

## 统计线性模型2023秋期中回忆版

1. (同2022年秋hw1.1) (40 marks) Suppose an appliance store conducts a 5-month experiment to determine the effect of advertising on sales revenue. The results are shown below.

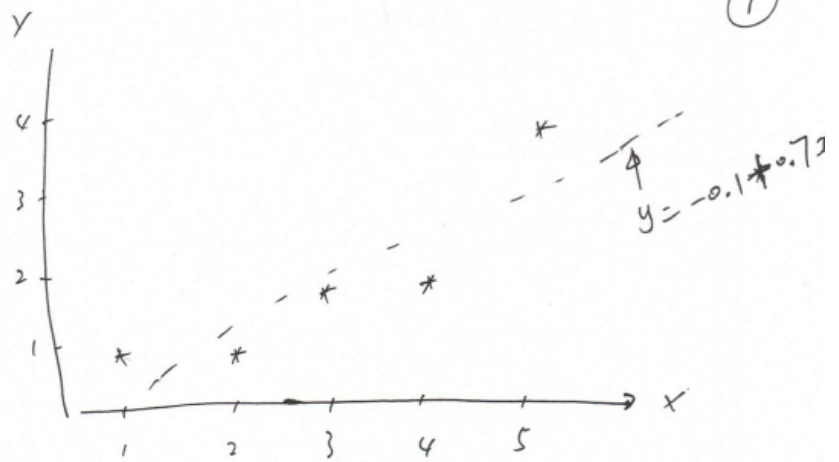
Advertising Expenditure $x$ (hundreds of dollars)	1	2	3	4	5	
Sales Revenue $y$ (thousands of dollars)	1	1	2	2	4	

- (a) Draw a scatterplot of the data and comment the relationship between  $y$  and  $x$ .
- (b) What is your linear regression model? State the necessary assumptions.
- (c) Find the least squares line from the data and plot it on your scatterplot.
- (d) Test the hypothesis that the Advertising Expenditure has no effect of the Sales Revenue when a linear model is used (use  $\alpha = 0.05$ ). State the null and alternative hypotheses. Draw the appropriate test conclusions.
- (e) Find a 95% confidence interval for  $\beta_1$  (slope of the linear regression model). Interpret your results.
- (f) Find the coefficient of determination for the linear regression model. Interpret your result.
- (g) Find a prediction for the mean Sales Revenue when 4 hundreds dollars are spent on advertising and its 95% interval. What is the 95% interval for the Sales Revenue?

# Assignment 1 Solutions

(1)

1. (a)



it shows a rough linear relationship between  $y$  and  $x$ , but the correlation is ~~not~~ quite strong.

(b) A linear regression model can be expressed as  
$$y_i = \beta_0 + \beta_1 x_i + \varepsilon_i, \quad \varepsilon_i \sim N(0, \sigma^2).$$

main assumptions are:

- (i) the relationship between  $y_i$  and  $x_i$  is linear
- (ii) the variance of  $y$  is a constant
- (iii) the observations are independent
- (iv) (optional)  $y$  is distributed normally.

(c)  $\bar{x} = 3$     $\bar{y} = 2$

$$S_{xx} = \sum (x_i - \bar{x})^2 = 10$$

$$S_{xy} = \sum (x_i - \bar{x})(y_i - \bar{y}) = 7$$

$$S_{yy} = \sum (y_i - \bar{y})^2 = 6$$

$$\hat{\beta}_1 = S_{xy} / S_{xx} = 7/10 = 0.7$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = 2 - 0.7 \times 3 = -0.1$$

thus the least squares line is  $y = -0.1 + 0.7x$ , which has been added to the scatterplot

(d)  $H_0: \beta_1 = 0$  v.s.  $H_1: \beta_1 \neq 0$   $n = 5$

(2)

$$SSE = \sum (y_i - \hat{y}_i)^2 = 1.1$$

$$s^2 = \frac{SSE}{n-2} = 0.3667$$

$$t = \frac{\hat{\beta}_1}{(S/S_{xx})^{1/2}} = \frac{0.7}{(0.3667/10)^{1/2}} = 3.6556$$

$$> t_{0.025, 3} = 3.1824$$

Thus, reject  $H_0$ , meaning the Advertising Expenditure has effect of the Sales Revenue.

(e)  $t = \frac{\hat{\beta}_1 - \beta_1}{S/S_{xx}^{1/2}} \sim t_3$ , the 95% C.I. of  $\beta_1$  is

$$\hat{\beta}_1 \pm t_{0.025, 3} \cdot S/S_{xx}^{1/2}$$

$$= 0.7 \pm 3.1824 * \left( \frac{0.3667}{10} \right)^{1/2}$$

$$= (0.0906, 1.3094)$$

There is 95% chance that  $\beta_1$  would take values between 0.0906 and 1.3094.

(f)  $R^2 = \frac{SSR}{SST} = 1 - \frac{SSE}{SST} = 1 - \frac{1.1}{6} = 0.8167$

meaning 81.67% percent of the variations of the Sales Revenue can be explained by the model.

