BUAN6356_Homework4_UdayakumarA

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```
library(ISLR)
## Warning: package 'ISLR' was built under R version 4.0.5
library(ggplot2)
library(leaps)
## Warning: package 'leaps' was built under R version 4.0.4
library(rpart)
## Warning: package 'rpart' was built under R version 4.0.4
library(caret)
## Warning: package 'caret' was built under R version 4.0.4
## Loading required package: lattice
## Warning: package 'lattice' was built under R version 4.0.4
library(rpart.plot)
## Warning: package 'rpart.plot' was built under R version 4.0.4
library(tree)
## Warning: package 'tree' was built under R version 4.0.5
library(gbm)
## Warning: package 'gbm' was built under R version 4.0.5
## Loaded gbm 2.1.8
```

```
library(randomForest)
## Warning: package 'randomForest' was built under R version 4.0.5
## randomForest 4.6-14
## Type rfNews() to see new features/changes/bug fixes.
## Attaching package: 'randomForest'
## The following object is masked from 'package:ggplot2':
##
##
      margin
library(tinytex)
tinytex::tlmgr_install("pdfcrop")
## tlmgr install pdfcrop
## tlmgr update --self
## A new version of TeX Live has been released. If you need to install or update any LaTeX packages, yo
## tlmgr install pdfcrop
Sys.setenv(R_GSCMD="C:/Program Files/gs/gs9.53.3/bin/gswin32c.exe")
#Question 1
str(Hitters)
## 'data.frame':
                   322 obs. of 20 variables:
## $ AtBat : int 293 315 479 496 321 594 185 298 323 401 ...
## $ Hits
              : int 66 81 130 141 87 169 37 73 81 92 ...
## $ HmRun : int 1 7 18 20 10 4 1 0 6 17 ...
             : int 30 24 66 65 39 74 23 24 26 49 ...
## $ Runs
## $ RBI
             : int 29 38 72 78 42 51 8 24 32 66 ...
## $ Walks : int 14 39 76 37 30 35 21 7 8 65 ...
## $ Years : int 1 14 3 11 2 11 2 3 2 13 ...
## $ CAtBat : int 293 3449 1624 5628 396 4408 214 509 341 5206 ...
## $ CHits
              : int 66 835 457 1575 101 1133 42 108 86 1332 ...
## $ CHmRun : int 1 69 63 225 12 19 1 0 6 253 ...
## $ CRuns : int 30 321 224 828 48 501 30 41 32 784 ...
## $ CRBI
             : int 29 414 266 838 46 336 9 37 34 890 ...
## $ CWalks : int 14 375 263 354 33 194 24 12 8 866 ...
## $ League : Factor w/ 2 levels "A", "N": 1 2 1 2 2 1 2 1 2 1 ...
## $ Division : Factor w/ 2 levels "E", "W": 1 2 2 1 1 2 1 2 2 1 ...
## $ PutOuts : int 446 632 880 200 805 282 76 121 143 0 ...
## $ Assists : int 33 43 82 11 40 421 127 283 290 0 ...
## $ Errors : int 20 10 14 3 4 25 7 9 19 0 ...
## $ Salary : num NA 475 480 500 91.5 750 70 100 75 1100 ...
```

\$ NewLeague: Factor w/ 2 levels "A", "N": 1 2 1 2 2 1 1 1 2 1 ...

is.na(Hitters\$Salary)

```
[1] TRUE FALSE FALSE
   [13] FALSE FALSE FALSE TRUE FALSE TRUE FALSE FALSE FALSE TRUE FALSE
   [25] FALSE FALSE FALSE FALSE FALSE TRUE FALSE TRUE FALSE FALSE FALSE
  [37] TRUE FALSE TRUE TRUE FALSE TRUE FALSE TRUE FALSE FALSE FALSE
  [49] TRUE FALSE FALSE TRUE FALSE FALSE FALSE FALSE TRUE FALSE FALSE
   [61] FALSE FALSE FALSE TRUE FALSE TRUE FALSE TRUE FALSE TRUE
##
   [73] FALSE FALSE FALSE FALSE TRUE FALSE FALSE TRUE FALSE FALSE
  [85] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE TRUE FALSE
## [97] FALSE TRUE FALSE FALSE TRUE FALSE TRUE TRUE TRUE TRUE TRUE FALSE
## [109] FALSE FALSE FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE FALSE
## [121] FALSE FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE FALSE
## [133] FALSE FALSE FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE FALSE
## [145] TRUE FALSE FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE
## [157] FALSE TRUE TRUE FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE
## [169] FALSE TRUE FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE
## [181] FALSE FALSE
## [193] FALSE FALSE FALSE FALSE TRUE FALSE TRUE FALSE FALSE TRUE
## [205] FALSE FALSE FALSE TRUE FALSE TRUE FALSE FALSE FALSE FALSE FALSE
## [217] FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE TRUE FALSE FALSE
## [229] TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## [241] FALSE FALSE FALSE FALSE FALSE TRUE FALSE FALSE TRUE FALSE
## [253] FALSE TRUE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## [265] FALSE FALSE FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE FALSE
## [277] FALSE FALSE
## [289] FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE TRUE FALSE
## [301] FALSE FALSE TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## [313] FALSE FALSE FALSE TRUE FALSE FALSE FALSE FALSE
```

