

Question1 (2 points)

Let $Y_i = \beta x_i + \varepsilon_i$ for $i = 1, 2$, where

$$\varepsilon_1 \sim \text{Normal}(0, \sigma^2) \quad \text{and} \quad \varepsilon_2 \sim \text{Normal}(0, 2\sigma^2)$$

Suppose ε_1 and ε_2 are statistically independent. For $x_1 = 1$ and $x_2 = -1$, find the weighted least squares estimate of β and find the variance of your estimate in terms of σ^2 .

Question2 (4 points)

Consider the AR(1) model

$$Y_t = 5 - 0.7Y_{t-1} + \varepsilon_t \quad \text{where} \quad \sigma_\varepsilon^2 = \text{Var}[\varepsilon_t] = 2 \quad \text{for all } t.$$

- (a) (1 point) Is this process stationary? Explain your answer.
- (b) (1 point) What is the mean of this process?
- (c) (1 point) What is the variance of this process?
- (d) (1 point) What is the covariance of this process?

Question3 (3 points)

Let Y_t be a stationary AR(2) process

$$(Y_t - \mu) = \phi_1 (Y_{t-1} - \mu) + \phi_2 (Y_{t-2} - \mu) + \varepsilon_t$$

- (a) (1 point) Show that the ACF of Y_t satisfies the following equation

$$\rho(k) = \phi_1 \rho(k-1) + \phi_2 \rho(k-2) \quad \text{for } k = 1, 2, \dots$$

- (b) (1 point) Show ϕ_1 and ϕ_2 solves the following system

$$\begin{bmatrix} \rho(1) \\ \rho(2) \end{bmatrix} = \begin{bmatrix} 1 & \rho(1) \\ \rho(1) & 1 \end{bmatrix} \begin{bmatrix} \phi_1 \\ \phi_2 \end{bmatrix}$$

- (c) (1 point) Suppose that $\rho(1) = 0.4$ and $\rho(2) = 0.2$, find $\rho(3)$.

Question4 (1 points)

Suppose that $\mathbb{E}[\mathbf{Y}] = \boldsymbol{\theta}$, $\mathbf{A}\boldsymbol{\theta} = \mathbf{0}$, and $\text{Var}[\mathbf{Y}] = \sigma^2 \mathbf{V}$, where \mathbf{A} is an $m \times n$ matrix of rank m and \mathbf{V} is a known $n \times n$ positive-definite matrix. Let $\hat{\boldsymbol{\theta}}$ be the generalised least squares estimate of $\boldsymbol{\theta}$; that is, $\hat{\boldsymbol{\theta}}$ minimises

$$(\mathbf{Y} - \boldsymbol{\theta})^T \mathbf{V}^{-1} (\mathbf{Y} - \boldsymbol{\theta}) \quad \text{subject to} \quad \mathbf{A}\boldsymbol{\theta} = \mathbf{0}$$

Show that

$$\mathbf{Y} - \hat{\boldsymbol{\theta}} = \mathbf{V}\mathbf{A}^T \hat{\boldsymbol{\gamma}}$$

where $\hat{\boldsymbol{\gamma}}$ is the generalised least squares estimate of $\boldsymbol{\gamma}$ for the model

$$\mathbb{E}[\mathbf{Y}] = \mathbf{V}\mathbf{A}^T \boldsymbol{\gamma} \quad \text{and} \quad \text{Var}[\mathbf{Y}] = \sigma^2 \mathbf{V}$$