(3)

$$(g) x_h = 4, \quad \hat{y}_h = \hat{\beta}_0 + \hat{\beta}_1 x_h = 2.7$$

C.I. of  $E(y_h)$  is

$$\hat{y}_h \pm t_{0.025, 3} \cdot s \cdot \sqrt{\frac{1}{n} + \frac{(x_h - \bar{x})^2}{S_{xx}}}$$

$$= (1.6445, 3.7559)$$

Predictive interval of  $y_h$  at  $x_h = 4$  is

$$\hat{y}_h \pm t_{0.025, 3} \cdot s \cdot \sqrt{1 + \frac{1}{n} + \frac{(x_h - \bar{x})^2}{S_{xx}}}$$

$$= (0.5028, 4.8972).$$

2. (20 marks) List at least two generalized inverse of A, verify it.

$$A = \begin{bmatrix} 2 & 2 & 6 \\ 1 & 0 & 2 \\ 3 & 2 & 8 \end{bmatrix}$$

3. (同2022年秋hw3.2) (20 marks) Let  $\mathbf{y} = (y_1, y_2, y_3)'$  be distributed as  $N_3(\boldsymbol{\mu}, \Sigma)$ , where

$$\boldsymbol{\mu} = \begin{pmatrix} 2 \\ -1 \\ 3 \end{pmatrix}, \quad \Sigma = \begin{pmatrix} 4 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 3 \end{pmatrix}.$$

- (a) Find the distribution of  $\begin{pmatrix} y_1 - y_2 + y_3 \\ 2y_1 + y_2 - y_3 \end{pmatrix}$ ;
- (b) The conditional distribution of  $(y_1, y_2)$  given  $y_3$ ;
- (c) The partial correlation between  $y_1$  and  $y_2$  given  $y_3$ .

Q2

$$(a) \begin{pmatrix} y_1 - y_2 + y_3 \\ 2y_1 + y_2 - y_3 \end{pmatrix} = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 1 & -1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} \sim \mathcal{N}(\underline{\mu}_1, \underline{\Sigma}_1).$$

$$\underline{\mu}_1 = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 1 & -1 \end{pmatrix} \begin{pmatrix} 2 \\ -1 \\ 3 \end{pmatrix} = \begin{pmatrix} 6 \\ 0 \end{pmatrix}$$

$$\begin{aligned} \underline{\Sigma}_1 &= \begin{pmatrix} 1 & -1 & 1 \\ 2 & 1 & -1 \end{pmatrix} \begin{pmatrix} 4 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 3 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ -1 & 1 \\ 1 & -1 \end{pmatrix} = \begin{pmatrix} 3 & 0 & 2 \\ 9 & 3 & -2 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ -1 & 1 \\ 1 & -1 \end{pmatrix} \\ &= \begin{pmatrix} 5 & 4 \\ 4 & 23 \end{pmatrix}. \end{aligned}$$

$$(b) \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} | y_3 \sim N(\mu_2, \Sigma_2)$$

$$\mu_2 = \mu_y + \Sigma_{yx} \Sigma_{xx}^{-1} (\bar{x} - \mu_x)$$

$$= \begin{pmatrix} 2 \\ -1 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} 3^{-1} (y_3 - 3)$$

$$= \begin{pmatrix} 2 \\ \frac{y_3}{3} - 2 \end{pmatrix}$$

$$\begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} \sim \begin{pmatrix} y \\ x \end{pmatrix}$$

$$\Sigma = \begin{pmatrix} 4 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 3 \end{pmatrix} = \begin{pmatrix} \Sigma_{yy} & \Sigma_{yx} \\ \Sigma_{xy} & \Sigma_{xx} \end{pmatrix}$$

$$\Sigma_2 = \Sigma_{yy} - \Sigma_{yx} \Sigma_{xx}^{-1} \Sigma_{xy}$$

$$= \begin{pmatrix} 4 & 1 \\ 1 & 2 \end{pmatrix} - \begin{pmatrix} 0 \\ 1 \end{pmatrix} 3^{-1} (0 \ 1) = \begin{pmatrix} 4 & 1 \\ 1 & \frac{5}{3} \end{pmatrix}$$

(c) The partial correlation between  $y_1$  and  $y_2$  given  $y_3$

is:

$$\frac{1}{\sqrt{4 * 5/3}} = 0.3873$$

4. (20 marks)  $y \sim N_p(\mu, \sigma^2 \Sigma)$ ,  $\mu, \sigma^2, \Sigma$  is known.

(a)  $E(y' \Sigma^{-1} y)$ ;

(b) Distribution of  $\frac{1}{\sigma^2} y' \Sigma^{-1} y$ .

## 4.1 Random vector and matrix

- Let  $\mathbf{Y}$  be a random vector with mean  $\boldsymbol{\mu} = E(\mathbf{Y})$  and  $\boldsymbol{\Sigma} = \text{Cov}(\mathbf{Y})$ , then

$$E(\mathbf{Y}'\mathbf{A}\mathbf{Y}) = \text{tr}(\mathbf{A}\boldsymbol{\Sigma}) + \boldsymbol{\mu}'\mathbf{A}\boldsymbol{\mu}$$

where  $\mathbf{A}$  is a symmetric matrix.

## Ch 5.2 Non-Central $\chi^2$ , $F$ and $t$ distributions

**Theorem 5.1** Let  $\mathbf{x}_{p \times 1} \sim N(\boldsymbol{\mu}, \boldsymbol{\Sigma})$  and  $\mathbf{A}$  be symmetric, then  $q = \mathbf{x}'\mathbf{A}\mathbf{x} \sim \chi^2_{(r, \lambda)}$  where  $r$  denoting the rank of  $\mathbf{A}$  and  $\lambda = \frac{\boldsymbol{\mu}'\mathbf{A}\boldsymbol{\mu}}{2}$  if and only if  $\mathbf{A}\boldsymbol{\Sigma}$  is idempotent.