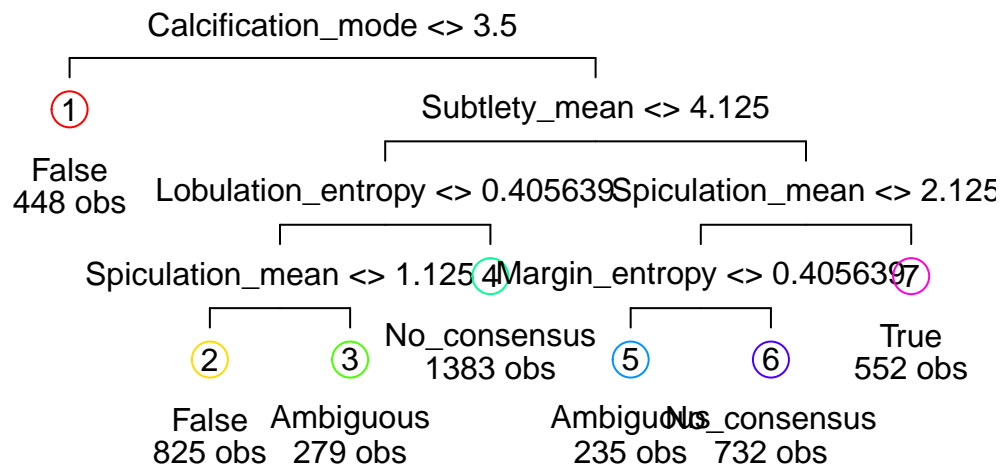


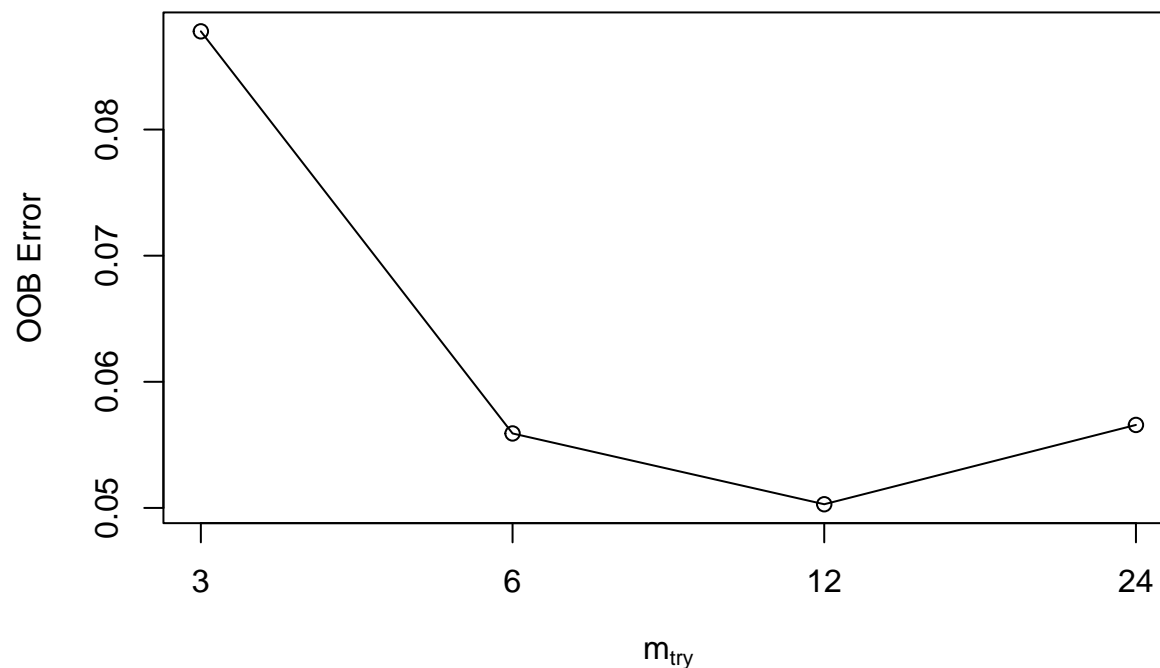
STAT 447 project

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```
## mtry = 6  OOB error = 5.59%
## Searching left ...
## mtry = 3  OOB error = 8.78%
## -0.5702811 0.05
## Searching right ...
## mtry = 12  OOB error = 5.03%
## 0.1004016 0.05
## mtry = 24  OOB error = 5.66%
## -0.125 0.05
```



```
##      mtry  OOBError
## 3.00B    3 0.08778626
## 6.00B    6 0.05590480
## 12.00B   12 0.05029187
## 24.00B   24 0.05657836
```

```
## Warning in confusionMatrix.default(predict(randomForest(Is_cancer ~ ., data =
## t), : Levels are not in the same order for reference and data. Refactoring data
## to match.
```

```
## Confusion Matrix and Statistics
```

```
##
##              Reference
## Prediction    True Ambiguous False No_consensus
## True          166      25    13         76
## Ambiguous      46     288   120        127
## False          37     172   562         72
## No_consensus   91     136   101        303
```

```
## Overall Statistics
```

```
##
##              Accuracy : 0.5649
##              95% CI : (0.5445, 0.5851)
##              No Information Rate : 0.3409
##              P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.401
```

