Exercise 9.5.R

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```
require(MASS) # to load the stepAIC function
## Loading required package: MASS
require(MPV) # to load the data
## Loading required package: MPV
##
## Attaching package: 'MPV'
## The following object is masked from 'package:MASS':
##
##
      cement
## The following object is masked from 'package:datasets':
##
##
      stackloss
# Excersice 9.5 from MPV
data(table.b3)
table.b3[22:26,] # Can you see the missing values?
              x1 x2 x3 x4 x5 x6 x7
                                          8x
                                               x9 x10 x11
         V
## 22 21.47 360.0 180 290 8.4 2.45 2 3 214.2 76.3 4250
## 23 16.59 400.0 185 NA 7.6 3.08 4 3 196.0 73.0 3850
## 24 31.90 96.9 75 83 9.0 4.30 2 5 165.2 61.8 2275
## 25 29.40 140.0 86 NA 8.0 2.92 2 4 176.4 65.4 2150
## 26 13.27 460.0 223 366 8.0 3.00 4 3 228.0 79.8 5430
datis <- table.b3[-c(23,25),]
# The full model -----
full.model \leftarrow lm(y \sim ., data = datis)
summary(full.model)
##
## Call:
## lm(formula = y ~ ., data = datis)
## Residuals:
               1Q Median
                               3Q
## -5.3441 -1.6711 -0.4486 1.4906 5.2508
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 17.339838 30.355375
                                   0.571 0.5749
## x1
             -0.075588
                        0.056347 -1.341
                                            0.1964
              -0.069163 0.087791 -0.788 0.4411
## x2
                          0.088113 1.306
## x3
              0.115117
                                            0.2078
                          3.101464 0.482 0.6357
              1.494737
## x4
```

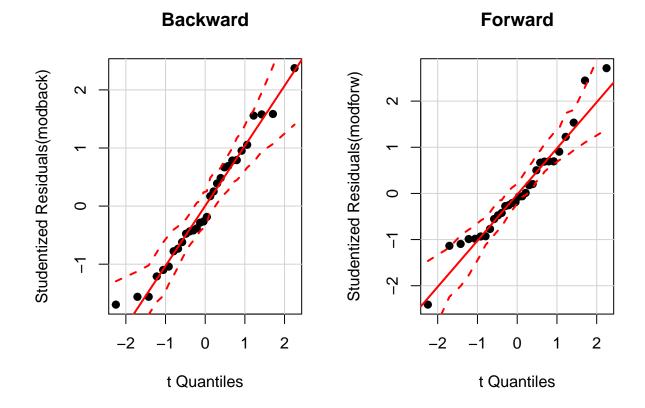
```
## x5
               5.843495
                          3.148438
                                    1.856
                                            0.0799 .
## x6
                          1.288967
                                   0.246
                                            0.8082
               0.317583
## x7
              -3.205390
                          3.109185 -1.031
                                            0.3162
               0.180811
## x8
                          0.130301
                                    1.388
                                            0.1822
                                  -1.230
## x9
              -0.397945
                          0.323456
                                            0.2344
              -0.005115
                          0.005896 -0.868
## x10
                                            0.3971
               0.638483
                                   0.211
## x11
                          3.021680
                                            0.8350
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 3.227 on 18 degrees of freedom
## Multiple R-squared: 0.8355, Adjusted R-squared: 0.7349
## F-statistic: 8.31 on 11 and 18 DF, p-value: 5.231e-05
# logLik and AIC for full.model
length(coef(full.model))
                                   # Number of betas
## [1] 12
logLik(full.model)
                                   # logLik with 13 df
## 'log Lik.' -70.04893 (df=13)
-2 * logLik(full.model) + 2 * 13
                                  # AIC manually
## 'log Lik.' 166.0979 (df=13)
AIC(full.model, k=2)
                                   # AIC automatically
## [1] 166.0979
AIC(full.model, k=log(30))
                                       # BIC automatically
## [1] 184.3134
# backward selection -----
modback <- stepAIC(full.model, trace=TRUE, direction="backward")</pre>
## Start: AIC=78.96
## y \sim x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11
##
##
         Df Sum of Sq
                         RSS
