	Q1	Q2	Q3	Q4	Q5
50 Points	10	14	8	8	10

Question 1

For multiple regression

$$y = X\beta + \epsilon, \ \epsilon \sim N(0, \ \sigma^2)$$

$$y = X \atop n \times 1 \qquad X \atop n \times p \qquad \beta \atop p \times 1 \qquad \epsilon \atop n \times 1$$

Derive or show that

a.
$$\hat{\beta} = (X'X)^{-1}X'Y$$

$$y = X\beta + \epsilon$$
 Minimize: $S(\beta) = \sum_{i=1}^n \epsilon_i^2 = \epsilon' \epsilon$

$$S(\beta) = (y - X\beta)'(y - X\beta)$$

$$= y'y - \beta'X'y - y'X\beta + \beta'X'X\beta$$
(since $\beta'X'y$ is 1 x 1, $\beta'X'y = y'X\beta$)
$$= y'y - 2\beta'X'y + \beta'X'X\beta$$

So,

$$\begin{split} \frac{\partial S}{\partial \beta} \Big|_{\hat{\beta}} &= -2X'y + 2X'X\hat{\beta} \\ -2X'y + 2X'X\hat{\beta} &= 0 \\ 2X'X\hat{\beta} &= 2X'y \\ X'X\hat{\beta} &= X'y \\ \hat{\beta} &= (X'X)^{-1}X'y \end{split}$$

b. $E[\hat{\beta}] = \beta$

$$\begin{split} \mathbf{E}[\hat{\beta}] &= \mathbf{E}[(X'X)^{-1}X'y] \\ &= (X'X)^{-1}X'\mathbf{E}[y] \\ &= (X'X)^{-1}X'(X\beta + 0) \\ &= (X'X)^{-1}X'X\beta \\ &= \beta \end{split}$$

c.
$$V[\hat{\beta}] = \sigma^2(X'X)^{-1}$$

$$\begin{split} V[\hat{\beta}] &= V[(X'X)^{-1}X'y] \\ &= (X'X)^{-1}X' \times V[y] \times ((X'X)^{-1}X')' \\ &= (X'X)^{-1}X' \times V[y] \times X((X'X)^{-1})' \\ &= (X'X)^{-1}X' \times V[y] \times X((X'X)')^{-1} \\ &= (X'X)^{-1}X' \times V[y] \times X(X'X)^{-1} \\ &= (X'X)^{-1}X' \times X(X'X)^{-1} \times V[y] \\ &= (X'X)^{-1}X'X(X'X)^{-1} \times V[y] \\ &= (X'X)^{-1}V[y] \\ &= \sigma^2(X'X)^{-1} \end{split}$$

d. $E[\hat{Y}] = X\beta$

$$\begin{split} \mathbf{E}[\hat{Y}] &= \mathbf{E}[\hat{\boldsymbol{\beta}}_0 + \hat{\boldsymbol{\beta}}_1 X_1 + \hat{\boldsymbol{\beta}}_2 X_2 ...] \\ &= \mathbf{E}[X \hat{\boldsymbol{\beta}}] \\ &= X \times \mathbf{E}[\hat{\boldsymbol{\beta}}] \\ &= X \boldsymbol{\beta} \end{split}$$

e. $V[\hat{Y}] = \sigma^2 H$, where H is the hat matrix and $H = X(X'X)^{-1}X'$

$$\begin{split} V[\hat{Y}] &= V[\hat{\beta}_0 + \hat{\beta}_1 X_1 + \hat{\beta}_2 X_2 \ldots] \\ &= V[X\hat{\beta}] \\ &= XV[\hat{\beta}] X' \\ &= X\sigma^2 (X'X)^{-1} X' \\ &= \sigma^2 X (X'X)^{-1} X' \\ &= \sigma^2 H \end{split}$$

Question 2 (problems 3.1 and 3.3 on page 121)

a. Fit a multiple linear regression model relating the number of games won to the team's passing yardage (x_2) , the percentage of rushing plays (x_7) , and the opponents' yards rushing (x_8) .

	Estimate	Std. Error	t value	$\Pr(> t)$
(Intercept)	-1.8084	7.9009	-0.23	0.8209
x\$x2	0.0036	0.0007	5.18	0.0000
x\$x7	0.1940	0.0882	2.20	0.0378
x\$x8	-0.0048	0.0013	-3.77	0.0009

b. Construct the analysis-of-variance table and test for significance of regression.

