

## CHAPTER 12: Multiple Regression

### 12.7 [LO 1]

- a. The plots show a linear (or somewhat linear) relationship between Price & Demand, IndPrice & Demand, PriceDiff & Demand, and AdvExp & Demand.
- b. The mean demand for the large size bottle of Fresh when the price of Fresh is \$3.70, the average industry price of competitors' similar detergents is \$3.90 and the advertising expenditure to promote Fresh is 6.50 (\$650,000).
- c.  $\beta_0$  = meaningless in practical terms  
 $\beta_1$  = the mean change in demand for each additional dollar in the price of Fresh holding all other predictor variables constant.  
 $\beta_2$  = the mean change in demand for each additional dollar in the average price of competitors' detergents holding all other predictor variables constant.  
 $\beta_3$  = the mean change in demand for each additional \$100,000 spent on advertising Fresh holding all other predictor variables constant.  
 $\varepsilon$  = all other factors that influence the demand for Fresh detergent
- d. The plots for Demand vs. AdvExp and Demand vs. PriceDif appear to be more linear than the other two plots.

### 12.15 [LO 4, 5]

- a.  $SSE = 1.4318$ ,  $s^2 = 1.4317/(30-4) = .0551$
- b. Total variation = 13.4586  
Explained variation = 12.0268
- c.  $R^2 = .894 = 12.0268/13.4586$   
Adjusted  $R^2 = .881 = (.894 - (3/29))(29/26)$   
Approximately 89% of the variance in demand is predicted by price, average price, and advertising, which drops to 88% when adjusted for the number of predictors.
- d.  $F = MS_{\text{explained}}/MSE = 72.80$
- e. at  $<.05$ , model is significant, F-critical = 2.98
- f. at  $<.01$ , model is significant, F-critical = 4.64
- g. output p = .000000000000888

### 12.19 [LO 3]

- a.  $b_0 = 1946.8020$ ,  $b_1 = 0.0386$ ,  $b_2 = 1.0394$ ,  $b_3 = -413.7578$   
 $b_0$  = labour hours when x-ray = 0, bed days = 0, and length of stay = 0 which is probably meaningless as the hospital has no patients staying there.

$b_1$  = implies that labour hours increases 0.04 for each unit increase in x-rays, when bed days and length of stay remain constant (predicted change).

$b_2$  = implies that labour hours increases by 1.04 for each unit of increase in bed days when x-rays and length of stay remain constant (predicted change).

$b_3$  = implies that labour hours decreases by 413.76 when length of stay decreases by one unit and both x-rays and bed days remain constant (predicted change).

b.  $\hat{y} = 1946.802 + .0386 (56194) + 1.0394 (14077.88) - 413.7578 (6.89) = 15897.65$

c. Therefore, actual hours were  $17207.31 - 15896.25 = 1311.06$  hours greater than predicted.

**12.29 [LO 6]**

$y = 17207.31$  is above the upper limit of the interval [14906.2, 16886.3]; this y-value is unusually high.

**12.33 [S 12.8]**

The shorter interval is from the model using  $x_4$ . This model is better.

**12.35 [S 12.9]**

Multiply:  $x_1 x_2$

**12.37 [S 12.9]**

$$\hat{y} = -2.3497 + 2.3611x_1 + 4.1831x_2 - 0.3489x_1x_2$$

$x_1 = \text{radio / TV}$                        $x_2 = \text{print}$

a.  $x_2 = 1$ , slope = 2.0122;  $x_2 = 2$ , slope = 1.6633;  $x_2 = 3$ , slope = 1.3144;  $x_2 = 4$ , slope = 0.9655;  $x_2 = 5$ , slope = 0.6166.

These slopes are the estimated average sales volume increase (in units of \$10,000) for every \$1,000 increase in radio and tv ads.

b.  $x_1 = 1$ , slope = 3.8342;  $x_1 = 2$ , slope 3.4853;  $x_1 = 3$ , slope = 3.1364,  $x_1 = 4$ , 2.7875;  $x_1 = 5$ , slope = 2.4386.

These slopes are the estimated average sales volume increase (in units of \$10,000) for every \$1,000 increase in print ads.

c. The smallest print slope is bigger than the largest radio/tv slope.

12.39 [S 12.10]

An independent variable, the levels of which are defined by describing them.

12.41 [S 12.10]

The effect of the qualitative independent variable on the dependent variable.

12.45 [S 12.10]

a. No interaction since  $p$ -values are so large.

b.  $\hat{y} = 8.61178$  (861,178 bottles)

95% prediction interval = [8.27089, 8.95266]—slightly bigger

12.47 [S 12.11]

$(k - g)$  denotes the number of regression parameters we have set equal to zero in  $H_0$ .  
 $[n - (k + 1)]$  denotes the denominator degrees of freedom.

12.49 [S 12.11]

Model 3—complete

Model 1—reduced

$$H_0 : \beta_4 = \beta_5 = \beta_6 = \beta_7 = 0$$

$$F = \frac{\frac{1.4318 - .5347}{4}}{\frac{.5347}{22}} = 9.228$$

$F_{.05} = 2.82$  based on 4 and 22 degrees of freedom.

$F_{.01} = 4.31$  based on 4 and 22 degrees of freedom.

Since  $9.228 > 4.31$ , reject  $H_0$  at  $\alpha = .05$  and  $.01$ ; Because the null hypothesis was that the equations have the same slope and intercept, rejecting the  $H_0$  means that at least one of these claims is false.

