

STAT 2509B4

Assignment 3

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1. Indicate whether or not each of the following models can be treated as an multiple linear regression (MLR) model:

(i) $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_1 x_2 + \epsilon$, can be treated as MLR

(ii) $y = (e^{\beta_0 + \beta_1 x_1 + \beta_2 x_2^2})\epsilon$, cannot be treated as MLR

(iii) $y = \beta_0 + \beta_1 x_1 + \beta_2 e^{x_1} + \epsilon$, can be treated as MLR

(iv) $y = \beta_0 + \beta_1 x_1 + \beta_2 x_1^2 + \beta_3 x_1^3 + \beta_4 x_2 + \epsilon$, can be treated as MLR

(v) $y = \beta_0 e^{\beta_1 x_1 + \beta_2 x_2} + \epsilon$, cannot be treated as MLR

2. A medical study was conducted to study the relationship between infants' systolic blood pressure and two explanatory variables, age (days) and weight (kg). The data for 25 infants are given below..:

Age (x_1)	Weight (x_2)	Systoli BP (y)
3	2.61	80
4	2.67	90
5	2.98	96
6	3.98	102
3	2.87	81
4	3.41	96
5	3.49	99
6	4.03	110
3	3.41	88
4	2.81	90
5	3.24	100
6	3.75	102
3	3.18	86
4	3.13	93
5	3.98	101
6	4.55	103
3	3.41	86
4	3.35	91
5	3.75	100
6	3.83	105
3	3.18	84
4	3.52	91
5	3.49	95
6	3.81	104
6	4.03	107

- (a) State all the assumptions that are necessary for the statistical inference under the MLR model.:

Model $y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \epsilon, n = 25$

Assumptions

- (1) X_1, X_2 are observed without error
- (2) ϵ 's are independently distributed
- (3) ϵ 's have common mean 0 in other words $E(\epsilon) = 0$ for all X_1, X_2 .
- (4) ϵ 's have common/constant variance σ^2 meaning $Var(\epsilon) = \sigma^2$ for all X_1, X_2
- (5) $\epsilon \sim N(0, \sigma^2)$ for any value of X_1, X_2

- (b) Use matrices to compute the least-squares estimates of the population parameters β_0, β_1 and β_2 , and obtain the fitted least-squares regression line:

Hint: $\mathbf{X}^T \mathbf{X} = \begin{bmatrix} 25.00 & 114.00 & 86.46 \\ 114.00 & 552.00 & 404.07 \\ 86.46 & 404.07 & 304.5062 \end{bmatrix}, \mathbf{X}^T \mathbf{Y} = \begin{bmatrix} 2380.00 \\ 11072.00 \\ 8306.16 \end{bmatrix},$

$(\mathbf{X}^T \mathbf{X})^{-1} \approx \begin{bmatrix} 2.3963567 & 0.11058177 & -0.8271483 \\ 0.1105818 & 0.06834592 & -0.1220909 \\ -0.8271483 & -0.12209090 & 0.4001512 \end{bmatrix},$

$\mathbf{Y}^T \mathbf{Y} = \sum_{i=1}^n y_i^2 = 228230$, and $\sum_{i=1}^n y_i = 2380$.

$$(\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{Y} = \begin{bmatrix} 2.3963567 & 0.11058177 & -0.8271483 \\ 0.1105818 & 0.06834592 & -0.1220909 \\ -0.8271483 & -0.12209090 & 0.4001512 \end{bmatrix} * \begin{bmatrix} 2380.00 \\ 11072.00 \\ 8306.16 \end{bmatrix} = \begin{bmatrix} 57.2642 \\ 5.80416 \\ 3.31649 \end{bmatrix} = \begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix} = \hat{\beta}$$

Therefore the fitted least squares regression line is $\hat{y} = 57.2642 + 5.80416x_1 + 3.31649x_2$

- (c) Set up the ANOVA table and test for significance of the model at the significance level of $\alpha = 0.05$

$$TSS = \mathbf{Y}^T \mathbf{Y} - \frac{(\sum_{i=1}^n y_i)^2}{n} = \sum_{i=1}^n y_i^2 - \frac{(\sum_{i=1}^n y_i)^2}{n} = 228230 - \frac{2380^2}{25} = 1654$$

$$SSR = \hat{\beta}^T (\mathbf{X}^T \mathbf{Y}) - \frac{(\sum_{i=1}^n y_i)^2}{n} = \begin{bmatrix} 57.2642 & 5.80416 & 3.31649 \end{bmatrix} * \begin{bmatrix} 2380.00 \\ 11072.00 \\ 8306.16 \end{bmatrix} - \frac{2380^2}{25} = 1523.752098$$

$$SSE = TSS - SSR = 1654 - 1523.752098 = 130.247902$$

$$MSR = \frac{SSR}{k} = \frac{1523.752098}{2} = 761.876$$

$$MSE = \frac{SSE}{n-(k+1)} = \frac{130.247902}{22} = 5.920359$$

$$F = \frac{MSR}{MSE} = \frac{761.876}{5.920359} = 128.687$$

Source	d.f	SS	MS	F
Regression	2	1523.752098	761.876	128.687
Error	22	130.247902	5.920359	
Total	24	1654		

