# Homework 4

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### Question 1

### Data preparation

```
rm(list=ls())
set.seed(1)
raw.data <-read.csv("Sepsis.csv")</pre>
n <- nrow(raw.data)</pre>
ntest <- round(n*0.3)</pre>
test.id = sample(1:n, ntest)
training.data <- raw.data[-test.id, ]</pre>
testing.data <- raw.data[test.id, ]</pre>
training.active <- subset(training.data, THERAPY==1)</pre>
training.control <- subset(training.data, THERAPY==0)</pre>
training.active.data <- training.active[, c("Health", "PRAPACHE", "AGE", "BLGCS", "ORGANNUM", "BLIL6", "BLLP.
training.active.best <- training.active[, "BEST"]</pre>
training.control.data <- training.control[,c("Health", "PRAPACHE", "AGE", "BLGCS", "ORGANNUM", "BLIL6", "BLL
training.control.best <- training.control[, "BEST"]</pre>
testing.data.data <- testing.data[, c("Health", "PRAPACHE", "AGE", "BLGCS", "ORGANNUM", "BLIL6", "BLLPLAT", "
testing.data.best <- testing.data[, c("BEST")]</pre>
#testing.data
```

#### Demo

```
library(randomForest)
```

```
## randomForest 4.6-14
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
active.forest <- randomForest(Health-., data=training.active.data, mtry=3, nodesize=100)
control.forest <- randomForest(Health-., data=training.control.data, mtry=3, nodesize=100)

predictions<-predict(active.forest, testing.data.data)

result <- sum((testing.data.best - predictions)^2)/length(predictions)

GetPrediction <- function(prediction.active, prediction.control)
{
    diff <- prediction.active - prediction.control
    result <- sapply(diff, function(x){
        if(x>0){return <- 1}
        else {return <- 0}
        })
    return <- result
}

mtrys <- c(1, 3, 6, 9)
nodesizes <- c(10, 100, 500, 1000)
errors<- matrix(rep(0, times=length(mtrys)*length(nodesizes)), nrow=length(mtrys), ncol=length(nodesize)</pre>
```