```
data <-Hitters
data.nona<- data[complete.cases(data[,19]),]
str(data.nona)</pre>
```

```
## 'data.frame':
                   263 obs. of 20 variables:
   $ AtBat
            : int 315 479 496 321 594 185 298 323 401 574 ...
  $ Hits
             : int 81 130 141 87 169 37 73 81 92 159 ...
##
              : int 7 18 20 10 4 1 0 6 17 21 ...
   $ HmRun
   $ Runs
             : int 24 66 65 39 74 23 24 26 49 107 ...
##
  $ RBI
             : int 38 72 78 42 51 8 24 32 66 75 ...
   $ Walks
           : int 39 76 37 30 35 21 7 8 65 59 ...
##
              : int 14 3 11 2 11 2 3 2 13 10 ...
   $ Years
             : int 3449 1624 5628 396 4408 214 509 341 5206 4631 ...
##
   $ CAtBat
##
   $ CHits
             : int 835 457 1575 101 1133 42 108 86 1332 1300 ...
  $ CHmRun : int 69 63 225 12 19 1 0 6 253 90 ...
##
   $ CRuns
              : int 321 224 828 48 501 30 41 32 784 702 ...
##
              : int 414 266 838 46 336 9 37 34 890 504 ...
   $ CRBI
   $ CWalks
             : int 375 263 354 33 194 24 12 8 866 488 ...
             : Factor w/ 2 levels "A", "N": 2 1 2 2 1 2 1 2 1 1 ...
   $ League
   $ Division : Factor w/ 2 levels "E","W": 2 2 1 1 2 1 2 2 1 1 ...
   $ PutOuts : int 632 880 200 805 282 76 121 143 0 238 ...
   $ Assists : int 43 82 11 40 421 127 283 290 0 445 ...
   $ Errors : int 10 14 3 4 25 7 9 19 0 22 ...
```

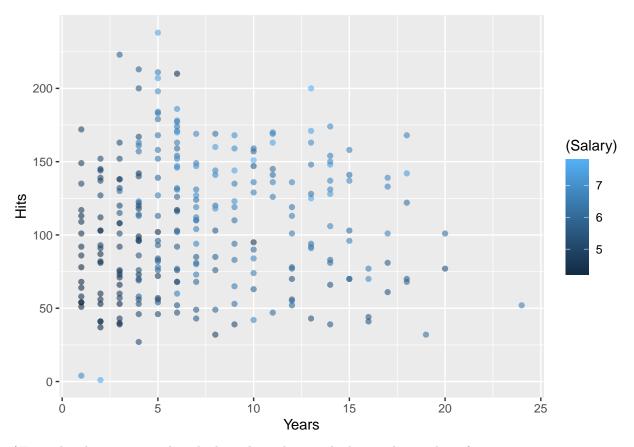
```
## $ Salary : num 475 480 500 91.5 750 ...
## $ NewLeague: Factor w/ 2 levels "A","N": 2 1 2 2 1 1 1 2 1 1 ...
```

1)59 Observations were removed by removing the observations with no salary record

```
#Question:2
data.nona$Salary <- log(data.nona$Salary)</pre>
str(data.nona)
## 'data.frame':
                   263 obs. of 20 variables:
## $ AtBat : int 315 479 496 321 594 185 298 323 401 574 ...
              : int 81 130 141 87 169 37 73 81 92 159 ...
## $ Hits
## $ HmRun : int 7 18 20 10 4 1 0 6 17 21 ...
             : int 24 66 65 39 74 23 24 26 49 107 ...
## $ Runs
## $ RBI
             : int 38 72 78 42 51 8 24 32 66 75 ...
             : int 39 76 37 30 35 21 7 8 65 59 ...
## $ Walks
## $ Years : int 14 3 11 2 11 2 3 2 13 10 ...
## $ CAtBat : int 3449 1624 5628 396 4408 214 509 341 5206 4631 ...
             : int 835 457 1575 101 1133 42 108 86 1332 1300 ...
## $ CHits
## $ CHmRun : int 69 63 225 12 19 1 0 6 253 90 ...
## $ CRuns : int 321 224 828 48 501 30 41 32 784 702 ...
## $ CRBI
            : int 414 266 838 46 336 9 37 34 890 504 ...
## $ CWalks : int 375 263 354 33 194 24 12 8 866 488 ...
   $ League : Factor w/ 2 levels "A", "N": 2 1 2 2 1 2 1 2 1 1 ...
##
##    $ Division : Factor w/ 2 levels "E", "W": 2 2 1 1 2 1 2 2 1 1 ...
## $ PutOuts : int 632 880 200 805 282 76 121 143 0 238 ...
## $ Assists : int 43 82 11 40 421 127 283 290 0 445 ...
## $ Errors : int 10 14 3 4 25 7 9 19 0 22 ...
## $ Salary : num 6.16 6.17 6.21 4.52 6.62 ...
## $ NewLeague: Factor w/ 2 levels "A", "N": 2 1 2 2 1 1 1 2 1 1 ...
```

2)Logarithmic transformations are carried out to normalize a highly skewed data variable.

