

1. We learned the general result:

If $\mathbf{Z} \sim N(\boldsymbol{\mu}, \Sigma)$, and if \mathbf{a} and M are nonrandom, then $\mathbf{a} + M\mathbf{Z} \sim N(\mathbf{a} + M\boldsymbol{\mu}, M\Sigma M^T)$

Apply the general result to specific cases by identifying \mathbf{Z} , $\boldsymbol{\mu}$, Σ , \mathbf{a} , and M for the specific cases.

(i) (2) $\boldsymbol{\varepsilon} \sim N(\mathbf{0}, \sigma^2 I_n)$, find distribution of $\mathbf{Y} = X\mathbf{b} + \boldsymbol{\varepsilon}$

$$\mathbf{Z} = \boldsymbol{\varepsilon} \quad \boldsymbol{\mu} = \mathbf{0} \quad \Sigma = \sigma^2 I \quad \mathbf{a} = X\mathbf{b} \quad M = I$$

$$\mathbf{Y} \sim N(X\mathbf{b} + I\mathbf{0}, I\sigma^2 I I) = N(X\mathbf{b}, \sigma^2 I)$$

(ii) (2) $\mathbf{Y} \sim N(?, ?)$, find distribution of $\hat{\mathbf{B}} = (X^T X)^{-1} X^T \mathbf{Y}$

$$\mathbf{Z} = \mathbf{Y} \quad \boldsymbol{\mu} = X\mathbf{b} \quad \Sigma = \sigma^2 I \quad \mathbf{a} = \mathbf{0} \quad M = (X^T X)^{-1} X^T$$

$$\hat{\mathbf{B}} \sim N((X^T X)^{-1} X^T X\mathbf{b}, (X^T X)^{-1} X^T \sigma^2 I X (X^T X)^{-1})$$

$$N(\mathbf{b}, \sigma^2 (X^T X)^{-1})$$

(iii) (2) $\mathbf{Y} \sim N(?, ?)$, find distribution of $\hat{\mathbf{Y}} = X(X^T X)^{-1} X^T \mathbf{Y} = P\mathbf{Y}$

$$\mathbf{Z} = \mathbf{Y} \quad \boldsymbol{\mu} = X\mathbf{b} \quad \Sigma = \sigma^2 I \quad \mathbf{a} = \mathbf{0} \quad M = P$$

$$\hat{\mathbf{Y}} \sim N(PX\mathbf{b}, P\sigma^2 I P^T)$$

$$N(X\mathbf{b}, \sigma^2 P P^T) = N(X\mathbf{b}, \sigma^2 P)$$

(iv) (2) $\mathbf{Y} \sim N(?, ?)$, find distribution of $\hat{\boldsymbol{\varepsilon}} = (I - X(X^T X)^{-1} X^T) \mathbf{Y} = (I - P)\mathbf{Y}$

$$\mathbf{Z} = \mathbf{Y} \quad \boldsymbol{\mu} = X\mathbf{b} \quad \Sigma = \sigma^2 I \quad \mathbf{a} = \mathbf{0} \quad M = I - P$$

$$\hat{\boldsymbol{\varepsilon}} \sim N((I - P)X\mathbf{b}, (I - P)\sigma^2 I (I - P)^T)$$

$$N(X\mathbf{b} - PX\mathbf{b}, \sigma^2 (I - P)) = N(\mathbf{0}, \sigma^2 (I - P))$$

2. Consider the model $\mathbf{Y} = X\mathbf{b} + \boldsymbol{\varepsilon}$, $\boldsymbol{\varepsilon} \sim N(\mathbf{0}, \sigma^2 I)$, with design matrix: $X = \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 1 & 1 \\ 1 & 1 \end{bmatrix}$
- (i) (3) Express the model by filling in the entries of the matrices below:

$$\begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \\ Y_4 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} b_0 \\ b_1 \end{bmatrix} + \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \varepsilon_3 \\ \varepsilon_4 \end{bmatrix}$$

- (ii) (4) Do the matrix operations from (i) in order to write the model for each response:

$$Y_1 = b_0 + \varepsilon_1$$

$$Y_2 = b_0 + \varepsilon_2$$

$$Y_3 = b_0 + b_1 + \varepsilon_3$$

$$Y_4 = b_0 + b_1 + \varepsilon_4$$

- (iii) (3) What is $E(Y_3) - E(Y_1)$ in terms of the elements of \mathbf{b} ?

$$E(Y_3) - E(Y_1) = b_0 + b_1 - b_0 = b_1$$

- (iv) (3) Given $X^T X = \begin{bmatrix} 4 & 2 \\ 2 & 2 \end{bmatrix}$ and $(X^T X)^{-1} = \begin{bmatrix} 1/2 & -1/2 \\ -1/2 & 1 \end{bmatrix}$, what is \hat{B}_0 in terms of elements of \mathbf{Y} ?

$$\begin{aligned} \hat{\mathbf{B}} &= \begin{bmatrix} \hat{B}_0 \\ \hat{B}_1 \end{bmatrix} = (X^T X)^{-1} X^T \mathbf{Y} = \begin{bmatrix} 1/2 & -1/2 \\ -1/2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \\ Y_4 \end{bmatrix} \\ &= \begin{bmatrix} 1/2 & 1/2 & 0 & 0 \\ -1/2 & -1/2 & 1/2 & 1/2 \end{bmatrix} \begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \\ Y_4 \end{bmatrix} = \begin{bmatrix} \frac{1}{2}Y_1 + \frac{1}{2}Y_2 \\ -\frac{1}{2}Y_1 - \frac{1}{2}Y_2 + \frac{1}{2}Y_3 + \frac{1}{2}Y_4 \end{bmatrix} = \hat{B}_0 \end{aligned}$$

(v) (3) What is the variance of \hat{B}_0 ?

$$\text{Cov}\left(\begin{bmatrix} \hat{B}_0 \\ \hat{B}_1 \end{bmatrix}\right) = \sigma^2 (X^T X)^{-1} = \sigma^2 \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & 1 \end{bmatrix}$$

$$\text{so } \text{Var}(\hat{B}_0) = \sigma^2 \cdot \frac{1}{2}$$

(vi) (3) What is the correlation between residuals \hat{e}_1 and \hat{e}_2 given that $P =$

$$P = \begin{bmatrix} 1/2 & 1/2 & 0 & 0 \\ 1/2 & 1/2 & 0 & 0 \\ 0 & 0 & 1/2 & 1/2 \\ 0 & 0 & 1/2 & 1/2 \end{bmatrix}$$

$$\text{Cov}(\hat{e}) = \sigma^2 (I - P) = \sigma^2 \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} & 0 & 0 \\ -\frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{2} & -\frac{1}{2} \\ 0 & 0 & -\frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$\text{Corr}(\hat{e}_1, \hat{e}_2) = \frac{\text{Cov}(\hat{e}_1, \hat{e}_2)}{\sqrt{\text{Var}(\hat{e}_1) \text{Var}(\hat{e}_2)}}$$

$$= \frac{-\sigma^2/2}{\sqrt{\frac{\sigma^2}{2} \cdot \frac{\sigma^2}{2}}} = -1$$

3. (3) $T = Z/\sqrt{W/k}$ has a t distribution with k degrees of freedom if what 3 things are true?

(1) $Z \sim N(0, 1)$

(2) $W \sim \chi_k^2$

(3) Z independent of W