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

$$H_a : \text{at least one of the } \beta_i \neq 0$$

$$\alpha = 0.05$$

$$\text{test-statistics: } F = \frac{MSR}{MSE} = 128.687$$

$$\text{Rejection region, we reject } H_0 \text{ if } F > F_{(k, n-(k+1)), \alpha} = F_{2, 22; 0.05} = 3.4434$$

Since $F = 128.687 > 3.4434$, we reject H_0 and conclude that at a 5% level of significance there is evidence to say there is a linear relationship between age, weight and the systolic BP.

- (d) Test whether age (x_1) contributes to explaining (or predicting) the systolic blood pressure (y) under the MLR model. Use t-test with $\alpha = 0.05$.

$$H_0 : \beta_1 = 0$$

$$H_a : \beta_1 \neq 0$$

$$\alpha \rightarrow \alpha/2 = 0.025$$

test statistics: $t = \frac{\hat{\beta}_1}{\sqrt{v_{11}MSE}} = \frac{5.80416}{\sqrt{0.0683459 \cdot 5.920359}} = 9.1245$

Rejection region, we reject H_0 if $|t| > t_{n-(k+1), \alpha/2} = t_{22, 0.025} = 2.07383$.

Since $t = 9.1245 > 2.07$, we reject H_0 and conclude that at a 5% level of significance there is evidence to say that the x_1 term contributes to the model.

- (e) **Find the values of the coefficient of determination, r^2 , and the adjusted r^2 . Interpret their meanings in this problem**

$$r^2 = \frac{SSR}{TSS} = \frac{1523.752098}{1654} = 0.9213 = 92.125\%$$

In other words approximately 92.13% of the total variation in the data is explained by the regression line. The rest is due to error.

$$r_{adj}^2 = 1 - \frac{SSE/n-(k+1)}{TSS/n-1} = 1 - \frac{MSE}{TSS/n-1} = 1 - \frac{5.920359}{1654/24} = 0.9141 = 91.41\%$$

Since both r^2 and r_{adj}^2 are quite high (above 80%), both have similar values around 90%, and since the x_1 term does contribute to the model, we can conclude that the model is good.

- (f) **Run SAS to verify your answers to the above questions. In addition, use the SAS output to answer subquestion (d) using the partial F-test with $\alpha = 0.05$. See attached SAS output**

$$H_0 : \beta_1 = 0$$

$$H_a : \beta_1 \neq 0$$

$$\alpha = 0.05$$

$$\text{full model: } y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \epsilon$$

$$\text{reduced model: } y = \beta_0 + \beta_2 x_2 + \epsilon$$

$$SSR_f = 1521.53295 \text{ with d.f.} = 2$$

$$SSE_f = 132.46705 \text{ with d.f.} = 22$$

$$SSR_r = 1028.63536 \text{ with d.f.} = 1$$

$$SSE_r = 625.36464 \text{ with d.f.} = 23$$

$$\text{test statistics: } F_{part} = \frac{[SSR_f - SSR_r]/[df_{SSR_f} - df_{SSR_r}]}{SSE_f/df_{SSE_f}} = \frac{(1521.53295 - 1028.63536)/(2-1)}{132.46705/22} = 81.85996$$

Rejection Region, we reject H_0 if $F_{part} > F_{(1,22);0.05} = 4.3009$

Since $F_{part} = 81.85996 > 4.3009$ we reject H_0 and conclude that at a 5% level of significance there is enough evidence to say that the X_1 (age) term contributes to the model.

