

$$r = 0.928, \quad S_Y = 2.507$$

$$\sum x_i = 14.7, \quad \sum x_i^2 = 29.15, \quad \sum y_i = 40, \quad \sum y_i^2 = 244, \quad \sum x_i y_i = 82.5$$

$$S_{YY} = \sum y_i^2 - \frac{(\sum y_i)^2}{n} = 244 - \frac{40^2}{8} = 44, \quad S_{XX} = \sum x_i^2 - \frac{(\sum x_i)^2}{n} = 29.15 - \frac{(14.7)^2}{8} = 2.13875$$

$$RSS = S_{YY} - \frac{S_{XY}^2}{S_{XX}} = 6.1274$$

$$S_{XY} = \sum x_i y_i - \frac{(\sum x_i)(\sum y_i)}{n} = 82.5 - \frac{(14.7)(40)}{8} = 9$$

ANOVA ~~df~~ ~~SS~~

Source	df	SS	MS	F
Regression	1	37.8726	37.8726	37.08637
Residuals	6	6.1274	1.0212	
Total	7	44		

$$T.S. = 37.0864 > F_{1,6,0.05}$$

$$t_{6,0.025} = 2.447, \quad F_{1,6,0.05} = 2.447^2 = 5.9878$$

$\therefore$ Dependence between X and Y.

$$y \sim N(\mu, \sigma^2)$$

$$y = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p + \epsilon$$

a).

$$\hat{y} = X\hat{\beta} = X(X^T X)^{-1} X^T Y$$

$$\begin{aligned} \text{Cov}\left(\frac{1}{n} \mathbf{1}^T Y, X(X^T X)^{-1} X^T Y\right) &= \text{Cov}\left(\frac{1}{n} Y, H Y\right) \\ &= \frac{1}{n} \text{Cov}(Y, Y) H^T \\ &= \sigma^2 \frac{1}{n} H \\ &= \frac{\sigma^2}{n} \end{aligned}$$

b).

$$\begin{aligned} E\left[\sum_{i=1}^n (\hat{y}_i - \bar{y})^2\right] &= E\left[Y^T \left(H - \frac{1}{n} \mathbf{1} \mathbf{1}^T\right) \left(H - \frac{1}{n} \mathbf{1} \mathbf{1}^T\right) Y\right] \\ &= E\left[Y^T \left(H - \frac{1}{n} \mathbf{1} \mathbf{1}^T\right) Y\right] \end{aligned}$$

$$= E\left[\text{tr}\left[Y^T \left(H - \frac{1}{n} \mathbf{1} \mathbf{1}^T\right) Y\right]\right]$$

$$= E\left[\text{tr}\left[\left(H - \frac{1}{n} \mathbf{1} \mathbf{1}^T\right) Y Y^T\right]\right]$$

$$= \text{tr}\left[\left(H - \frac{1}{n} \mathbf{1} \mathbf{1}^T\right) (\sigma^2 I + \mu \mu^T)\right]$$

$$= \text{tr}\left[\sigma^2 \left(H - \frac{1}{n} \mathbf{1} \mathbf{1}^T\right) + \mu^2 \left(H - \frac{1}{n} \mathbf{1} \mathbf{1}^T\right)\right]$$

$$= \sigma^2 \text{tr}\left(H - \frac{1}{n} \mathbf{1} \mathbf{1}^T\right)$$

$$= \sigma^2 (p+1 - 1) = \sigma^2 p$$

$$\begin{aligned} E(Y Y^T) &= \text{Var}(Y) + E(Y) E(Y)^T \\ &= \sigma^2 I + \mu \mu^T \\ &= \sigma^2 I + \mathbf{1} \mu^2 \end{aligned}$$

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$n=10,$

$$BIC = n \log\left(\frac{RSS}{n}\right) + p_c \log(n)$$

RSS	$p_c(p+1)$	BIC
18.8	3	13.22
14.3	4	12.79
4.8	5	4.17
2.4.	5	-2.76.

Model 4.

AIC, Mallaw's C_p Statistics.

$$Y = X\beta + e, \quad e \sim N(0, \sigma^2)$$

$$\hat{\beta} = (X^T X + kI)^{-1} X^T Y,$$

(a). $E(\hat{\beta}) = E[(X^T X + kI)^{-1} X^T Y]$

$$= (X^T X + kI)^{-1} X^T X \beta$$

$$= (X^T X + kI)^{-1} (X^T X + kI - kI) \beta$$

$$= [I - k(X^T X + kI)^{-1}] \beta$$

$$= \beta - k(X^T X + kI)^{-1} \beta.$$

If $k=0 \Rightarrow \hat{\beta}$ is unbiased.

