

## Assignment - 4

Obtain mean, variance, ACCVF and ACF of following models,

i)  $AR(1)$

ii)  $AR(3)$

iii)  $AR(p)$

iv) Draw ACF and PACF plot

Sol<sup>n</sup>

i)  $AR(1) :-$

$$y_t = \delta + \phi_1 y_{t-1} + u_t \rightarrow (1) \quad \left\{ u_t \sim N(0, \sigma^2) \right\}$$

Reversing back with lag 1

$$y_{t-1} = \delta + \phi_1 y_{t-2} + u_{t-1} \rightarrow (2)$$

putting (2) in (1), we get

$$\begin{aligned} y_t &= \delta + \phi_1 (\delta + \phi_1 y_{t-2} + u_{t-1}) + u_t \\ &= \delta + \phi_1 \delta + \phi_1^2 y_{t-2} + \phi_1 u_{t-1} + u_t \rightarrow (3) \end{aligned}$$

Similarly,  $y_{t-2} = \delta + \phi_1 y_{t-3} + u_{t-2} \rightarrow (4)$

Putting eq (4) in eq (3)

$$y_t = \delta + \phi_1 \delta + \phi_1^2 \delta + \phi_1^3 y_{t-3} + \phi_1^2 u_{t-1} + \phi_1 u_{t-1} + u_t$$

Similarly after  $p^{\text{th}}$  period recursion,

$$y_t = \delta \left[ \sum_{j=0}^p \phi_1^j \right] + \phi_1^{p+1} y_{t-p-1} + \sum_{j=0}^p \phi_1^j u_{t-j} \rightarrow (5)$$

Thus this will be invertible if  $|\phi_1| < 1$  as  $p \rightarrow \infty$

$$y_t = \delta \left[ \sum_{j=0}^{\infty} \phi_1^j \right] + 0 + \left[ \sum_{j=0}^{\infty} \phi_1^j u_{t-j} \right]$$

$$= \frac{\delta}{1-\phi_1} + \left[ \sum_{j=0}^{\infty} \phi_1^j u_{t-j} \right] \rightarrow (6)$$

From eq (6), we can say that  $MA(\infty)$  represents stationary  $AP(1)$

Say:  $MA(\infty)$

$$y_t = \alpha + \sum_{k=0}^{\infty} \theta_k \varepsilon_{t-k} \rightarrow (7) \quad \left\{ \theta_k = \phi_1^k, \varepsilon_t = u_t \right\}$$

Comparing (6) and (7)

$$\bullet E(y_t) = \alpha = \frac{\delta}{1-\phi_1}$$



$$\bullet V(y_t) = \left[ \sum_{k=0}^{\infty} \theta_k^2 \right] \sigma^2 = \sigma^2 \sum_{j=0}^{\infty} (\phi_1^j)^2$$

$$= \sigma^2 (1 + \phi_1^2 + \dots)$$

$$V(y_t) = \frac{\sigma^2}{1 - \phi_1^2}$$

$$\bullet \text{ACVF } \gamma(k) = \sigma^2 \sum_{i=0}^{\infty} \theta_i \theta_{i+k}$$

$$= \sigma^2 \sum_{j=0}^{\infty} \phi_1^j \phi_1^{j+k}$$

$$= \sigma^2 \phi_1^k \sum_{i=0}^{\infty} \phi_1^{2i}$$

$$\gamma(0) = \frac{\sigma^2}{1 - \phi_1^2}$$

$$\gamma(k) = \phi_1^k \gamma(0)$$

$$\bullet \text{ACF } \rho(k) = \frac{\gamma(k)}{\gamma(0)}$$

$$\rho(k) = \phi_1^k$$

(ii) AR(3) :-

$$y_t = \delta + \phi_1 y_{t-1} + \phi_2 y_{t-2} + \phi_3 y_{t-3} + u_t$$

$$\{u_t \sim N(0, \sigma^2)\}$$

• Mean  $E(y_t) = \delta + \phi_1 E(y_{t-1}) + \phi_2 E(y_{t-2}) + \phi_3 E(y_{t-3})$

