

Chapter 12 Selected Problem Solutions

Section 12-1

12-1. a) $\mathbf{X}'\mathbf{X} = \begin{bmatrix} 10 & 223 & 553 \\ 223 & 5200.9 & 12352 \\ 553 & 12352 & 31729 \end{bmatrix}$

$$\mathbf{X}'\mathbf{y} = \begin{bmatrix} 1916.0 \\ 43550.8 \\ 104736.8 \end{bmatrix}$$

b) $\hat{\beta} = \begin{bmatrix} 171.054 \\ 3.713 \\ -1.126 \end{bmatrix}$, so $\hat{y} = 171.054 + 3.714x_1 - 1.126x_2$

c) $\hat{y} = 171.054 + 3.714(18) - 1.126(43) = 189.481$

12-5.

Predictor	Coeff	StDev	T	P
Constant	33.449	1.576	21.22	0.000
x1	-0.054349	0.006329	-8.59	0.000
x6	1.0782	0.6997	1.54	0.138
S = 2.834	R-Sq = 82.9%		R-Sq(adj) = 81.3%	
Analysis of Variance				
Source	DF	SS	MS	F
Regression	2	856.24	428.12	53.32
Error	22	176.66	8.03	
Total	24	1032.90		

a) $\hat{y} = 33.4491 - 0.05435x_1 + 1.07822x_2$

b) $\hat{\sigma}^2 = 8.03$

c) $\hat{y} = 33.4491 - 0.05435(300) + 1.07822(2) = 19.30$ mpg.

12-7.

Predictor	Coeff	SE Coef	T	P
Constant	383.80	36.22	10.60	0.002
X1	-3.6381	0.5665	-6.42	0.008
X2	-0.11168	0.04338	-2.57	0.082
S = 12.35	R-Sq = 98.5%		R-Sq(adj) = 97.5%	
Analysis of Variance				
Source	DF	SS	MS	F
Regression	2	29787	14894	97.59
Residual Error	3	458	153	
Total	5	30245		

a) $\hat{y} = 383.80 - 3.6381x_1 - 0.1119x_2$

b) $\hat{\sigma}^2 = 153.0$, $se(\hat{\beta}_0) = 36.22$, $se(\hat{\beta}_1) = 0.5665$, and $se(\hat{\beta}_2) = .04338$

c) $\hat{y} = 383.80 - 3.6381(25) - 0.1119(1000) = 180.95$

d)

Predictor	Coeff	SE Coef	T	P
Constant	484.0	101.3	4.78	0.041
X1	-7.656	3.846	-1.99	0.185
X2	-0.2221	0.1129	-1.97	0.188
X1*X2	0.004087	0.003871	1.06	0.402
S = 12.12	R-Sq = 99.0%		R-Sq(adj) = 97.6%	
Analysis of Variance				
Source	DF	SS	MS	F
Regression	3	29951.4	9983.8	67.92
Residual Error	2	294.0	147.0	
Total	5	30245.3		

$$\hat{y} = 484.0 - 7.656x_1 - 0.222x_2 - 0.0041x_{12}$$

e) $\hat{\sigma}^2 = 147.0$, $se(\hat{\beta}_0) = 101.3$, $se(\hat{\beta}_1) = 3.846$, $se(\hat{\beta}_2) = 0.113$ and $se(\hat{\beta}_{12}) = 0.0039$

f) $\hat{y} = 484.0 - 7.656(25) - 0.222(1000) - 0.0041(25)(1000) = -31.3$

The predicted value is smaller

Predictor	Coef	SE Coef	T	P
Constant	47.17	49.58	0.95	0.356
x1	-9.735	3.692	-2.64	0.018
x2	0.4283	0.2239	1.91	0.074
x3	18.237	1.312	13.90	0.000
S	3.480	R-Sq = 99.4%	R-Sq(adj) = 99.3%	
Analysis of Variance				
Source	DF	SS	MS	F P
Regression	3	30532	10177	840.55 0.000
Residual Error	16	194	12	
Total	19	30725		

a) $y = 47.174 - 9.735x_1 + 0.4283x_2 + 18.2375x_3$

b) $\hat{\sigma}^2 = 12$

c) $se(\hat{\beta}_0) = 49.5815$, $se(\hat{\beta}_1) = 3.6916$, $se(\hat{\beta}_2) = 0.2239$, and $se(\hat{\beta}_3) = 1.312$