## - x11
                0.465 187.87 77.036
          1
## - x6
          1
                0.632 188.03 77.063
                2.418 189.82 77.346
## - x4
         1
## - x2
         1
                6.462 193.86 77.979
## - x10 1
               7.836 195.24 78.190
## - x7
               11.065 198.47 78.683
          1
## <none>
                      187.40 78.962
## - x9
               15.758 203.16 79.384
          1
## - x3
               17.770 205.17 79.679
          1
               18.736 206.14 79.820
## - x1
          1
## - x8
              20.047 207.45 80.011
        1
## - x5
               35.864 223.26 82.215
##
## Step: AIC=77.04
## y \sim x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10
##
```

```
Df Sum of Sq
                      RSS AIC
## - x6
        1 0.536 188.40 75.121
               2.363 190.23 75.411
## - x4
## - x2
                6.642 194.51 76.078
          1
## - x10
          1
               7.985 195.85 76.285
                     187.87 77.036
## <none>
## - x7
             14.124 201.99 77.211
          1
## - x9
             16.914 204.78 77.622
          1
## - x3
         1
             17.815 205.68 77.754
## - x1
         1
            18.280 206.15 77.822
## - x8 1 20.301 208.17 78.114
             36.370 224.24 80.345
## - x5
        1
##
## Step: AIC=75.12
## y \sim x1 + x2 + x3 + x4 + x5 + x7 + x8 + x9 + x10
##
##
         Df Sum of Sq
                      RSS
                               AIC
## - x4
             3.451 191.85 73.666
         1
## - x2
               6.932 195.33 74.205
          1
## - x10
          1
               9.351 197.75 74.574
## <none>
                     188.40 75.121
## - x7
             14.473 202.87 75.342
## - x3
             17.802 206.20 75.830
          1
## - x9
          1
             18.146 206.55 75.880
## - x1
         1 18.780 207.18 75.972
## - x8
        1 21.244 209.65 76.326
## - x5
          1
             39.332 227.73 78.809
##
## Step: AIC=73.67
## y \sim x1 + x2 + x3 + x5 + x7 + x8 + x9 + x10
##
##
         Df Sum of Sq
                      RSS
                               AIC
          1 10.780 202.63 73.306
## - x2
## - x7
               11.113 202.97 73.355
          1
## <none>
                     191.85 73.666
## - x10 1
             14.988 206.84 73.923
## - x1
          1
             16.602 208.46 74.156
## - x9
          1
             18.072 209.92 74.366
             21.314 213.17 74.826
## - x3
          1
## - x8
          1 28.835 220.69 75.867
## - x5
        1 40.323 232.18 77.389
##
## Step: AIC=73.31
## y \sim x1 + x3 + x5 + x7 + x8 + x9 + x10
##
         Df Sum of Sq
                      RSS
                               AIC
## - x7
          1
             10.457 213.09 72.815
## - x3
               10.595 213.23 72.835
          1
## - x1
          1
              11.998 214.63 73.032
## - x9
          1
              12.643 215.28 73.122
## - x10
              13.887 216.52 73.295
          1
## <none>
                     202.63 73.306
## - x8
          1
             27.665 230.30 75.145
## - x5
             30.191 232.82 75.472
          1
```

```
##
## Step: AIC=72.82
## y \sim x1 + x3 + x5 + x8 + x9 + x10
##
         Df Sum of Sq RSS
## - x3
        1 4.8720 217.96 71.494
## - x9
        1 5.2049 218.29 71.539
## - x1
         1 5.3212 218.41 71.555
                     213.09 72.815
## <none>
## - x10 1 18.3677 231.46 73.296
## - x5
        1 23.3458 236.44 73.934
## - x8
        1 26.0316 239.12 74.273
##
## Step: AIC=71.49
## y \sim x1 + x5 + x8 + x9 + x10
##
##
         Df Sum of Sq
                      RSS
                               AIC
       1 0.765 218.73 69.599
## - x1
## - x9
              5.863 223.82 70.290
          1
