

Data Mining Hw2 Draft

Blue Team 16

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Data Reading

```
setwd(dirname(getActiveDocumentContext())$path))

# Reading in data
trainingBin <- read.csv("insurance_t_bin.csv")
training <- read.csv("insurance_t.csv")

validationBin <- read.csv("insurance_v_bin.csv")
validation <- read.csv("insurance_v.csv")

# Fixing Separations and NAs
trainingBin <- trainingBin %>% mutate(across(everything(), ~ as.character(.x))) %>%
  mutate(across(everything(), ~ replace_na(.x,"M"))) %>%
  mutate(across(everything(), ~ as.factor(.x)))

validationBin <- validationBin %>% mutate(across(everything(), ~ as.character(.x))) %>%
  mutate(across(everything(), ~ replace_na(.x,"M"))) %>%
  mutate(across(everything(), ~ as.factor(.x)))
```

Old Logistic Regression Model

```
finalModel <- glm(INS ~ NSF + MTG + INV + ILSBAL_BIN + IRA + DDA + TELLER_BIN + CC + ATMAMT_BIN + CHECKS
```

Decision Tree Models

```
# Making a large tree to prune later. The values I selected are what I came to after playing around with
# Only LR variables
lrTree <- rpart(INS ~ NSF + MTG + INV + ILSBAL + IRA + DDA + TELLER + CC + ATMAMT + CHECKS + MMBAL + CDBAL + DDABAL + SAVBAL,
               control = rpart.control(minsplit = 30, cp = .001, maxdepth = 6))
# All variables
bigTree <- rpart(INS ~ ., data=training, method='class', parms = list(split="gini"),
               control = rpart.control(minsplit = 30, cp = .001, maxdepth = 6))
```

Pruning

Subset Variable Model

```
printcp(lrTree)

##
## Classification tree:
## rpart(formula = INS ~ NSF + MTG + INV + ILSBAL + IRA + DDA +
##       TELLER + CC + ATMAMT + CHECKS + MMBAL + CDBAL + DDABAL +
##       SAVBAL, data = training, method = "class", parms = list(split = "gini"),
##       control = rpart.control(minsplit = 30, cp = 0.001, maxdepth = 6))
##
## Variables actually used in tree construction:
## [1] ATMAMT CDBAL CHECKS DDA DDABAL INV IRA MMBAL MTG SAVBAL
## [11] TELLER
##
## Root node error: 2918/8495 = 0.3435
##
## n= 8495
##
##      CP nsplit rel error  xerror   xstd
## 1 0.1329678     0  1.00000 1.00000 0.014999
## 2 0.0277587     1  0.86703 0.87286 0.014472
## 3 0.0099383     2  0.83927 0.84202 0.014321
## 4 0.0056546     5  0.80946 0.82728 0.014246
## 5 0.0054832    10  0.78033 0.81837 0.014199
## 6 0.0049692    11  0.77485 0.81426 0.014177
## 7 0.0035984    13  0.76491 0.80363 0.014120
## 8 0.0032557    15  0.75771 0.79575 0.014077
## 9 0.0027416    17  0.75120 0.79678 0.014083
## 10 0.0023989   18  0.74846 0.79472 0.014071
## 11 0.0020562   19  0.74606 0.79575 0.014077
## 12 0.0017135   23  0.73783 0.79130 0.014052
## 13 0.0010281   24  0.73612 0.79164 0.014054
## 14 0.0010000   28  0.73201 0.79164 0.014054
```

Only want to include first 6 layers based on oneSE

```
lrTree <- prune(lrTree, cp=0.0049692)
```

