

1. Consider a drug trial where the first two subjects are sisters, and the third and fourth subjects are unrelated to each other and to the two sisters. Consider the following model:

$$Y_i = b_0 + b_2 x_{i2} + \varepsilon_i$$

where $x_{i2} = 1$ if subject i got the drug and 0 otherwise, and $\varepsilon_i \sim N(0, \sigma^2)$, that is, all variances are σ^2 . However, the covariance between the two sisters is $0.5\sigma^2$, and all other covariances are zero.

- (i) (3) Write down the covariance matrix Σ for $(Y_1, Y_2, Y_3, Y_4)^T$.

$$\Sigma = \begin{bmatrix} \sigma^2 & \sigma^2/2 & 0 & 0 \\ \sigma^2/2 & \sigma^2 & 0 & 0 \\ 0 & 0 & \sigma^2 & 0 \\ 0 & 0 & 0 & \sigma^2 \end{bmatrix}$$

- (ii) (3) Subjects 1 and 3 got a placebo. Subjects 2 and 4 got the drug. What is the design matrix X ?

$$X = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 0 \\ 1 & 1 \end{bmatrix}$$

- (iii) (4) Show that the following estimator is unbiased for b_2 :

$$\hat{B}_2 = \frac{2}{3}(Y_2 - Y_1) + \frac{1}{3}(Y_4 - Y_3)$$

$$E(\hat{B}_2) = \frac{2}{3}(E(Y_2) - E(Y_1)) + \frac{1}{3}(E(Y_4) - E(Y_3))$$

$$= \frac{2}{3}(b_0 + b_2 - b_0) + \frac{1}{3}(b_0 + b_2 - b_0)$$

$$= \frac{2}{3}b_2 + \frac{1}{3}b_2 = b_2 \checkmark$$

(iv) (4) Calculate the following variances. Hint: $\text{Cov}(Y_2 - Y_1, Y_4 - Y_3) = 0$

$$\begin{aligned}\text{Var}(Y_2 - Y_1) &= \text{Var}(Y_2) + \text{Var}(Y_1) - 2 \text{Cov}(Y_2, Y_1) \\ &= \sigma^2 + \sigma^2 - 2 \frac{\sigma^2}{2} = \sigma^2\end{aligned}$$

$$\begin{aligned}\text{Var}(Y_4 - Y_3) &= \text{Var}(Y_4) + \text{Var}(Y_3) - 2 \text{Cov}(Y_4, Y_3) \\ &= \sigma^2 + \sigma^2 + 0 = 2\sigma^2\end{aligned}$$

$$\begin{aligned}\text{Var}\left(\frac{2}{3}(Y_2 - Y_1) + \frac{1}{3}(Y_4 - Y_3)\right) &= \frac{4}{9} \text{Var}(Y_2 - Y_1) + \frac{1}{9} \text{Var}(Y_4 - Y_3) + 0 \\ &= \frac{4}{9} \sigma^2 + \frac{1}{9} 2\sigma^2 = \frac{6}{9} \sigma^2 = \frac{2}{3} \sigma^2\end{aligned}$$

2. Consider this dataset and model for y_i :

i	$j(i)$	y_i
1	1	6.2
2	1	9.7
3	2	4.6
4	3	10.2

$$Y_i = b_0 + B_{j(i)} + \varepsilon_i, \quad B_1, \dots, B_3 \stackrel{\text{ind}}{\sim} N(0, \sigma_1^2), \quad \varepsilon_1, \dots, \varepsilon_4 \stackrel{\text{ind}}{\sim} N(0, \sigma_2^2)$$

(i) (3) What is the design matrix X ?

$$X = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

(ii) (3) What is the covariance matrix Σ

$$\Sigma = \begin{bmatrix} \sigma_1^2 + \sigma_2^2 & \sigma_1^2 & 0 & 0 \\ \sigma_1^2 & \sigma_1^2 + \sigma_2^2 & 0 & 0 \\ 0 & 0 & \sigma_1^2 + \sigma_2^2 & 0 \\ 0 & 0 & 0 & \sigma_1^2 + \sigma_2^2 \end{bmatrix}$$

3. In the moving average model, we have

$$Y_i = b_0 + \rho \varepsilon_{i-1} + \varepsilon_i, \quad \varepsilon_i \stackrel{\text{ind}}{\sim} N(0, \sigma^2)$$

Let $\underline{\varepsilon} = (\varepsilon_0, \varepsilon_1, \varepsilon_2, \varepsilon_3)^T$, and $\mathbf{Y} = (Y_1, Y_2, Y_3)^T$.

