

$$\underline{\underline{Q.2}} \quad y_t = \beta_1 \sin(2\pi\omega t) + \beta_2 \cos(2\pi\omega t) + \varepsilon_t + 0.5 \varepsilon_{t-1}$$

$$\varepsilon_t \sim WN(0, \sigma^2)$$

ω, β_1, β_2 are constants.

Mean:

$$E(y_t) = E[\beta_1 \sin(2\pi\omega t) + \beta_2 \cos(2\pi\omega t) + \varepsilon_t + 0.5 \varepsilon_{t-1}]$$

$$= \beta_1 \sin(2\pi\omega t) + \beta_2 \cos(2\pi\omega t) + E(\varepsilon_t + 0.5 \varepsilon_{t-1})$$

[\because Due to linearity of Expectation and the first two terms are constant]

$$= \beta_1 \sin(2\pi\omega t) + \beta_2 \cos(2\pi\omega t) + E(\varepsilon_t) + 0.5 E(\varepsilon_{t-1})$$

$$\text{Again, } \varepsilon_t \sim WN(0, \sigma^2)$$

$$\therefore E(\varepsilon_t) = E(\varepsilon_{t-1}) = 0$$

$$\text{So } E(y_t) = \beta_1 \sin(2\pi\omega t) + \beta_2 \cos(2\pi\omega t)$$

Variance:-

$$V(y_t) = V[\underbrace{\beta_1 \sin(2\pi\omega t) + \beta_2 \cos(2\pi\omega t)}_{\text{non stochastic}} + \varepsilon_t + 0.5 \varepsilon_{t-1}]$$

$$= V(\varepsilon_t + 0.5 \varepsilon_{t-1}) \quad [\because \text{Var}(a+x) = \text{Var}(x)]$$

$$= V(\varepsilon_t) + 0.5^2 V(\varepsilon_{t-1}) + 2 \times 0.5 \times \text{Cov}(\varepsilon_t, \varepsilon_{t-1})$$

$$= \sigma^2 + 0.25\sigma^2 = 1.25\sigma^2 \quad \left[\begin{array}{l} \because \varepsilon_t \sim WN(0, \sigma^2) \\ \Rightarrow \text{Cov}(\varepsilon_t, \varepsilon_{t-1}) = 0 \end{array} \right]$$

$$\gamma(1) = \text{Cov}(y_{t+1}, y_t)$$

$$= \text{Cov}(a + \varepsilon_{t+1} + 0.5 \varepsilon_t, b + \varepsilon_t + 0.5 \varepsilon_{t-1}) \quad [a, b \text{ constants}]$$

$$= \text{Cov}(a, b) + \text{Cov}(a, \varepsilon_t + 0.5 \varepsilon_{t-1}) + \text{Cov}(\varepsilon_{t+1} + 0.5 \varepsilon_t, b)$$

$$+ \text{Cov}(\varepsilon_{t+1} + 0.5 \varepsilon_t, \varepsilon_t + 0.5 \varepsilon_{t-1})$$

[~~∴~~ By the distributive property of Covariance]

$$\begin{aligned}
 &= \text{Cov}(E_{t+1} + 0.5 E_t, E_t + 0.5 E_{t-1}) \quad [\because \text{Cov}(a, x) = 0 \text{ as } a \text{ is const}] \\
 &= \text{Cov}(E_{t+1}, E_t) + 0.5 \text{Cov}(E_t, E_t) + 0.5 \text{Cov}(E_{t+1}, E_{t-1}) \\
 &\quad + 0.5^2 \text{Cov}(E_t, E_{t-1}) \\
 &= 0.5 \sigma^2 \quad [\because \text{Cov}(E_t, E_{t+h}) = 0 \quad \forall h \in \mathbb{Z}]
 \end{aligned}$$

$$S(2) = \frac{\gamma(2)}{\gamma(0)}.$$

$$\gamma(2) = \text{Cov}(Y_{t+2}, Y_t).$$

$$= \text{Cov}(a + E_{t+2} + 0.5 E_{t+1}, b + E_t + 0.5 E_{t-1})$$

$$= \text{Cov}(E_{t+2} + 0.5 E_{t+1}, E_t + 0.5 E_{t-1})$$

$$\begin{aligned}
 &= \text{Cov}(E_{t+2}, E_t) + 0.5 \text{Cov}(E_{t+1}, E_t) + 0.5 \text{Cov}(E_{t+2}, E_{t-1}) \\
 &\quad + 0.5^2 \text{Cov}(E_{t+1}, E_{t-1}).
 \end{aligned}$$

$$= 0.$$

$$\text{So } S(2) = \frac{0}{\gamma(0)} = 0 \quad [\gamma(0) = \text{Var}(Y_t) \neq 0]$$

3) x_t, y_t are two stationary process.

$$z_t = x_t + y_t.$$

$$\begin{aligned}
 \textcircled{1} E(z_t) &= E(\cancel{x_t} + \cancel{x_t}) E(x_t + y_t) \\
 &= E(x_t) + E(y_t)
 \end{aligned}$$

[~~is~~ independent of t as x_t, y_t stationary]

$$\textcircled{1} E(Z_t^2) = E[(x_t + y_t)^2] = E[x_t^2] + E[y_t^2] + 2E[x_t, y_t]$$

$$\text{If } E[x_t, y_t] < \infty$$

$$\text{then } E[Z_t^2] < \infty$$

x_t, y_t are stationary process.

$$\begin{aligned} \textcircled{1} \text{Cov}(Z_{t+h}, Z_t) &= \text{Cov}(x_{t+h} + y_{t+h}, x_t + y_t) \\ &= \text{Cov}(x_{t+h}, x_t) + \text{Cov}(y_{t+h}, x_t) + \text{Cov}(x_{t+h}, y_t) \\ &\quad + \text{Cov}(y_{t+h}, y_t). \end{aligned}$$

It is independent of t if $\text{Cov}(x_{t+h}, y_t)$ and $\text{Cov}(y_{t+h}, y_t)$ are independent of t .

If x_t and y_{t+h} are independent then

$$\text{Cov}(x_t, y_{t+h}) = 0$$

$\Rightarrow \text{Cov}(Z_{t+h}, Z_t)$ is independent of t .

So Z_t will be stationary based on finite $E[x_t, y_t]$ and independence of $\text{Cov}(x_t, y_{t+h}) \forall h \in \mathbb{Z}$ with

If x_t, y_t independent, then Z_t is stationary.