```
#Question 3
scatter_plot <- ggplot(data.nona,aes(y=Hits,x=Years,color =(Salary)))+geom_point(alpha =0.6)
scatter_plot</pre>
```



3)From the plot we notice that the log salaries become higher as the number of years increase.

```
#Question 4
hitters.lm <- lm(Salary~.,data=data.nona)
summary(hitters.lm)
##
## Call:
## lm(formula = Salary ~ ., data = data.nona)
```

```
Min
##
                   1Q
                        Median
                                      3Q
                                              Max
                       0.09424
##
   -2.22870 -0.45350
                                0.40474
                                          2.77223
##
## Coefficients:
##
                  Estimate Std. Error t value Pr(>|t|)
                4.618e+00
                            1.765e-01
                                        26.171
                                                < 2e-16 ***
##
   (Intercept)
                            1.232e-03
## AtBat
                -2.984e-03
                                        -2.421
                                                0.01620 *
## Hits
                 1.308e-02
                            4.622e-03
                                                0.00503 **
                                         2.831
## HmRun
                 1.179e-02
                            1.205e-02
                                         0.978
                                                0.32889
                                                0.80670
                -1.419e-03
                            5.794e-03
                                        -0.245
## Runs
## RBI
                -1.675e-03
                            5.056e-03
                                        -0.331
                                                0.74063
## Walks
                 1.096e-02
                            3.554e-03
                                         3.082
                                                0.00229 **
## Years
                 5.696e-02
                            2.413e-02
                                         2.361
                                                0.01902 *
## CAtBat
                 1.283e-04
                            2.629e-04
                                                0.62596
```

1.311e-03

3.144e-03

-4.414e-04

-7.809e-05

##

Residuals:

CHits

CHmRun

0.73670

0.98020

0.488

-0.337

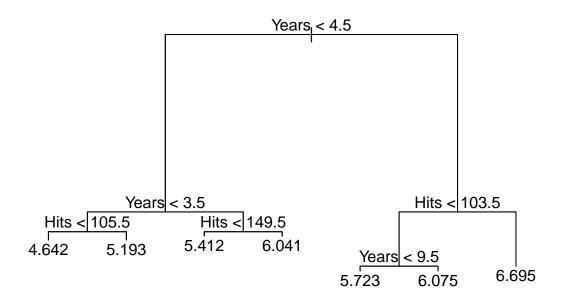
-0.025

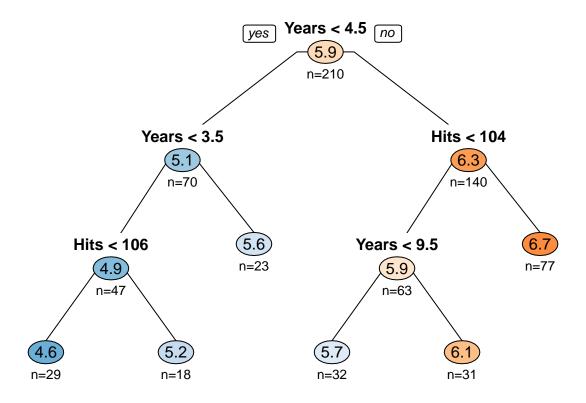
```
## CRuns
                1.513e-03 1.459e-03
                                        1.037 0.30072
## CRBI
                1.312e-04
                           1.346e-03
                                        0.097
                                               0.92246
                                       -2.298
## CWalks
               -1.466e-03
                           6.377e-04
                                               0.02239 *
                2.825e-01
                           1.541e-01
                                        1.833
                                               0.06797
## LeagueN
## DivisionW
               -1.656e-01
                           7.847e-02
                                       -2.111
                                               0.03580
## PutOuts
                3.389e-04
                           1.505e-04
                                        2.251
                                               0.02526 *
## Assists
                6.214e-04
                           4.300e-04
                                        1.445
                                               0.14970
## Errors
               -1.197e-02
                           8.537e-03
                                       -1.402
                                               0.16225
## NewLeagueN
              -1.742e-01
                           1.536e-01
                                       -1.134
                                               0.25788
##
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.6135 on 243 degrees of freedom
## Multiple R-squared: 0.5586, Adjusted R-squared: 0.524
## F-statistic: 16.18 on 19 and 243 DF, p-value: < 2.2e-16
search <- regsubsets(Salary~.,data=data.nona,nbest=1,</pre>
                     nvmax=dim(data.nona)[2],method="exhaustive")
sum <- summary(search)</pre>
sum$which
```