d) $y = 47.174 - 9.735(14.5) + 0.4283(220) + 18.2375(5) = 91.43$

Section 12-2

12-13. $n = 10$, $k = 2$, $p = 3$, $\alpha = 0.05$

$$H_0: \beta_1 = \beta_2 = \dots = \beta_k = 0$$

$$H_1: \beta_j \neq 0 \quad \text{for at least one } j$$

$$SS_T = 371595.6 - \frac{(1916)^2}{10} = 4490$$

$$\mathbf{X}'\mathbf{y} = \begin{bmatrix} \sum y_i \\ \sum x_{i1}y_i \\ \sum x_{i2}y_i \end{bmatrix} = \begin{bmatrix} 1030 \\ 21310 \\ 44174 \end{bmatrix}$$

$$\hat{\mathbf{a}}'\mathbf{X}'\mathbf{y} = [171.054 \ 3.713 \ -1.126] \begin{bmatrix} 1916 \\ 43550.8 \\ 104736.8 \end{bmatrix} = 371535.9$$

$$SS_R = 371535.9 - \frac{1916^2}{10} = 4430.38$$

$$SS_E = SS_T - SS_R = 4490 - 4430.38 = 59.62$$

$$f_0 = \frac{\frac{SS_R}{k}}{\frac{SS_E}{n-p}} = \frac{4430.38/2}{59.62/7} = 260.09$$

$$f_{0.05,2,7} = 4.74$$

$$f_0 > f_{0.05,2,7}$$

Reject H_0 and conclude that the regression model is significant at $\alpha = 0.05$.

b) $H_0: \beta_1 = 0 \quad \beta_2 = 0$

$$\begin{aligned}
H_1 : \beta_1 &\neq 0 & \beta_2 &\neq 0 \\
t_0 &= \frac{\hat{\beta}_1}{se(\hat{\beta}_1)} & t_0 &= \frac{\hat{\beta}_2}{se(\hat{\beta}_2)} \\
&= \frac{3.713}{0.1934} = 19.20 & &= \frac{-1.126}{0.0861} = -13.08
\end{aligned}$$

$$t_{\alpha/2,7} = t_{.025,7} = 2.365$$

Reject H_0
Both regression coefficients are significant

$$12-17. \quad a) H_0 : \beta_1 = \beta_6 = 0$$

$$H_1 : \text{at least one } \beta \neq 0$$

$$f_0 = 53.3162$$

$$f_{\alpha,2,22} = f_{.05,2,22} = 3.44$$

$$f_0 > f_{\alpha,2,22}$$

Reject H_0 and conclude regression model is significant at $\alpha = 0.05$

$$b) H_0 : \beta_1 = 0$$

$$H_1 : \beta_1 \neq 0$$

$$t_0 = -8.59$$

$$t_{.025,25-3} = t_{.025,22} = 2.074$$

$|t_0| > t_{\alpha/2,22}$, Reject H_0 and conclude β_1 is significant at $\alpha = 0.05$

$$H_0 : \beta_6 = 0$$

$$H_1 : \beta_6 \neq 0$$

$$\alpha = 0.05$$

$$t_0 = 1.5411$$

$|t_0| \not> t_{\alpha/2,22}$, Do not reject H_0 , conclude that evidence is not significant to state β_6 is significant at $\alpha = 0.05$.

No, only x_1 contributes significantly to the regression.

$$12-21. \quad a) H_0 : \beta_1 = \beta_2 = \beta_{12} = 0$$

$$H_I : \text{at least one } \beta_j \neq 0$$

$$\alpha = 0.05$$

$$f_0 = 67.92$$

$$f_{\alpha,3,2} = f_{.05,3,2} = 19.16$$

$$f_0 \not> f_{\alpha,3,2}$$

Reject H_0

$$b) H_0 : \beta_{12} = 0$$

$$H_1 : \beta_{12} \neq 0$$

$$\alpha = 0.05$$

$$SSR(\beta_{12} | \beta_1, \beta_2) = 29951.4 - 29787 = 164.4$$

$$f_0 = \frac{SSR}{MS_E} = \frac{164.4}{153} = 1.07$$

$$f_{.05,1,2} = 18.51$$

$$f_0 > f_{\alpha,1,2}$$

Do not reject H_0

c) $\hat{\sigma}^2 = 147.0$

$$\hat{\sigma}^2 \text{ (no interaction term)} = 153.0$$

$MS_E(\hat{\sigma}^2)$ was reduced in the interaction term model due to the addition of this term.

