## CS771A Assignment 1: Decision Trees

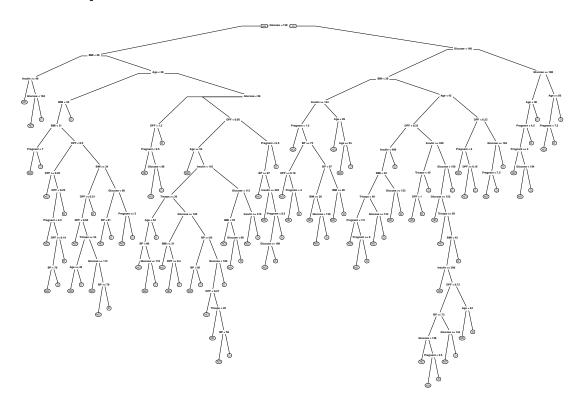
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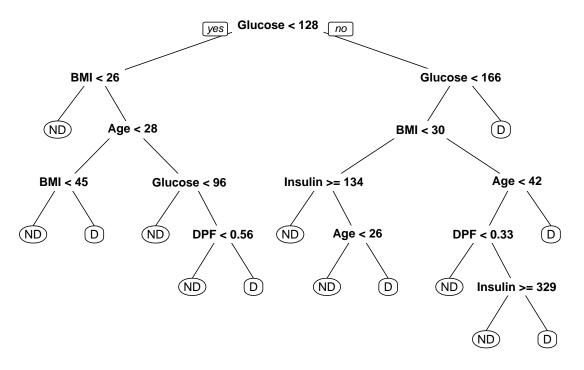
```
library(rpart)
library(rpart.plot)
set.seed(10)
rawData = read.csv(file="data", header=F, sep=",")
originalData = rawData[sample(nrow(rawData)),]
colnames(originalData) = c("PregnantCount", "Glucose", "BP", "Triceps",
                           "Insulin", "BMI", "DPF", "Age", "Class")
N = nrow(originalData)
K = 5
foldWidth = floor(N/K)
# Missing Values can be in Glucose, BP, Triceps, Insulin, BMI
Accuracy = 0
for (i in (1:K))
   data = originalData
    data$Glucose[data$Glucose==0] = NA  # Missing Data # c
   data\$BP[data\$BP==0] = NA
   data$Triceps[data$Triceps==0] = NA
   data$Insulin[data$Insulin==0] = NA
   data$BMI[data$BMI==0] = NA
   start = as.integer((i-1)*foldWidth)+1
   end = as.integer(i*foldWidth)
   if(i==K)
        end = N
   }
   testData = data[c(start:end),]
   learnData = data[c(-start:-end),]
   diabStat = factor(learnData$Class, levels=0:1, labels=c('ND','D'))
    cfit = rpart(
                    diabStat ~ PregnantCount+Glucose+BP+Triceps+Insulin+BMI+DPF+Age,
                    data = learnData,
                    na.action = na.rpart,
          method ='class',
                    parms = list(split = "information"),
                    control = rpart.control(
                                                cp = 0.0, # Threshold complexity parameter
                                                minsplit = 1, # Min no. of obs. for which the routine
                                                minbucket = 1, # Min no. of obs in leaf. Default = min
                                            )
    opt = cfit$cptable[which.min(cfit$cptable[,"xerror"]),"CP"]
                                                                                         # Pruning
   prunedTree = prune(cfit, cp = opt)
   predictedFactor = predict(prunedTree, testData, type="class")
   predictedFrame = as.data.frame.factor(predictedFactor)
   predicted = c(predictedFrame[ ,1]) - 1
```

```
actual = testData$Class
TP = sum(predicted & actual)
TN = nrow(testData) - sum(predicted | actual)
# Accuracy
print((TP+TN)/nrow(testData))
Accuracy = Accuracy + (TP+TN)/nrow(testData)
print("Unpruned Tree")
rpart.plot(cfit)
print("Pruned Tree")
rpart.plot(prunedTree)
}
```

```
## [1] 0.7254902
## [1] "Unpruned Tree"
```

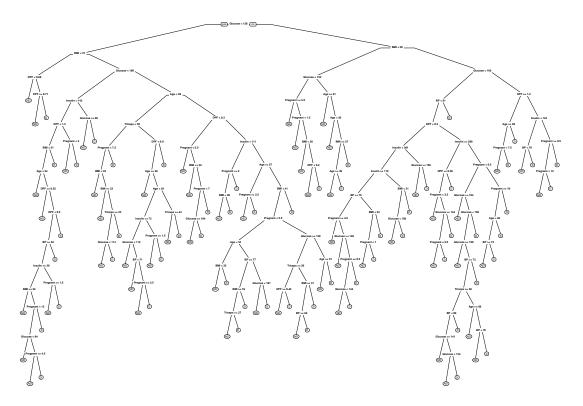


## [1] "Pruned Tree"

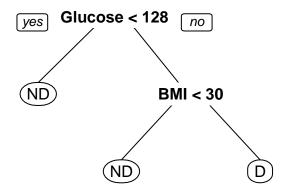


## [1] 0.7777778

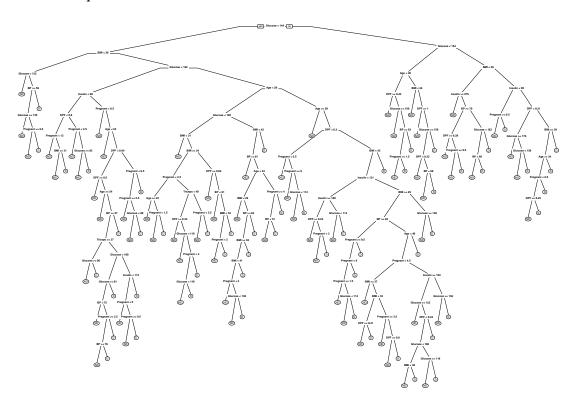
## [1] "Unpruned Tree"



