

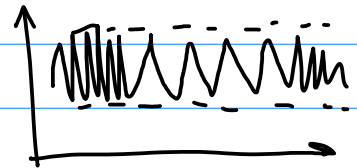
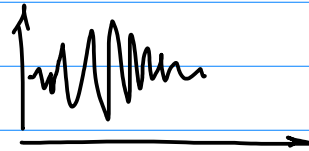
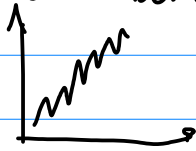
Non-Stationary Time Series

Weak Stationary / 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) for finite samples

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

$$0 < \beta_1 < 1$$

2. MA(1), MA(2)

$$X_t = \epsilon_t + \alpha_1 \epsilon_{t-1}$$

$$X_t = \epsilon_t + \alpha_1 \epsilon_{t-1} + \alpha_2 \epsilon_{t-2}$$

Not-Stationary:

1. Trend-Stationary (time trend)

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

2. RW

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

RW with drift

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

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

$$1) E(X_t) = \alpha + \beta t$$

Problem 4. $X_t = X_{t-1} + \varepsilon_t$

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

$$E(X_0 + \varepsilon_1 + \dots + \varepsilon_t) = E(X_0)$$

$$\text{if } X_0 \text{ z.v. } E(X_t) = \mu$$

$$2) \text{Var}(X_t) = \text{Var}(X_0 + \varepsilon_1 + \dots + \varepsilon_t) =$$

$$= \sigma_{X_0}^2 + t \cdot \sigma_{\varepsilon}^2$$

$$\Rightarrow X_t - \text{not-stationary}$$

Problem 5: AR(1) for finite sample

and X_0 : $E(X_0) = 0$

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

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

$$= \beta_2 (\beta_2 X_{t-2} + \varepsilon_{t-1}) + \varepsilon_t = \dots = \beta_2^t \cdot X_0 + \beta_2^{t-1} \varepsilon_1 + \dots + \beta_2 \varepsilon_{t-1} + \varepsilon_t$$

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

if $E(X_0) = 0 \Rightarrow 1^{st}$ prop. not violated

$$2) \text{Var}(X_t) = \beta_2^{2t} \sigma_{X_0}^2 + \beta_2^{2t-2} \sigma_\varepsilon^2 + \dots + \beta_2^2 \sigma_\varepsilon^2 + \sigma_\varepsilon^2$$

$$= \beta_2^{2t} \sigma_{X_0}^2 + \frac{1 - \beta_2^{2t}}{1 - \beta_2^2} \sigma_\varepsilon^2 = \dots$$

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

$$\dots = \beta_2^{2t} \cdot \frac{1}{1 - \beta_2^2} \sigma_\varepsilon^2 + \frac{1 - \beta_2^{2t}}{1 - \beta_2^2} \cdot \sigma_\varepsilon^2 = \frac{1}{1 - \beta_2^2} \sigma_\varepsilon^2$$

$$3) X_t = \beta_2 X_{t-1} + \varepsilon_t$$

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

$$\text{Cov}(X_t, X_{t+s}) = \beta_2^s \text{Var}(X_t) + \text{Cov}(X_t, \beta_2^{s-1} \varepsilon_{t+1} + \dots + \beta_2 \varepsilon_{t+s-1} + \varepsilon_{t+s})$$

$$= \beta_2^s \cdot \sigma_\varepsilon^2 + 0$$

Problem 7. MA(1) :