12-23. a) $H_0 : \beta_1 = \beta_2 = \beta_3 = 0$ for all j

$$H_1 : \beta_j \neq 0 \quad \text{for at least one } j$$

$$f_0 = 840.55$$

$$f_{.05,3,16} = 3.24$$

$$f_0 > f_{\alpha,3,16}$$

Reject H_0 and conclude regression is significant at $\alpha = 0.05$

b) $\alpha = 0.05 \quad t_{\alpha/2,n-p} = t_{.025,16} = 2.12$

$$H_0 : \beta_1 = 0 \quad \beta_2 = 0 \quad \beta_3 = 0$$

$$H_1 : \beta_1 \neq 0 \quad \beta_2 \neq 0 \quad \beta_3 \neq 0$$

$$t_0 = -2.637 \quad t_0 = 1.91 \quad t_0 = 13.9$$

$$|t_0| > t_{\alpha/2,16} \quad |t_0| > t_{\alpha/2,16} \quad |t_0| > t_{\alpha/2,16}$$

Reject H_0 Do not reject H_0 Reject H_0

Sections 12-3 and 12-4

12-27. a) $-0.00657 \leq \beta_8 \leq -0.00122$

b) $\sqrt{\hat{\sigma}^2 \mathbf{x}_0' (\mathbf{X}' \mathbf{X})^{-1} \mathbf{x}_0} = 0.497648 = se(\hat{\mu}_{Y|x_0})$

c) $\hat{\mu}_{Y|x_0} = -7.63449 + 0.00398(2000) + 0.24777(60) - 0.00389(1800) = 8.19$

$$\hat{\mu}_{Y|x_0} \pm t_{.025,24} se(\hat{\mu}_{Y|x_0})$$

$$8.19 \pm (2.064)(0.497648)$$

$$8.19 \pm 1.03$$

$$7.16 \leq \mu_{Y|x_0} \leq 9.22$$

12-29. a) 95 % CI on coefficients

$$\beta_1 \pm t_{\alpha/2,n-p} (\hat{\beta}_1)$$

$$0.0972 \leq \beta_1 \leq 1.4174$$

$$-1.9646 \leq \beta_2 \leq 17.0026$$

$$-1.7953 \leq \beta_3 \leq 6.7613$$

$$-1.7941 \leq \beta_4 \leq 0.8319$$

b) $\hat{\mu}_{Y|x_0} = 290.44$ $se(\hat{\mu}_{Y|x_0}) = 7.61$ $t_{.025,7} = 2.365$

$$\hat{\mu}_{Y|x_0} \pm t_{\alpha/2,n-p} se(\hat{\mu}_{Y|x_0})$$

$$290.44 \pm (2.365)(7.61)$$

$$272.44 \leq \mu_{Y|x_0} \leq 308.44$$

c) $\hat{y}_0 \pm t_{\alpha/2,n-p} \sqrt{\hat{\sigma}^2 (1 + \mathbf{x}_0' (\mathbf{X}'\mathbf{X})^{-1} \mathbf{x}_0)}$

$$290.44 \pm 2.365(14.038)$$

$$257.25 \leq y_0 \leq 323.64$$

12-31 a) 95% Confidence Interval on coefficients

$$-0.595 \leq \beta_2 \leq 0.535$$

$$0.229 \leq \beta_3 \leq 0.812$$

$$-0.216 \leq \beta_4 \leq 0.013$$

$$-7.982 \leq \beta_5 \leq 2.977$$

b) $\hat{\mu}_{Y|x_0} = 8.99568$ $se(\hat{\mu}_{Y|x_0}) = 0.472445$ $t_{.025,14} = 2.145$

$$\hat{\mu}_{Y|x_0} \pm t_{\alpha/2,n-p} se(\hat{\mu}_{Y|x_0})$$

$$8.99568 \pm (2.145)(0.472445)$$

$$7.982 \leq \mu_{Y|x_0} \leq 10.009$$

c) $y_0 = 8.99568$ $se(\hat{y}_0) = 1.00121$

$$8.99568 \pm 2.145(1.00121)$$

$$6.8481 \leq y_0 \leq 11.143$$

12-35. a) $0.3882 \leq \beta_{pts} \leq 0.5998$

b) $\hat{y} = -5.767703 + 0.496501x_{pts}$

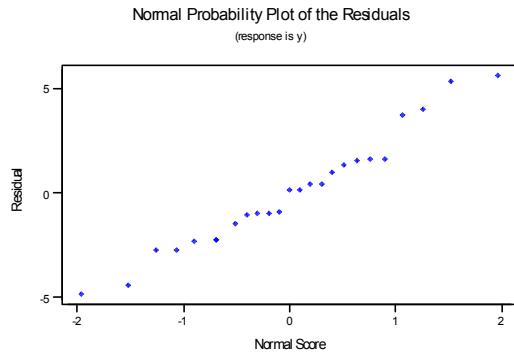
c) $0.4648 \leq \beta_{pts} \leq 0.5282$

d) The simple linear regression model has the shorter interval. Yes, the simple linear regression model in this case is preferable.

