

STAT 3008 Applied Regression Analysis

Midterm Solution

2012-2013 1st term, Oct. 29, 2012

Q1-11.(By Yang Ping @ LSB G30)

1. C) is incorrect, since

$$\frac{(\bar{X} - \mu)\sqrt{n}}{S} \sim t(n-1). \quad (3 \text{ marks})$$

2. Mean function = 0; (1 mark) Variance function = constant; (1 mark) No separated points. (1 mark)

3. Since there are only two data points, the fitted line must pass through these two points.

$$\frac{y-4}{x+2} = \frac{3-4}{-1+2}. \quad (2 \text{ marks})$$

$$\Rightarrow y = 2 - x. \quad (1 \text{ mark})$$

Hence, $\hat{\beta}_0 = 2$ (1 mark), $\hat{\beta}_1 = -1$ (1 mark).

4. a), b), d), and e) are correct. (4 correct answers for 3 marks, 3 for 2 and so on)

Error term is a random variable and it can not be consistently estimated.

5. $Var(\hat{e}) = Var[(I-H)Y] = (I-H)Var(Y)(I-H) = \sigma^2(I-H) \neq \sigma^2I$.

(2 marks for each of the first two equality signs; 1 mark for each of the others)

6.

$$\hat{e}_5 = 0 - \sum_{i=1}^4 4\hat{e}_i = -0.1. \quad (2 \text{ marks})$$

$$\therefore RSS = 3 \times 0.1^2 + 0.2^2 + 0.3^2 = 0.16. \quad (2 \text{ marks})$$

$$R^2 = 1 - \frac{RSS}{SYY} = 1 - \frac{0.16}{3.3} = 0.952. \quad (2 \text{ marks})$$

7. c). (1 mark) Compare their absolute value.

8. b). (6 marks)

$$\hat{\beta}_{1(new)} = \frac{c \sum (x_i - \bar{x})(y_i - \bar{y})}{c^2 \sum (x_i - \bar{x})^2} = \frac{1}{c} \hat{\beta}_1.$$

$$\hat{\beta}_{0(new)} = \bar{y} - \left(\frac{1}{c}\hat{\beta}_1\right)(c\bar{x}) = \bar{y} - \hat{\beta}_1\bar{x} = \hat{\beta}_0.$$

$$\hat{e}_{i(new)} = y_i - \hat{\beta}_0 - \hat{\beta}_{1(new)}x_{i(new)} = y_i - \hat{\beta}_0 - \left(\frac{1}{c}\hat{\beta}_1\right)(cx_i) = y_i - \hat{\beta}_0 - \hat{\beta}_1x_i = \hat{e}_i,$$

$$\therefore \hat{\sigma}_{(new)}^2 = \frac{\sum \hat{e}_i^2}{n-2} = \hat{\sigma}^2.$$

$$\therefore \hat{e}_{i(new)} = \hat{e}_i, RSS_{(new)} = RSS.$$

$$\therefore SY^2 = \sum (y_i - \bar{y})^2 \text{ is unaffected,}$$

$$\therefore R_{(new)}^2 = 1 - \frac{RSS}{SY^2} = R^2.$$

$$t_{(new)} = \frac{\hat{\beta}_0}{se(\hat{\beta}_0)_{(new)}}, \text{ where } se(\hat{\beta}_0)_{(new)} = \hat{\sigma} \left(\frac{1}{n} + \frac{\bar{x}_{(new)}^2}{SXX_{(new)}} \right)^{1/2}.$$

$$\therefore \frac{\bar{x}_{(new)}^2}{SXX_{(new)}} = \frac{c^2\bar{x}^2}{c^2 \sum (x_i - \bar{x})^2} = \frac{\bar{x}^2}{\sum (x_i - \bar{x})^2} = \frac{\bar{x}^2}{SXX},$$

$$\therefore t_{(new)} = t.$$

9. $tr(I_n - X(X'X)^{-1}X') = tr(I_n) - tr(X(X'X)^{-1}X') = n - tr(X(X'X)^{-1}X')$

$$tr(X(X'X)^{-1}X') = tr[(X'X)^{-1}(X'X)], \quad (1 \text{ mark})$$

$$tr[(X'X)^{-1}(X'X)] = tr(I_r) = r, \quad (1 \text{ mark})$$

$$\therefore tr(I_n - X(X'X)^{-1}X') = n - r. \quad (1 \text{ mark})$$

10.