```
##
      (Intercept) AtBat Hits HmRun Runs
                                              RBI Walks Years CAtBat CHits CHmRun
## 1
             TRUE FALSE FALSE FALSE FALSE FALSE FALSE
                                                                FALSE FALSE
                                                                             FALSE
## 2
             TRUE FALSE
                          TRUE FALSE FALSE FALSE FALSE
                                                                 TRUE FALSE
                                                                             FALSE
## 3
             TRUE FALSE
                         TRUE FALSE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                FALSE FALSE
                                                                             FALSE
                  TRUE
## 4
             TRUE
                          TRUE FALSE FALSE FALSE
                                                   TRUE FALSE
                                                                 TRUE FALSE
                                                                             FALSE
## 5
             TRUE FALSE
                          TRUE FALSE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                FALSE
                                                                       TRUE
                                                                             FALSE
## 6
             TRUE
                   TRUE
                          TRUE FALSE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                FALSE
                                                                       TRUE
                                                                             FALSE
## 7
                   TRUE
                          TRUE FALSE FALSE FALSE
                                                   TRUE
                                                         TRUE
             TRUE
                                                                FALSE FALSE
                                                                             FALSE
## 8
             TRUE
                   TRUE
                          TRUE FALSE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                FALSE FALSE
## 9
             TRUE
                   TRUE
                          TRUE FALSE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                FALSE FALSE
                                                                             FALSE
## 10
             TRUE
                   TRUE
                          TRUE FALSE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                FALSE FALSE
                                                                             FALSE
## 11
             TRUE
                   TRUE
                         TRUE
                               TRUE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                FALSE FALSE
                                                                             FALSE
## 12
             TRUE
                   TRUE
                          TRUE
                                TRUE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                FALSE FALSE
                                                                             FALSE
                   TRUE
                          TRUE
                                TRUE FALSE FALSE
                                                                FALSE FALSE
## 13
             TRUE
                                                   TRUE
                                                         TRUE
                                                                             FALSE
## 14
             TRUE
                   TRUE
                          TRUE
                                TRUE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                 TRUE FALSE
                                                                             FALSE
                   TRUE
## 15
             TRUE
                          TRUE
                                TRUE FALSE FALSE
                                                   TRUE
                                                         TRUE
                                                                 TRUE
                                                                       TRUE
                                                                             FALSE
                   TRUE
##
  16
             TRUE
                          TRUE
                                TRUE FALSE
                                             TRUE
                                                   TRUE
                                                         TRUE
                                                                 TRUE
                                                                       TRUE
                                                                             FALSE
             TRUE
                   TRUE
                          TRUE
                                TRUE
                                      TRUE
                                             TRUE
                                                   TRUE
                                                         TRUE
                                                                 TRUE
                                                                       TRUE
##
  17
                                                                             FALSE
##
  18
             TRUE
                   TRUE
                          TRUE
                                TRUE
                                      TRUE
                                             TRUE
                                                   TRUE
                                                         TRUE
                                                                 TRUE
                                                                       TRUE
                                                                             FALSE
##
  19
             TRUE
                   TRUE
                          TRUE
                               TRUE
                                      TRUE
                                             TRUE
                                                   TRUE
                                                         TRUE
                                                                 TRUE
                                                                       TRUE
                                                                              TRUE
##
            CRBI CWalks LeagueN DivisionW PutOuts Assists Errors NewLeagueN
      CRuns
       TRUE FALSE
## 1
                   FALSE
                            FALSE
                                      FALSE
                                               FALSE
                                                       FALSE
                                                              FALSE
                                                                          FALSE
## 2
      FALSE FALSE
                   FALSE
                            FALSE
                                      FALSE
                                               FALSE
                                                       FALSE
                                                              FALSE
                                                                          FALSE
##
  3
      FALSE FALSE
                   FALSE
                            FALSE
                                      FALSE
                                               FALSE
                                                       FALSE
                                                              FALSE
                                                                          FALSE
## 4
                                                              FALSE
                                                                          FALSE
      FALSE FALSE
                   FALSE
                            FALSE
                                      FALSE
                                               FALSE
                                                       FALSE
## 5
      FALSE FALSE
                   FALSE
                            FALSE
                                       TRUE
                                               FALSE
                                                       FALSE
                                                               FALSE
                                                                          FALSE
## 6
      FALSE FALSE
                   FALSE
                            FALSE
                                       TRUE
                                                       FALSE
                                                              FALSE
                                                                          FALSE
                                               FALSE
## 7
       TRUE FALSE
                    TRUE
                            FALSE
                                      FALSE
                                                TRUE
                                                       FALSE
                                                              FALSE
                                                                          FALSE
## 8
       TRUE FALSE
                    TRUE
                            FALSE
                                       TRUE
                                                TRUE
                                                       FALSE
                                                              FALSE
                                                                          FALSE
## 9
       TRUE FALSE
                    TRUE
                             TRUE
                                        TRUE
                                                TRUE
                                                       FALSE
                                                               FALSE
                                                                          FALSE
## 10
       TRUE FALSE
                    TRUE
                             TRUE
                                       TRUE
                                                TRUE
                                                       FALSE
                                                              FALSE
                                                                           TRUE
                                                       FALSE
                                                                           TRUE
## 11
       TRUE FALSE
                    TRUE
                             TRUE
                                       TRUE
                                                TRUE
                                                               FALSE
                                                TRUE
## 12
      TRUE FALSE
                    TRUE
                             TRUE
                                       TRUE
                                                        TRUE
                                                                TRUE
                                                                          FALSE
```

