

# 第四次作业.

6.1. 证明如下:  $\underline{A}\underline{\theta} + \underline{E} = \underline{x}$

$$\therefore \underline{E} = \underline{x} - \underline{A}\underline{\theta}$$

代入加权误差函数  $Q(\theta)$

$$\text{得 } Q(\theta) = (\underline{x} - \underline{A}\underline{\theta})^T \underline{W} (\underline{x} - \underline{A}\underline{\theta})$$

$$= (\underline{x}^T \underline{W} - \underline{\theta}^T \underline{A}^T \underline{W}) (\underline{x} - \underline{A}\underline{\theta})$$

$$= \underline{x}^T \underline{W} \underline{x} - \underline{\theta}^T \underline{A}^T \underline{W} \underline{x} - \underline{x}^T \underline{W} \underline{A} \underline{\theta} + \underline{\theta}^T \underline{A}^T \underline{W} \underline{A} \underline{\theta}$$

$\therefore$  用  $\theta$  对  $Q(\theta)$  求导可得.

$$\begin{aligned} \frac{\partial Q(\theta)}{\partial \theta} &= -\underline{A}^T \underline{W} \underline{x} - (\underline{x}^T \underline{W} \underline{A})^T + \underline{A}^T \underline{W} \underline{A} \underline{\theta} + (\underline{\theta}^T \underline{A}^T \underline{W} \underline{A})^T \\ &= 0 \end{aligned}$$

$\therefore W$  对称.

$$\therefore 2\underline{A}^T \underline{W} \underline{x} = 2\underline{A}^T \underline{W} \underline{A} \underline{\theta}$$

$$\therefore \underline{\theta}_{WLS} = (\underline{A}^T \underline{W} \underline{A})^{-1} \underline{A}^T \underline{W} \underline{x} \quad \text{得证}$$

$\therefore$  代入  $\underline{x} = \underline{A}\underline{\theta} + \underline{E}$  到  $\underline{\theta}_{WLS}$  中可得

$$\begin{aligned} \underline{\theta}_{WLS} &= (\underline{A}^T \underline{W} \underline{A})^{-1} \underline{A}^T \underline{W} (\underline{A}\underline{\theta} + \underline{E}) \\ &= \underline{\theta} + (\underline{A}^T \underline{W} \underline{A})^{-1} \underline{A}^T \underline{W} \underline{E} \end{aligned}$$

$\therefore$  求协方差矩阵.

$$\begin{aligned} C &= E \{ (\underline{\theta} - \underline{\theta}_{WLS}) (\underline{\theta} - \underline{\theta}_{WLS})^T \} \\ &= (\underline{A}^T \underline{W} \underline{A})^{-1} \underline{A}^T \underline{W} R \underline{W} \underline{A} (\underline{A}^T \underline{W} \underline{A})^{-1} \end{aligned}$$

$\therefore$  用  $W$  对称可得

$$\text{当 } \frac{\partial \text{tr}(C)}{\partial W} = 0 \text{ 时, } \underline{A} (\underline{A}^T \underline{W} \underline{A})^{-1} \underline{A}^T \underline{W} R \underline{W} \underline{A} (\underline{A}^T \underline{W} \underline{A})^{-1} \underline{A}^T = R \underline{W} \underline{A} (\underline{A}^T \underline{W} \underline{A})^{-1} \underline{A}^T$$

$\therefore$  此时可得  $W = R^{-1}$

$$6.4. \therefore \text{求} \min \frac{1}{2} \|\underline{A}\underline{x} - \underline{b}\|_2^2 - \frac{1}{2} \lambda \|\underline{x}\|_2^2$$

$$\therefore \frac{1}{2} \|\underline{A}\underline{x} - \underline{b}\|_2^2 - \frac{1}{2} \lambda \|\underline{x}\|_2^2$$

$$= \frac{1}{2} [(\underline{A}\underline{x} - \underline{b})^H (\underline{A}\underline{x} - \underline{b}) - \lambda \underline{x}^H \underline{x}]$$

$\therefore$  对上式用  $\underline{x}^H$  求导可得

$$\frac{\partial \frac{1}{2} \|\underline{A}\underline{x} - \underline{b}\|_2^2 - \frac{1}{2} \lambda \|\underline{x}\|_2^2}{\partial \underline{x}^H} = \underline{A}^H \underline{A} \underline{x} - \underline{A}^H \underline{b} - \lambda \underline{x} = 0$$

$$\therefore \underline{x} = (\underline{A}^H \underline{A} - \lambda \underline{I})^{-1} \underline{A}^H \underline{b} \quad \text{得证}$$

6.7. 数据点为 (1,3), (3,1), (5,7), (4,6), (7,4)

① 总误差最小 = 求拟合:

$$\therefore \bar{x} = 4, \bar{y} = 4.2$$

$$\therefore M =$$

$$\begin{bmatrix} x_1 - \bar{x} & y_1 - \bar{y} \\ x_2 - \bar{x} & y_2 - \bar{y} \\ x_3 - \bar{x} & y_3 - \bar{y} \\ x_4 - \bar{x} & y_4 - \bar{y} \\ x_5 - \bar{x} & y_5 - \bar{y} \end{bmatrix} = \begin{bmatrix} -3 & -1.2 \\ -1 & -3.2 \\ 1 & 2.8 \\ 0 & 1.8 \\ 3 & -0.2 \end{bmatrix}$$

$$\therefore M^T M = \begin{bmatrix} 20 & 9 \\ 9 & 22.8 \end{bmatrix}, \text{再对其进行特征值分解得}$$

$$\text{特征多项式 } \lambda^2 - 42.8\lambda + 37.5 = 0$$

$$\text{解得 } \lambda_1 = 30.508, \lambda_2 = 12.292$$

$$\therefore M^T M = \begin{bmatrix} 0.6505 & -0.7595 \\ 0.7595 & 0.6505 \end{bmatrix} \begin{bmatrix} 30.508 & 0 \\ 0 & 12.292 \end{bmatrix} \begin{bmatrix} 0.6505 & -0.7595 \\ 0.7595 & 0.6505 \end{bmatrix}$$

$$\therefore \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} -0.7595 \\ 0.6505 \end{bmatrix}$$

$$\therefore \text{直线方程为 } -0.7595(x-4) + 0.6505(y-4.2) = 0$$

$$\therefore D_{ILS} = 12.2918.$$

② 一般误差最小 = 求法

$$D_{LS}^{(1)} = \sum_{i=1}^5 [m(x_i - \bar{x}) + (y_i - \bar{y})]^2$$

$$= (-3m - 1.2)^2 + (-m - 3.2)^2 + (m + 2.8)^2 + (1.8)^2 + (3m - 0.2)^2$$

$$\therefore \frac{\partial D_{LS}^{(1)}}{\partial m} = 40m + 18 = 0, \quad m = -0.45$$

$$\therefore d_1^2 = \frac{D_{LS}^{(1)}}{1+m^2} = 15.5945$$

$$\therefore D_{LS}^{(1)} = \sum_{i=1}^5 [(x-\bar{x}) + m(y-\bar{y})]^2$$

$$= (-3-1.2m)^2 + (-1-3.2m)^2 + (1+2.8m)^2 + (1-8m)^2 + (3-0.2m)^2$$

$$\therefore \frac{dD_{LS}^{(1)}}{dm} = 45.6m + 18 = 0 \Rightarrow m = -0.3947$$

$$\therefore \text{Regression line } (x-4) - 0.3947(y-4.4) = 0$$

$$\therefore d_2^2 = \frac{D_{LS}^{(1)}}{1+m^2} = 14.2305$$