$$E(y_t) = \delta + (\phi_1 + \phi_2 + \phi_3) \mu$$

$$\mu = \delta + (\phi_1 + \phi_2 + \phi_3) \mu$$

$$\mu - (\phi_1 + \phi_2 + \phi_3) \mu = \delta$$

$$\mu (1 - (\phi_1 + \phi_2 + \phi_3)) = \delta$$

$$\mu = \frac{\delta}{1 - (\phi_1 + \phi_2 + \phi_3)}$$

• ACVF:

$$r(k) = \text{Cov}(y_t, y_{t-k})$$

$$= \text{Cov}(\delta + \phi_1 y_{t-1} + \phi_2 y_{t-2} + \phi_3 y_{t-3} + u_t, y_{t-k})$$

$$= \phi_1 \text{Cov}(y_{t-1}, y_{t-k}) + \phi_2 \text{Cov}(y_{t-2}, y_{t-k}) +$$

$$\phi_3 \text{Cov}(y_{t-3}, y_{t-k}) + \text{Cov}(u_t, y_{t-k})$$

$$= \phi_1 r(k-1) + \phi_2 r(k-2) + \phi_3 r(k-3) + C$$

$$\text{where } C = \begin{cases} \sigma^2 & \text{if } k=0 \\ 0 & \text{if } k \neq 0 \end{cases}$$

$$r(0) = \text{Variance} = \phi_1 r(1) + \phi_2 r(2) + \phi_3 r(3) + \sigma^2$$

$$r(k) = \sum_{i=1}^3 \phi_i r(k-i) \rightarrow \textcircled{1}$$



Similarly, 
$$r(0) = \sum_{i=1}^3 \phi_i r(i) + \sigma^2$$

$$\sigma^2 = r(0) \left[ 1 - \sum_{i=1}^3 \phi_i \frac{r(i)}{r(0)} \right]$$

Variance: 
$$\sigma^2 = r(0) \left[ 1 - \sum_{i=1}^3 \phi_i \rho(i) \right]$$

• ACF: 
$$\rho(k) = \frac{r(k)}{r(0)}$$

$$= \frac{\sum_{i=1}^3 \phi_i r(k-i)}{\sum_{i=1}^3 \phi_i r(i) + \sigma^2}$$

$$\rho(k) = \sum_{i=1}^3 \phi_i \rho(k-i); \quad k=1, 2, \dots$$

iii) AR (P) :-

$$y_t = \delta + \phi_1 y_{t-1} + \dots + \phi_p y_{t-p} + u_t$$

$$E(y_t) = \delta + \phi_1 E(y_{t-1}) + \dots + \phi_p E(y_{t-p}) + 0$$

$$\mu = \delta + (\phi_1 + \phi_2 + \dots + \phi_p) \mu$$

$$\mu = \frac{\delta}{1 - (\phi_1 + \phi_2 + \dots + \phi_p)}$$

• ACVF :-

$$\gamma(k) = \text{Cov}(y_t, y_{t-k})$$

$$\therefore \gamma(k) = \text{Cov}(\phi_0 + \phi_1 y_{t-1} + \dots + \phi_p y_{t-p} + u_t, y_{t-k})$$

$$= \sum_{i=1}^p \phi_i \text{Cov}(y_{t-i}, y_{t-k}) + \text{Cov}(u_t, y_{t-k})$$

$$= \sum_{i=1}^p \phi_i \gamma(k-i) + \begin{cases} \sigma^2 & k=0 \\ 0 & k>0 \end{cases}$$

$$\text{Now, } \gamma(0) = \sum_{i=1}^p \phi_i \gamma(i) + \sigma^2$$

$$= \frac{\sigma^2}{1 - \sum_{i=1}^p \phi_i \rho(i)}$$

$$\gamma(k) = \sum_{i=1}^p \phi_i \gamma(k-i) \quad k=1, 2, \dots$$

• ACF :

$$\rho(k) = \frac{\gamma(k)}{\gamma(0)}$$

$$= \sum_{i=1}^p \phi_i \rho(k-i) \quad ; \quad k=1, 2, \dots$$