$$\frac{df(\beta)}{d\beta} = X^TY - 2X^TX\beta, \quad (1 \text{ mark})$$

$$\Rightarrow \hat{\beta} = \frac{1}{2}(X^TX)^{-1}X^TY, \quad (1 \text{ mark})$$

$$\begin{aligned} \therefore \text{The minimum value of } f(\beta) &= \frac{1}{2}Y^TX(X^TX)^{-1}X^TY \\ &\quad - \frac{1}{4}Y^TX(X^TX)^{-1}(X^TX)(X^TX)^{-1}X^TY \\ &= \frac{1}{4}Y^TX(X^TX)^{-1}X^TY. \end{aligned} \quad (1 \text{ mark})$$

11. C.I. for $E(Y|X)$ is for a point, while C.B. for $E(Y|X)$ is C.I. for the whole line.

(2 marks)

C.I. for $E(Y|X)$ uses t_{n-p-1} , while C.B. for $E(Y|X)$ uses $\sqrt{(p+1)F_{(p+1, n-p-1)}}$.

(1 mark)

Q12.(By Pan Deng@LSB126)

$$\bar{Y} = 12/4 = 3, SYY = 25 + 16 + 9 + 36 = 86.$$

ANOVA Table				
Source	Sum of Squares	d.f.	Mean Square	F-statistics
Regression	74(half mark)	2(1 mark)	37(1 mark)	37/12(1 mark)
Residuals	12	1(1 mark)	12(1 mark)	
Total	86(half mark)			

Q13.(By Pan Deng@LSB126)

Let $H = X(X'X)^{-1}X'$, then $\hat{Y} = HY$ and $\sum_{i=1}^n \hat{y}_i^2 = \hat{Y}'\hat{Y}$.

$$E\left(\sum_{i=1}^n \hat{y}_i^2\right) = E[(HY)'(HY)] \quad (1 \text{ mark})$$

$$= E(Y'HY) \quad (1 \text{ mark})$$

$$= E[\text{tr}(HYY')] \quad (1 \text{ mark})$$

$$= \text{tr}[H \cdot E(YY')] \quad (1 \text{ mark})$$

$$= \text{tr}[H(X\beta\beta'X' + \sigma^2I)]$$

$$= \text{tr}(\beta'X'X\beta) + \text{tr}(\sigma^2H)$$

$$= \beta'X'X\beta + p\sigma^2 \quad (1 \text{ mark})$$

Q14.(By Feng Xiangnan @ LSB G24)

(1.)

$$\exp(t) = Ax^B$$

$$t = \log(Ax^B) = \log A + B \log x \dots (2\text{pt})$$

$$t = a + b \log x$$

$$a = \log A; \quad b = B \dots (2\text{pt})$$

(2.)

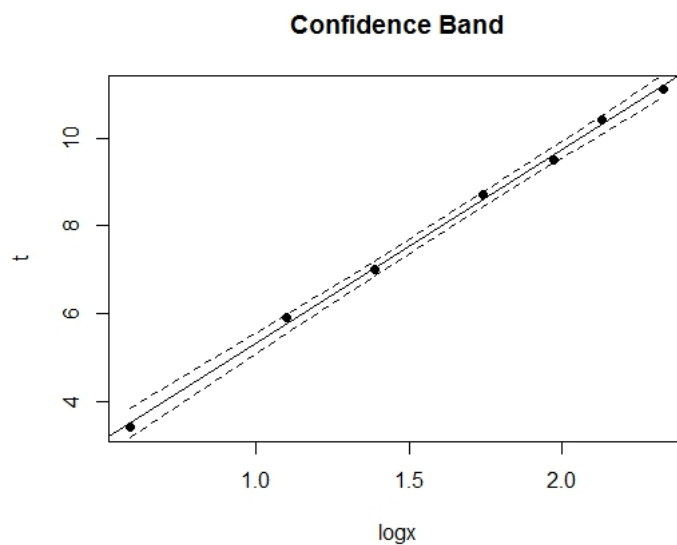
$$t = a + b \log x$$

$$\hat{b} = \frac{\sum t \log x - n \bar{t} \bar{\log x}}{\sum (\log x)^2 - n \bar{\log x}^2} \dots (2\text{pt})$$

$$\hat{b} = 4.4074 \dots (1\text{pt})$$

$$\hat{a} = \bar{t} - \bar{\log x} \hat{b} \dots (2\text{pt})$$

$$\hat{a} = 0.9181 \dots (1\text{pt})$$



plot.....(1pt)

line.....(1pt)

This fitting is appropriate as the data set is approximately a straight line.....(2pt)

You may answer this question with R^2 or other relevant concepts.

