

04/08

## Warm up

**Exercise:** Using the exercise 1.27 of the book (muscle mass) and the 6 observations:

Observation	1	2	3	4	5	6
$X_i$	43	39	41	86	72	76
$Y_i$	106	106	97	60	70	80

- Obtain a Table in order to compute the least square estimates. Similar to Table 1.1 (Tocula company example) page 19.
- Compute  $b_0$  and  $b_1$  using the equations (1.10a) and (1.10b).
- Consider and explain a, b and c in 2.27 (Book) for this exercise.

Figure 1: Midterm bonus problem.

- Given  $X_h = 40$ , compute the point estimator of the mean response  $E[Y_h]$  and the corresponding 95% confidence interval.

$i$	$X_i$	$Y_i$	$X_i - \bar{X}$	$Y_i - \bar{Y}$	$(X_i - \bar{X})(Y_i - \bar{Y})$	$(X_i - \bar{X})^2$	$(Y_i - \bar{Y})^2$
1	43	106	-16.5	19.5			
2	39	106	-20.5	19.5			
3	41	97	-18.5	10.5			
4	86	60	26.5	-26.5			
5	72	70	12.5	-16.5			
6	76	80	16.5	-6.5			
Total	357	519					
Mean	59.5	86.5					

a

$$b_1 = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^n (X_i - \bar{X})^2}$$

$$b_0 = \bar{Y} - b_1 \bar{X}$$

b

c

d

$$\hat{Y}_h = b_0 + b_1 X_h$$

$$(\hat{Y}_h - t(1 - \alpha/2; n - 2) * s(\hat{Y}_h), \hat{Y}_h + t(1 - \alpha/2; n - 2) * s(\hat{Y}_h))$$

## 1.2

$Y = 2X + 300$ . Functional relation.

**1.8**

Yes. Not necessarily

$$E[Y] = \beta_0 + \beta_1 * X$$

**1.12**

a

b

c

d

**1.27**

a

b

**1.30**

**04/15**

### Warm up

Given  $X_h = 40$ , compute the 90% prediction interval for the mean of the  $m = 3$  new  $Y$  observations. See Page 60 in the textbook.

$$\hat{Y}_h \pm t(1 - \alpha/2; n - 2)s\{\text{predmean}\},$$

where

$$s^2\{\text{predmean}\} = \frac{MSE}{m} + s^2(\hat{Y}_h) = MSE \left[ \frac{1}{m} + \frac{1}{n} + \frac{(X_h - \bar{X})^2}{\sum_{i=1}^n (X_i - \bar{X})^2} \right].$$

In the example provided by the textbook (Page 61), we have  $b_0 = 62.37$ ,  $b_1 = 3.5702$ ,  $\bar{X} = 70$ ,  $\sum (X_i - \bar{X})^2 = 19800$ , and  $MSE = 2384$ . Thus

$$\hat{Y}_h = b_0 + b_1 X_h = 62.37 + 3.5702 \times 40 = 205.2,$$

and

$$s^2(\hat{Y}_h) = 2384 \left[ \frac{1}{25} + \frac{(40 - 70)^2}{19800} \right] = 203.72.$$

Therefore

$$s^2\{\text{predmean}\} = \frac{2384}{3} + 203.72 = 998.4.$$

Note that  $t(0.95; 23) = 1.714$ . We obtain

$$\begin{aligned} 205.2 - 1.714 \times 31.60 &\leq \bar{Y}_{h(\text{new})} \leq 205.2 + 1.714 \times 31.60 \\ 151.0 &\leq \bar{Y}_{h(\text{new})} \leq 259.4. \end{aligned}$$

### 2.9

Formula (2.30).  $s(\hat{Y}_h)$  depends on  $X_h$ , which needs to be specified.

### 2.10

(a)

(b)

(c)

### 2.12

(a) (2.37)

$$\sigma^2\{\text{pred}\} = \sigma^2\{Y_{h(\text{new})} - \hat{Y}_h\} = \sigma^2\{Y_{h(\text{new})}\} + \sigma^2(\hat{Y}_h) = \sigma^2 + \sigma^2(\hat{Y}_h),$$

where  $\hat{Y}_h = b_0 + b_1 X_h$ .

(b) (2.29b)

$$\sigma^2(\hat{Y}_h) = \sigma^2 \left[ \frac{1}{n} + \frac{(X_h - \bar{X})^2}{\sum_{i=1}^n (X_i - \bar{X})^2} \right].$$

**2.27**

(a)

$$H_0 : \beta_1 \geq 0 \quad \text{vs} \quad H_a : \beta_1 < 0.$$

$$t^* < t(\alpha; n - 2).$$

$$p(T < t^*) < \alpha,$$

where  $T$  is a random variable which is  $t$  distributed with degrees of freedom  $n - 2$ .

(b)

(c)

$$b_0 + b_1 X; b_0 + b_1(X + 1)$$

$$b_1 \pm t(0.975; 58)s(b_1)$$

04/29

**Exercise:** Using the exercise 1.27 of the book (muscle mass) and the 6 observations:

Observation	1	2	3	4	5	6
$X_i$	43	39	41	86	72	76
$Y_i$	106	106	97	60	70	80

- Estimate a 95% confidence interval for the parameters  $\beta_0$  and  $\beta_1$
- Use the p-value for testing  $H_0: \beta_1 = 0$ , with  $\alpha = 0.01$ .
- Compute a 99% confidence interval for the mean response  $E\{Y_h\}$  using  $X_h = 30$ .
- Test whether  $\beta_1 = 0$  using an F test with  $\alpha = 0.01$ . State the alternatives, decision rule, and conclusion.