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

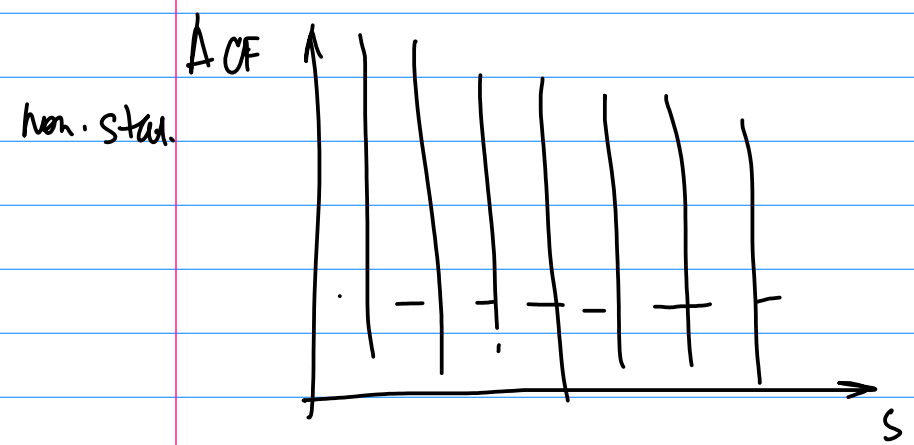
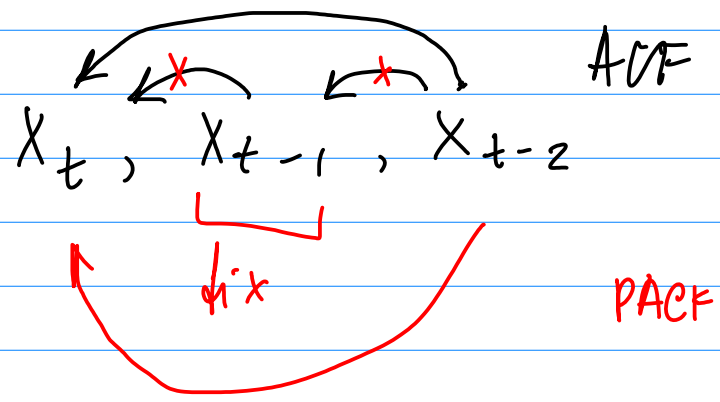
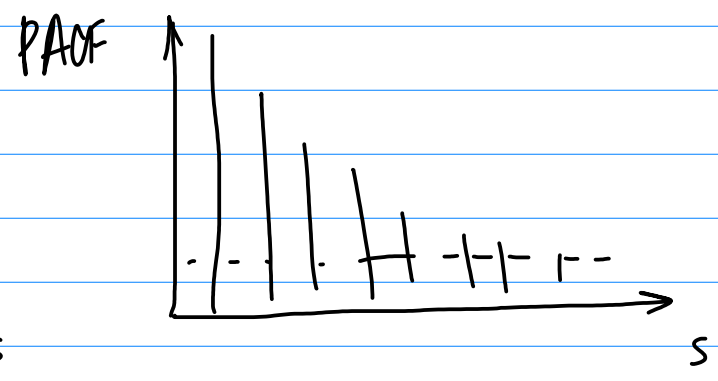
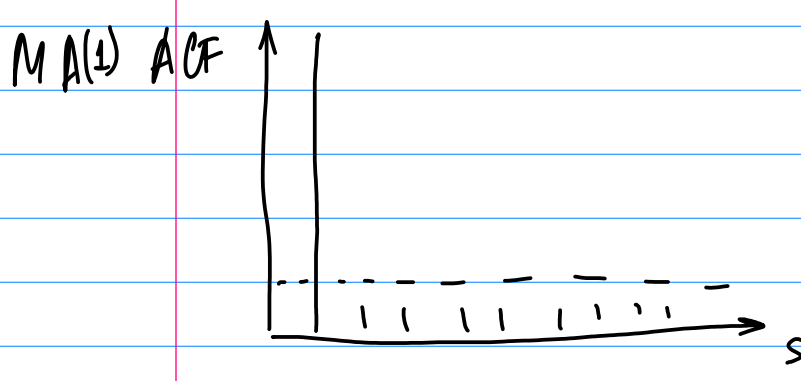
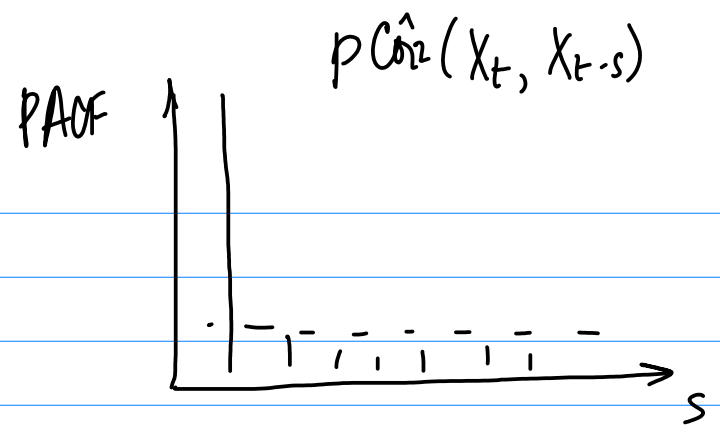
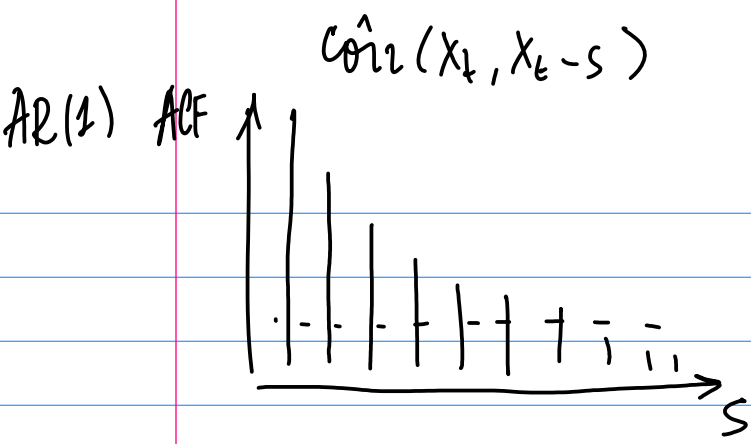
1. $E(X_t) = 0$

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

3. $s=1$: $\text{Cov}(X_t, X_{t+1}) = \text{Cov}(\varepsilon_t + \alpha_1 \cdot \varepsilon_{t-1}, \varepsilon_{t+1} + \alpha_1 \cdot \varepsilon_t) =$
 $= \alpha_1 \cdot \sigma_\varepsilon^2$

$s > 1$: $\text{Cov}(X_t, X_{t+s}) = 0$

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



$$1) \text{ ACF : } X_t = \varepsilon_t + \alpha_1 \varepsilon_{t-1}$$

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

$$s = 1 \quad \text{Cov}(X_t, X_{t+s}) = \alpha_1 \cdot \sigma_\varepsilon^2$$

$$s > 1 : \text{Cov}(X_t, X_{t+s}) = 0$$

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

$$2) \text{ ACF : } X_t = \beta_2 X_{t-1} + \varepsilon_t$$

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