0.988"

#reducing the features only increased prediction rate by 0.007