

Unless stated otherwise, for all significance tests, use $\alpha = 0.05$ significance level.

Q.1. A regression model was fit, relating estimated cost of de-commissioning oil platforms (Y, in millions of \$) to 2 predictors: Total number of piles/legs (X_1) and water depth (X_2 , in 100s of feet). The model was fit, based on $n = 17$ oil platforms. Consider the models:

Model 1: $E\{Y\} = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_{11} X_1^2 + \beta_{22} X_2^2 + \beta_{12} X_1 X_2$ Model 2: $E\{Y\} = \beta_0 + \beta_1 X_1 + \beta_2 X_2$

ANOVA	Model 1		ANOVA	Model 2	
	df	SS		df	SS
Regression	5	7397	Regression	2	7163
Residual	11	551	Residual	14	785
Total	16	7948	Total	16	7948
Coefficients			Coefficients		
		Standard Error			Standard Error
Intercept	-20.399	21.040	Intercept	-26.064	5.335
totpiles	-0.357	0.821	totpiles	0.957	0.274
wtrdph	6.155	7.048	wtrdph	4.401	1.095
tp2	0.047	0.046			
wtrdph2	-0.077	0.598			
tp*wd	-0.070	0.233			

p.1.a. Test whether the linear (main effects) model is appropriate. $H_0: \beta_{11} = \beta_{22} = \beta_{12} = 0$

⑧ $F_{obs} = \frac{\frac{785 - 551}{14 - 11}}{\frac{551}{11}} = \frac{234(11)}{3(551)} = \frac{2574}{1653} \approx 1.557$

RR: $F_{obs} \geq F_{.05, 3, 11} = 3.587$

③ Do not reject H_0 . ②

p.1.b. For each model, obtain the predicted value and residual for rig 17 ($Y = 78.5, X_1 = 44, X_2 = 13$)

Model 1: $\hat{Y} = -20.399 - 0.357(44) + 6.155(13) + 0.047(44^2) - 0.077(13^2) - 0.070(44)(13)$
 $= 81.85 \Rightarrow e_1 = 78.5 - 81.85 = -3.35$ ②

Model 2: $\hat{Y} = -26.064 + 0.957(44) + 4.401(13) = 73.26$ ②
 $\Rightarrow e_2 = 78.5 - 73.26 = 5.24$ ②

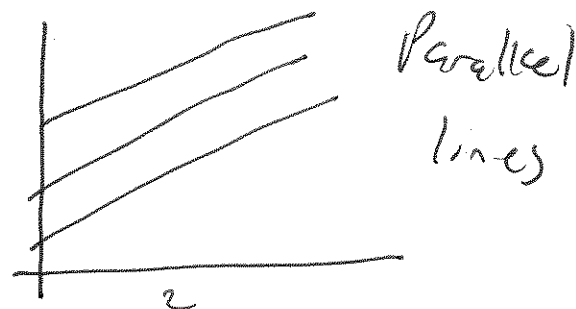
Model 1: Predicted = 81.85 Residual = -3.35 Model 2: Predicted = 73.26 Residual = 5.24

Q.2. A study related subsidence rate (Y) to water table depth (X_1) for 3 crops: pasture ($X_2 = 0, X_3 = 0$), truck crop ($X_2 = 1, X_3 = 0$), and sugarcane ($X_2 = 0, X_3 = 1$). Note the total sum of squares is TSS = 35.686, and $n = 24$.

p.2.a. Give a model that allows **separate intercepts** for each crop type, with a **common slope for water table depth** among crop types. Sketch the graph for this model. For this model, SSE = 1.853. Give the error degrees of freedom.

$$4 \quad E\{Y\} = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3$$

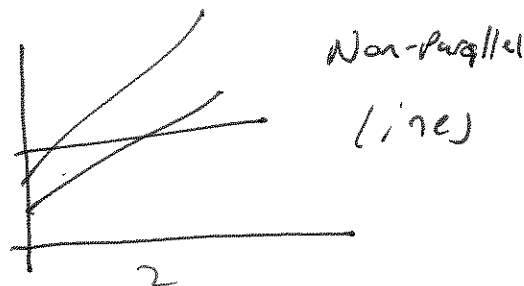
$$3 \quad df_e = 24 - 4 = 20$$



p.2.b. Give a model that allows **separate intercepts** for among crop types, with **separate slopes for water table depth** among crop types. Sketch the graph for this model. For this model, SSE = 1.261. Give the error degrees of freedom.

