

$$\text{AR}(1) : Y_t = \phi Y_{t-1} + \varepsilon_t$$

$$|\phi| < 1 \rightarrow \text{NON-STATIONARITY}$$

$$|\phi| = 1 \rightarrow \text{NON-STATIONARITY}$$

$$\phi > 1 \rightarrow \text{NON-STATIONARITY}$$

$$\text{AR}(1) : Y_t = Y_{t-1} + \varepsilon_t \quad (\text{RANDOM WALK})$$

$$Y_t = \mu + Y_{t-1} + \varepsilon_t \quad (\text{RANDOM WALK WITH DRIFT})$$

$$X_t = \mu + X_{t-1} + \varepsilon_t$$

where  $X_0 = 0$

$$\varepsilon_t \sim WN(0, \sigma_\varepsilon^2)$$

$$\rightarrow X_t = \mu + X_{t-1} + \varepsilon_t$$

$$= \mu + \underbrace{(\mu + X_{t-2} + \varepsilon_{t-1})}_{\downarrow} + \varepsilon_t =$$

$$= 2\mu + X_{t-2} + \varepsilon_{t-1} + \varepsilon_t$$

$$= 2\mu + \underbrace{(\mu + X_{t-3} + \varepsilon_{t-2})}_{\downarrow} + \varepsilon_{t-1} + \varepsilon_t$$

$$= 3\mu + X_{t-3} + \varepsilon_{t-2} + \varepsilon_{t-1} + \varepsilon_t$$

= ...

$$= t\mu + \underbrace{X_0}_{\downarrow} + \varepsilon_t + \varepsilon_{t-1} + \varepsilon_{t-2} + \dots + \underbrace{\varepsilon_1}_{\downarrow}$$

$X_{t-t} = X_0 = 0$   $= \varepsilon_{t-(t-1)}$

$$= t\mu + \varepsilon_t + \varepsilon_{t-1} + \varepsilon_{t-2} + \dots + \varepsilon_1$$

$$(i) \quad E(X_t) = ?$$

$$\begin{aligned} E(X_t) &= E(t\mu + \varepsilon_t + \varepsilon_{t-1} + \varepsilon_{t-2} + \dots + \varepsilon_1) \\ &= \underbrace{E(t\mu)}_{= t\mu} + \underbrace{E(\varepsilon_t)}_{= 0} + \underbrace{E(\varepsilon_{t-1})}_{= 0} + \underbrace{E(\varepsilon_{t-2})}_{= 0} + \dots \end{aligned}$$

BECAUSE  $\varepsilon_t$  IS WN

$$E(X_t) = t\mu \quad \rightarrow \text{SERIES IS NOT STATIONARY}$$

$$(ii) \quad \text{Var}(X_t) = ?$$

$$\begin{aligned} \text{Var}(X_t) &= E[(X_t - E(X_t))^2] \\ &= E\left[\left(\underbrace{(t\mu + \varepsilon_t + \varepsilon_{t-1} + \dots)}_{\downarrow} - \underbrace{t\mu}_{\downarrow}\right)^2\right] \\ &= E[(\varepsilon_t + \varepsilon_{t-1} + \varepsilon_{t-2} + \dots + \varepsilon_1)^2] \\ &= \underbrace{E(\varepsilon_t^2)}_{= \text{Var}(\varepsilon_t)} + \underbrace{E(\varepsilon_{t-1}^2)}_{\sigma_\varepsilon^2} + \underbrace{E(\varepsilon_{t-2}^2)}_{\sigma_\varepsilon^2} + \dots \\ &= E[(\varepsilon_t - E(\varepsilon_t))^2] = E[\varepsilon_t^2] = \sigma_\varepsilon^2 \\ &= \sigma_\varepsilon^2 + \sigma_\varepsilon^2 + \sigma_\varepsilon^2 + \dots \\ &= t\sigma_\varepsilon^2 \quad \rightarrow \text{SERIES IS NOT STAT.} \end{aligned}$$

