

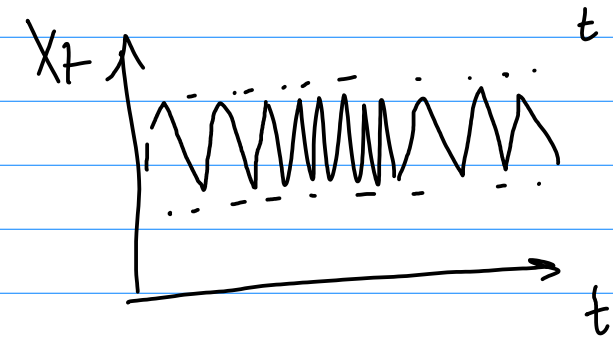
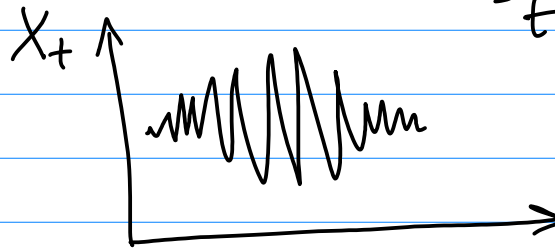
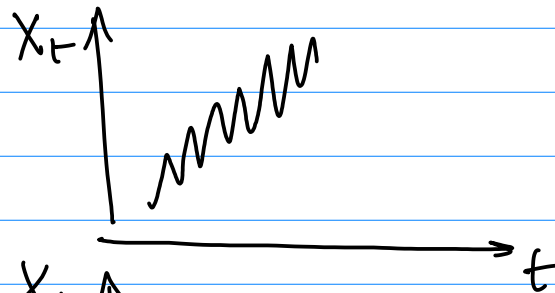
Not-Stationary Time Series

Weak (Covariance) Stationary

1. $E(X_t) = \text{const}$

2. $\text{Var}(X_t) = \text{const}$

3. $\text{Cov}(X_t, X_{t+s}) = f(s)$



Stationary Processes

1. AR(1) in finite samples

$$X_t = \beta_1 \cdot X_{t-1} + \varepsilon_t$$

$$0 < \beta_1 < 1$$

2. MA(1), MA(2)

$$X_t = \varepsilon_t + \alpha_1 \varepsilon_{t-1}$$

$$X_t = \varepsilon_t + \alpha_1 \varepsilon_{t-1} + \alpha_2 \varepsilon_{t-2}$$

Non-Stationary Processes

1. Trend Stationary

$$X_t = \alpha + \beta t + \varepsilon_t$$

2. RW

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

RW with drift

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

Problem 3. $X_t = \alpha + \beta t + \varepsilon_t$

1) $E(X_t) = \alpha + \beta t \Rightarrow X_t$ - non-stationary

Problem 4. $X_t = \underbrace{\hat{X}_{t-1}}_{\leftarrow} + \varepsilon_t$ $X_{t-1} = X_{t-2} + \varepsilon_{t-1}$

$$X_t = X_{t-2} + \varepsilon_{t-1} + \varepsilon_t = \dots = X_0 + \varepsilon_1 + \dots + \varepsilon_t$$

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

$$\dots = E(X_0) + 0 = E(X_0)$$

\downarrow X_0 i.v. with $E(X_0) = \mu$

2) $Var(X_t) = \underbrace{\sigma^2}_{X_0} + t \cdot \underbrace{\sigma^2}_{\varepsilon}$

$\Rightarrow X_t$ non-stationary

Problem 5. $X_t = \beta_2 X_{t-1} + \varepsilon_t$ finite sample

(+) X_0 i.i.d. with $E(X_0) = 0$

$$\text{Var}(X_0) = \frac{1}{1 - \beta_2^2} \sigma_\varepsilon^2$$

$$\begin{aligned} X_t &= \beta_2 (\beta_2 X_{t-2} + \varepsilon_{t-1}) + \varepsilon_t = \dots \\ &= \beta_2^t X_0 + \beta_2^{t-1} \varepsilon_1 + \dots + \beta_2 \varepsilon_{t-1} + \varepsilon_t \end{aligned}$$

$$1) E(X_t) = \beta_2^t E(X_0)$$

if $E(X_0) = 0 \Rightarrow 1^{\text{st}}$ prop. isn't violated

$$\begin{aligned} 2) \text{Var}(X_t) &= \beta_2^{2t} \sigma_{X_0}^2 + \beta_2^{2t-2} \sigma_\varepsilon^2 + \dots + \beta_2^2 \sigma_\varepsilon^2 + \sigma_\varepsilon^2 \\ &= \beta_2^{2t} \sigma_{X_0}^2 + \frac{1 - \beta_2^{2t}}{1 - \beta_2^2} \sigma_\varepsilon^2 \end{aligned}$$