```
## 13 TRUE FALSE
                    TRUE
                            TRUE
                                       TRUE
                                               TRUE
                                                       TRUE
                                                              TRUE
                                                                          TRUE
## 14 TRUE FALSE
                    TRUE
                            TRUE
                                       TRUE
                                               TRUE
                                                       TRUE
                                                              TRUE
                                                                         TRUE
## 15 TRUE FALSE
                    TRUE
                            TRUE
                                      TRUE
                                               TRUE
                                                       TRUE
                                                              TRUE
                                                                         TRUE
## 16 TRUE FALSE
                    TRUE
                            TRUE
                                      TRUE
                                               TRUE
                                                       TRUE
                                                              TRUE
                                                                         TRUE
## 17
      TRUE FALSE
                    TRUE
                            TRUE
                                      TRUE
                                               TRUE
                                                       TRUE
                                                              TRUE
                                                                         TRUE
## 18 TRUE TRUE
                    TRUE
                            TRUE
                                               TRUE
                                                       TRUE
                                                              TRUE
                                                                         TRUE
                                      TRUE
## 19 TRUE TRUE
                    TRUE
                            TRUE
                                               TRUE
                                                       TRUE
                                                                         TRUE
                                      TRUE
                                                              TRUE
sum$rsq
  [1] 0.3857520 0.4822942 0.4986075 0.5090077 0.5190638 0.5270507 0.5355590
   [8] 0.5436891 0.5473898 0.5501579 0.5524819 0.5552470 0.5577193 0.5579177
## [15] 0.5582361 0.5583376 0.5584807 0.5585572 0.5585583
sum$adjr2
## [1] 0.3833985 0.4783118 0.4927999 0.5013954 0.5097071 0.5159660 0.5228097
## [8] 0.5293171 0.5312890 0.5323071 0.5328696 0.5338989 0.5346284 0.5329615
## [15] 0.5314083 0.5296116 0.5278447 0.5259917 0.5240423
sum$bic
  [1] -117.0304 -156.4291 -159.2777 -159.2182 -159.0885 -157.9207 -157.1229
## [8] -156.1954 -152.7649 -148.8061 -144.5962 -140.6541 -136.5480 -131.0939
## [15] -125.7112 -120.1995 -114.7125 -109.1859 -103.6145
4) The 3rd model gives us the lowest BIC hence that is considered to be the best subset. The predictor
variables included in the best model are Hits, Walks and years.
#Question 5
set.seed(42)
train.index <- sample(c(1:263),210)
train.df <- data.nona[train.index,]</pre>
valid.df <- data.nona[-train.index,]</pre>
#Question 6
#using tree package
tree.hitters <- tree(Salary~Hits+Years,data.nona,subset = train.index)</pre>
summary(tree.hitters)
##
## Regression tree:
## tree(formula = Salary ~ Hits + Years, data = data.nona, subset = train.index)
## Number of terminal nodes: 7
## Residual mean deviance: 0.2436 = 49.45 / 203
## Distribution of residuals:
       Min. 1st Qu.
                       Median
                                  Mean 3rd Qu.
                                                     Max.
## -2.19500 -0.29810 -0.03641 0.00000 0.22790 2.18200
```

```
plot(tree.hitters)
text(tree.hitters,pretty = 0)
```

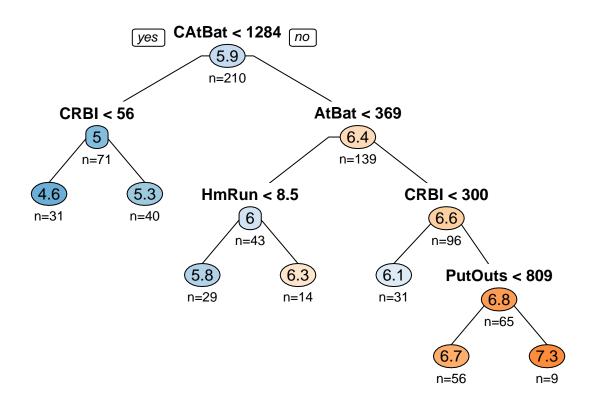




rpart.rules(reg_tree, cover = TRUE)

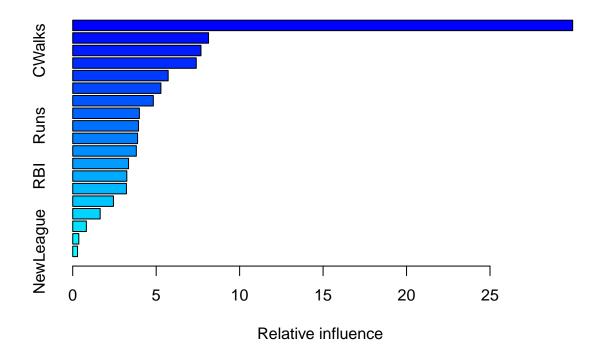
```
##
    Salary
                                                   cover
##
       4.6 when Years < 4
                                  & Hits < 106
                                                     14%
##
       5.2 \text{ when Years} < 4
                                  & Hits >= 106
                                                      9%
       5.6 when Years is 4 to 5
                                                     11%
       5.7 when Years is 5 to 10 & Hits < 104
                                                     15%
##
##
       6.1 when Years >=
                               10 & Hits < 104
                                                     15%
##
       6.7 when Years >=
                                5 & Hits >= 104
                                                     37%
```

6)When the player has more than or equal to 4.5 years of experience and hits more than or equal to 104 he gets high salary

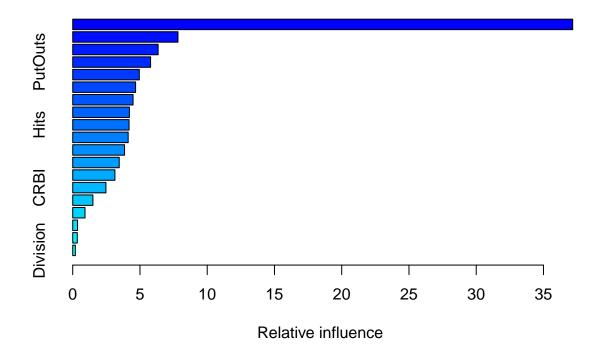


rpart.rules(reg_tree_all, cover = TRUE)