$$4 \quad E\{Y\} = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + \beta_{12} X_1 X_2 + \beta_{13} X_1 X_3$$

$$3 \quad df_e = 24 - 6 = 18$$



p.2.c. Test the null hypothesis that the (simpler) model in p.2.a. is appropriate. That is, the extra parameters in the second model are not significantly different from 0.

$$F_{obs} = \frac{\frac{1.853 - 1.261}{20 - 18}}{\frac{1.261}{18}} = \frac{18(0.592)}{2(1.261)} = 8.225$$

$$RN = F_{obs} \geq F_{.05, 2, 18} = 3.555$$

Reject H_0

p.2.d. For the model in p.2.a., compute R^2

$$1 - \frac{1.853}{35.686} = 1 - .052 = .948$$

Q.3. A study investigated meteorological effects on condition of wheat yield in Ohio (Y), based on a series of $n = 24$ years. The predictors were: Average October/November temperature (X_1), September precipitation (X_2), October/November precipitation (X_3), and percent September sunshine (X_4). The best 1-, 2-, 3-, and 4-variable models (minimum SSE) are given below.

Model	p'	SSE	C_p	AIC	BIC
X3	2	1738	4.27	106.78	109.13
X2,X3	3	1531	3.37	105.73	109.27
X1,X2,X3	4	1395	3.48	105.50	110.21
X1,X2,X3,X4	5	1361	5.00	106.91	112.80

p.3.a. Complete the table. Use MSE of the full (X_1, X_2, X_3, X_4) models as the estimate of σ^2 when computing C_p

$$\textcircled{5} C_p = \frac{SSE}{MSE_c} + 2p' \approx n = \frac{1531}{71.63} + 2(3) \approx 24 = 3.37$$

$$\frac{1361}{19} = 71.63$$

$$\textcircled{5} AIC = n \ln\left(\frac{SSE}{n}\right) + 2p' = 24 \ln\left(\frac{1361}{24}\right) + 2(5) =$$

$$\textcircled{5} BIC = 24 \ln\left(\frac{1395}{24}\right) + \ln(24) \approx 110.21$$

p.3.b. Give the best model based on each criteria. C_p X_1, X_2, X_3 AIC X_1, X_2, X_3 BIC X_3

p.3.c. The following output gives the regression coefficients for the X_1, X_2, X_3 model. Give the fitted value and residual for the first year ($Y = 92, X_1 = 46, X_2 = 1.6, X_3 = 6.3$).

Coefficients	
Intercept	16.59
tempon.x1	1.06
rains.x2	2.38
rainon.x3	2.96

$$\hat{Y} = 16.59 + 1.06(46) + 2.38(1.6) + 2.96(6.3) = 87.8$$

Fitted Value 87.8 Residual $92 - 87.8 = 4.2$

p.3.d. For the model in p.3.c., we obtain:

$$\sum_{t=2}^{24} (e_t - e_{t-1})^2 = 3206 \quad d_L(n=24, p=3) = 1.10 \quad d_U(n=24, p=3) = 1.66 \quad DW = \frac{3206}{1395} = 2.30$$

Test H_0 : Errors are not autocorrelated versus H_A : Errors are autocorrelated

Circle the Best Answer Reject H_0 Accept H_0 Test is inconclusive

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Q.4. A simple linear regression model was fit relating Weight (Y, in pounds) to Height (X, in inches) for a random sample of n=52 National Hockey League players. The total sum of squares, TSS = 9237, and R² = 0.262.

p.4.a. Complete the following ANOVA table.

ANOVA					
Source	df	SS	MS	F	F(0.05)
Regression	1 ⁽²⁾	2420.0 ⁽¹⁾	2420.0 ⁽²⁾	17.75 ⁽³⁾	4.034 ⁽³⁾
Residual	50 ⁽²⁾	6816.7 ⁽¹⁾	136.3 ⁽³⁾	#N/A	#N/A
Total	51 51	9236.7	#N/A	#N/A	#N/A

$$\frac{SSR}{TSS} = 0.262 \Rightarrow SSR = 0.262(9236.7) =$$

p.4.b. Complete the following table and use it to conduct the F-test for Lack-of-Fit (there are c = 9 distinct heights).

$$H_0: \mu_j = \beta_0 + \beta_1 X_j \quad j = 1, \dots, c \quad H_A: \mu_j \neq \beta_0 + \beta_1 X_j$$

$$SSLF = \sum_{j=1}^c n_j (\bar{Y}_j - \hat{Y}_j)^2 \quad df_{LF} = c - 2 \quad SSPE = \sum_{j=1}^c (n_j - 1) S_j^2 \quad df_{PE} = n - c$$