$$\text{Var}(X_0) = \frac{1}{1 - \beta_2^2} \sigma_\varepsilon^2$$

$$\begin{aligned} \text{Var}(X_t) &= \beta_2^{2t} \cdot \frac{1}{1 - \beta_2^2} \sigma_\varepsilon^2 + \frac{1 - \beta_2^{2t}}{1 - \beta_2^2} \sigma_\varepsilon^2 = \\ &= \frac{1}{1 - \beta_2^2} \sigma_\varepsilon^2 \end{aligned}$$

3)

$$X_t = \beta_2 X_{t-1} + \varepsilon_t$$

$$X_{t+s} = \beta_2^s X_t + \beta_2^{s-1} \varepsilon_{t+1} + \dots + \varepsilon_{t+s}$$

$$\begin{aligned} \text{Cov}(X_t, X_{t+s}) &= \beta_2^s \text{Var}(X_t) + \text{Cov}(X_t; \beta_2^{s-1} \varepsilon_{t+1} + \dots + \varepsilon_{t+s}) \\ &= \beta_2^s \text{Var}(X_t) + 0 \end{aligned}$$

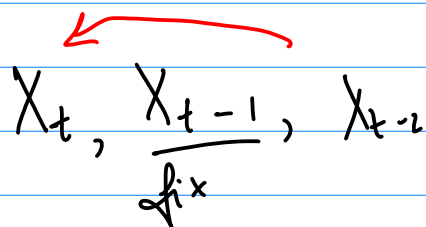
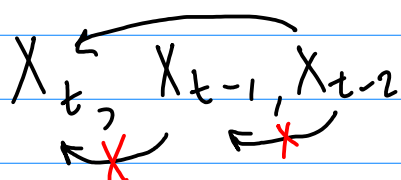
Hence X_t is stationary