12.51 [LO 6]

$$\hat{y} = 30,626 + 3.893(28000) - 29,607(1.56) + 86.52(1821.7) \cong 251,056$$

12.53 [LO 5]

a. Output for all:

SUMMARY OUTPUT                      All

<i>Regression Statistics</i>	
Multiple R	0.878394
R Square	0.771577
Adjusted R Square	0.754006
Standard Error	1.319372
Observations	29

ANOVA					
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	2	152.8786	76.43932	43.91193	4.61E-09
Residual	26	45.25929	1.740742		
Total	28	198.1379			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>	<i>Lower 95%</i>	<i>Upper 95%</i>	<i>Lower 95.0%</i>	<i>Upper 95.0%</i>
Intercept	16.94219	1.435079	11.80575	6.01E-12	13.99234	19.89204	13.99234	19.89204
Age(x1)	-0.00066	0.013029	-0.05035	0.96023	-0.02744	0.026126	-0.02744	0.026126
Price(x2)	-0.05548	0.006086	-9.11638	1.4E-09	-0.06799	-0.04297	-0.06799	-0.04297

Significant regression model. Price is the only significant predictor.

**b. Outputs:**

SUMMARY OUTPUT                      Males

<i>Regression Statistics</i>	
Multiple R	0.817165
R Square	0.667758
Adjusted R Square	0.607351
Standard Error	1.408836
Observations	14

ANOVA					
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	2	43.88127	21.94063	11.05422	0.002333
Residual	11	21.83302	1.98482		
Total	13	65.71429			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>	<i>Lower 95%</i>	<i>Upper 95%</i>	<i>Lower 95.0%</i>	<i>Upper 95.0%</i>
Intercept	13.23223	2.361521	5.603267	0.00016	8.034556	18.42991	8.034556	18.42991
Age(x1)	-0.07728	0.035932	-2.15066	0.054585	-0.15636	0.001808	-0.15636	0.001808
Price(x2)	-0.01943	0.017359	-1.11942	0.286809	-0.05764	0.018774	-0.05764	0.018774

SUMMARY OUTPUT Females

<i>Regression Statistics</i>	
Multiple R	0.948477
R Square	0.899609
Adjusted R Square	0.882877
Standard Error	1.000841
Observations	15

ANOVA					
	<i>df</i>	<i>SS</i>	<i>MS</i>	<i>F</i>	<i>Significance F</i>
Regression	2	107.7131	53.85657	53.7661	1.02E-06
Residual	12	12.02019	1.001683		
Total	14	119.7333			

	<i>Coefficients</i>	<i>Standard Error</i>	<i>t Stat</i>	<i>P-value</i>	<i>Lower 95%</i>	<i>Upper 95%</i>	<i>Lower 95.0%</i>	<i>Upper 95.0%</i>
Intercept	13.83504	4.13341	3.347125	0.005811	4.829112	22.84097	4.829112	22.84097
X Variable 1	0.028102	0.02989	0.940197	0.365658	-0.03702	0.093226	-0.03702	0.093226
X Variable 2	-0.04714	0.013665	-3.44982	0.004807	-0.07691	-0.01737	-0.07691	-0.01737

Models are significant for both men and women. For men, Age has a slight negative relationship with interest (younger more interested,  $p < .10$ ). For women, Price is the significant predictor (greater interest with lower prices,  $p < .01$ ).

12.55 [S 12.9]

- a. Interaction term is not a significant predictor ( $p > .10$ ).
- b. Introducing the interaction term decreases the F-value but increases the Multiple R slightly.

12.57 [LO 5]

- a.  $\beta_5$ :  $b_5 = 0.2137$ , Confidence Interval =  $[0.0851, 0.3423]$ ,  $p$ -value = .0022, significant at 0.01 but not 0.001 so we have very strong evidence.

$\beta_5$ :  $b_6 = 0.3818$ , Confidence Interval =  $[0.2551, 0.5085]$ ,  $p$ -value  $< .001$ , significant at 0.001 so we have extremely strong evidence.

- b.  $b_6 = .1681$  Confidence Interval:  $[-.0363, .29]$ ,  $p$ -value = .0147, strong evidence.

c. 
$$\begin{aligned} \mu_{[d, a, C]} - \mu_{[d, a, A]} &= [\beta_0 + \beta_1 d + \beta_2 a + \beta_3 a^2 + \beta_4 da + \beta_5(0) + \beta_6(1) + \beta_7 a(0) + \beta_8 a(1)] \\ &\quad - [\beta_0 + \beta_1 d + \beta_2 a + \beta_3 a^2 + \beta_4 da + \beta_5(0) + \beta_6(0) + \beta_7 a(0) + \beta_8 a(0)] \\ &= \beta_6 + \beta_8 a \\ &= -.9351 + .2035(6.2) = .3266 \\ &= -.9351 + .2035(6.6) = .408 \end{aligned}$$

$$\begin{aligned} \mu_{[d, a, C]} - \mu_{[d, a, B]} &= [\beta_0 + \beta_1 d + \beta_2 a + \beta_3 a^2 + \beta_4 da + \beta_5(0) + \beta_6(1) + \beta_7 a(0) + \beta_8 a(1)] \\ &\quad - [\beta_0 + \beta_1 d + \beta_2 a + \beta_3 a^2 + \beta_4 da + \beta_5(1) + \beta_6(0) + \beta_7 a(1) + \beta_8 a(0)] \\ &= \beta_6 + \beta_8 a - \beta_5 - \beta_7 a \\ &= \beta_6 - \beta_5 + \beta_8 a - \beta_7 a \\ &= -.9351 - (-.4807) + .2035(6.2) - .1072(6.2) = .14266 \\ &= -.9351 - (-.4807) + .2035(6.6) - .1072(6.6) = .18118 \end{aligned}$$

Both differences increased with the larger value of  $a$ .

d. The prediction interval for the third model is slightly shorter.

The differences between campaign A and campaigns B & C change as volume level changes.