

T₁. $\hat{y} = \hat{w}x + b$. 由最小二乘法定义知 $\sum (\hat{y} - y_i)^2$ 最小.

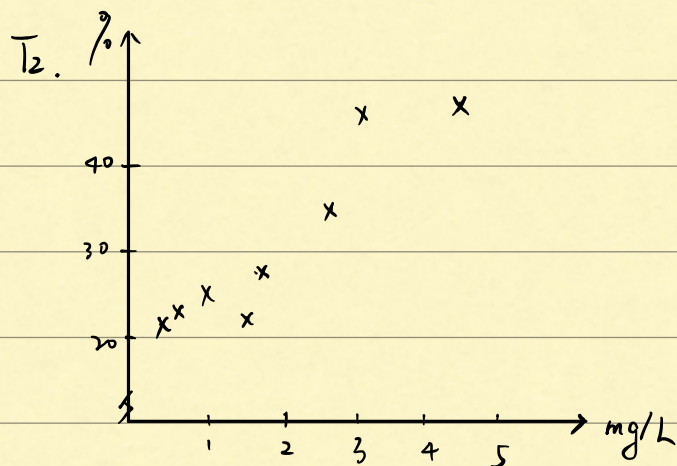
$$\therefore 2 \sum (\hat{w}x_i + b - y_i) = 0 \quad 2x_i \sum (\hat{w}x_i + b - y_i) = 0$$

$$\therefore \sum (\hat{y}_i - y_i) = 0 \quad \sum y_i (\hat{y}_i - y_i) = 0$$

$$\therefore \sum (\bar{y} - y_i) (\hat{y}_i - y_i) = 0$$

$$\text{原式左例} = \sum (y_i - \hat{y}_i)^2 + \sum (\hat{y}_i - \bar{y})^2 + 2 \sum (y_i - \hat{y}_i) (\hat{y}_i - \bar{y})$$

$$\text{将上式代入} = \sum (y_i - \hat{y}_i)^2 + \sum (\hat{y}_i - \bar{y})^2 = \text{右边. 得证}$$



$$\bar{x} = 1.995 \quad \bar{y} = 31.126$$

$$s_{xx} = \sum (x_i - \bar{x})^2 = 15.18 \quad s_{xy} = \sum (x_i - \bar{x})(y_i - \bar{y}) = 97.48$$

$$s_{yy} = \sum (y_i - \bar{y})^2 = 718.03$$

$$\hat{w} = \frac{s_{xy}}{s_{xx}} = 6.42 \quad b = \bar{y} - \hat{w}\bar{x} = 18.3$$

$$\therefore \hat{y} = 6.42x + 18.3 \quad r^2 = \frac{s_{xy}^2}{s_{xx}s_{yy}} = 0.87$$

T₄. 1) $J(\beta) = (X\beta - Y)^T (X\beta - Y) + \lambda \beta^T \beta$

$$\text{令 } \frac{\partial J(\beta)}{\partial \beta} = 0 \quad \therefore 2X^T(X\beta - Y) + 2\lambda \beta = 0$$

$$\therefore (X^T X + \lambda I)\beta = X^T Y \quad \therefore \beta = (X^T X + \lambda I)^{-1} X^T Y$$

2) $\lambda = 1$ 时 $\beta = (0.6143 \quad 0.548 \quad 0.0662)^T$

$\lambda = 5$ 时 $\beta = (0.3909 \quad 0.3721 \quad 0.0188)^T$

$\lambda = 10$ 时 $\beta = (0.2687 \quad 0.2669 \quad 0.0019)^T$