Full Variable Model Tree

```
printcp(bigTree)
```

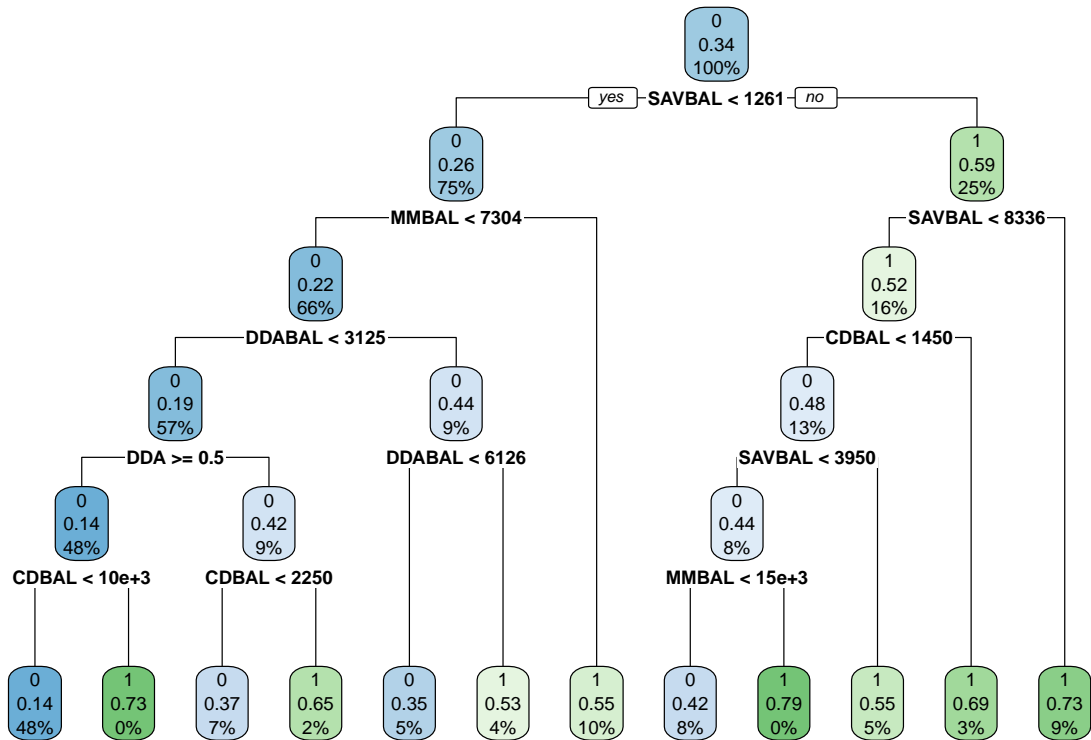
```
##
## Classification tree:
## rpart(formula = INS ~ ., data = training, method = "class", parms = list(split = "gini"),
##       control = rpart.control(minsplit = 30, cp = 0.001, maxdepth = 6))
##
## Variables actually used in tree construction:
## [1] ACCTAGE ATMAMT  BRANCH  CDBAL   CHECKS  CRSCORE DDA      DDABAL  DEP
## [10] MM            SAVBAL
##
## Root node error: 2918/8495 = 0.3435
##
## n= 8495
##
##      CP nsplit rel error  xerror    xstd
## 1  0.1329678      0  1.00000 1.00000 0.014999
## 2  0.0277587      1  0.86703 0.87286 0.014472
## 3  0.0119945      2  0.83927 0.83516 0.014286
## 4  0.0111378      3  0.82728 0.81871 0.014201
## 5  0.0090816      5  0.80500 0.82180 0.014217
## 6  0.0065798      7  0.78684 0.81768 0.014196
## 7  0.0065113     12  0.75394 0.81083 0.014159
## 8  0.0034270     13  0.74743 0.79267 0.014060
## 9  0.0030843     14  0.74400 0.79438 0.014069
## 10 0.0020562     15  0.74092 0.79472 0.014071
## 11 0.0017135     17  0.73681 0.80226 0.014113
## 12 0.0015422     19  0.73338 0.80226 0.014113
## 13 0.0013708     22  0.72824 0.80226 0.014113
## 14 0.0010281     24  0.72550 0.80260 0.014115
## 15 0.0010000     25  0.72447 0.80672 0.014137
```