$$\text{iii) } \text{COV}(X_t, X_s) = ?$$

$$\text{WE KNOW } X_t = \mu + X_{t-1} + \varepsilon_t$$

$$X_t = t\mu + \varepsilon_t + \varepsilon_{t-1} + \dots + \varepsilon_1$$

LET'S DEFINE  $X_s$

$$X_s = \mu + X_{s-1} + \varepsilon_s$$

$$X_s = s\mu + \varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1$$

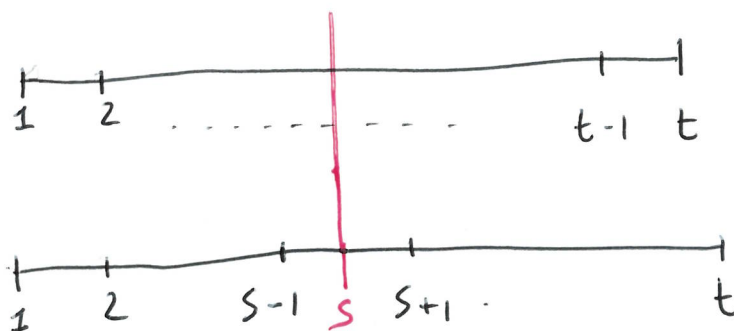
WE ASSUME  $s < t$

$$\text{COV}(X_t, X_s) =$$

$$= E[(X_t - E(X_t))(X_s - E(X_s))]$$

$$= E\left[\left(\left(t\mu + \varepsilon_t + \dots + \varepsilon_1\right) - t\mu\right)\left(\left(s\mu + \varepsilon_s + \dots + \varepsilon_1\right) - s\mu\right)\right]$$

$$= E\left[\left(\varepsilon_t + \varepsilon_{t-1} + \dots + \varepsilon_1\right)\left(\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1\right)\right]$$



$$= E \left[ \underbrace{(\varepsilon_t + \varepsilon_{t-1} + \dots + \varepsilon_1)}_{\text{we can write as}} \underbrace{(\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1)}_{\text{we can write as}} \right] =$$

$$\underbrace{(\varepsilon_t + \varepsilon_{t-1} + \dots + \varepsilon_{s+1}) + (\varepsilon_s + \varepsilon_{s+1} + \dots + \varepsilon_1)}$$

$$= E \left[ \left\{ (\varepsilon_t + \varepsilon_{t-1} + \dots + \varepsilon_{s+1}) + (\varepsilon_s + \varepsilon_{s+1} + \dots + \varepsilon_1) \right\} \cdot (\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1) \right] =$$

$$= E \left[ (\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1)^2 + (\varepsilon_t \varepsilon_s + \varepsilon_t \varepsilon_{s-1} + \dots + \varepsilon_{t-1} \varepsilon_s + \dots) \right]$$

$$= E \left[ (\varepsilon_s + \varepsilon_{s-1} + \dots + \varepsilon_1)^2 \right] + E \left[ (\varepsilon_t \varepsilon_s + \varepsilon_t \varepsilon_{s-1} + \dots) \right] =$$

$$= E(\varepsilon_s^2) + E(\varepsilon_{s-1}^2) + \dots$$

$$\begin{aligned} \downarrow & \quad \downarrow \\ = \text{VAR OF } \varepsilon_s & = \sigma_\varepsilon^2 \\ = \sigma_\varepsilon^2 & \end{aligned}$$

$$\begin{aligned} \downarrow \\ = E(\varepsilon_t \varepsilon_s) + E(\varepsilon_t \varepsilon_{s-1}) + \dots = \\ \underbrace{= 0} \quad \underbrace{= 0} \end{aligned}$$

BECAUSE  $\varepsilon$  IS WHITE NOISE WITH ZERO COVARIANCE

$$= \sigma_\varepsilon^2 + \sigma_\varepsilon^2 + \dots + 0 + 0 + 0 + \dots$$

$$= S \sigma_\varepsilon^2$$

$$= S \sigma_\varepsilon^2 \rightarrow \text{COV}(X_t, X_s) = S \sigma_\varepsilon^2 = \min(t, s) \sigma_\varepsilon^2$$

## Problem 7.2

$$X_t = X_{t-1} + \varepsilon_t$$

where  $\varepsilon_t \sim WN(0, \sigma_\varepsilon^2)$

THEN:

$$X_{t+1} = X_t + \varepsilon_{t+1}$$

$$X_{t+2} = X_{t+1} + \varepsilon_{t+2}$$

⋮

$$X_{t+k} = X_{t+k-1} + \varepsilon_{t+k}$$

i)  $\hat{X}_t(1) = ?$

$$\hat{X}_t(1) = E[X_{t+1} | F_t] = [X_{t+1}]$$

$$\hat{X}_t(1) = [X_t + \varepsilon_{t+1}]$$

$$= \underbrace{[X_t]}_{= X_t} + \underbrace{[\varepsilon_{t+1}]}_{= 0}$$

$$E(\varepsilon_{t+1} | F_t) = 0$$

BECAUSE PAST VALUES  
CANNOT PREDICT FUTURE  
- ERROR TERMS

$$\hat{X}_t(1) = X_t$$

$$\text{ii) } \hat{X}_t(2) = ?$$

$$\begin{aligned}\hat{X}_t(2) &= [X_{t+2}] \\ &= [X_{t+1} + \varepsilon_{t+2}] \\ &= \underbrace{[X_{t+1}]}_{\downarrow} + \underbrace{[\varepsilon_{t+2}]}_{=0} \\ &\quad \hat{X}_t(1) = X_t\end{aligned}$$

