Guided Question Set 9 Solutions

1)

```
Data<-read.table("nfl.txt", header=TRUE)</pre>
allreg <- regsubsets(y ~., data=Data, nbest=2)</pre>
summary(allreg)
## Subset selection object
## Call: regsubsets.formula(y ~ ., data = Data, nbest = 2)
## 9 Variables (and intercept)
               Forced in Forced out
##
## x1
                         FALSE
                                                     FALSE
## x2
                         FALSE
                                                     FALSE
## x3
                         FALSE
                                                     FALSE
## x4
                         FALSE
                                                     FALSE
                         FALSE
                                                     FALSE
## x5
                         FALSE
## x6
                                                     FALSE
## x7
                         FALSE
                                                     FALSE
## x8
                         FALSE
                                                     FALSE
## x9
                         FALSE
                                                     FALSE
## 2 subsets of each size up to 8
## Selection Algorithm: exhaustive
##
                              x1 x2 x3 x4 x5 x6
                                                                                        x7
                ## 1
               ## 2
             ## 2
               ## 3
               (1) " " " * " " " " " " " " * " " * " " * " " * " " * " " * " " * " " * " " * " " * " " * " " * " " * " " * " " * " " * " " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * " * "
## 4
               ## 4
             ## 5
             ## 5
               (1)""*""*""*""""""*""*""*"
## 6
            ## 6
```

2a)

The regression equation with the highest adjusted R^2 is $\hat{y} = -1.8217 + 0.0038x_2 + 0.2169x_7 - 0.0040x_8 - 0.0016x_9$.

```
coef(allreg, which.max(summary(allreg)$adjr2))
```

```
## (Intercept) x2 x7 x8 x9
## -1.821703427 0.003818572 0.216894094 -0.004014887 -0.001634926
```

2b)

The regression equation with the lowest Mallow's C_p is $\hat{y} = -1.8084 + 0.0036x_2 + 0.1940x_7 - 0.0048x_8$.

```
coef(allreg, which.min(summary(allreg)$cp))
```

```
## (Intercept) x2 x7 x8
## -1.808372059 0.003598070 0.193960210 -0.004815494
```

2c)

The regression equation with the lowest BIC is $\hat{y} = -1.8084 + 0.0036x_2 + 0.1940x_7 - 0.0048x_8$.

```
coef(allreg, which.min(summary(allreg)$bic))
```

```
## (Intercept) x2 x7 x8
## -1.808372059 0.003598070 0.193960210 -0.004815494
```

3)

The regression equation from forward selection is $\hat{y} = -1.8217 + 0.0038x_2 + 0.2169x_7 - 0.0040x_8 - 0.0016x_9$.

```
##intercept only model
regnull <- lm(y~1, data=Data)
##model with all predictors
regfull <- lm(y~., data=Data)</pre>
step(regnull, scope=list(lower=regnull, upper=regfull), direction="forward")
## Start: AIC=70.81
## y ~ 1
##
##
          Df Sum of Sq
                          RSS
                                  AIC
           1
               178.092 148.87 50.785
## + x8
               115.068 211.90 60.669
## + x1
           1
## + x7
           1
                97.238 229.73 62.931
## + x5
                86.116 240.85 64.255
           1
## + x2
                76.193 250.77 65.385
           1
## + x9
           1
                30.167 296.80 70.104
## <none>
                       326.96 70.814
## + x4
                21.844 305.12 70.878
           1
## + x6
                16.411 310.55 71.372
           1
## + x3
                 2.135 324.83 72.631
           1
##
## Step:
          AIC=50.78
## y ~ x8
##
##
          Df Sum of Sq
                           RSS
                                   AIC
                64.934 83.938 36.741
## + x2
           1
## + x5
           1
                11.607 137.265 50.512
## <none>
                       148.872 50.785
## + x1
                 6.636 142.236 51.508
           1
## + x3
           1
                 6.368 142.504 51.561
## + x4
                 6.345 142.527 51.565
           1
## + x7
           1
                 0.974 147.898 52.601
## + x6
                 0.487 148.385 52.693
           1
## + x9
           1
                 0.008 148.864 52.783
##
## Step: AIC=36.74
## y ~ x8 + x2
##
          Df Sum of Sq
##
                          RSS
                                  AIC
## + x7
           1
               14.0682 69.870 33.604
## + x1
           1
               11.1905 72.748 34.734
## + x3
                8.9010 75.037 35.602
           1
## + x5
           1
                5.8147 78.124 36.730
```