(3.)

$$W = \sqrt{2 * F(\alpha, 2, 5)} = 3.401804$$

sefit:

$$> W * \sqrt{\text{sigma2} * \text{t}(\text{xmatrix}[1]) \% \% \text{invM} \% \% \text{xmatrix}[1]}$$

0.3030883

$$> W * \sqrt{\text{sigma2} * \text{t}(\text{xmatrix}[2]) \% \% \text{invM} \% \% \text{xmatrix}[2]}$$

0.1724965

$$> W * \sqrt{\text{sigma2} * \text{t}(\text{xmatrix}[3]) \% \% \text{invM} \% \% \text{xmatrix}[3]}$$

0.1691638

$$> W * \sqrt{\text{sigma2} * \text{t}(\text{xmatrix}[4]) \% \% \text{invM} \% \% \text{xmatrix}[4]}$$

0.2087156

$$> \sqrt{\text{sigma2} * \text{t}(\text{xmatrix}[1]) \% \% \text{invM} \% \% \text{xmatrix}[1]}$$

0.08909634

$$> \sqrt{\text{sigma2} * \text{t}(\text{xmatrix}[2]) \% \% \text{invM} \% \% \text{xmatrix}[2]}$$

0.05070736

$$> \sqrt{\text{sigma2} * \text{t}(\text{xmatrix}[3]) \% \% \text{invM} \% \% \text{xmatrix}[3]}$$

0.04972767

$$> \sqrt{\text{sigma2} * \text{t}(\text{xmatrix}[4]) \% \% \text{invM} \% \% \text{xmatrix}[4]}$$

0.0613544

$$> \text{yhat} [1] 3.973127 7.028126 8.815186 10.083125$$

.....Some points are given based on the above information

$$\hat{y} \pm \hat{\sigma} \sqrt{(p+1)F(\alpha, p+1, n-(p+1))(\mathbf{x}'(X'X)^{-1}\mathbf{x})} \dots (2\text{pt})$$

$$(X'X)^{-1} = \begin{pmatrix} 1.2673 & -0.6998 \\ -0.6998 & 0.4355 \end{pmatrix}$$

$$\hat{\sigma} = 0.1252;$$

$$\mathbf{x}_1 = (1, \log(2))'$$

$$\hat{y}_1 \pm \hat{\sigma} \sqrt{2F(95\%, 2, 5)(\mathbf{x}'_1(X'X)^{-1}\mathbf{x}_1)} \dots (1\text{pt})$$

$$(3.6700, 4.2762)$$

$$\mathbf{x}_2 = (1, \log(4))'$$

$$\hat{y}_2 \pm \hat{\sigma} \sqrt{2F(95\%, 2, 5)(\mathbf{x}'_2(X'X)^{-1}\mathbf{x}_2)} \dots (1\text{pt})$$

$$(6.8556, 7.2006)$$

$$\mathbf{x}_3 = (1, \log(6))'$$

$$\hat{y}_3 \pm \hat{\sigma} \sqrt{2F(95\%, 2, 5)(\mathbf{x}'_3(X'X)^{-1}\mathbf{x}_3)} \dots (1\text{pt})$$

$$(8.6460, 8.9844)$$

$$\mathbf{x}_4 = (1, \log(8))'$$

$$\hat{y}_4 \pm \hat{\sigma} \sqrt{2F(95\%, 2, 5)(\mathbf{x}'_4(X'X)^{-1}\mathbf{x}_4)} \dots (1\text{pt})$$

$$(9.8744, 10.2918)$$

If all the calculation and plot(should be a curve) are right(2pt)

(4.)

$$\frac{(\hat{\beta} - \beta)'(X'X)(\hat{\beta} - \beta)}{(p + 1)\hat{\sigma}^2} \leq F(\alpha, p + 1, n - p - 1) \dots (4\text{pt})$$

$$(X'X) = \begin{pmatrix} 7 & 11.2476 \\ 11.2476 & 20.3687 \end{pmatrix}$$

$$\hat{\sigma}^2 = 0.01568$$

$$223.285 * a^2 - 3572.52 * a + 15706.8 + 717.548 * a * b - 6385.91 * b + 649.718 * b^2 \leq F(0.95, 2, 5)$$

$$= 5.786$$

.....(2pt)

plot and three points.....(2pt)

