STAT 652 - Assignment 2

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Question 3

3. Use boosting to model the relationship between Ozone and all ONLY THE THREE ORIGINAL VARIABLES. Tune on an initial grid of shrink = 0.001, 0.005, 0.025, 0.125 and d = 2, 4, 6, and select trees optimally using twice the number suggested by OOB error. Use two reps of 5-fold CV (refer to the lecture note and R code to understand how to do this).

```
# Helper Functions
get.MSPE = function(Y, Y.hat){
   return(mean((Y - Y.hat)^2))
}

# Create k CV folds for a AQset of size n
get.folds = function(n, K) {
   ### Get the appropriate number of fold labels
   n.fold = ceiling(n / K) # Number of observations per fold (rounded up)
   fold.ids.raw = rep(1:K, times = n.fold) # Generate extra labels
   fold.ids = fold.ids.raw[1:n] # Keep only the correct number of labels
   # Shuffle the fold labels
   folds.rand = fold.ids[sample.int(n)]
   return(folds.rand)
}
```

3. (a) Report the mean root-MSPE for each combination of shrink and depth? Answer: Root-MSPEs for each combination of shrink and depth are: $2|0.001\ 2|0.005\ 2|0.025\ 2|0.125\ 4|0.001\ 4|0.005\ 4|0.025\ 4|0.125\ 6|0.001\ 6|0.005\ 6|0.025\ 6|0.125\ 30.30477\ 23.41988\ 18.68153\ 18.84860\ 30.03603\ 22.63194\ 18.76087\ 18.93884\ 30.03713\ 22.61434\ 18.59552\ 18.94697$ and mean of root-MSPEs for each combination of shrink and depth is 22.65137.

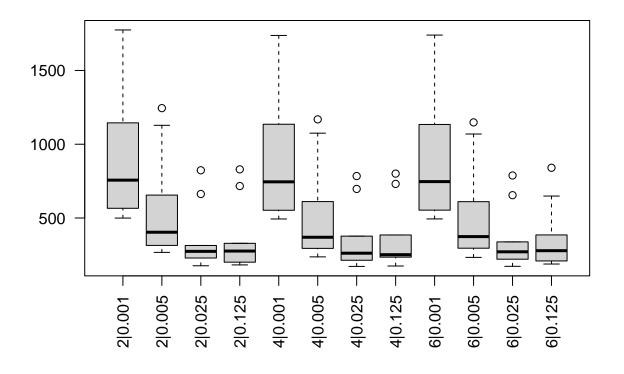
```
library(gbm)
set.seed(301471961)
AQ = airquality[1:4]

# Removing Null values
AQ = na.omit(AQ)

# Setting parameter values
trees = 200
shrink = c(0.001,0.005,0.025,0.125)
depth = c(2,4,6)
```

```
V=5
R=2
n = nrow(AQ)
# Create the folds and save in a matrix
folds = matrix(NA, nrow = n, ncol = R)
for(r in 1:R){
  folds[,r]=floor((sample.int(n)-1)*V/n) + 1
}
NS = length(shrink)
ND = length(depth)
gb.cv = matrix(NA, nrow=ND*NS, ncol=V*R)
opt.tree = matrix(NA, nrow=ND*NS, ncol=V*R)
qq = 1
for(r in 1:R){
  for(v in 1:V){
    AQ.train = AQ[folds[,r]!=v,]
    AQ.test = AQ[folds[,r]==v,]
    counter=1
    for(d in depth){
      for(s in shrink){
        AQ.gbm <- gbm(data = AQ.train, Ozone~., distribution = "gaussian", n.trees = trees,
                      interaction.depth = d, shrinkage = s,bag.fraction=0.8)
        treenum = min(trees, 2*gbm.perf(AQ.gbm, method = "00B", plot.it = FALSE))
        opt.tree[counter,qq] = treenum
        preds = predict(AQ.gbm, newdata = AQ.test, n.trees=treenum)
        gb.cv[counter,qq] = mean((preds - AQ.test$0zone)^2)
        counter=counter+1
    }
    qq = qq+1
  }
parms = expand.grid(shrink,depth)
row.names(gb.cv) = paste(parms[,2], parms[,1], sep="|")
row.names(opt.tree) = paste(parms[,2], parms[,1], sep="|")
(mean.tree = apply(opt.tree, 1, mean))
```

```
## 2|0.001 2|0.005 2|0.025 2|0.125 4|0.001 4|0.005 4|0.025 4|0.125 6|0.001 6|0.005
                            32.6
                                   200.0
    200.0
          200.0
                   172.8
                                           200.0
                                                   149.8
                                                           29.6
                                                                  200.0
                                                                          200.0
## 6|0.025 6|0.125
##
    158.8
             29.8
(sq.mean.cv = sqrt(apply(gb.cv, 1, mean)))
## 2|0.001 2|0.005 2|0.025 2|0.125 4|0.001 4|0.005 4|0.025 4|0.125
## 30.30477 23.41988 18.68153 18.84860 30.03603 22.63194 18.76087 18.93884
## 6|0.001 6|0.005 6|0.025 6|0.125
## 30.03713 22.61434 18.59552 18.94697
sq.mean.cv
## 2|0.001 2|0.005 2|0.025 2|0.125 4|0.001 4|0.005 4|0.025 4|0.125
## 30.30477 23.41988 18.68153 18.84860 30.03603 22.63194 18.76087 18.93884
## 6|0.001 6|0.005 6|0.025 6|0.125
## 30.03713 22.61434 18.59552 18.94697
```