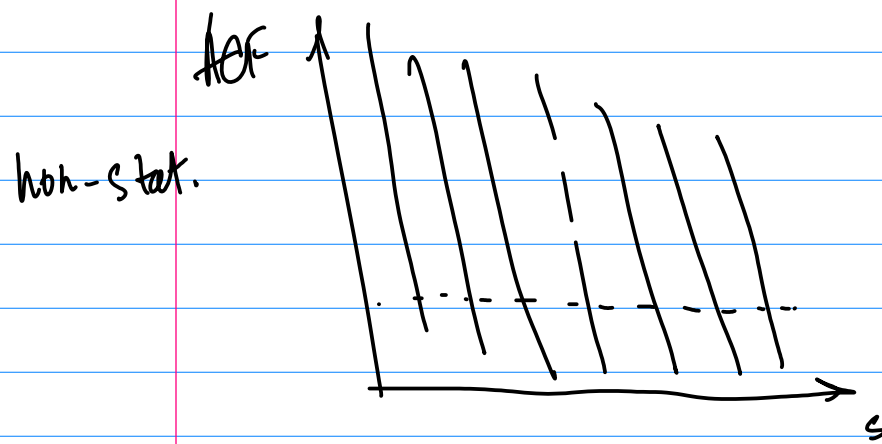
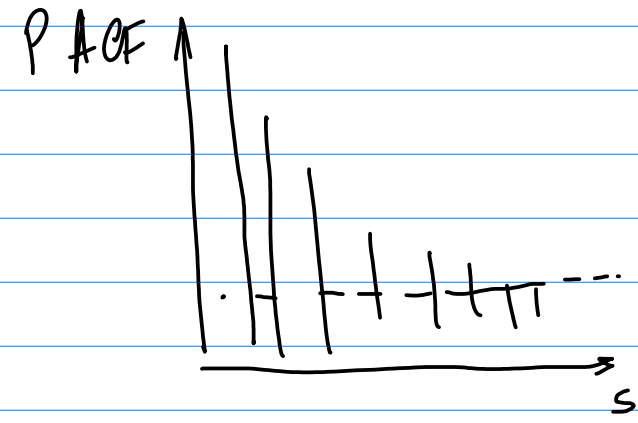
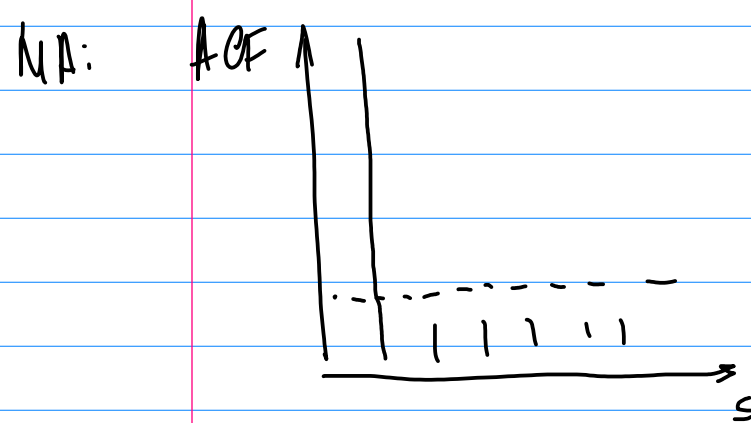
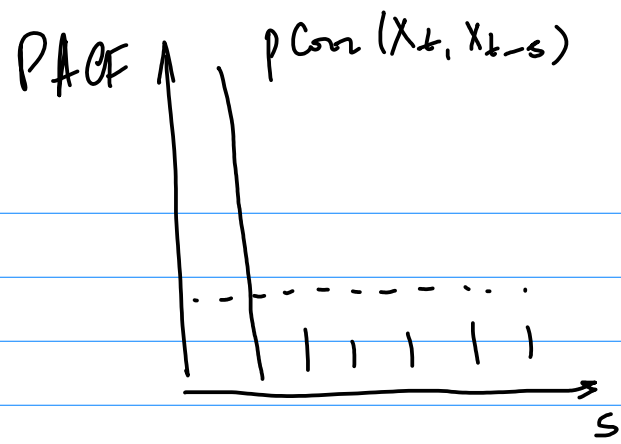
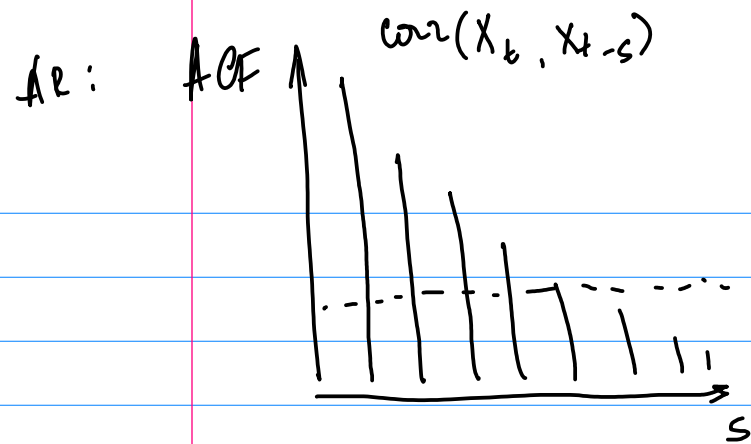
Problem 7 $X_t = \varepsilon_t + \alpha_1 \cdot \varepsilon_{t-1}$

$$1) E(X_t) = 0$$

$$2) \text{Var}(X_t) = \sigma_\varepsilon^2 + \alpha_1^2 \cdot \sigma_\varepsilon^2$$

$$3) \text{Cov}(X_t, X_{t+s}) = \begin{cases} s=1: \text{Cov}(\varepsilon_t + \alpha_1 \varepsilon_{t-1}, \varepsilon_{t+1} + \alpha_1 \varepsilon_t) = \alpha_1 \sigma_\varepsilon^2 \\ s>1: 0 \end{cases}$$





$$a) X_t = \varepsilon_t + \alpha_1 \cdot \varepsilon_{t-1}$$

$$1) E(X_t) = 0$$

$$2) \text{Var}(X_t) = \sigma_\varepsilon^2 + \alpha_1^2 \cdot \sigma_\varepsilon^2$$

$$3) \text{Cov}(X_t, X_{t+s}) = \begin{cases} s=1: \alpha_1 \cdot \sigma_\varepsilon^2 \\ s>1: 0 \end{cases}$$

$$\text{Corr}(X_t, X_{t+s}) = \begin{cases} s=1: \frac{\alpha_1 \cdot \sigma_\varepsilon^2}{(1+\alpha_1^2)\sigma_\varepsilon^2} = \frac{\alpha_1}{1+\alpha_1^2} \\ s>1: 0 \end{cases}$$

$$b) \text{AR}(1): X_t = \beta_2 X_{t-1} + \varepsilon_t$$

$$\text{Corr}(X_t, X_{t+s}) = \rho^s = \frac{\beta_2^s \sigma_\varepsilon^2 / (1-\beta_2^2)}{\sigma_\varepsilon^2 / (1-\beta_2^2)} = \beta_2^s$$