```
##
    Salary
                                                                                               cover
##
       4.6 when CAtBat < 1284 & CRBI <
                                           56
                                                                                                 15%
##
       5.3 when CAtBat < 1284 & CRBI >=
                                                                                                 19%
##
       5.8 \text{ when CAtBat} >= 1284
                                              & AtBat < 369 & HmRun < 9
                                                                                                 14%
##
       6.1 when CAtBat >= 1284 & CRBI < 300 & AtBat >= 369
                                                                                                 15%
##
       6.3 when CAtBat >= 1284
                                              & AtBat < 369 & HmRun >= 9
                                                                                                 7%
##
       6.7 when CAtBat >= 1284 & CRBI >= 300 & AtBat >= 369
                                                                           & PutOuts < 809
                                                                                                 27%
##
       7.3 when CAtBat >= 1284 & CRBI >= 300 & AtBat >= 369
                                                                           & PutOuts >= 809
                                                                                                  4%
boost.hitters1 <- gbm(Salary~.,data = train.df,distribution = "gaussian",
                      shrinkage = 0.2,n.trees = 1000,interaction.depth = 4)
summary(boost.hitters1)
```



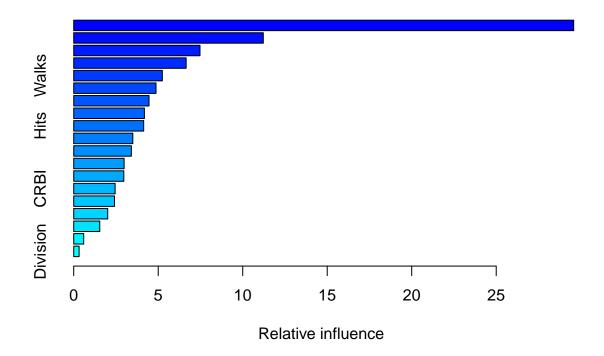
```
rel.inf
                    var
                 CAtBat 29.9536299
## CAtBat
## CRBI
                   CRBI
                         8.1380122
## CWalks
                 CWalks
                         7.6828786
## CHmRun
                 CHmRun
                         7.4019172
## AtBat
                  AtBat
                         5.7189766
## PutOuts
                         5.2836204
               PutOuts
## CRuns
                  CRuns
                         4.8299937
## HmRun
                  HmRun
                         3.9941825
## Runs
                  Runs
                         3.9418922
## Walks
                  Walks
                         3.8863946
                         3.8109649
## Years
                  Years
                Assists
                         3.3449673
## Assists
## RBI
                    RBI
                         3.2407968
## Errors
                Errors
                         3.2176918
## Hits
                  Hits
                         2.4405844
## CHits
                  CHits
                         1.6446300
## League
                League
                         0.8171153
## Division
              Division
                         0.3647234
## NewLeague NewLeague
                         0.2870281
boost.hitters2 <- gbm(Salary~.,data = train.df,distribution = "gaussian",</pre>
                       shrinkage = 0.4,n.trees = 1000,interaction.depth = 4)
summary(boost.hitters2)
```



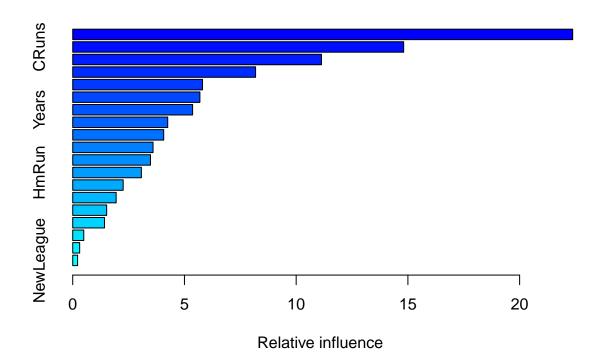
```
CAtBat 37.1717184
## CAtBat
## CHmRun
                 {\tt CHmRun}
                        7.8251839
## CRuns
                  CRuns
                         6.3515200
## PutOuts
                PutOuts
                         5.7890263
## HmRun
                  HmRun
                         4.9519234
## Walks
                         4.6738234
                  Walks
## AtBat
                  AtBat
                         4.4869908
                         4.2131627
## Assists
                Assists
## Hits
                   Hits
                         4.1890760
## Years
                  Years
                         4.1225387
## CWalks
                 CWalks
                         3.8541090
## RBI
                    RBI
                         3.4578867
## Errors
                 Errors
                         3.1360647
## CRBI
                   CRBI
                         2.4697327
## Runs
                   Runs
                         1.5006629
## CHits
                  CHits
                         0.9142512
## NewLeague NewLeague
                         0.3581985
## League
                 League
                         0.3386043
## Division
              Division 0.1955265
boost.hitters3 <- gbm(Salary~.,data = train.df,distribution = "gaussian",</pre>
                       shrinkage = 0.6,n.trees = 1000,interaction.depth = 4)
summary(boost.hitters3)
```

rel.inf

var



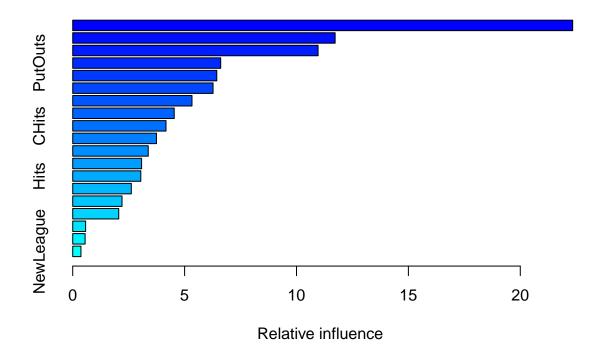
```
rel.inf
                    var
## CRuns
                  CRuns 29.5840844
## CAtBat
                 CAtBat 11.2183324
## AtBat
                  AtBat
                        7.4711804
## PutOuts
               PutOuts
                         6.6536370
## Walks
                  Walks
                         5.2436526
## Errors
                         4.8691959
                 Errors
## Assists
                Assists
                         4.4561914
## RBI
                    RBI
                         4.1829045
## Hits
                   Hits
                         4.1401195
## CHmRun
                 CHmRun
                         3.4965370
## CWalks
                 CWalks
                         3.4114881
## Runs
                         2.9831977
                   Runs
## HmRun
                  {\tt HmRun}
                         2.9639143
## CRBI
                   CRBI
                         2.4498169
## Years
                  Years
                         2.4129578
## CHits
                  CHits
                         2.0116112
## League
                 League
                         1.5407362
## NewLeague NewLeague
                         0.5908351
## Division
              Division 0.3196076
boost.hitters4 <- gbm(Salary~.,data = train.df,distribution = "gaussian",</pre>
                       shrinkage = 0.01,n.trees = 1000,interaction.depth = 4)
summary(boost.hitters4)
```