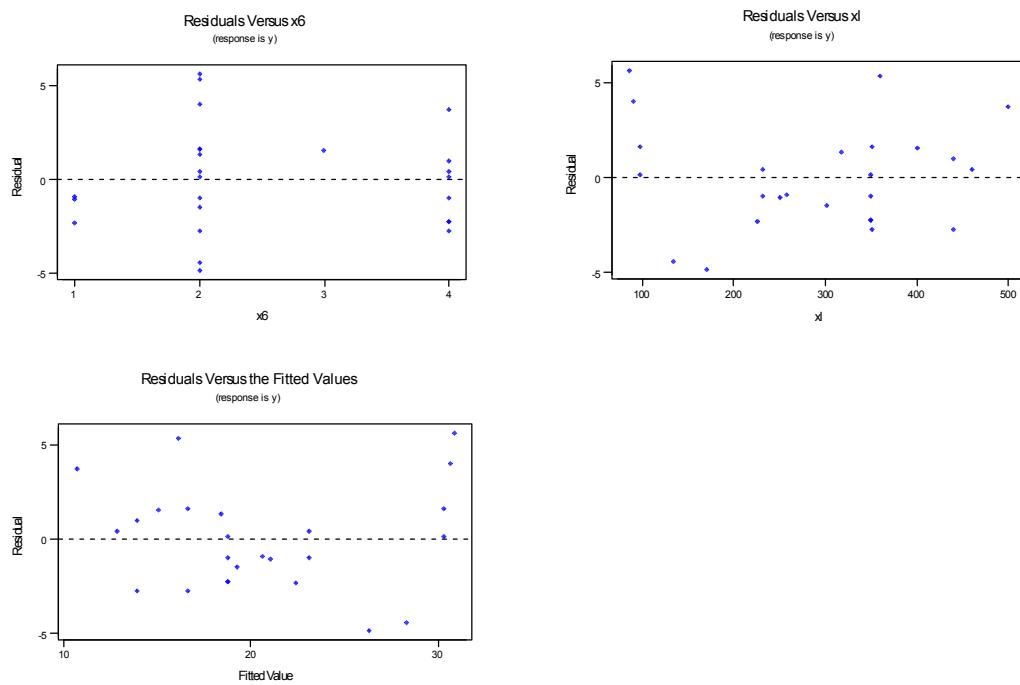
Section 12-5

12-37. a) $r^2 = 0.82897$

b) Normality assumption appears valid.



c) Assumption of constant variance appears reasonable.



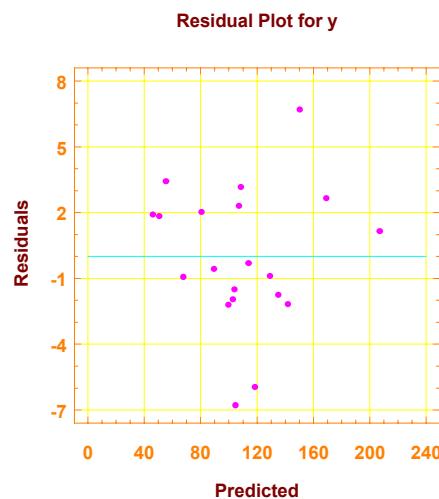
d) Yes, observations 7, 10, and 18

12-39. a) $r^2 = 0.985$

b) $r^2 = 0.990$

r^2 increases with addition of interaction term. No, adding additional regressor will always increase r^2

12-41 a) There is some indication of nonconstant variance since the residuals appear to “fan out” with increasing values of y .



b)

Source	Sum of Squares	DF	Mean Square	F-Ratio	P-value
Model	30531.5	3	10177.2	840.546	.0000
Error	193.725	16	12.1078		
Total (Corr.)	30725.2	19			
R-squared = 0.993695					Stnd. error of est. = 3.47963
R-squared (Adj. for d.f.) = 0.992513					Durbin-Watson statistic = 1.77758

$$R^2 = 0.9937 \text{ or } 99.37\%;$$

$$R_{Adj}^2 = 0.9925 \text{ or } 99.25\%;$$

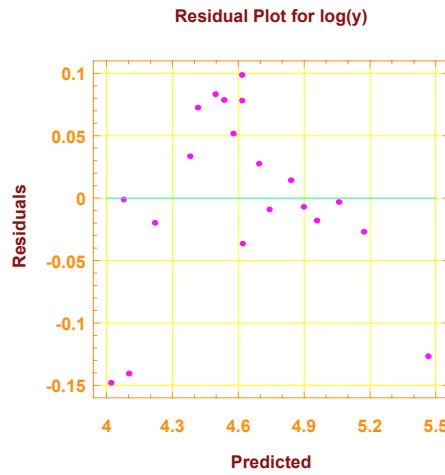
c)

