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MA 641 HW 1

Q6] If x and y are dependent
but $V(x) = V(y)$
 $\Rightarrow \text{Cor}(x+y, x-y)$

$$\therefore \text{Cor}(x+y, x-y) = E[(x+y)(x-y)] - E(x+y)E(x-y)$$

$$\text{which gives us} \\ \Rightarrow E(x^2 - y^2) - [E(x^2) - E(y^2)]$$

$$\text{but as we know} \\ E(x) = E(y)$$

therefore we get

$$\Rightarrow E(x^2 - y^2) - 0 \Rightarrow 0$$

$$\because E(x^2) = E(y^2)$$

$$V(x) = E(x^2) - [E(x)]^2$$

$$V(y) = E(y^2) - [E(y)]^2$$

$$\therefore \text{Cor}(x+y, x-y) = E(x^2 - y^2) - [E(x^2) - E(y^2)] \\ = 0$$

Q4] To show that $\{y_t\}$ is stationary
we need to show $\mu_t = E(y_t) = \mu$ is constant for all t and
 $y_{t+k} = \text{Cor}(y_t, y_{t+k})$

a] To find $\rightarrow E(y_t)$

$$E(y_t) = E\left(\sum_{s=-\infty}^{\infty} a_s e^{t-s}\right)$$

$$= \sum_{s=-\infty}^{\infty} a_s E(e^{t-s})$$

but from the given in the Q we get

$$E(Y_t) = \sum_{s=-\infty}^{\infty} a_s \times 0 = 0$$

which tells us $E(Y_t) = 0$ for all t and is stationary.

b] $\text{Cov}(Y_t, Y_{t-k}) = \sigma^2 \sum_{i=-\infty}^{\infty} a_{k+i} a_i$

To find :- Is the process $\{Y_t\}$ stationary

$$\Rightarrow \text{Cov}(Y_t, Y_{t-k}) = \text{Cov}\left(\sum_{s=-\infty}^{\infty} a_s e_{t-s}, \sum_{s=-\infty}^{\infty} a_s e_{t-k-s}\right)$$

i) for $t \neq s$ $E(e_t) = 0$

\therefore we get cov between e_t and e_s will be 0.

ii) for $t = s$

$$\begin{aligned} \text{Cov}(Y_t, Y_{t-k}) &= \text{Cov}(a_0 e_t, a_k e_{t-k}) \\ &\Rightarrow a_0 a_k \text{Cov}(e_t, e_{t-k}) \end{aligned}$$

as we know e_t and e_{t-k} are iid which means that cov depends on lag k

$$\text{Cov}(e_t, e_{t-k}) = \sigma^2 \gamma_k$$

simply giving us

$$\text{Cov}(Y_t, Y_{t-k}) = \begin{cases} a_0 a_k \sigma^2 & k \neq 0 \\ 0 & k = 0 \end{cases}$$

$$\therefore \text{Cov}(Y_t, Y_{t-k}) = \sigma^2 \sum_{i=-\infty}^{\infty} a_{k+i} a_i$$

Hence, we get to know that it clearly depends on the lag k and not on time t.

$\therefore \{Y_t\}$ is Stationary

(cov of Y_t and $E(Y_t)$ is not dependent on t.)

Q7] a) $y_t = x$

$$E(x) = \mu \text{ and } \text{Var}(x) = \sigma^2$$

Let us assume that x does not depend on t

then $E(y_t) = E(x) = \mu$

$$\text{Var}(y_t) = \text{Var}(x) = \sigma^2$$

$$\text{Cor}(y_t, y_{t-k}) = \text{Cor}(x, x)$$

giving us $\text{Var}(x)$

which is nothing but σ^2

As we can understand from alone that none of the alone are dependent on t , the process $\{y_t\}$ is stationary