83.938 36.741

<none>

```
## + x9
           1
                2.0256 81.913 38.057
                1.3216 82.617 38.296
## + x6
## + x4
           1
                0.0161 83.922 38.735
##
## Step: AIC=33.6
## y \sim x8 + x2 + x7
##
          Df Sum of Sq
##
                           RSS
                                  AIC
## + x9
                4.8657 65.004 33.583
## <none>
                       69.870 33.604
## + x3
           1
                1.3873 68.483 35.043
## + x4
           1
                0.9792 68.891 35.209
                0.9022 68.968 35.240
## + x1
           1
## + x6
                0.4879 69.382 35.408
           1
## + x5
                0.2987 69.571 35.484
           1
##
## Step:
          AIC=33.58
## y \sim x8 + x2 + x7 + x9
##
##
          Df Sum of Sq
                           RSS
                                  AIC
                       65.004 33.583
## <none>
               1.86452 63.140 34.768
## + x1
## + x4
               1.74260 63.262 34.822
## + x3
           1
               0.70148 64.303 35.279
               0.45071 64.554 35.388
## + x6
           1
## + x5
               0.32667 64.678 35.442
           1
##
## Call:
## lm(formula = y \sim x8 + x2 + x7 + x9, data = Data)
##
## Coefficients:
## (Intercept)
                         8x
                                       x2
                                                     x7
                                                                  x9
     -1.821703
                  -0.004015
                                 0.003819
                                               0.216894
##
                                                           -0.001635
```

4)

Backward elimination pick the same model as forward selection.

```
step(regfull, scope=list(lower=regnull, upper=regfull), direction="backward")
```

5)

Stepwise regression picks the same model as forward selection and backward elimination.

```
step(regnull, scope=list(lower=regnull, upper=regfull), direction="both")
```

6)

```
PRESS <- function(linear.model) {
    ## get the residuals from the linear.model.
    ## extract hat from lm.influence to obtain the leverages
    pr <- residuals(linear.model)/(1-lm.influence(linear.model)$hat)
    ## calculate the PRESS by squaring each term and adding them up
    PRESS <- sum(pr^2)
    return(PRESS)
}</pre>
```

7)

The PRESS statistic is 87.46. The $R^2_{prediction}$ is 0.7325. The R^2 is 0.7863.

The model might be able to explain 73.25% of the variability in the new observations. The R^2 is 0.7863. Both values are fairly high and close to each other, so the model has good predictive ability.

```
result<-lm(y~x2+x7+x8, data=Data)
PRESS(result)</pre>
```

[1] 87.46123

```
##Find SST
anova_result<-anova(result)
SST<-sum(anova_result$"Sum Sq")
##R2 pred
Rsq_pred<-1-PRESS(result)/SST
Rsq_pred</pre>
```

```
## [1] 0.7325052
```

summary(result)

```
##
## Call:
## lm(formula = y \sim x2 + x7 + x8, data = Data)
##
## Residuals:
##
     Min
             1Q Median
                          3Q
                               Max
## -3.0370 -0.7129 -0.2043 1.1101 3.7049
##
## Coefficients:
##
             Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.808372
                    7.900859 -0.229 0.820899
## x2
            ## x7
             ## x8
            ## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.706 on 24 degrees of freedom
## Multiple R-squared: 0.7863, Adjusted R-squared: 0.7596
## F-statistic: 29.44 on 3 and 24 DF, p-value: 3.273e-08
```