                217.96 71.494
## <none>
## - x10 1
            20.291 238.25 72.164
## - x5 1 23.020 240.98 72.506
## - x8
        1 31.634 249.59 73.559
## Step: AIC=69.6
## y \sim x5 + x8 + x9 + x10
##
         Df Sum of Sq RSS
        1 5.097 223.82 68.290
## - x9
                     218.73 69.599
## <none>
             40.404 259.13 72.684
## - x5
         1
## - x8
        1 57.407 276.13 74.591
## - x10 1 135.105 353.83 82.029
##
## Step: AIC=68.29
## y \sim x5 + x8 + x10
##
##
         Df Sum of Sq RSS
## <none>
                     223.82 68.290
## - x5
             36.314 260.14 70.800
          1
## - x8
          1 52.960 276.78 72.661
          1 194.838 418.66 85.076
## - x10
modback$anova
## Stepwise Model Path
## Analysis of Deviance Table
## Initial Model:
## y \sim x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11
##
## Final Model:
## y \sim x5 + x8 + x10
##
##
```

```
Step Df
             Deviance Resid. Df Resid. Dev
##
                                   187.4007 78.96155
## 1
                               18
## 2 - x11 1 0.4648362
                               19
                                   187.8655 77.03587
## 3 - x6 1 0.5356445
                                  188.4012 75.12128
                               20
## 4 - x4
          1 3.4514854
                               21
                                   191.8526 73.66591
## 5 - x2 1 10.7796848
                               22
                                  202.6323 73.30587
## 6 - x7 1 10.4571693
                               23
                                   213.0895 72.81545
## 7 - x3 1 4.8720101
                                  217.9615 71.49363
                               24
## 8 - x1 1 0.7654631
                               25
                                  218.7270 69.59881
## 9 - x9 1 5.0970905
                               26
                                  223.8241 68.28989
summary(modback)
##
## Call:
## lm(formula = y \sim x5 + x8 + x10, data = datis)
## Residuals:
      Min
               1Q Median
                               3Q
                                     Max
## -4.6101 -1.9868 -0.6613 2.0369 5.8811
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 4.590404 11.771925 0.390
                                            0.6998
## x5
               2.597240
                         1.264562 2.054
                                            0.0502 .
## x8
               0.217814
                          0.087817
                                   2.480
                                            0.0199 *
## x10
              -0.009485
                        0.001994 -4.757 6.38e-05 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.934 on 26 degrees of freedom
## Multiple R-squared: 0.8035, Adjusted R-squared: 0.7808
## F-statistic: 35.44 on 3 and 26 DF, p-value: 2.462e-09
# forward selection -----
empty.model <- lm(y \sim 1, data = datis)
horizonte <- formula( lm(y ~ ., data = datis) )
horizonte
## y \sim x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11
modforw <- stepAIC(empty.model, trace=FALSE, direction="forward",</pre>
                  scope=horizonte)
modforw$anova
## Stepwise Model Path
## Analysis of Deviance Table
## Initial Model:
## y ~ 1
##
## Final Model:
## y \sim x1 + x4
##
##
##
    Step Df Deviance Resid. Df Resid. Dev
                                                AIC
```

```
29 1139.1050 111.10402
## 1
## 2 + x1 1 866.49528
                            28 272.6097 70.20532
## 3 + x4 1 18.57161
                            27 254.0381 70.08861
summary(modforw)
##
## Call:
## lm(formula = y \sim x1 + x4, data = datis)
## Residuals:
##
      Min
              1Q Median
                              3Q
                                     Max