```
CAtBat 22.3739580
## CAtBat
## CRuns
                  CRuns 14.8105227
## CRBI
                   CRBI 11.1319227
## CWalks
                 CWalks
                         8.1834583
## CHits
                  CHits
                         5.8079736
                         5.6910572
## PutOuts
               PutOuts
## Years
                  Years
                         5.3653720
## CHmRun
                 \tt CHmRun
                         4.2515434
## AtBat
                  AtBat
                         4.0719784
## Walks
                  Walks
                         3.5942114
## Hits
                         3.4796736
                   Hits
## HmRun
                  {\tt HmRun}
                         3.0730736
## Errors
                 Errors
                         2.2579230
## RBI
                    RBI
                         1.9438171
## Assists
                Assists
                         1.5182732
## Runs
                   Runs
                         1.4215855
## League
                 League
                         0.4947809
## Division
              Division
                         0.3106599
## NewLeague NewLeague
                         0.2182155
boost.hitters5 <- gbm(Salary~.,data = train.df,distribution = "gaussian",
                       shrinkage = 0.02,n.trees = 1000,interaction.depth = 4)
summary(boost.hitters5)
```

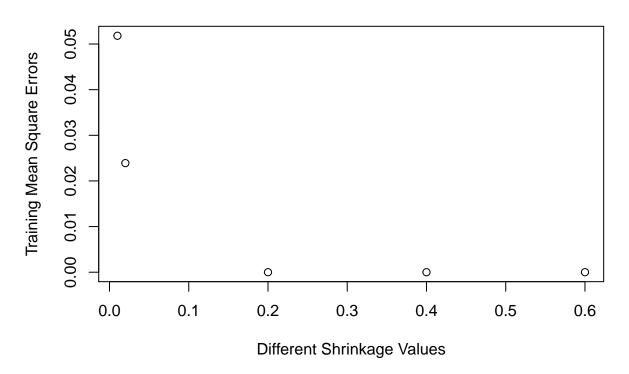
rel.inf

var



