

Sheet 12

① Ω : sample space

Random variable $X: \Omega \rightarrow \mathbb{R}$

Stochastic Process $Y: \Omega \xrightarrow{\text{measurable}} \mathbb{R}^T$

Probability
Not expon
Relevant

\mathbb{R}^T is a space of real valued functions
defined on some index set T .

In TSA typically: $T = \mathbb{Z}$ or $T = \mathbb{N}$

a-) $T = \{t \in \mathbb{R}, t > 0\} = \mathbb{R}^+$

b-) At a fixed time point t $y_t | = e_t \sim N(0, 1)$

For the next parts, we have to assume i.i.d ness
of the e_t 's

$$c-) F_{t_1, \dots, t_n}(y_1, \dots, y_n) = \underset{\substack{\text{i.i.d} \\ \text{ness}}}{F_{N(0,1)}(y_1)} \cdot \dots \cdot F_{N(0,1)}(y_n)$$

d-) Now we take y_n

$$\lim_{\substack{t_i \rightarrow \infty \\ y_k \rightarrow \infty}} F_{t_1, \dots, t_k}(y_1, \dots, y_k)_{\text{i.i.d}} = f_{N(0,1)}(y_1) \cdot \dots \cdot f_{N(0,1)}(y_{k-1})$$

$$\lim_{y_k \rightarrow \infty} F_{N(0,1)}(y_k) = 1 \text{ since } F \text{ is a cdf}$$

\Rightarrow The process is consistent

Problem 2

$$a-) X_t = \phi_1 X_{t-1} + \phi_2 X_{t-2} + \varepsilon_t$$



$$X_t - \phi_1 X_{t-1} - \phi_2 X_{t-2} = \varepsilon_t$$



$$W X_t = \varepsilon_t, \quad W := (1 - \phi_1 B - \phi_2 B^2)$$

$$b-) \text{Cov}(X_t, X_{t+h}) \stackrel{\text{hint}}{=} E[X_t, X_{t+h}]$$

$$\stackrel{\text{Definition}}{=} E[(\Phi_1 X_{t-1} + \Phi_2 X_{t-2} + \varepsilon_t) X_{t+h}]$$

$$\stackrel{\text{expectation rules}}{=} \Phi_1 E[X_{t-1} X_{t+h}] + \Phi_2 E[X_{t-2} X_{t+h}] + \underbrace{E[\varepsilon_t X_{t+h}]}_{=0 \text{ if } h > 0}$$

$$\stackrel{\text{Stationarity}}{=} \Phi_1 \gamma(h-1) + \Phi_2 \gamma(h-2) + \sigma^2 \Big|_{\{h=0\}} \quad \left\{ \begin{array}{l} \text{means } h \text{ is 0} \\ \text{when } h \neq 0 \end{array} \right.$$

$$\rho(h) = \frac{\Phi_1 \gamma(h-1) + \Phi_2 \gamma(h-2) + \sigma^2}{\Phi_1 \gamma(1) + \Phi_2 \gamma(2) + \sigma^2} \Big|_{\{h=0\}}$$

$$c-) \text{ AR}(2) - \text{filter : } W \stackrel{a)}{=} (1 - \Phi_1 B - \Phi_2 B^2)$$

characteristic polynomial of W :

$$P(z) = 1 - \Phi_1 z - \Phi_2 z^2, z \in \mathbb{C}$$

$$\left[\begin{smallmatrix} \text{sf} & \text{cpt} \\ \text{slide} & \text{zg} \end{smallmatrix} \right]$$

$$\text{Roots: } z^* = \frac{\Phi_1 + \sqrt{\Phi_1^2 + 4\Phi_2}}{-2\Phi_2} \quad (\text{of cpt g slide 13})$$

Problem 3

$$a) X_t = \phi_1 X_{t-1} + \varepsilon_t + \theta \varepsilon_{t-1}$$

$$\iff X_t - \phi_1 X_{t-1} = \varepsilon_t + \theta \varepsilon_{t-1}$$

$$\iff W(B)X_t = V(B)\varepsilon_t, \quad W := I - \phi_1 B$$

$$V(B) := I + \theta B$$

$$\iff X_t = \underbrace{\sum_{j=0}^{\infty} \phi^j B^j}_{\text{slide 33}} \left(\sum_{j=0}^{\infty} \phi^j B^j \right) (I + \theta B) \varepsilon_t$$

$$\begin{aligned} & \stackrel{\text{lin}}{=} \left[\underbrace{\sum_{j=0}^{\infty} \phi^j B^j}_{\text{operators}} + \underbrace{\sum_{j=0}^{\infty} \phi^j \theta B^{j+1}}_{= \theta \sum_{j=1}^{\infty} \phi^{j-1} B^j} \right] \varepsilon_t \\ & = I + \sum_{j=1}^{\infty} \phi^j B^j \end{aligned}$$