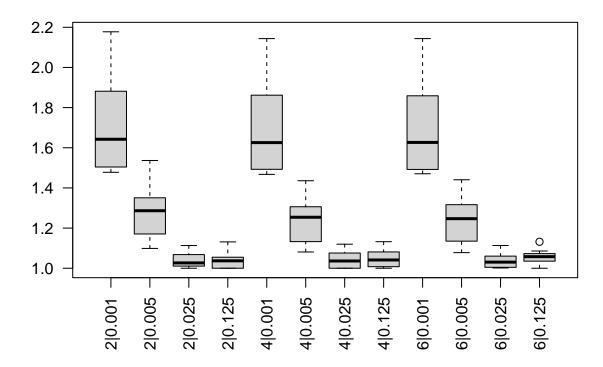
mean(sq.mean.cv)

boxplot(gb.cv, use.cols=FALSE, las=2)

[1] 22.65137

3. (b) Show relative root-MSPE boxplots for each combination of shrink and depth?

Relative root-MSPE boxplots for each combination of shrink and der



3. (c) What combination of shrink and depth do you prefer?

Answer: Based on the relative MSPE plot, shrinkage = 0.025 and depth = 6 looks like the best fitted model. And relatively, Temp is the most important among Wind and Solar.R.

```
set.seed(301471961)

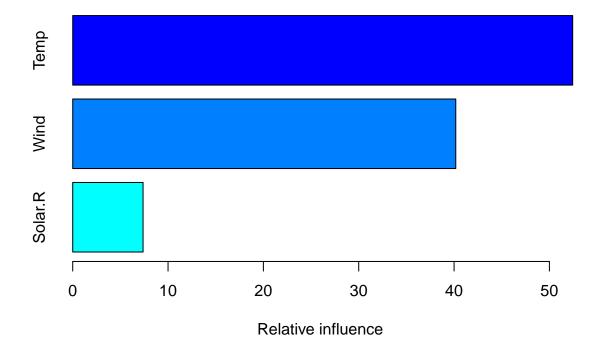
fit.gbm.best = gbm(Ozone ~ ., data = AQ, distribution = "gaussian", n.trees = 125, interaction.depth = n.trees.best = gbm.perf(fit.gbm.best, plot.it = F) * 2

pred.gbm.best = predict(fit.gbm.best, AQ.test, n.trees.best)

MSPE.gbm = get.MSPE(AQ.test$Ozone, pred.gbm.best)

MSPE.gbm
```

[1] 86.23719



var rel.inf
Temp Temp 52.442337
Wind Wind 40.179455
Solar.R Solar.R 7.378207