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
x\$x2	1	76.19	76.19	26.17	0.0000
x\$x7	1	139.50	139.50	47.92	0.0000
x\$x8	1	41.40	41.40	14.22	0.0009
Residuals	24	69.87	2.91		

To test for significance of regression, we establish H_0 and H_a :

$$H_0: \beta_2 = \beta_7 = \beta_8 = 0$$

 H_a : $\beta_j \neq 0$ for at least one of j = 2, 7, 8

We reject H_0 if $F_{0,j} > F_{0.05 = \alpha}$, $g_0 = 18 = (28 - 9 - 1)$ for any $g_0 = 18 = 19$ for an

$$F_{0,2} = 26.17 > 2.4563$$

$$F_{0,7} = 47.92 > 2.4563$$

$$F_{0.8} = 14.22 > 2.4563$$

So, reject H_0 . There is evidence to conclude that there is a linear relationship for $y \sim x_2$, $y \sim x_7$, and $y \sim x_8$

- c. Calculate t statistics for testing the hypotheses H_0 : $\beta_2=0$, H_0 : $\beta_7=0$, H_0 : $\beta_8=0$. What conclusions can you draw about the roles the variables x_2 , x_7 , and x_8 play in the model?
 - (1) R:

i) H₀:
$$\beta_2 = 0$$

 $\beta_2 = 0.003598$, t = 5.177, t_{0.05,24} = 2.064 \longrightarrow |5.117| > 2.064 \Rightarrow Reject H₀

ii) H₀:
$$\beta_7 = 0$$

 $\beta_7 = 0.193960$, t = 2.198, t_{0.05/224} = 2.064 \longrightarrow |2.198| > 2.064 \Longrightarrow Reject H₀

iii) H₀:
$$\beta_8=0$$
 $\beta_8=-0.004816,$ t = -3.771, t_{0.05.24} = 2.064 \longrightarrow |-3.771| > 2.064 \Longrightarrow Reject H₀

(2) Manual:

i) H₀:
$$\beta_2 = 0$$

 $\beta_2 = 0.003598$, $t_{\frac{0.05}{2},24} = 2.064$

$$t = \frac{\hat{\beta}_2 - 0}{se(\hat{\beta}_2)}$$
$$= \frac{0.003598}{0.000695}$$
$$= 5.177$$

$$|5.117| > 2.064 \Rightarrow \text{Reject H}_0$$

ii)
$$H_0$$
: $\beta_7 = 0$
 $\beta_7 = 0.193960$, $t_{\frac{0.05}{2}.24} = 2.064$

$$t = \frac{\hat{\beta}_7 - 0}{se(\hat{\beta}_7)}$$
$$= \frac{0.193960}{0.088233}$$
$$= 2.198$$

$$|2.198| > 2.064 \Rightarrow \text{Reject H}_0$$

iii) H₀:
$$\beta_{~8}=0$$

 $\beta_{~8}=$ -0.004816, t_{0.05,24} = 2.064

$$t = \frac{\hat{\beta}_8 - 0}{se(\hat{\beta}_8)}$$
$$= \frac{-0.004816}{0.001277}$$
$$= -3.771$$

$$\mid$$
 - 3.771 \mid > 2.064 \Rightarrow Reject H₀

- d. Calculate R^2 and R^2_{adj} for this model.
 - (1) R:

 $R^2 \longrightarrow summary(model)$ \$r.squared yields **0.7863069**

 $R^2_{adj} \longrightarrow summary(model)$ \$adj.r.squared yields **0.7595953**

(2) Manual:

Knowing: $SS_T = SS_R + SS_{res}$

From anova(model) in R:

$$SS_T = (76.193 + 139.501 + 41.400) (SS_R) + 69.870 (SS_{res}) = 326.964$$

$$R^{2} = 1 - \frac{SS_{res}}{SS_{T}}$$
$$= 1 - \frac{69.870}{326.964}$$
$$= 0.7863067$$

$$R_{\text{adj}}^2 = \frac{1 - \frac{SS_{\text{res}}}{(n-p)}}{\frac{SS_{\text{T}}}{(n-1)}}$$

$$= 1 - \frac{SS_{res}(n-1)}{SS_T(n-k-1)}$$

$$= 1 - \frac{69.870(27)}{326.964(24)}$$

$$= 0.7595951$$

e. Using the partial F test, determine the contribution of x_7 to the model. How is this partial F statistic related to the t test for β 7 calculated in part c above?

Knowing:

The partial F-test is the most common method of testing for a nested normal linear regression model. "Nested" model is just a fancy way of saying a reduced model in terms of variables included.

If $F_0 > F_{\alpha,r,n-p}$, we reject H_0 , concluding that at least one of the parameters in β_2 is not zero, and consequently at least one of the regressors $x_{k-r+1}, x_{k-r+2}, \ldots, x_k$ in X_2 contribute significantly to the regression model. Some authors call the test in (3.35) a partial F test because it measures the contribution of the regressors in X_2 given that the other regressors in X_1 are in the model.