(b). $\text{Var}(\hat{\beta}) = (X^T X + kI)^{-1} X^T \text{Var}(Y) X (X^T X + kI)^{-1}$

$$= \sigma^2 (X^T X + kI)^{-1} (X^T X) (X^T X + kI)^{-1}$$

$$= \sigma^2 (X^T X + kI)^{-1} - k\sigma^2 (X^T X + kI)^{-2}.$$

(c). Multicollinearity.

$$\rightarrow (X^T X + kI)^{-1}.$$

~~$$\text{Var}(\hat{\beta}) = (X^T X + kI)^{-1} \sigma^2.$$~~

k makes $(X^T X + kI)$ invertible even though $X^T X$ is invertible

(a) $Y = \beta_0 + X_1\beta_1 + X_2\beta_2 + C\beta_3 + e, \quad e \sim N(0, \sigma^2 W)$

$$C = \begin{pmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 1 \\ 0 \end{pmatrix}, \quad W = \begin{pmatrix} \frac{1}{4} & & & & & \\ & \frac{1}{6} & & & & \\ & & \frac{1}{10} & & & \\ & & & \frac{1}{5} & & \\ & & & & \frac{1}{9} & \\ 0 & & & & & \frac{1}{4} \end{pmatrix}$$

(b) $X = \begin{pmatrix} 1 & 60 & 170 & 1 \\ 1 & 45 & 160 & 0 \\ 1 & 90 & 190 & 1 \\ 1 & 65 & 175 & 1 \\ 1 & 55 & 170 & 0 \\ 1 & 70 & 180 & 1 \\ 1 & 60 & 155 & 0 \end{pmatrix}, \quad Y = \begin{pmatrix} 7 \\ 6 \\ 9 \\ 7.5 \\ 6.5 \\ 8 \\ 4 \end{pmatrix}, \quad W^{-1} = \begin{pmatrix} 4 & & & & & \\ & 6 & & & & \\ & & 10 & & & \\ & & & 5 & & \\ & & & & 9 & \\ 0 & & & & & 4 \end{pmatrix}$

$$\hat{\beta} = (X^T W^{-1} X)^{-1} X^T W^{-1} Y$$

$$X^T W^{-1} X = \begin{pmatrix} 39 & 2540 & 6745 & 20 \\ 2540 & 175200 & 445825 & 1535 \\ 6745 & 445825 & 1171925 & 3635 \\ 20 & 1535 & 3635 & 20 \end{pmatrix}$$

$$X^T W^{-1} Y = \begin{pmatrix} 274 \\ 18575 \\ 480475 \\ 16435 \end{pmatrix}$$

(c) $H_0: E(Y|X) = \beta_0$
 $H_1: E(Y|X) = \beta_0 + C\beta_3$

$$X = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}, \quad W = \begin{pmatrix} 4 & & & & & \\ & 6 & & & & \\ & & 10 & & & \\ & & & 5 & & \\ & & & & 9 & \\ & & & & & 4 \end{pmatrix}$$

$$W^{-\frac{1}{2}} Y = W^{-\frac{1}{2}} X \beta + W^{-\frac{1}{2}} e$$

$$Z = M\beta + d, \quad E(d) = 0, \quad \text{Var}(d) = \sigma^2 I$$

$$RSS_1 = Z^T (I - H) Z = 32.2428, \quad df_1 = 7 - 2 = 5$$

$$RSS_0 = Z^T (I - \frac{J}{n}) Z = 300.8606, \quad df_0 = 7 - 1 = 6$$

$$F = \frac{(300.8606 - 32.2428) / (6 - 5)}{32.2428 / 5} = 41.6555$$

$$F_{0.05, 1, 5} = 6.6079$$

H_0 is Rejected

$$(d) \quad \frac{\partial \hat{Y}}{\partial X_1} = 0$$

$$\hat{X}_1 = 180,$$

$$\frac{\partial \hat{Y}}{\partial X_2} = 0$$

$$\hat{X}_2 = 75,$$

$$\frac{\partial \hat{Y}}{\partial C} = 2.4 > 0.$$

$\therefore$ A man with 180 kg of weight and 75 cm of height has the highest running speed on average.

$$X_i = x_i + a_i, \quad E(a_i) = 0, \quad \text{Var}(a_i) = \sigma_a^2$$

$$y_i = \beta_0 + \beta_1 x_i + e_i, \quad \text{Var}(e_i) = \sigma^2, \quad E(e_i) = 0$$