$$\stackrel{\text{lin}}{=} \text{operator} \left[I + \sum_{j=1}^{\infty} B^j [\phi^j + \phi^{j-1} \theta] \right] \varepsilon_t$$

with $V_0 = I$

$$V_1 = (\phi + \theta) \phi^{j-1} \quad \text{iid } \varepsilon_t$$

$$= \varepsilon_t + \sum_{j=1}^{\infty} \phi^{j-1} (\phi + \theta) \varepsilon_{t-j}$$

$$= \varepsilon_t + (\phi + \theta) \sum_{j=1}^{\infty} \phi^{j-1} \varepsilon_{t+j} + \underbrace{\theta \varepsilon_t}_{\text{MA-1}}$$

$$= \sum_{j=0}^{\infty} V_j \varepsilon_{t-j}$$

$$b-) E[X_t] = E \left[\sum_{j=0}^{\infty} v_j \varepsilon_{t-j} \right] = \sum_{j=0}^{\infty} v_j \underbrace{E[\varepsilon_{t-j}]}_{=0} = 0$$

$$\text{Var}(X_t) = \text{Var} \left(\sum_{j=0}^{\infty} v_j \cdot \varepsilon_{t-j} \right)^{\text{WN}} = \sum_{j=0}^{\infty} v_j^2 \cdot \text{Var}(\varepsilon_{t-j})$$

$$= \sum_{j=0}^{\infty} v_j^2 \cdot \sigma^2$$

$$\sum_{j=0}^{\infty} v_j^2 = 1 + \sum_{j=1}^{\infty} (\phi^j (1 + \phi^{-1} \theta))^2$$

$$= \begin{matrix} \\ \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \end{matrix}$$

$$= \frac{1 + \theta^2 + 2\phi\theta}{1 - \phi^2}$$

$$\Rightarrow \text{Var}(X_t) = \frac{1 + \theta^2 + 2\phi\theta}{1 - \phi^2} \cdot \sigma^2$$

Given $k > 0$

$$\text{Cov}(X_t, X_{t+k}) = \text{Cov} \left(\sum_{j=0}^{\infty} v_j \varepsilon_{t-j}, \sum_{j=0}^{\infty} v_j \varepsilon_{t+k-j} \right)$$

$$WN = \sum_{j=0}^{\infty} v_j v_{j+k} \cdot \sigma^2$$

$$= \sigma^2 \left(v_k + \sum_{j=1}^{\infty} v_j v_{j+k} \right)$$

$$= \sigma^2 \left(v_k + \sum_{j=1}^{\infty} (\phi^j (1 + \phi^{-1} \theta) \phi^{j+k} (1 + \phi^{-1} \theta)) \right)$$

$$= \sigma^2 \left(v_k + \sum_{j=1}^{\infty} \phi^{j-1} (\phi + \theta)^2 \phi^{j+k-1} \right) = \dots =$$

$$= \frac{\sigma^2 \cdot \phi^k + \theta \phi^{k-1} + \phi^{k+1} + \theta + \phi^k \theta^2}{1 - \phi^2}$$

→ This is important for exam

problem 4

$$\textcircled{1} X_t = \frac{4}{5} X_{t-1} - \frac{3}{20} + \varepsilon_t - \frac{3}{10} \varepsilon_{t-1}$$

↔

$$(1 - \frac{4}{5} B) X_t = \frac{-3}{20} + \left(1 - \frac{3}{10} B\right) \varepsilon_t$$

⇒ ARMA(1,1) with intercept

$(1 - \frac{4}{5} B)$ as well as $\left(1 - \frac{3}{10} B\right)$ are invertible

(since $| \frac{4}{5} | < 1$ and $| \frac{3}{10} | < 1$) (of chapter 7)

$\Rightarrow Y_t$ is causal and invertible

$$\textcircled{2} \quad Y_t = Y_{t-1} - \frac{1}{2} Y_{t-2} + \varepsilon_t + \varepsilon_{t-1}$$

$$\Leftrightarrow \underbrace{\left(1 - B + \frac{1}{2}B^2\right)}_{\rightarrow z\text{-transform}} Y_t = \underbrace{(1-B)\varepsilon_t}_{\rightarrow z\text{-transform}}$$
$$f(z) = 1 - z = 0$$

$$f(z) = 1 - z + \frac{1}{2}z^2 \stackrel{!}{=} 0 \quad \Rightarrow z = 1$$

$$\Rightarrow z^2 - 2z + 2 = 0 \quad \rightarrow \text{not invertible}$$

$$\Leftrightarrow z \frac{1}{2} = -\frac{2}{2} \pm \sqrt{1-2} \rightarrow \text{not invertible}$$

$- = 1 \pm i \Rightarrow |z \frac{1}{2}| > 1 \Rightarrow$ AR-part is invertible, Y_t is causal,
but not invertible