Model fitting results for: log(y)

Independent variable	coefficient	std. error	t-value	sig.level
CONSTANT	6.22489	1.124522	5.5356	0.0000
x1	-0.16647	0.083727	-1.9882	0.0642
x2	-0.000228	0.005079	-0.0448	0.9648
x3	0.157312	0.029752	5.2875	0.0001
R-SQ. (ADJ.) = 0.9574	SE= 0.078919	MAE= 0.053775	DurbWat= 2.031	
Previously: 0.0000	0.000000	0.000000		0.000
20 observations fitted, forecast(s) computed for 0 missing val. of dep. var.				

$$\hat{y}^* = 6.22489 - 0.16647x_1 - 0.000228x_2 + 0.157312x_3$$

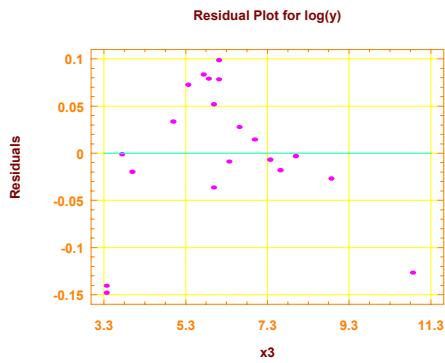
d)



Plot exhibits curvature

There is curvature in the plot. The plot does not give much more information as to which model is preferable.

e)



Plot exhibits curvature

Variance does not appear constant. Curvature is evident.

f)

Model fitting results for: log(y)

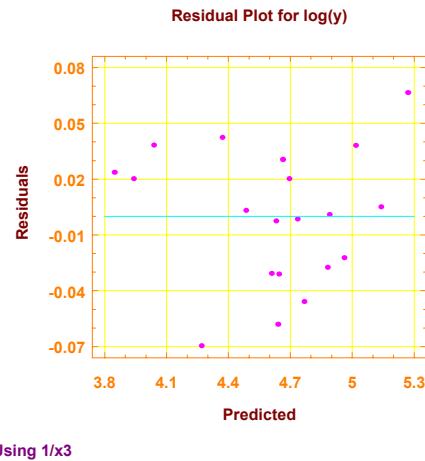
Independent variable	coefficient	std. error	t-value	sig.level
CONSTANT	6.222045	0.547157	11.3716	0.0000
x1	-0.198597	0.034022	-5.8374	0.0000
x2	0.009724	0.001864	5.2180	0.0001
1/x3	-4.436229	0.351293	-12.6283	0.0000

R-SQ. (ADJ.) = 0.9893 SE= 0.039499 MAE= 0.028896 DurbWat= 1.869
Previously: 0.9574 0.078919 0.053775 2.031
20 observations fitted, forecast(s) computed for 0 missing val. of dep. var.

Analysis of Variance for the Full Regression

Source	Sum of Squares	DF	Mean Square	F-Ratio	P-value
Model	2.75054	3	0.916847	587.649	.0000
Error	0.0249631	16	0.00156020		
Total (Corr.)	2.77550	19			

R-squared = 0.991006 Stnd. error of est. = 0.0394993
R-squared (Adj. for d.f.) = 0.98932 Durbin-Watson statistic = 1.86891