(a)

$$y_i = \beta_0 + \beta_1 X_i + \delta_i$$

$$= \beta_0 + \beta_1 (x_i + a_i) + \delta_i$$

$$= \beta_0 + \beta_1 x_i + \beta_1 a_i + \delta_i$$

$$\therefore e_i = \beta_1 a_i + \delta_i$$

$$\delta_i = e_i - \beta_1 a_i$$

$$E(\delta_i) = E(e_i - \beta_1 a_i)$$

$$= 0$$

$$\text{Var}(\delta_i) = \text{Var}(e_i - \beta_1 a_i)$$

$$= \text{Var}(e_i) + \beta_1^2 \text{Var}(a_i)$$

$$= \sigma^2 + \beta_1^2 \sigma_a^2$$

(b) $\text{Cov}(X_i, \delta_i) = \text{Cov}(x_i + a_i, e_i - \beta_1 a_i)$

$$= -\beta_1 \sigma_a^2$$

errors and regressor should be uncorrelated.

c) ~~$\hat{\beta} \rightarrow \frac{\text{Cov}(X, Y)}{\text{Var}(X)} = \frac{\sigma_a^2}{\sigma_a^2}$~~

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(c). $\hat{\beta}_1 \xrightarrow{P} \frac{\text{Cov}(X, Y)}{\text{Var}(X)} = \frac{\beta_1 \sigma_x^2}{\sigma_x^2 + \sigma_a^2}$ where $\sigma_x^2 = \lim_{n \rightarrow \infty} \frac{1}{n} \sum (x_i - \bar{x})^2$

$\therefore \hat{\beta}_1$ is biased.

(a)

$$Y = (2, 6, 0, 9, 7, 9)^T, \quad X = \begin{pmatrix} 1 & 1 & 2 \\ 1 & -1 & 1 \\ 1 & 1 & 3 \\ 1 & -1 & 4 \\ 1 & 1 & -5 \\ 1 & -1 & -5 \end{pmatrix}$$

$$X^T X = \begin{pmatrix} 6 & 0 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 80 \end{pmatrix}, \quad (X^T X)^{-1} = \begin{pmatrix} 1/6 & 0 & 0 \\ 0 & 1/6 & 0 \\ 0 & 0 & 1/80 \end{pmatrix}$$

$$X^T Y = \begin{pmatrix} 33 \\ -15 \\ -36 \end{pmatrix}, \quad \hat{\beta} = \begin{pmatrix} 11/2 \\ -5/2 \\ -9/20 \end{pmatrix}$$

$$\hat{Y} = \begin{pmatrix} 21/10 \\ 13/20 \\ 33/20 \\ 31/5 \\ 21/4 \\ 41/4 \end{pmatrix}, \quad \hat{e} = \begin{pmatrix} -0.1 \\ -1.55 \\ -1.65 \\ 2.8 \\ 1.75 \\ -1.25 \end{pmatrix}, \quad \hat{\sigma}^2 = \frac{\hat{e}^T \hat{e}}{6-3} = \frac{88}{15} = 5.8667$$

(b) $X_* = \begin{pmatrix} 1 \\ -1 \\ 3 \end{pmatrix}$

$$X_*^T (X^T X)^{-1} X_* = 107/240$$

$$\text{sepred}(\hat{y}_* | X_*) = \sqrt{\hat{\sigma}^2 (1 + 107/240)} = 2.9124$$

$$\hat{y}_* | X_* = 133/20, \quad t_{0.005, 3} = 4.032$$

$$= [-5.0928, 18.3928]$$

(c) Test $H_0: \beta_1 = \beta_2$ vs. $H_0: \beta_1 \neq \beta_2$

$$\Rightarrow H_0: \beta_1 - \beta_2 = 0 \quad \text{vs.} \quad H_A: \beta_1 - \beta_2 \neq 0$$

Find $\hat{\beta}_1 - \hat{\beta}_2$, $\text{Var}(\hat{\beta}_1 - \hat{\beta}_2) = \text{Var}(\hat{\beta}_1) + \text{Var}(\hat{\beta}_2) - 2\text{Cov}(\hat{\beta}_1, \hat{\beta}_2)$

$$T.S. = \frac{\hat{\beta}_1 - \hat{\beta}_2 - 0}{\sqrt{\text{Var}(\hat{\beta}_1 - \hat{\beta}_2)}} \sim t_3$$