## -6.5011 -2.1243 -0.3884 1.9964 6.9582
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 7.179421 18.787955
                                  0.382
## x1
              -0.044479
                        0.005225 -8.513 3.98e-09 ***
## x4
               3.077228
                         2.190294
                                  1.405
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.067 on 27 degrees of freedom
## Multiple R-squared: 0.777, Adjusted R-squared: 0.7605
## F-statistic: 47.03 on 2 and 27 DF, p-value: 1.594e-09
# Comparing -----
coef(modback)
## (Intercept)
                        x5
                                     8x
                                                x10
## 4.590404189 2.597240342 0.217814237 -0.009485195
AIC(modback)
## [1] 155.4262
coef(modforw)
## (Intercept)
                      x1
                                  x4
## 7.17942062 -0.04447915 3.07722832
AIC(modforw)
## [1] 157.2249
# In a graphical way -----
par(mfrow=c(1,2))
require(car)
## Loading required package: car
qqPlot(modback, main="Backward", pch=19)
qqPlot(modforw, main="Forward", pch=19)
```

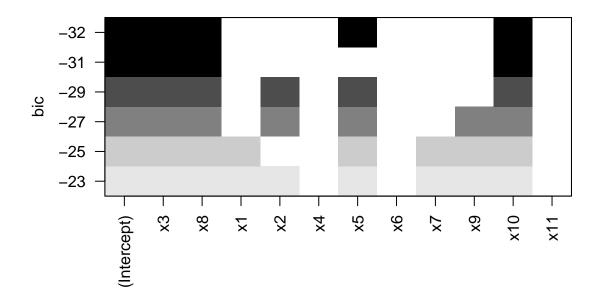


```
library(leaps)
prueba <- regsubsets(y ~ ., data = datis, nbest=2, intercept=T)</pre>
summary(prueba)
## Subset selection object
## Call: regsubsets.formula(y ~ ., data = datis, nbest = 2, intercept = T)
## 11 Variables (and intercept)
##
       Forced in Forced out
## x1
           FALSE
                      FALSE
## x2
           FALSE
                      FALSE
           FALSE
                      FALSE
## x3
           FALSE
                      FALSE
## x4
## x5
           FALSE
                      FALSE
           FALSE
                      FALSE
## x6
           FALSE
                      FALSE
## x7
                      FALSE
## x8
           FALSE
## x9
           FALSE
                      FALSE
           FALSE
                      FALSE
## x10
           FALSE
                      FALSE
## x11
## 2 subsets of each size up to 8
## Selection Algorithm: exhaustive
##
                                 x6
## 1
      (1)
```

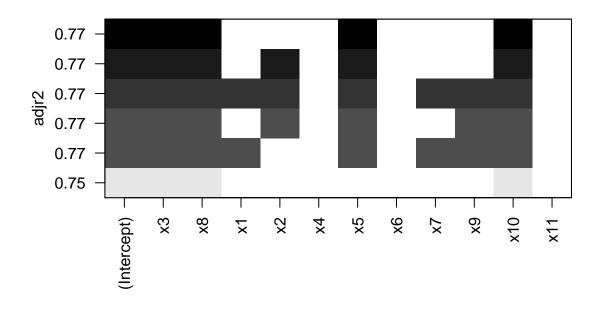
```
(1) " " " " " " " *" " " " " " *" " *"
        (2) " " " " " * " * " " " " " * " " " * " " * " " * " " * " " * " " * " " * " " * " " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " *
        (1) "*" " "*" " "*" "*" "*" "*" "*"
        ## 8 (1) "*" "*" "*" " " "*" "*" "*" "*" "*" "
        (2) "*" " " "*" "*" "*" "*" "*" "*" "*" "
summary(prueba)
## Subset selection object
## Call: regsubsets.formula(y \sim ., data = datis, nbest = 2, intercept = T)
## 11 Variables (and intercept)
          Forced in Forced out
                FALSE
## x1
                                 FALSE
                FALSE
                                 FALSE
## x2
                FALSE
## x3
                                FALSE
## x4
                FALSE
                                FALSE
                FALSE
                                FALSE
## x5
## x6
                FALSE
                                FALSE
                FALSE
                                FALSE
## x7
## x8
                FALSE
                                FALSE
## x9
                FALSE
                                FALSE
                FALSE
                                FALSE
## x10
## x11
                FALSE
                                FALSE
## 2 subsets of each size up to 8
## Selection Algorithm: exhaustive
                  x1 x2 x3 x4 x5 x6 x7 x8 x9 x10 x11
(1) " " " " " " * " " " " * " " " * " " "
(2)""""*""*"""""*"""
        ## 6 (1) " " " " " " " *" " *" " *" " *" " *"
        (1) "*" " "*" " "*" " "*" "*" "*" "*"
        (2) "*" " "*" "*" "*" "*" "*" "*" "*" "