```
##
                           rel.inf
                    var
                 CAtBat 22.3415257
## CAtBat
## CRuns
                  CRuns 11.7261489
## CRBI
                   CRBI 10.9647167
## PutOuts
               PutOuts
                         6.6121951
## CWalks
                 CWalks
                         6.4370365
                         6.2708049
## CHmRun
                 CHmRun
## Years
                 Years
                         5.3295279
## AtBat
                  AtBat
                         4.5360799
## CHits
                  CHits
                         4.1686028
## Walks
                  Walks
                         3.7420610
## HmRun
                  HmRun
                         3.3753418
## RBI
                    RBI
                         3.0776240
## Hits
                  Hits
                         3.0428474
## Errors
                Errors
                         2.6181498
## Assists
                Assists
                         2.2027160
## Runs
                   Runs
                         2.0564324
## League
                League
                         0.5769805
## Division
              Division
                         0.5533210
## NewLeague NewLeague
                         0.3678877
MSE_train <- c(boost.hitters1$train.error[1000],boost.hitters2$train.error[1000]
               ,boost.hitters3\$train.error[1000],boost.hitters4\$train.error[1000]
               ,boost.hitters5\$train.error[1000])
MSE_train
```

Train MSE and the shrinkage values



```
#Question 8
Shrinkage_values <- c(0.2,0.4,0.6,0.01,0.02)
hitter.test <- data.nona[-train.index,"Salary"]
yhat.boost1 <- predict(boost.hitters1,newdata = valid.df,n.trees = 1000)
a <- mean((yhat.boost1-hitter.test)^2)
a

## [1] 0.3829805

yhat.boost2 <- predict(boost.hitters2,newdata = valid.df,n.trees = 1000)
b <- mean((yhat.boost2-hitter.test)^2)
b</pre>
```

[1] 0.4186236

```
yhat.boost3 <- predict(boost.hitters3,newdata = valid.df,n.trees = 1000)
c <- mean((yhat.boost3-hitter.test)^2)
c

## [1] 0.4649825

yhat.boost4<- predict(boost.hitters4,newdata = valid.df,n.trees = 1000)
d <- mean((yhat.boost1-hitter.test)^2)
d

## [1] 0.3829805

yhat.boost5 <- predict(boost.hitters5,newdata = valid.df,n.trees = 1000)
e <- mean((yhat.boost1-hitter.test)^2)
e

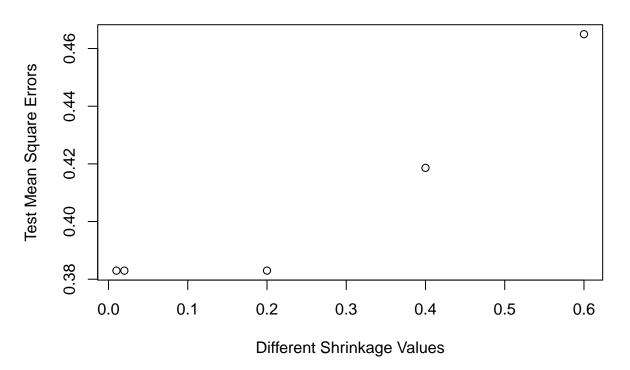
## [1] 0.3829805

MSE_test <- c(a,b,c,d,e)
MSE_test

## [1] 0.3829805 0.4186236 0.4649825 0.3829805 0.3829805

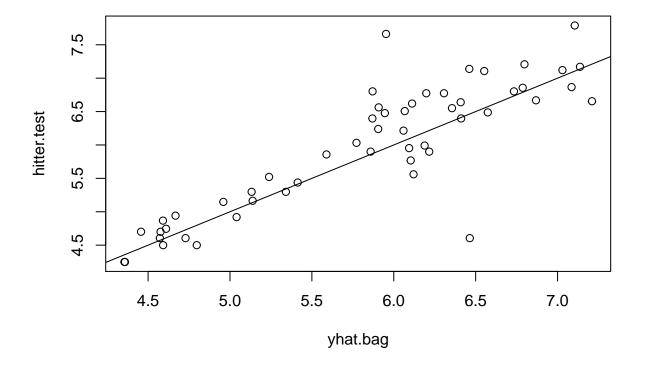
plot(Shrinkage_values,MSE_test,xlab = "Different Shrinkage Values",ylab = "Test Mean Square Errors",main = "Test MSE and shrinkage values")</pre>
```

Test MSE and shrinkage values



9) By altering te shrinkage parameters to different values we see that the most important predictors are CAtBat and CRuns

```
#Question 10
bag.hitters <- randomForest(Salary~., data=train.df,</pre>
                            mtry = 19, importance = TRUE)
bag.hitters
##
## Call:
    randomForest(formula = Salary ~ ., data = train.df, mtry = 19,
                                                                           importance = TRUE)
##
##
                  Type of random forest: regression
                         Number of trees: 500
##
## No. of variables tried at each split: 19
##
             Mean of squared residuals: 0.2033149
##
                        % Var explained: 73.15
##
yhat.bag <- predict(bag.hitters, newdata=valid.df)</pre>
plot(yhat.bag, hitter.test)
abline(0,1)
```



```
MSE_test <- mean((yhat.bag-hitter.test)^2)
MSE_test</pre>
```

[1] 0.2369779

10) The MSE_test is 0.24

R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see http://rmarkdown.rstudio.com.

When you click the \mathbf{Knit} button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

Note that the echo = FALSE parameter was added to the code chunk to prevent printing of the R code that generated the plot.