```

##
## McNemar's Test P-Value : 8.37e-06
##
## Statistics by Class:
##
##           Class: True Class: Ambiguous Class: False
## Sensitivity      0.48824      0.4638      0.7060
## Specificity      0.94286      0.8291      0.8174
## Pos Pred Value   0.59286      0.4957      0.6667
## Neg Pred Value   0.91533      0.8101      0.8432
## Prevalence       0.14561      0.2660      0.3409
## Detection Rate   0.07109      0.1233      0.2407
## Detection Prevalence 0.11991      0.2488      0.3610
## Balanced Accuracy 0.71555      0.6464      0.7617
##
##           Class: No_consensus
## Sensitivity      0.5242
## Specificity      0.8133
## Pos Pred Value   0.4802
## Neg Pred Value   0.8386
## Prevalence       0.2475
## Detection Rate   0.1298
## Detection Prevalence 0.2702
## Balanced Accuracy 0.6688

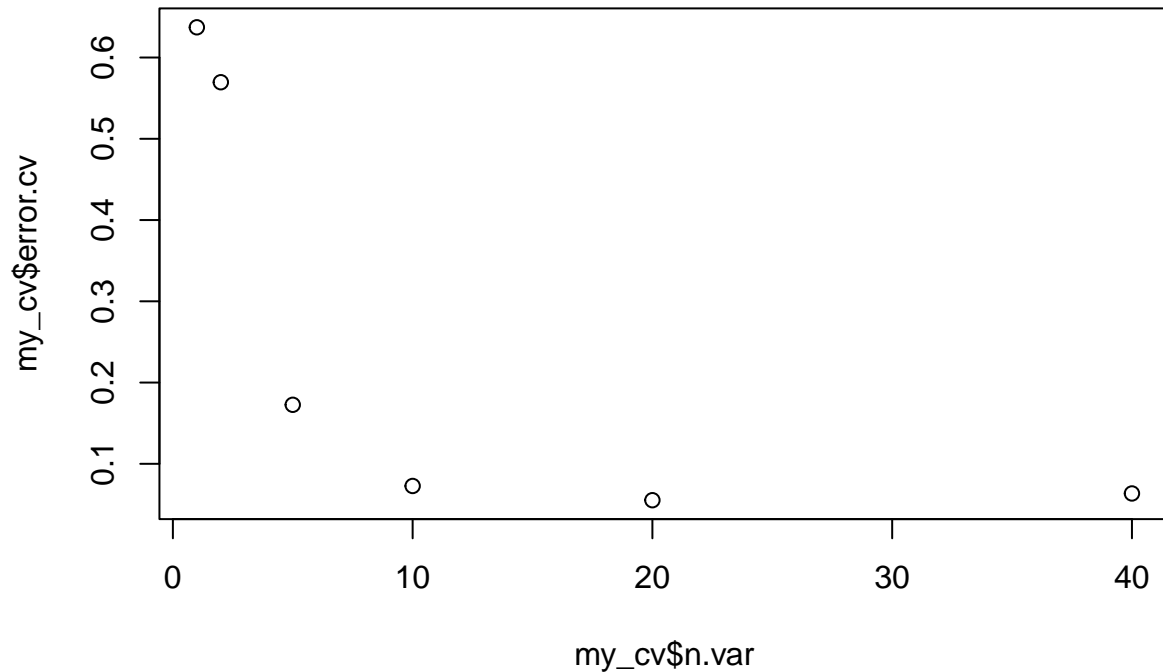
## Warning in confusionMatrix.default(predict(randomForest(Is_cancer ~ ., data =
## t, : Levels are not in the same order for reference and data. Refactoring data
## to match.

## Confusion Matrix and Statistics
##
##           Reference
## Prediction   True Ambiguous False No_consensus
##   True       169      29    16      80
##   Ambiguous   49     286   132     130
##   False       32     171   548      66
##   No_consensus 90     135   100     302
##
## Overall Statistics
##
##           Accuracy : 0.5589
##           95% CI : (0.5385, 0.5792)
##   No Information Rate : 0.3409
##   P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.394
##
## McNemar's Test P-Value : 0.0007548
##
## Statistics by Class:
##
##           Class: True Class: Ambiguous Class: False
## Sensitivity      0.49706      0.4605      0.6884
## Specificity      0.93734      0.8186      0.8252
## Pos Pred Value   0.57483      0.4791      0.6707
## Neg Pred Value   0.91622      0.8072      0.8366

```

```
## Prevalence          0.14561          0.2660          0.3409
## Detection Rate      0.07238          0.1225          0.2347
## Detection Prevalence 0.12591          0.2557          0.3499
## Balanced Accuracy   0.71720          0.6396          0.7568
##
## Class: No_consensus
## Sensitivity          0.5225
## Specificity          0.8150
## Pos Pred Value       0.4817
## Neg Pred Value       0.8384
## Prevalence           0.2475
## Detection Rate        0.1293
## Detection Prevalence  0.2685
## Balanced Accuracy     0.6688
```

```
plot(my_cv$n.var, my_cv$error.cv)
```



```
# Number of training set misclassifications
t_error_ct <- function(rf) {
  cm <- rf$confusion
  sum(cm[1:4,1:4]) - sum(diag(cm))
}

# Number of holdout set misclassifications
h_error_ct <- function(rf) {
  pr <- predict(rf,h) == h$Is_cancer
  length(pr[!pr])
}
```

```

mt <- c(40,20,10,5,2,1)

te <- c()
he <- c()
for (i in mt) {
  rf <- randomForest(Is_cancer~.,data=t,mtry = i)
  tr_error <- t_error_ct(rf)
  ho_error <- h_error_ct(rf)
  te <- c(te,tr_error)
  he <- c(he,ho_error)
}

info <- tibble(mtry = mt,
               training_misclass = te,
               holdout_misclass = he,
               total_misclass = te + he) %>%
  pivot_longer(-mtry)

ggplot(info,aes(x=mtry,y=value,group=name,color=name)) +
  geom_line() +
  geom_point()

```