Height	n	Y-bar	Y-hat	n*(YB-YH)	(n-1)S^2
70	2	193.00	190.31	14.42	128.00
71	6	198.50	194.12	114.95	785.50
72	7	197.86	197.93	0.04	330.86
73	13	199.15	201.74	86.87	1357.69
74	11	202.27	205.55	117.91	1782.18
75	6	211.50	209.35	27.61	433.50
76	4	219.50	213.16	160.65	1451.00
77	2	216.50	216.97	0.44	24.50
78	1	222.00	220.78	1.52	0.00
Sum	52	#N/A	#N/A	524.82	6293.23

$$F = \frac{SSLF / (c - 2)}{SSPE / (n - c)} = \frac{524.82 / 7}{6293.23 / 43} = \frac{43(524.82)}{7(6293.23)} = 0.522 \quad 0.513$$

Test Statistic: 0.522

Reject H₀ if Test Stat falls in Range: ≥ F_{0.05, 7, 43} ≈ 2.232

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(3) $\frac{40}{50} \frac{2.249}{2.199} \approx 2.232$

Q.5. A simple linear regression model is fit, relating Orlando June Total Precipitation (Y) to Mean Temperature (X) over an n = 45 year period. The following table gives the results.

ANOVA		
	df	SS
Regression	1	32.59982
Residual	43	489.2068
Total	44	521.8067

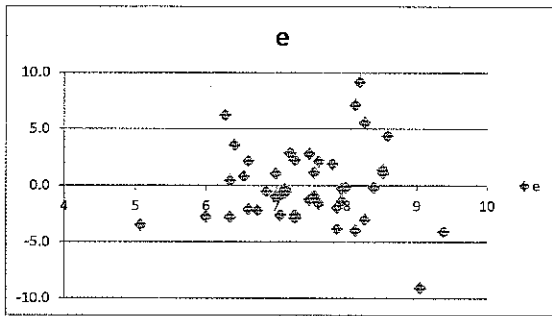
Coefficients and Standard Error		
Intercept	61.3280	31.8335
meanTemp	-0.6626	0.3915

p.5.a. Use the t-test to test $H_0: \beta_1 = 0$ vs $H_A: \beta_1 \neq 0$ at $\alpha = 0.10$ significance level

Test Statistic: $t = \frac{-0.6626}{0.3915} = -1.692$ (4)

Reject H_0 if Test Stat falls in range: $|t| \geq t_{.05, 43} \approx 1.681$ (3) $\frac{1.684}{1.676}$

A plot of the residuals displays a possible case of unequal variances. The regression of the squared residuals on X is given below:



	df	SS
Regression	1	1326.60
Residual	43	14436.63
Total	44	15763.24

p.5.b. Conduct the Breusch-Pagan test to test H_0 : Equal Variances H_A : Variance is related to X. (Use $\alpha = 0.05$):

$$\chi^2_{BP} = \frac{SSR_{reg}^2 / 2}{(SSE/n)^2} = \frac{1326.60/2}{\left(\frac{489.2068}{45}\right)^2} = \frac{663.30}{118.18} = 5.61$$
 (7)

Test Statistic: $\chi^2_{BP} = 5.61$ (3)

Reject H_0 if Test Statistic falls in the Range: $\geq \chi^2_{.05, 1} = 3.841$

Q.6. A multiple regression model is fit with 3 predictors and an intercept, based on a sample of $n=25$ observations. How large must $R^2/(1-R^2)$ be to reject $H_0: \beta_1 = \beta_2 = \beta_3 = 0$?

(8)

$$F = \frac{R^2/p}{1-R^2/(n-p)} = \frac{R^2}{1-R^2} \left(\frac{n-p}{p} \right) \geq F_{.05, 3, 21} = 3.072$$

$$\Rightarrow \frac{R^2}{1-R^2} \left(\frac{21}{3} \right) \geq 3.072 \Rightarrow \frac{R^2}{1-R^2} \geq \frac{3.072}{7} = .4389$$

Q.7. It is possible to fail to reject $H_{01}: \beta_1 = 0$ and $H_{02}: \beta_2 = 0$ based on t-tests in multiple linear regression model with $p > 2$ predictors, but still reject $H_{012}: \beta_1 = \beta_2 = 0$, controlling for the remaining $p-2$ predictors. **True** or **False**

(4)

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