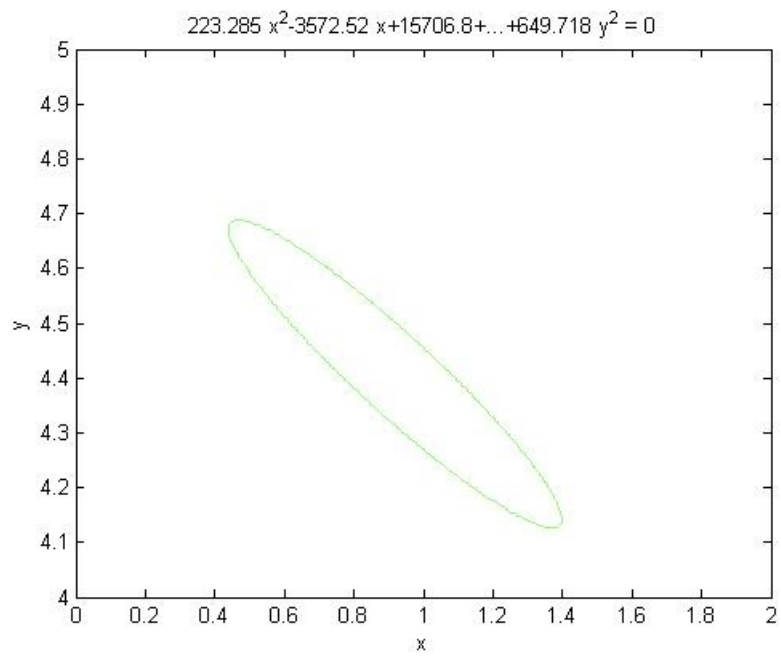
$$x=1$$

$$4.4552$$

$$4.2692$$

x=0.5
4.6844
4.5921

x=0.7
4.6119
4.4438



Q15.(By Pan Deng@LSB126)

Set $Y = \beta_1 X_1 + \beta_2 X_2 + e$, then

(i)

$$\hat{e}_{Y|X_2} = Y - \bar{Y} \quad (2 \text{ marks})$$

$$\hat{e}_{X_1|X_2} = X_1 - \bar{X}_1 \quad (2 \text{ marks})$$

$$\text{slope} = \frac{\sum_{i=1}^n (y_i - \bar{Y})(x_i - \bar{X}_1)}{\sum_{i=1}^n (x_i - \bar{X}_1)^2} = \hat{\beta}_1 \quad (1 \text{ mark})$$

(ii)

$$\hat{e}_{Y|X_1} = Y - \frac{\sum_{i=1}^n x_i y_i}{\sum_{i=1}^n x_i^2} X_1 \quad (2 \text{ marks})$$

$$\hat{e}_{X_2|X_1} = X_2 - \frac{\sum_{i=1}^n x_i}{\sum_{i=1}^n x_i^2} X_1 \quad (2 \text{ marks})$$

$$\begin{aligned} \text{slope} &= \frac{\sum_{i=1}^n (y_i - \frac{\sum_{i=1}^n x_i y_i}{\sum_{i=1}^n x_i^2} x_i)(1 - \frac{\sum_{i=1}^n x_i}{\sum_{i=1}^n x_i^2} x_i)}{\sum_{i=1}^n (1 - \frac{\sum_{i=1}^n x_i}{\sum_{i=1}^n x_i^2} x_i)^2} \\ &= \bar{Y} - \hat{\beta}_1 \bar{X} \end{aligned} \quad (1 \text{ mark})$$

Q16.(By Pan Deng@LSB126)

$$X'Y = |Y| \cdot |X| \cdot \cos \theta \quad (1 \text{ mark})$$

where θ is the angle between the vectors Y and X .

$$OA = |Y| \cdot \cos \theta = \frac{X'Y}{|X|} = (\sqrt{X'X})^{-1} X'Y \quad (2 \text{ marks})$$

$$\hat{Y} = OA \cdot \frac{X}{|X|} = X(X'X)^{-1} X'Y = cX \quad (2 \text{ marks})$$

Where $c = (X'X)^{-1} X'Y$.