## CS771A Assignment 1: Decision Trees

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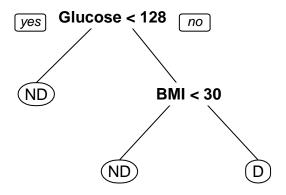
January 19, 2014

```
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)
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$Class, levels=0:1, labels=c('ND','D'))
    #' Any split that does not decrease the overall lack of fit by a factor
    #' of cp is not attempted. By setting appropriate (here, 0.03), we are
    #' controlling the growth of tree by thresholding impurity decrease.
    cfit = rpart(
                    diabStat ~ PregnantCount+Glucose+BP+Triceps+Insulin+BMI+DPF+Age,
                    data = learnData,
                    na.action = na.rpart,
                    method ='class',
                    parms = list(split = "gini"),
                    control = rpart.control(
                                                cp = 0.03,
   predictedFactor = predict(cfit, testData, type="class")
   predictedFrame = as.data.frame.factor(predictedFactor)
   predicted = c(predictedFrame[ ,1]) - 1 # gives 1 for ND, 2 for D
   actual = testData$Class
   TP = sum(predicted & actual)
   TN = nrow(testData) - sum(predicted | actual)
```

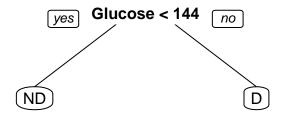
```
# Accuracy
print((TP+TN)/nrow(testData))
Accuracy = Accuracy + (TP+TN)/nrow(testData)
rpart.plot(cfit)
}
```

## [1] 0.751634

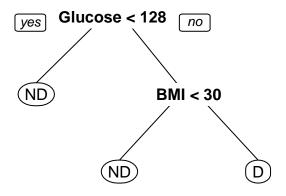
**##** [1] 0.7777778



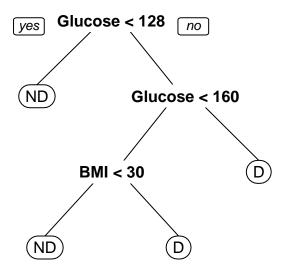
## [1] 0.6993464



## [1] 0.7647059



## [1] 0.8012821



# Mean Accuracy
print(Accuracy/K)

## [1] 0.7589492