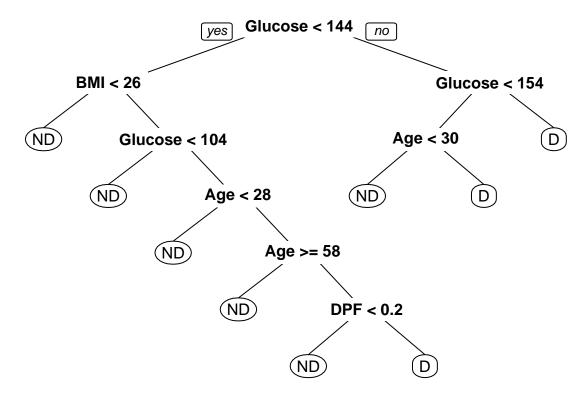
## [1] "Pruned Tree"



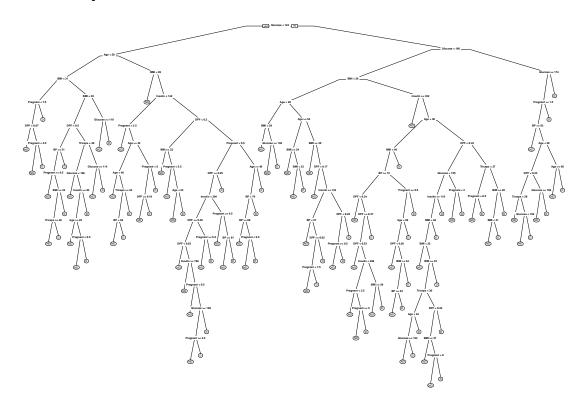
## [1] 0.745098 ## [1] "Unpruned Tree"



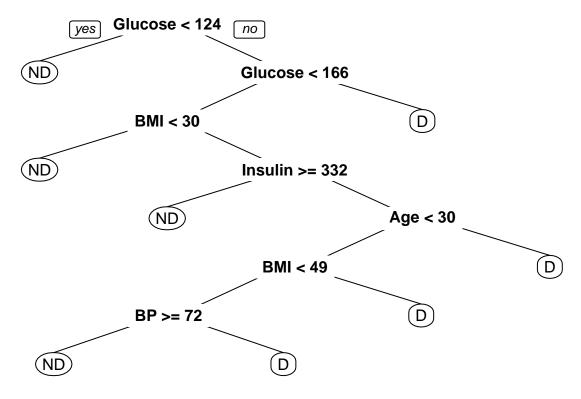
## [1] "Pruned Tree"



- ## [1] 0.7254902
- ## [1] "Unpruned Tree"

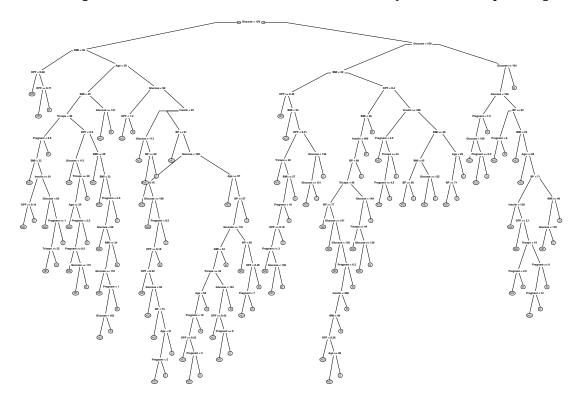


## [1] "Pruned Tree"

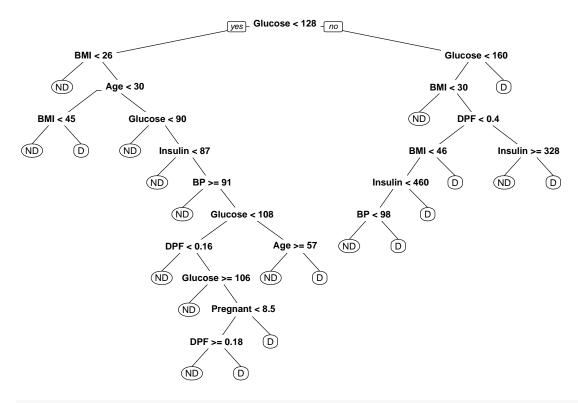


- ## [1] 0.7628205
- ## [1] "Unpruned Tree"

## Warning: labs do not fit even at cex 0.15, there may be some overplotting



## [1] "Pruned Tree"



#Mean Accuracy
print(Accuracy/K)

## [1] 0.7473353