### Train random forest

```
for(iter in c(1:100))
{
    cat(iter, " ")
    if(iter%,10=0)
    {
        cat("\n")
    }

set.seed(iter)

test.id = sample(1:n, ntest)

training.data <- raw.data[-test.id, ]
    testing.data <- raw.data[test.id, ]

training.active <- subset(training.data, THERAPY==1)
    training.control <- subset(training.data, THERAPY==0)

training.active.data <- training.active[, c("Health", "PRAPACHE", "AGE", "BLGCS", "ORGANNUM", "BLIL6", "BL training.active.best <- training.active[, "BEST"]</pre>
training.control.data <- training.control[,c("Health", "PRAPACHE", "AGE", "BLGCS", "ORGANNUM", "BLIL6", "BL training.control.data <- training.control[,c("Health", "PRAPACHE", "AGE", "BLGCS", "ORGANNUM", "BLIL6", "BLI
```

```
training.control.best <- training.control[, "BEST"]</pre>
 testing.data.data <- testing.data[, c("Health", "PRAPACHE", "AGE", "BLGCS", "ORGANNUM", "BLIL6", "BLLPLAT"
 testing.data.best <- testing.data[, c("BEST")]</pre>
 for(i in 1:length(mtrys))
   for(j in 1:length(nodesizes))
     active.forest <- randomForest(Health~., data=training.active.data, mtry=mtrys[i], nodesize=nodesi
     control.forest <- randomForest(Health~., data=training.control.data, mtry=mtrys[i], nodesize=node</pre>
     predictions.active <- predict(active.forest, testing.data.data)</pre>
     predictions.control <- predict(control.forest, testing.data.data)</pre>
     predictions <- GetPrediction(predictions.active, predictions.control)</pre>
     error <- sum((testing.data.best - predictions)^2)/length(predictions)</pre>
     errors[i,j] = errors[i,j] + error
   }
 }
}
     2 3 4 5 6 7 8
                          9 10
## 11 12 13 14 15 16 17
                              18
                                  19
                                      20
## 21 22 23 24 25
                      26
                          27
                              28
                                  29
                                      30
## 31 32 33 34 35
                      36
                          37
                              38
                                  39
                                      40
## 41 42 43 44 45 46
                         47 48
                                  49
                                      50
## 51 52 53 54 55 56 57
                              58
                                  59
                                      60
## 61 62
          63 64 65 66
                              68
                                  69
                                      70
                          67
## 71 72 73 74 75 76
                          77
                              78
                                  79
                                      80
## 81 82
          83 84
                  85 86
                          87
                              88
                                  89
                                      90
## 91 92 93 94 95 96 97 98 99
                                      100
Check Errors
```

```
errors.average <- errors/100
errors.average
```

```
## [,1] [,2] [,3] [,4]

## [1,] 0.2601418 0.2869504 0.3489362 0.3489362

## [2,] 0.2325532 0.2304255 0.2620567 0.2594326

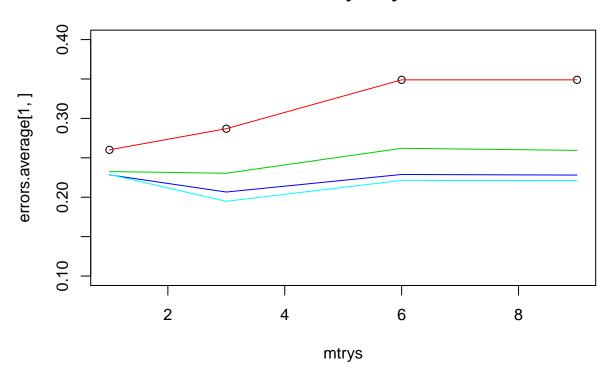
## [3,] 0.2283688 0.2065957 0.2288652 0.2280851

## [4,] 0.2287943 0.1948227 0.2213475 0.2211348
```

### Demo Errors

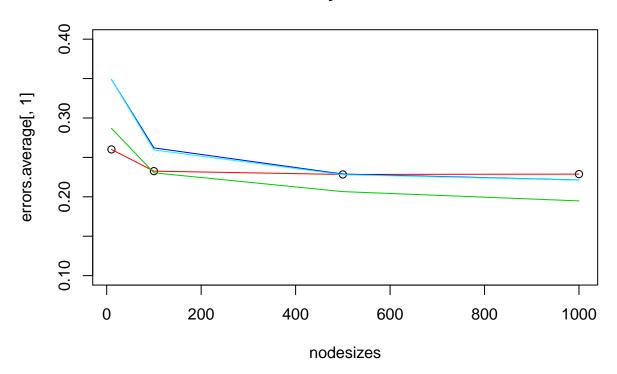
```
plot(mtrys, errors.average[1,],ylim=c(0.1,0.4), main="Error by mtry")
for(i in 1:length(mtrys))
{
    lines(mtrys, errors.average[i,], col=i+1)
}
```

# **Error by mtry**



```
plot(nodesizes, errors.average[,1],ylim=c(0.1,0.4), main="Error by nodesize")
for(i in 1:length(nodesizes))
{
    lines(nodesizes, errors.average[,i], col=i+1)
}
```

# **Error by nodesize**



Based on the plots above, we can discover that the best prediction model is with parameters:

mtry: 6

nodesizes: 100

the general trend of different parameters are:

- as mtry increase, prediction error decrease.
- the best nodesize is about 100.