Only want first 10 layers

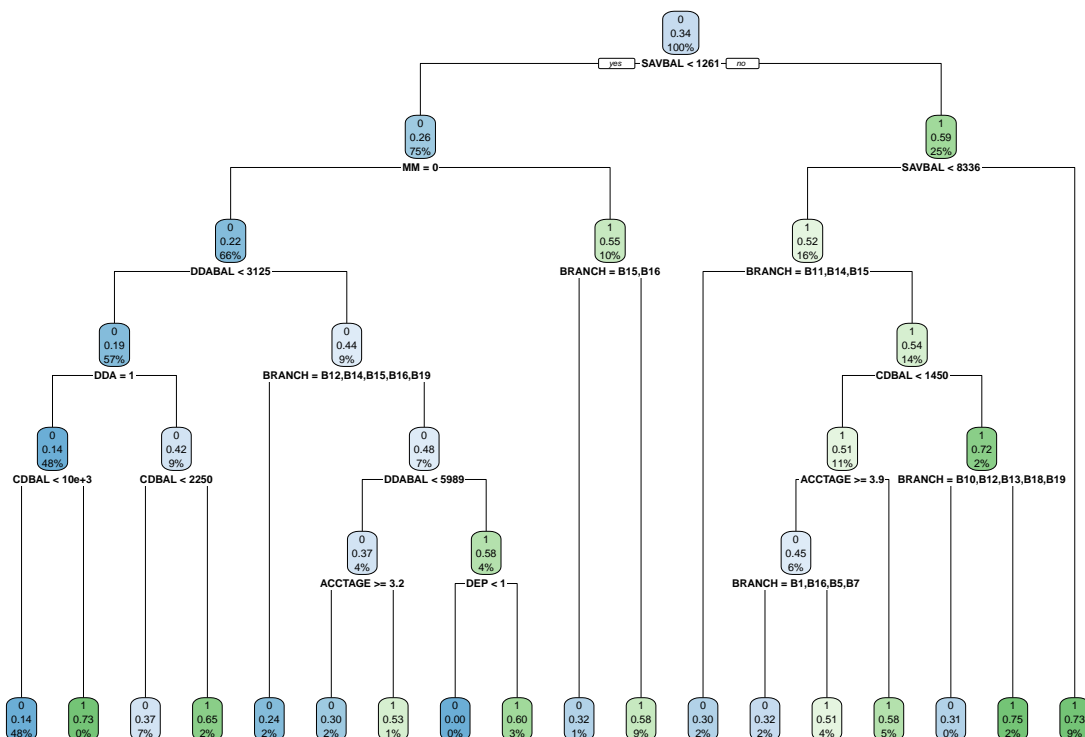
```
bigTree <- prune(bigTree,cp=0.0020562)
```

Visualizing

Subset Variable Model



Full Variable Model



F # Predicting

Predictions and Fitted Values

```
probLRTree <- predict(lrTree,validation,type = "prob")
probBigTree <- predict(bigTree,validation,type = "prob")

predLRTree <- predict(lrTree,validation,type = "class")
predBigTree <- predict(bigTree,validation,type = "class")

fittedLRTree <- predict(lrTree,training,type = "prob")
fittedBigTree <- predict(bigTree,training,type = "prob")
```

Accuracy scores

Subset Model

```
lrAccuracy <- (length(which(predLRTree == validation$INS))) / nrow(validation)

lrAccuracy
```

```
## [1] 0.7123352
```

Full Variable Model

```
bigAccuracy <- (length(which(predBigTree == validation$INS)) / nrow(validation))  
  
bigAccuracy  
  
## [1] 0.7306968
```

Logistic Regression Accuracy

```
# Taken from logistic regression ROC curve  
cutoff <- 0.2970672  
pred <- predict(finalModel, validationBin, type = "response")  
pred <- data.frame(pred = pred) %>% mutate(pred = if_else(pred > cutoff, 1, 0))  
pred <- pred$pred  
  
# Create accuracy vector  
accDF <- data.frame(pred = pred, observed = validation$INS) %>% mutate(accuracy = if_else(pred == observed, 1, 0))  
  
accuracy <- round(mean(accDF$accuracy), 4)  
# Accuracy  
accuracy  
  
## [1] 0.702
```