Figure 2

What we need from the data:

$$n = 6$$

$$t(1 - 0.05/2; n - 2) = t(0.975, n - 2) = 2.776$$

$$t(1 - 0.01/2; n - 2) = t(0.995, n - 2) = 4.604$$

$$\bar{X} = \frac{43 + 39 + 41 + 86 + 72 + 76}{6} = 59.5$$

$$\bar{Y} = \frac{106 + 106 + 97 + 60 + 70 + 80}{6} = 86.5$$

$$\sum (X_i - \bar{X})^2 = 2165.5$$

$$SSTO = \sum (Y_i - \bar{Y})^2 = 1887.5$$

$$\sum (X_i - \bar{X})(Y_i - \bar{Y}) = -1931.5$$

What we can derive:

$$b_1 = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{\sum (X_i - \bar{X})^2} = \frac{-1931.5}{2165.5} = -0.892$$

$$b_0 = \bar{Y} - b_1 \bar{X} = 86.5 - (-0.892) \times 59.5 = 139.57$$

$$SSR = b_1^2 \sum (X_i - \bar{X})^2 = 1722.79$$

$$MSR = SSR/1 = 1722.79$$

$$SSE = SSTO - SSR = 164.71$$

$$MSE = SSE/(n - 2) = 41.18$$

$$s^2(b_0) = MSE \left[ \frac{1}{n} + \frac{\bar{X}^2}{\sum (X_i - \bar{X})^2} \right] = 74.19$$

$$s^2(b_1) = \frac{MSE}{\sum(X_i - \bar{X})^2} = 0.019$$

$$t^* = \frac{b_1}{s(b_1)} = \frac{-0.892}{\sqrt{0.019}} = -6.471$$

$$F^* = \frac{MSR}{MSE} = \frac{1722.79}{41.18} = 41.84$$

$$F^* = (t^*)^2$$

a P45 and P49

$$b_0 \pm t(1 - \alpha/2; n - 2)s(b_0)$$

$$139.57 \pm 2.776 \times \sqrt{74.19} = [115.66, 163.48]$$

$$b_1 \pm t(1 - \alpha/2; n - 2)s(b_1)$$

$$-0.892 \pm 2.776 \times \sqrt{0.019} = [-1.127, -0.509]$$

b

$$H_a : \beta_1 \geq 0 \quad P(t(4) > t^*)$$

$$H_a : \beta_1 \leq 0 \quad P(t(4) < t^*)$$

*p*-value:

$$2P(t(4) > |t^*|) = 2P(t(4) > |-6.471|) = 0.00294$$

R code: `2*(1-pt(6.471, 4))` Since the *p*-value is less than  $\alpha = 0.01$ , we conclude  $H_a$ .

c

$$\hat{Y}_h = b_0 + b_1 X_h = 139.57 + (-0.892) * 30 = 112.81$$

$$s^2(\hat{Y}_h) = MSE \left[ \frac{1}{n} + \frac{(X_h - \bar{X})^2}{\sum(X_i - \bar{X})^2} \right] = 23.41$$

$$\hat{Y}_h \pm t(1 - \alpha/2; n - 2)s(\hat{Y}_h)$$

$$112.81 \pm 4.604 \times \sqrt{23.41} = [90.53, 135.09]$$

d

$$H_a : \beta_1 \neq 0$$

*p*-value:  $P(F(1, n - 2) > F^*) = 0.0029 < \alpha = 0.01$

R code: `1-pf(41.84, 1, 4)`

quantile:  $F(1 - \alpha; 1, n - 2) = 21.2 < F^*$

R code: `qf(0.99, 1, 4)`

Decision rule: If  $F^* \leq F(1 - \alpha; 1, n - 2)$ , conclude  $H_0$ . if  $F^* > F(1 - \alpha; 1, n - 2)$ , conclude  $H_a$ .

Conclusion:  $H_a$

## Midterm II Bonus (Due May 21 at 4pm)

Consider the following regression model with two predictors  $X_1$  and  $X_2$  ( $p = 3$  in the general linear regression model)

$$Y_i = \beta_0 + \beta_1 X_{i1} + \beta_2 X_{i2} + \varepsilon_i, i = 1, 2, \dots, n.$$

In matrix form,

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$$

where

$$\mathbf{Y} = \begin{pmatrix} Y_1 \\ Y_2 \\ \vdots \\ Y_n \end{pmatrix}, \mathbf{X} = \begin{pmatrix} 1 & X_{11} & X_{12} \\ 1 & X_{21} & X_{22} \\ \vdots & \vdots & \vdots \\ 1 & X_{n1} & X_{n2} \end{pmatrix}, \boldsymbol{\beta} = \begin{pmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \end{pmatrix}, \boldsymbol{\varepsilon} = \begin{pmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \vdots \\ \varepsilon_n \end{pmatrix}.$$

- Compute  $(\mathbf{X}'\mathbf{X})^{-1}$ . Show all the steps and formulas. **Hint: Equation 5.23.**
- Show  $\mathbf{H}\mathbf{H} = \mathbf{H}$ , where  $\mathbf{H} = \mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'$  is the hat matrix. Show all the steps and formulas.
- Compute the ANOVA table, formulas. Write the model. **Hint: the second column of the ANOVA table (Table 6.1 *SS*) should be expressed using  $Y_i, X_{ij}, i = 1, 2, \dots, n, j = 1, 2, 3$ . Equations 6.74 and 6.75 may be useful.**