(i) (3) Find $\underline{\mu} = E(\mathbf{Y})$

$$E(\underline{Y}) = \begin{bmatrix} b_0 \\ b_0 \\ b_0 \\ \cancel{b_0} \end{bmatrix}$$

$$\begin{aligned} \text{b.c. } E(Y_i) &= b_0 + \rho E(\varepsilon_{i-1}) + E(\varepsilon_i) \\ &= b_0 + 0 + 0 \end{aligned}$$

(ii) (4) Find a matrix M such that $\mathbf{Y} = \underline{\mu} + M\underline{\varepsilon}$. Pay attention to the dimensions.

$$Y_1 = b_0 + \rho \varepsilon_0 + \varepsilon_1$$

$$Y_2 = b_0 + \rho \varepsilon_1 + \varepsilon_2$$

$$Y_3 = b_0 + \rho \varepsilon_2 + \varepsilon_3$$

$$\cancel{Y_4 = b_0 + \rho \varepsilon_3 + \varepsilon_4}$$

$$M\underline{\varepsilon} = \begin{bmatrix} \rho & 1 & 0 & 0 \\ 0 & \rho & 1 & 0 \\ 0 & 0 & \rho & 1 \end{bmatrix} \begin{bmatrix} \varepsilon_0 \\ \varepsilon_1 \\ \varepsilon_2 \\ \varepsilon_3 \end{bmatrix}$$

(iii) (3) What is the covariance matrix for \mathbf{Y} ? (Hint, you can use your previous answer).

$$\begin{aligned} \text{Cov}(\underline{Y}) &= M \text{Cov}(\underline{\varepsilon}) M^T = M \sigma^2 I M^T = \sigma^2 M M^T \\ &= \sigma^2 \begin{bmatrix} \rho & 1 & 0 & 0 \\ 0 & \rho & 1 & 0 \\ 0 & 0 & \rho & 1 \end{bmatrix} \begin{bmatrix} \rho & 0 & 0 \\ 1 & \rho & 0 \\ 0 & 1 & \rho \\ 0 & 0 & 1 \end{bmatrix} = \sigma^2 \begin{bmatrix} \rho^2 + 1 & \rho & 0 \\ \rho & \rho^2 + 1 & \rho \\ 0 & \rho & \rho^2 + 1 \end{bmatrix} \end{aligned}$$

4. (3) Bonus: In the autoregressive model, we have (in simplified form):

$$Y_1 = \frac{\sigma}{\sqrt{1-\rho^2}}\varepsilon_1, \quad Y_i = \rho Y_{i-1} + \varepsilon_i \quad \text{for } i = 2, 3, 4, \quad \varepsilon \sim N(0, \sigma^2 I_4)$$

Find a matrix M such that $\mathbf{Y} = (Y_1, Y_2, Y_3, Y_4)^T = M^{-1}\varepsilon$.

$$Y_1 = \frac{\sigma}{\sqrt{1-\rho^2}} \varepsilon_1$$

$$Y_1 \frac{\sqrt{1-\rho^2}}{\sigma} = \varepsilon_1$$

$$Y_2 - \rho Y_1 = \varepsilon_2$$

$$Y_3 - \rho Y_2 = \varepsilon_3$$

$$Y_4 - \rho Y_3 = \varepsilon_4$$

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$$M \underline{Y} = \underline{\varepsilon}$$

$$\begin{bmatrix} \frac{\sigma}{\sqrt{1-\rho^2}} & 0 & 0 & 0 \\ -\rho & 1 & 0 & 0 \\ 0 & -\rho & 1 & 0 \\ 0 & 0 & -\rho & 1 \end{bmatrix} \begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \\ Y_4 \end{bmatrix} = \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \varepsilon_3 \\ \varepsilon_4 \end{bmatrix}$$

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$$\underline{Y} = M^{-1} \underline{\varepsilon}$$