

# Time Series Assignments 3

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Q3) Obtain  $MA(1)$ ,  $MA(2)$  and  $MA(q)$  process.

Sol) (i)  $MA(1)$  : 1<sup>st</sup> Order Moving Average  
$$y_t = \mu + u_t - \theta_1 u_{t-1}$$

$$E(y_t) = \mu$$

$$\text{Var}(y_t) = (1 + \theta_1^2) \sigma^2$$

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

• for  $n=1$

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

$$= E[(y_t - \mu)(y_{t-1} - \mu)]$$

$$= E[(\mu + u_t - \theta_1 u_{t-1} - \mu)(\mu + u_{t-1} - \theta_1 u_{t-2} - \mu)]$$

$$= E[(u_t - \theta_1 u_{t-1})(u_{t-1} - \theta_1 u_{t-2})]$$

$$= E[u_t u_{t-1} - \theta_1 (u_{t-1})^2 - \theta_1 u_{t-1} u_{t-2} + \theta_1^2 u_{t-1} u_{t-2}]$$

$$= -\theta_1 \sigma^2$$

• for  $n=2$ ,

$$\gamma(2) = \text{Cov}(y_t, y_{t-2})$$

$$= E[(u_t - \theta_1 u_{t-1})(u_{t-2} - \theta_1 u_{t-3})]$$

$$= 0$$

Similarly  $\gamma(k) = 0$  for  $k > 1$

$$\therefore \gamma(k) = \gamma(-k)$$

$$\gamma(k) = \begin{cases} -\theta_1 \sigma^2 & \text{if } k = \pm 1 \\ 0 & \text{o.w.} \end{cases}$$

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

$$\rho(k) = \frac{\gamma(k)}{\gamma(0)} = \begin{cases} -\theta_1 \sigma^2 & \text{if } k = 1 \\ \frac{(1+\theta_1^2)\sigma^2}{(1+\theta_1^2)\sigma^2} & \text{if } k = 0 \\ 0 & \text{Other wise} \end{cases}$$

Therefore, Auto correlation becomes 0 when  $k > 0$  in MA(1) process.

(ii) MA(2): 2<sup>nd</sup> Order Moving Average

$$y_t = \mu + u_t - \theta_1 u_{t-1} - \theta_2 u_{t-2}$$

$$E(y_t) = \mu, \quad V(y_t) = (1 + \theta_1^2 + \theta_2^2) \sigma^2$$

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

$\rightarrow$  for  $n=1$ ,

$$\text{Cov}(y_t, y_{t-1}) = E \left[ \begin{matrix} (u_t - \theta_1 u_{t-1} - \theta_2 u_{t-2}) \\ (u_{t-1} - \theta_1 u_{t-2} - \theta_2 u_{t-3}) \end{matrix} \right]$$



$$\begin{aligned}
 &= -\theta_1 \sigma^2 + \theta_1 \theta_2 \sigma^2 \\
 &= (-\theta_1 + \theta_1 \theta_2) \sigma^2
 \end{aligned}$$

$$\rightarrow \text{for } n=2, \quad \text{Cov}(y_t, y_{t-2}) = -\theta_2 \sigma^2$$

$$\rightarrow \text{for } n=3, \quad \text{Cov}(y_t, y_{t-3}) = 0$$

$$\text{So, ACVF} = \left\{ \begin{array}{ll} (-\theta_1 + \theta_1 \theta_2) \sigma^2 & k = \pm 1 \\ -\theta_2 \sigma^2 & k = \pm 2 \\ 0 & \text{Otherwise} \end{array} \right\}$$

this shows that ~~the~~ autocovariance depends on lag not on time.

$$\text{ACF} = \left\{ \begin{array}{ll} 1 & \text{for } k=0 \\ \frac{\theta_1 + \theta_1 \theta_2}{1 + \theta_1^2 + \theta_2^2} & \text{for } k = \pm 1 \\ \frac{\theta_2}{1 + \theta_1^2 + \theta_2^2} & \text{for } k = \pm 2 \\ 0 & k > 2 \end{array} \right\}$$

iii) MA(q):

$$y_t = \mu + u_t - \theta_1 u_{t-1} - \dots - \theta_q u_{t-q}$$

$$E(y_t) = \mu ; \text{Var}(y_t) = (1 + \theta_1^2 + \dots + \theta_q^2) \sigma^2$$

$$\Rightarrow \text{ACVF} : \gamma(k) = E[y_t, y_{t-k}]$$

$$= E[(y_t - E(y_t))(y_{t-k} - E(y_{t-k}))]$$

$$= E[(u_t - \theta_1 u_{t-1} - \dots - \theta_q u_{t-q})$$

$$(u_{t-k} - \theta_1 u_{t-k-1} - \dots - \theta_q u_{t-k-q})]$$

$$= E[u_t (u_{t-k} - \theta_1 u_{t-k-1} - \dots - \theta_q u_{t-k-q})$$

$$- \theta_1 u_{t-2} (u_{t-k} - \dots - \theta_q u_{t-k-q})$$

$$- \dots - \theta_q u_{t-q} (u_{t-k} - \dots - \theta_q u_{t-k-q})]$$

$$= E[(u_t - \theta_1 u_{t-1} - \dots - \theta_q u_{t-q}) u_{t-k}$$

$$+ (u_t - \theta_1 u_{t-1} - \dots - \theta_q u_{t-q}) (-\theta_1 u_{t-k-1})$$

$$+ \dots + (u_t - \theta_1 u_{t-1} - \dots - \theta_q u_{t-q})$$

$$(-\theta_q u_{t-k-q})]$$



$$= -\theta_k \sigma^2 + \theta_1 \theta_{k+1} \sigma^2 + \dots + \theta_q \theta_{q+k} \sigma^2 ; k \leq q$$

$$\text{ACVF} = \begin{cases} (-\theta_k + \theta_1 \theta_{k+1} + \dots + \theta_q \theta_{q+k}) \sigma^2 & \text{for } 0 \leq k \leq q \\ 0 & \text{for } k > q \end{cases}$$

$$\text{ACF} = \begin{cases} \frac{(-\theta_k + \sum_{i=1}^q \theta_i * \theta_{k+i}) \sigma^2}{(1 + \sum_{i=1}^q \theta_i^2) \sigma^2} & ; 0 \leq k \leq q \\ 0 & ; k > q \end{cases}$$