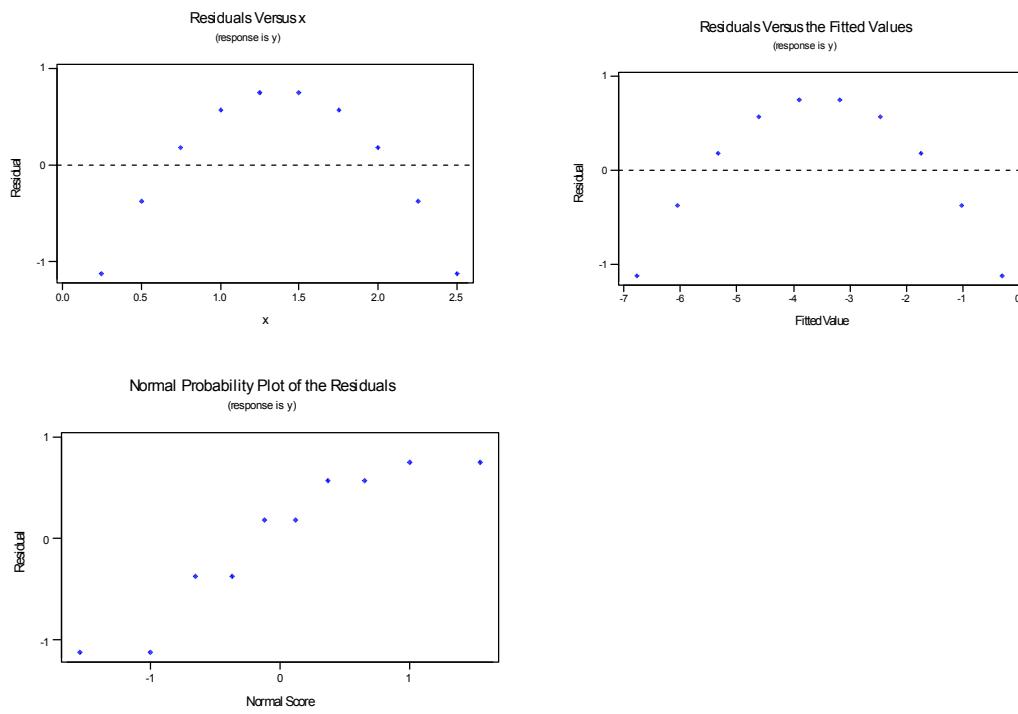


The residual plot indicates better conformance to assumptions.

Curvature is removed when using $1/x_3$ as the regressor instead of x_3 and the log of the response data.

Section 12-6

- 12-47. a) $\hat{y} = -1.633 + 1.232x - 1.495x^2$
 b) $f_0 = 1858613$, reject H_0
 c) $t_0 = -601.64$, reject H_0
 d) Model is acceptable, observation number 10 has large leverage.



12-49. $\hat{y} = 759.395 - 90.783x' - 47.166(x')^2$, where $x' = \frac{x - \bar{x}}{S_x}$

a) At $x = 285$ $x' = \frac{285 - 297.125}{11.9336} = -1.016$

$$\hat{y} = 759.395 - 90.783(-1.016) - 47.166(-1.016)^2 = 802.943 \text{ psi}$$

b) $\hat{y} = 759.395 - 90.783\left(\frac{x-297.125}{11.9336}\right) - 47.166\left(\frac{x-297.125}{11.9336}\right)^2$

$$\hat{y} = 759.395 - 7.607(x - 297.125) - 0.331(x - 297.125)^2$$

$$\hat{y} = -26204.14 + 189.09x - 0.331x^2$$

c) They are the same.

d) $\hat{y}' = 0.385 - 0.847x' - 0.440(x')^2$

where $y' = \frac{y - \bar{y}}{S_y}$ and $x' = \frac{x - \bar{x}}{S_x}$

The "proportion" of total variability explained is the same for both standardized and un-standardized models. Therefore, R^2 is the same for both models.

$$y' = \beta_0^* + \beta_1^*x' + \beta_{11}^*(x')^2 \quad \text{where } y' = \frac{y - \bar{y}}{S_y} \text{ and } x' = \frac{x - \bar{x}}{S_x} \quad y' = \beta_0^* + \beta_1^*x' + \beta_{11}^*(x')^2$$

12-51

a)	Predictor	Coef	SE Coef	T	P
	Constant	-1.769	1.287	-1.37	0.188
	x1	0.4208	0.2942	1.43	0.172
	x2	0.2225	0.1307	1.70	0.108
	x3	-0.12800	0.07025	-1.82	0.087
	x1x2	-0.01988	0.01204	-1.65	0.118
	x1x3	0.009151	0.007621	1.20	0.247
	x2x3	0.002576	0.007039	0.37	0.719
	x1^2	-0.01932	0.01680	-1.15	0.267
	x2^2	-0.00745	0.01205	-0.62	0.545
	x3^2	0.000824	0.001441	0.57	0.575