        do.call(cbind,(summary(prueba)[c("rsq","rss","adjr2","cp","bic")]))
##
                                                adjr2
                                    rss
     [1,] 0.7606808 272.6097 0.7521337 0.1844047 -36.09631
      [2,] 0.7274044 310.5150 0.7176688 3.8252404 -32.19058
```

```
## [3,] 0.7769845 254.0381 0.7604648 0.4005851 -34.81182
## [4,] 0.7741259 257.2944 0.7573944 0.7133507 -34.42973
## [5,] 0.8035088 223.8241 0.7808368 -0.5015020 -35.20935
## [6,] 0.7886898 240.7045 0.7643079 1.1198744 -33.02806
## [7,] 0.8079835 218.7270 0.7772608 1.0089180 -32.49923
## [8,] 0.8059055 221.0940 0.7748504 1.2362726 -32.17632
## [9,] 0.8112433 215.0137 0.7719190 2.6522501 -29.61171
## [10,] 0.8091718 217.3733 0.7694159 2.8789005 -29.28427
## [11,] 0.8145684 211.2261 0.7661949 4.2884510 -26.74369
## [12,] 0.8142238 211.6186 0.7657605 4.3261490 -26.68800
## [13,] 0.8221127 202.6323 0.7655122 5.4630135 -24.58857
## [14,] 0.8219563 202.8105 0.7653060 5.4801255 -24.56221
## [15,] 0.8315760 191.8526 0.7674144 6.4276153 -22.82734
## [16,] 0.8285201 195.3336 0.7631944 6.7619658 -22.28790
# To have x8 and x3 in the model
prueba <- regsubsets(y ~ ., data = datis, nbest=1, intercept=T,</pre>
                 force.in=c("x8", "x3"))
summary(prueba)
## Subset selection object
## Call: regsubsets.formula(y ~ ., data = datis, nbest = 1, intercept = T,
     force.in = c("x8", "x3"))
## 11 Variables (and intercept)
##
     Forced in Forced out
## x3
         FALSE
                  FALSE
## x8
         FALSE
                  FALSE
                  FALSE
## x1
          TRUE
## x2
         FALSE
                  FALSE
         FALSE
## x4
                  FALSE
## x5
         FALSE
                  FALSE
## x6
         FALSE
                  FALSE
## x7
          TRUE
                  FALSE
## x9
         FALSE
                  FALSE
## x10
         FALSE
                  FALSE
## x11
         FALSE
                  FALSE
## 1 subsets of each size up to 8
## Selection Algorithm: exhaustive
          x3 x8 x1 x2 x4 x5 x6 x7 x9 x10 x11
do.call(cbind,(summary(prueba)[c("rsq","rss","adjr2","cp","bic")]))
##
                          adjr2
           rsq
                   rss
                                    ср
                                            bic
## [1,] 0.7752010 256.0697 0.7492627 2.595716 -31.17167
## [2,] 0.8049501 222.1823 0.7737422 1.340803 -32.02901
## [3,] 0.8088112 217.7841 0.7689803 2.918351 -29.22764
## [4,] 0.8142238 211.6186 0.7657605 4.326149 -26.68800
## [5,] 0.8221127 202.6323 0.7655122 5.463013 -24.58857
## [6,] 0.8315760 191.8526 0.7674144 6.427615 -22.82734
```

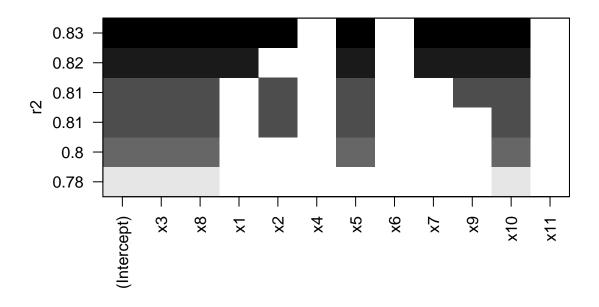
```
\# Coefficients for the first 3 models
coef(prueba, 1:3)
## [[1]]
## (Intercept)
                          xЗ
                                       8x
## 17.860760992 -0.014720296  0.187416711 -0.008456783
##
## [[2]]
## (Intercept)
                        xЗ
                                    8x
                                                x5
                                                           x10
## 1.88960198 0.01092218 0.23700836 2.93057655 -0.01069613
## [[3]]
## (Intercept)
                       xЗ
                                    8x
                                                x2
                                                            x5
                                                                       x10
## -0.55381414 0.04401768 0.24735136 -0.04262137 3.72400540 -0.01161920
# plot a table of models showing variables in each model.
# models are ordered by the selection statistic.
par(mfrow=c(1,1))
plot(prueba, scale="bic")
```



plot(prueba, scale="adjr2")



plot(prueba, scale="r2")



plot(prueba, scale="Cp")