# Question 2

### Get predictions

```
rm(list=ls())
set.seed(1)

raw.data <- read.csv("Sepsis.csv")

data.cols <- c("Health", "PRAPACHE", "AGE", "BLGCS", "ORGANNUM", "BLIL6", "BLLPLAT", "BLLBILI", "BLLCREAT", "TIM"
raw.data.data <- raw.data[, data.cols]</pre>
```

```
raw.active <- subset(raw.data, THERAPY==1)
raw.active.data <- raw.active[, data.cols]

raw.control <- subset(raw.data, THERAPY==0)
raw.control.data <- raw.control[, data.cols]

GetPrediction <- function(prediction.active, prediction.control)
{
    diff <- prediction.active - prediction.control
    result <- sapply(diff, function(x){
        if(x>0){return <- 1}
        else {return <- 0}
        })
        return <- result
}

active.forest <- randomForest(Health~., data=raw.active.data, mtry=6, nodesize=100)
control.forest <- randomForest(Health~., data=raw.control.data, mtry=6, nodesize=100)

predictions.active <- predict(active.forest, raw.data.data)
predictions.control <- predict(control.forest, raw.data.data)
predictions <- GetPrediction(predictions.active, predictions.control)</pre>
```

#### Train decision tree

```
library(rpart)
library(rpart.plot)

data.cols <- c("PRAPACHE","AGE","BLGCS","ORGANNUM","BLIL6","BLLPLAT","BLLBILI","BLLCREAT","TIMFIRST","B

rpart.fit <- rpart(as.factor(predictions)~., data = data.frame(raw.data.data[,data.cols],predictions))

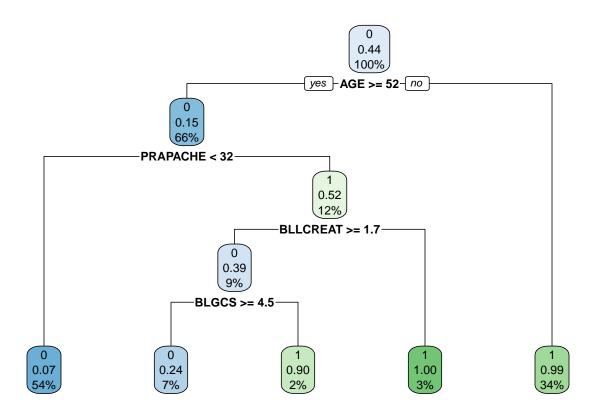
rpart.plot(rpart.fit)

## Warning: Bad 'data' field in model 'call' (expected a data.frame or a matrix).

## To silence this warning:

## Call rpart.plot with roundint=FALSE,

## or rebuild the rpart model with model=TRUE.</pre>
```

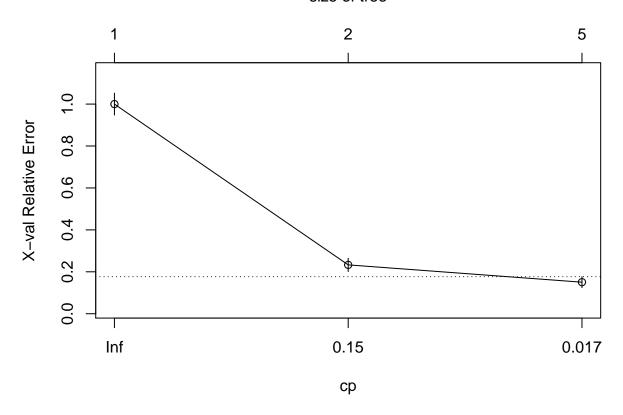


# rpart.fit\$cptable

```
## CP nsplit rel error xerror xstd
## 1 0.76699029 0 1.0000000 1.0000000 0.05221790
## 2 0.02912621 1 0.2330097 0.2330097 0.03186843
## 3 0.01000000 4 0.1359223 0.1504854 0.02612143
```

# plotcp(rpart.fit)

# size of tree



# prune(rpart.fit, cp=0.03)

```
## n= 470
##
## node), split, n, loss, yval, (yprob)
## * denotes terminal node
##
## 1) root 470 206 0 (0.5617021 0.4382979)
## 2) AGE>=51.7395 310 47 0 (0.8483871 0.1516129) *
## 3) AGE< 51.7395 160 1 1 (0.0062500 0.9937500) *</pre>
```