S = 0.06092 R-Sq = 91.7% R-Sq(adj) = 87.0%

Analysis of Variance

Source	DF	SS	MS	F	P
Regression	9	0.655671	0.072852	19.63	0.000
Residual Error	16	0.059386	0.003712		
Total	25	0.715057			

$$\begin{aligned} \hat{y} = & -1.769 + 0.421x_1 + 0.222x_2 - 0.128x_3 - 0.02x_1x_2 + 0.009x_1x_3 + \\ & 0.003x_2x_3 - 0.019x_1^2 - 0.007x_2^2 + 0.001x_3^2 \end{aligned}$$

b) $H_0 : \beta_1 = \beta_2 = \beta_3 = K = \beta_{33} = 0$

H_1 : at least 1 $\beta_j \neq 0$

$f_0 = 19.628$

$f_{.05, 9, 16} = 2.54$

$f_0 > f_{\alpha, 9, 16}$

Reject H_0 and conclude that the model is significant at $\alpha = 0.05$

c) Model is acceptable.

d) $H_0 : \beta_{11} = \beta_{22} = \beta_{33} = \beta_{12} = \beta_{13} = \beta_{23} = 0$

H_1 : at least one $\beta_{jj} \neq 0$

$$f_0 = \frac{SS_R(\beta_{11}, \beta_{22}, \beta_{33}, \beta_{12}, \beta_{13}, \beta_{23} | \beta_1, \beta_2, \beta_3, \beta_0) / r}{MS_E} = \frac{\frac{0.0359}{6}}{0.003712} = 1.612$$

$$f_{.05,6,16} = 2.74$$

$$f_0 > f_{.05,6,16}$$

Do not reject H_0

$$SS_R(\beta_{12}, \beta_{13}, \beta_{23}, \beta_{11}, \beta_{22}, \beta_{33} | \beta_1, \beta_2, \beta_3, \beta_0) = SS_R(\beta_1, \beta_2, \beta_3, \beta_{12}, \beta_{13}, \beta_{23}, \beta_{11}, \beta_{22}, \beta_{33} | \beta_0) -$$

$$\begin{aligned} SS_R(\beta_1 \beta_2 \beta_3 | \beta_0) \\ = 0.65567068 - 0.619763 \\ = 0.0359 \end{aligned}$$

$$\text{Reduced Model: } y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3$$

- 12-55. a) The min. MS_E equation is $x_1, x_2, x_3, x_4, x_5, x_6, x_7, x_8$

$$MS_E = 6.58 \quad C_p = 5.88$$

The min. C_p x_5, x_8, x_{10}

$$C_p = 5.02 \quad MS_E = 7.97$$

b) $\hat{y} = 34.434 - 0.048x_1$

$$MS_E = 8.81 \quad C_p = 5.55$$

c) Same as part b.

d) $\hat{y} = 0.341 + 2.862x_5 + 0.246x_8 - 0.010x_{10}$

$$MS_E = 7.97 \quad C_p = 5.02$$

e) Minimum C_p and backward elimination result in the same model. Stepwise and forward selection result in the same model. Because it is much smaller, the minimum C_p model seems preferable.

- 12-61. a) Min. C_p

$$\hat{y} = -3.517 + 0.486x_1 - 0.156x_9$$

$$C_p = -1.67$$

b) Min MS_E model is x_1, x_7, x_9 , $MS_E = 1.67$, $C_p = -0.77$

$$y = -5.964 + 0.495x_1 + 0.025x_7 - 0.163x_9$$

c) Max. adjusted R^2 model is x_1, x_7, x_9 , Adj. $R^2 = 0.98448$ Yes, same as Min. MS_E model.

Supplemental Exercises

- 12-65. a) $H_0 : \beta_3^* = \beta_4 = \beta_5 = 0$

$$H_1 : \beta_j \neq 0 \quad \text{for at least one } j$$

$$\alpha = 0.01$$

$$f_0 = 1323.62$$

$$f_{01,3,36} = 4.38$$

$$f_0 >> f_{\alpha,3,36}$$

Reject H_0 and conclude regression is significant.

P-value < 0.00001

b) $\alpha = 0.01$	$t_{0.005,36} = 2.72$	
$H_0: \beta_3^* = 0$	$H_0: \beta_4 = 0$	$H_0: \beta_5 = 0$
$H_1: \beta_3^* \neq 0$	$H_1: \beta_4 \neq 0$	$H_1: \beta_5 \neq 0$
$t_0 = -1.32$	$t_0 = 19.97$	$t_0 = 2.48$
$ t_0 > t_{\alpha/2,36}$	$ t_0 > t_{\alpha/2,36}$	$ t_0 > t_{\alpha/2,36}$
Do not reject H_0	Reject H_0	Do not reject H_0
Only regressor x_4 is significant		

c) Curvature is evident in the residuals vs. regressor plots from this model.