$$\rightarrow \hat{X}_t(2) = X_t$$

$$\begin{aligned}\rightarrow \hat{X}_t(3) &= [X_{t+3}] \\ &= \underbrace{[X_{t+2}]}_{\hat{X}_t(2) = X_t} + \underbrace{[\varepsilon_{t+3}]}_{=0} \\ &= X_t\end{aligned}$$

$$\text{iii) } \hat{X}_t(k) ?$$

$$\begin{aligned}\hat{X}_t(k) &= [X_{t+k}] \\ &= \underbrace{[X_{t+k-1}]}_{\downarrow} + \underbrace{[\varepsilon_{t+k}]}_{=0}\end{aligned}$$

$$\downarrow \\ \hat{X}_t(k-1) = X_t$$

$$\rightarrow \hat{X}_t(k) = X_t$$

THE SERIES IS NOT MEAN REVERTING

$$E(X_t) = E(X_{t-1} + \varepsilon_t)$$

$$E(X_t) = E(X_{t-1}) + E(\varepsilon_t)$$

$$\underbrace{\phantom{E(X_t)}}_{\mu} \quad \underbrace{\phantom{E(X_{t-1})}}_{\downarrow} \quad \underbrace{\phantom{E(\varepsilon_t)}}_{=0}$$

$$E(X_{t-2} + \varepsilon_{t-1})$$

⋮

$$E(X_0) = 0$$

$$\mu = 0$$



## Problem 7.3

$$X_t = X_{t-1} + Z_t$$

$$\text{where } Z_t = 3 + 0.5 Z_{t-1} + 0.1 Z_{t-2} + \varepsilon_t$$

$$\varepsilon_t \sim \text{WN}(0, \sigma_\varepsilon^2)$$

$$\hat{X}_t(1) = ?$$

- FIRST WE DEFINE  $X_{t+1}$

$$X_{t+1} = X_t + Z_{t+1}$$

- LET'S FIND  $Z_{t+1}$  AND 1-STEP AHEAD FORECAST  $\hat{Z}_t(1)$

$$Z_{t+1} = 3 + 0.5 Z_t + 0.1 Z_{t-1} + \varepsilon_t$$

$$\hat{Z}_t(1) = E[Z_{t+1} | F_t] = [Z_{t+1}]$$

$$= [3 + 0.5 Z_t + 0.1 Z_{t-1} + \varepsilon_t]$$

$$= 3 + 0.5 \underbrace{[Z_t]}_{Z_t} + 0.1 \underbrace{[Z_{t-1}]}_{= Z_{t-1}} + \underbrace{[\varepsilon_t]}_{= 0}$$

$$= 3 + 0.5 Z_t + 0.1 Z_{t-1}$$

WE KNOW THAT

$$X_t = X_{t-1} + z_t \quad \Leftrightarrow \quad z_t = X_t - X_{t-1}$$

THEREFORE

$$\begin{aligned} & 3 + 0.5 z_t + 0.1 z_{t-1} = \\ & = 3 + 0.5 (X_t - X_{t-1}) + 0.1 (X_{t-1} - X_{t-2}) \\ & = 3 + 0.5 X_t - 0.4 X_{t-1} + 0.1 X_{t-2} \end{aligned}$$

• FIND  $\hat{X}_t(1)$

$$\begin{aligned} \hat{X}_t(1) &= \cancel{X_{t+1}} = E[X_{t+1} | F_t] = [X_{t+1}] \\ &= [X_t + z_{t+1}] = \\ &= [X_t] + [z_{t+1}] \\ &= \downarrow \quad \downarrow \\ &= X_t + \hat{z}_t(1) \\ &= X_t + (3 + 0.5 X_t - 0.4 X_{t-1} - 0.1 X_{t-2}) \\ &= 3 + 1.5 X_t - 0.4 X_{t-1} - 0.1 X_{t-2} \end{aligned}$$