3. **An experimenter wished to compare the potencies of three different drug products. To do this, 12 test tubes were inoculated with a culture of the virus under study and incubated for 2 days at 35C. Four dosage levels (0.2, 0.4, 0.8, and 1.6 mg per tube) were to be used from each of the three drug products (A, B and C), with only one dose-drug product combination for each of the 12 test-tube cultures. The data are shown in the following table**

Dose	Drug potency (y)		
	Drug A	Drug B	Drug C
0.2	2.0	1.8	1.3
0.4	4.3	4.1	2.0
0.8	6.5	4.9	2.8
1.6	8.9	5.7	3.4

Let

$$x_1 = \ln(\text{dose}), x_2 = \begin{cases} 1, & \text{if drug B} \\ 0, & \text{otherwise} \end{cases}, x_3 = \begin{cases} 1, & \text{if drug C} \\ 0, & \text{otherwise} \end{cases} \text{ and } y = \text{drug potency. Consider the following}$$

MLR model. $y = \beta_0 + \beta_1x_1 + \beta_2x_2 + \beta_3x_3 + \beta_4x_1x_2 + \beta_5x_1x_3 + \epsilon$. **Run SAS to test whether the 3 lines corresponding to the effects of the 3 drugs are parallel (i.e. whether these 3 lines have the same slope). Use $\alpha = 0.05$.**

Full model: $y = \beta_0 + \beta_1x_1 + \beta_2x_2 + \beta_3x_3 + \beta_4x_1x_2 + \beta_5x_1x_3 + \epsilon$

If drug A: $y = \beta_0 + \beta_1x_1 + \epsilon$

If drug B: $y = \beta_0 + \beta_1x_1 + \beta_2 + \beta_4x_1 + \epsilon = (\beta_0 + \beta_2) + (\beta_1 + \beta_4)x_1 + \epsilon$

If drug C: $y = \beta_0 + \beta_1x_1 + \beta_3 + \beta_5x_1 + \epsilon = (\beta_0 + \beta_3) + (\beta_1 + \beta_5)x_1 + \epsilon$

Test if any of the 3 drug lines are parallel or have the same slope:

$H_0 : \beta_4 = \beta_5 = 0$

$H_a : \text{at least one of the } \beta\text{'s} \neq 0.$

$\alpha = 0.05$

Reduced model $y = \beta_0 + \beta_1x_1 + \beta_2x_2 + \beta_3x_3 + \epsilon$

$SSR_f = 55.29350$ with d.f. = 5

$SSE_f = 0.68900$ with d.f. = 6

$SSR_r = 48.84417$ with d.f. = 3

$SSE_r = 7.13833$ with d.f. = 8

$$F_{part} = \frac{[SSR_f - SSR_r]/[df_{SSR_f} - df_{SSR_r}]}{SSE_f/df_{SSE_f}} = \frac{(55.29350 - 48.84417)/(5-3)}{0.68900/6} = 28.081$$

Rejection Region, we reject H_0 if $F_{part} > F_{(2,6);0.05} = 5.14$.

Since $F_{part} = 28.081 > 5.14$ we reject H_0 and conclude that at a 5% level of significance there is enough evidence to say that the slopes of the 3 drug lines are not parallel.

```
Footnote 'Krystian Wojcicki, 101001444';
```

```
□ Data BloodData;
```

```
Input Bp Age Weight;
```

```
Cards;
```

```
80 3 2.61
90 4 2.67
96 5 2.98
102 6 3.98
81 3 2.87
96 4 3.41
99 5 3.49
110 6 4.03
88 3 3.41
90 4 2.81
100 5 3.24
102 6 3.75
86 3 3.18
93 4 3.13
101 5 3.98
103 6 4.55
86 3 3.41
91 4 3.35
100 5 3.75
105 6 3.83
84 3 3.18
91 4 3.52
95 5 3.49
104 6 3.81
107 6 4.03
```

```
Run;
```

```
ods pdf file="a3-output.pdf";
```

```
ods graphics off;
```

```
□ Proc Reg;
```

```
Model Bp=Age Weight;
```

```
Model Bp=Weight;
```

```
Run;
```

❏ **Data Drug;**

```
Input dose X2 X3 potency;
```

```
  X1=log(dose);
```

```
  interact12=X1*X2;
```

```
  interact13=X1*X3;
```

```
Cards;
```

```
  0.2 0 0 2.0
```

```
  0.4 0 0 4.3
```

```
  0.8 0 0 6.5
```

```
  1.6 0 0 8.9
```

```
  0.2 1 0 1.8
```

```
  0.4 1 0 4.1
```

```
  0.8 1 0 4.9
```

```
  1.6 1 0 5.7
```

```
  0.2 0 1 1.3
```

```
  0.4 0 1 2.0
```

```
  0.8 0 1 2.8
```

```
  1.6 0 1 3.4
```

```
Run;
```

❏ **Proc Reg;**

```
Model potency=X1 X2 X3 interact12 interact13;
```

```
Model potency=X1 X2 X3;
```

```
Run;
```

```
ods pdf close
```