12-67. a) $\hat{y} = -0.908 + 5.482x_1^* + 1.126x_2^* - 3.920x_3^* - 1.143x_4^*$

b) $H_0: \beta_1 = \beta_2 = \beta_3 = \beta_4 = 0$
 $H_1: \beta_j \neq 0 \quad \text{for at least one } j$

$\alpha = 0.05$

$f_0 = 109.02$

$f_{0.05,4,19} = 2.90$

$f_0 >> f_{\alpha,4,19}$

Reject H_0 and conclude regression is significant at $\alpha = 0.05$.

$\alpha = 0.05$	$t_{0.025,19} = 2.093$		
$H_0: \beta_1 = 0$	$H_0: \beta_2 = 0$	$H_0: \beta_3 = 0$	$H_0: \beta_4 = 0$
$H_1: \beta_1 \neq 0$	$H_1: \beta_2 \neq 0$	$H_1: \beta_3 \neq 0$	$H_1: \beta_4 \neq 0$
$t_0 = 11.27$	$t_0 = 14.59$	$t_0 = -6.98$	$t_0 = -8.11$
$ t_0 > t_{\alpha/2,19}$			
Reject H_0	Reject H_0	Reject H_0	Reject H_0

c) The residual plots are more pleasing than those in Exercise 12-66.

12-69. a) $\hat{y} = -3982.1 + 1.0964x_1 + 0.1843x_3 + 3.7456x_4 + 0.8343x_5 - 16.2781x_6$

$MS_E(p) = 694.93 \quad C_p = 5.62$

b) $\hat{y} = -4280.2 + 1.442x_1 + 0.209x_3 + 0.6467x_5 - 17.5103x_6$

$MS_E(p) = 714.20 \quad C_p = 5.57$

c) Same as model b.

d) Models from parts b. and c. are identical. Model in part a. is the same with x_4 added in.

MS_E model in part a. = 694.93 $C_p = 5.62$

MS_E model in parts b.&c. = 714.20 $C_p = 5.57$

12-71. a) $VIF(\hat{\beta}_3^*) = 51.86$

$VIF(\hat{\beta}_4) = 9.11$

$VIF(\hat{\beta}_5) = 28.99$

* Yes, VIFs for X_3^* and X_5 exceed 10.

b) Model from Exercise 12-65: $\hat{y} = 19.69 - 1.27x_3^* + 0.005x_4 + 0.0004x_5$

12-73. a) $R^2 = \frac{SS_R}{SS_T}$

$$SS_R = R^2(SS_T) = 0.94(0.50) = 0.47$$

$$SS_E = SS_T - SS_R = 0.5 - 0.47 = 0.03$$

$$H_0: \beta_1 = \beta_2 = \dots = \beta_6 = 0$$

$$H_1: \beta_j \neq 0 \text{ for at least one } j.$$

$$\alpha = 0.05$$

$$f_0 = \frac{SS_R/k}{SS_E/(n-p)} = \frac{0.47/6}{0.03/7} = 18.28$$

$$f_{0.05,6,7} = 3.87$$

$$f_0 > f_{\alpha,6,7}$$

Reject H_0 .

b) $k = 5 \quad n = 14 \quad p = 6 \quad R^2 = 0.92$

$$SS_R' = R^2(SS_T) = 0.92(0.50) = 0.46$$

$$SS_E' = SS_T - SS_R' = 0.5 - 0.46 = 0.04$$

$$SS_R(\beta_j, \beta_{i,i=1,2,\dots,6,i \neq j} | \beta_0) = SS_R(\text{full}) - SS_R(\text{reduced})$$

$$= 0.47 - 0.46$$

$$= 0.01$$

$$f_0 = \frac{SS_R(\beta_j | \beta_{i,i=1,2,\dots,6,i \neq j} | \beta_0)/r}{SS_E'/(n-p)} = \frac{0.01/1}{0.04/8} = 2$$

$$f_{0.05,1,8} = 5.32$$

$$f_0 > f_{\alpha,1,8}$$

Do not reject H_0 and conclude that the evidence is insufficient to claim that the removed variable is significant at $\alpha = 0.05$

c) $MS_E(\text{reduced}) = \frac{SS_E}{n-p} = \frac{0.04}{8} = 0.005$

$$MS_E(\text{full}) = \frac{0.03}{7} = 0.00429$$

No, the MS_E is larger for the reduced model, although not by much. Generally, if adding a variable to a model reduces the MS_E it is an indication that the variable may be useful in explaining the response variable. Here the decrease in MS_E is not very great because the added variable had no real explanatory power.