b) To find - γ_k for $\{y_t\}$

$$\gamma_k = \text{Cor}(y_t, y_{t-k})$$

$$= \text{Cor}(x, x)$$

giving us $\text{Var}(x) = \sigma^2$

$$\therefore \gamma_k = \sigma^2$$

Q2) Given - first difference $w_t = y_t - y_{t-1}$

Second difference $z_t = w_t - w_{t-1}$

To write the model in the form

$$y_t = \phi_1 y_{t-1} + \phi_2 y_{t-2} + e_t$$

Now, $z_t = w_t - w_{t-1}$

$$= y_t - y_{t-1} - (y_{t-1} - y_{t-2})$$

$$= y_t - 2y_{t-1} + y_{t-2}$$

$$y_t = z_t + 2y_{t-1} - y_{t-2}$$

$$y_t = e_t + 2y_{t-1} - y_{t-2}$$

where: $\phi_1 = 2$, $\phi_2 = -1$, $e_t = 2t$

∴ Model, $y_t = 2y_{t-1} - y_{t-2} + e_t$

$$Q5(a) \quad y_t = \sum_{i=1}^t e_i \quad E(e_t) = u > 0$$

Now we get

$$E(y_t) = E\left(\sum_{i=1}^t e_i\right) \Rightarrow \sum_{i=1}^t E(e_i)$$

which gives us nothing but

$$E(y_t) = t u$$

which tells us the process $\{y_t\}$ is not stationary because $E(y_t)$ depends on t

$$b) \quad E(y_t) = \sum_{i=1}^t E(e_i) = 0$$

$$\therefore \text{Var}(y_t) = \text{Var}\left(\sum_{i=1}^t e_i\right) \\ = \sum_{i=1}^t \text{Var}(e_i)$$

From the given information as we know $\text{Var}(e_t) = \sigma^2$

$$\text{Hence } \Rightarrow \sum_{i=1}^t \text{Var}(e_i) = t \sigma^2$$

which tells us the process $\{y_t\}$ is not stationary because $\text{Var}(y_t)$ depends on t .

Q) Cauchy-Schwarz Inequality says that for any $R^V X \in \mathbb{R}$

$$[E(XY)]^2 \leq E(X^2) E(Y^2)$$

if $E(X^2) = 0$ then $P(X=0) = 1$
and $E(XY) = 0$

a) Now, $0 \leq E(X^2) < \infty$ and $0 \leq E(Y^2) < \infty$

For $a \neq b$, we get

$$\textcircled{1} \quad 0 \leq E[(ax+by)^2] = a^2 E(X^2) + b^2 E(Y^2) + 2ab E(XY)$$

$$\textcircled{2} \Rightarrow 0 \leq E[(ax-by)^2] = a^2 E(X^2) + b^2 E(Y^2) - 2ab E(XY)$$

For \textcircled{1}

$$0 \leq E[(\sqrt{E(Y^2)}X + \sqrt{E(X^2)}Y)^2] = E(Y^2) E(X^2) + E(X^2) E(Y^2) + 2\sqrt{E(Y^2) E(X^2)} E(XY)$$

giving us

$$0 \leq E(X^2) E(Y^2) + \sqrt{E(Y^2) E(X^2)} E(XY)$$

\textcircled{3}

In the same way,

$$0 \leq E(ax-by)^2 \text{ will give}$$

$$0 \leq E(X^2) E(Y^2) - \sqrt{E(Y^2) E(X^2)} E(XY)$$

\textcircled{4}

Solving \textcircled{3}

$$-\sqrt{E(Y^2) E(X^2) E(XY)} \leq E(X^2) E(Y^2)$$

$$E(XY) \leq \sqrt{E(X^2) E(Y^2)}$$

$$\text{Similarly, } E(XY) \geq -\sqrt{E(X^2) E(Y^2)}$$

Hence we get

$$-\sqrt{E(X^2) E(Y^2)} \leq E(XY) \leq \sqrt{E(X^2) E(Y^2)}$$

$$\text{Simplifying } (E(XY))^2 \leq E(X^2) E(Y^2)$$

telling us that the Cauchy-Schwarz holds

b) $[E(uw)]^2 \leq E(u^2)E(w^2)$
 $\sqrt{E(u^2)E(w^2)} \leq E(uw) \leq \sqrt{E(u^2)E(w^2)}$

$u = x - E(x)$
 $u^2 = (x - E(x))^2$

$E(u^2) = E[(x - E(x))^2] \Rightarrow \text{Var}(x)$

$E(w^2) = \text{Var} Y$

$-\sqrt{\text{Var}(x)\text{Var}(Y)} \leq \text{corr}(xY) \leq \sqrt{\text{Var}(x)\text{Var}(Y)}$

$-1 \leq \frac{\text{corr}(xY)}{\sqrt{\text{Var}(x)\text{Var}(Y)}} \leq 1$

giving us $-1 \leq \text{corr}(xY) \leq 1$