Partial F-Test:

$$H_0: \beta_2 = 0$$

$$F_0 = \frac{SS_R(\beta_1|\beta_2)}{r \times MS_{res}}$$

where $\beta_1 = \beta - \{\beta_7\}, \beta_2 = \beta_7$

$$SS_R(\beta_2|\beta_1) = SS_R(\beta) - SS_R(\beta_1)$$
$$SS_R(\beta_2|\beta_1) = (76.193 + 139.501 + 41.400) - (76.193 + 41.400)$$
$$= 139.501$$

r = 1

 $MS_{res} = 2.911$

$$F_0 = \frac{139.501}{1 \times 2.911}$$
$$= 47.92202$$

$$F_{\alpha,r,n-p} = F_{0.05,1,(28-(3+1)=24} = 4.2597$$

Reject H_0 if $F_0 > F_{0.05,1,24}$
 $47.92202 > 4.2597 \longrightarrow \text{reject } H_0$
anova(lm(y \sim x7))\$F yields 11.00524
 $qf(0.025, df1 = 1, df2 = 24, \text{lower.tail} = F)$ yields 5.713369

- f. Find a 95% CI on β_7 . (This is part a of problem 3.3, and the following one is part b of problem 3.3.)
 - (1) R:
 - (2) Manual:

A CI for
$$\beta_j$$
 is $\hat{\beta}_j$ (+ or -) $\mathbf{t}_{\frac{\alpha}{2},n-p} SE(\hat{\beta}_j)$
 $\hat{\beta}_7 = 0.193960$

$$\rho_{7} = 0.130300$$

$$t_{\frac{\alpha}{2},n-p} = t_{0.025,28-4=24} = 2.064$$

$$SE(\hat{\beta}_i = 0.088233)$$

$$(0.193960 - (2.064 \times 0.088233), 0.193960 + (2.064 \times 0.088233)$$

- g. Find a 95% CI on the mean number of games won by a team when $x_2 = 2300$, $x_7 = 56.0$, and $x_8 = 2100$.
 - (1) R:

prediction(

(2) Manual:

Note: For c, d, f, and g, please show two versions of your results: (1) obtained using R code and (2) based on your manual calculation (please show detailed step for your manual calculation. You can use the partial output from the lm or ANOVA, e.g., the SS_{reg} , SS_{res} , the estimated value of β and its variance or standard deviation). If you can show how to get the t-statistics (or CI, R-square) based on part of the output obtained from R, that will be fine.

Question 3 (Exercise 3.4 on page 122)

Reconsider the National Football League data from Problem 3.1. Fit a model to this data using only x_7 and x_8 as the regressors.

- a. Test for significance of the regression.
- b. Calculate R^2 and R^2_{adj} . How do these quantities compare to the values computed for the model in problem 3.1, which included an additional regressor (x^2) ?
- c. Calculate a 95% CI on β 7. Also, find a 95% CI on the mean number of games won by a team when $x_7 = 56.0$ and $x_8 = 2100$. Compare the lengths of these CIs to the lengths of the corresponding CIs from problem 3.3 (that is, the above part f and g in question 2)
- d. What conclusions can you draw from this problem about the consequences of omitting an important regressor from a model?

Question 4 (exercise 4.2 on page 165)

Consider the multiple regression model fit to the National Football League (NFL) team performance data in problem 3.1.

- a. Construct a normal probability plot of the residuals. Does there seem to be any problem with the normality assumption?
- b. Construct and interpret a plot of the residuals versus the predicted response.
- c. Construct plots of the residuals versus each of the regressor variables. Do these plots imply that the regressor is correctly specified?
- d. Construct the partial regression plots for this model. Compare the plots with the plots of residuals versus regressors from part c above. Discuss the type of information provided by these plots.

Question 5

Show that the hat matrix $H = X(X'X)^{-1}X'$ and I - H (where I is the identity matrix) are symmetric and idempotent. That is, please show:

a. H' = H and HH = H (H' means the transpose of H, HH means H * H)

$$H = X(X'X)^{-1}X'$$

$$H' = (X(X'X)^{-1}X')'$$

$$= X((X'X)^{-1})'X'$$

$$= X((X'X)')^{-1}X'$$

$$= X(X'X)^{-1}X'$$

$$= H$$

$$H = X(X'X)^{-1}X'$$

$$HH = (X(X'X)^{-1}X')(X(X'X)^{-1}X')$$

$$HH = X(X'X)^{-1}X'X(X'X)^{-1}X'$$

$$= X(X'X)^{-1}X'$$

$$= H$$

b.
$$(I - H)' = I - H$$
 and $(I - H)(I - H) = I - H$

$$(I - H)' = (I - X(X'X)^{-1}X')'$$

$$= I' - (X(X'X)^{-1}X')')'$$

$$= I - (X(X'X)^{-1}X')'$$

$$= I - X(X'X)^{-1}X'$$

$$= I - H$$

$$\begin{split} (I-H)(I-H) &= (I-X(X'X)^{-1}X')(I-X(X'X)^{-1}X') \\ &= I-2X(X'X)^{-1}X' + (X(X'X)^{-1}X')(X(X'X)^{-1}X') \\ &= I-2X(X'X)^{-1}X' + X(X'X)^{-1}X' \text{ by (a)} \\ &= I-X(X'X)^{-1}X' \\ &= I-H \end{split}$$

Hint: A = X'X is a symmetric matrix, and for a symmetric matrix, $(A')^{-1} = (A^{-1})'$. You can use this property directly in your proof of (a) and (b). If you are interested in the proof of this property, you may check the following web page:

https://math.stackexchange.com/questions/325082/is-the-inverse-of-a-symmetric-matrix-also-symmetric