

# Non-stationary TS

(p1)

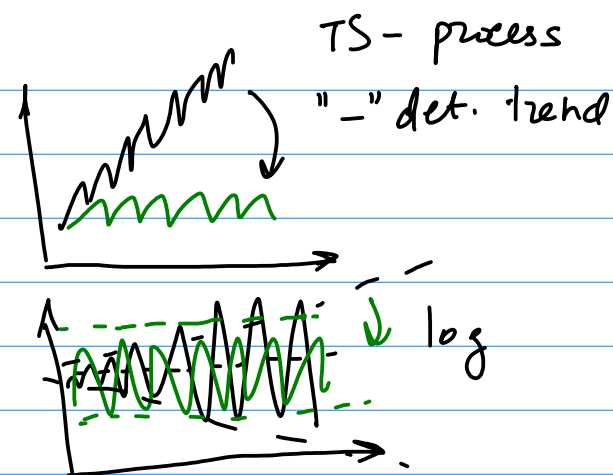
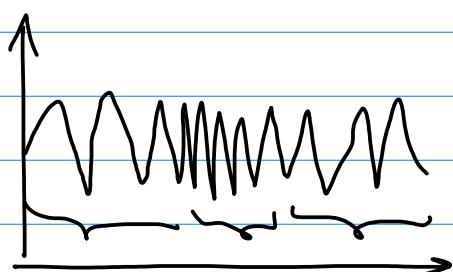
Weakly - Stationary

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

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

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

Linear  
or SV



Box-Cox Transf.

$$y_t^{(\lambda)} = \begin{cases} \frac{y_t^\lambda - 1}{\lambda} & \lambda \neq 0 \\ \ln y_t & \lambda = 0 \end{cases}$$

(p2)

Stationary Processes:

$MA(q)$   $\forall q$  - always stationary

$AR(p), ARMA(p,q)$  equations

has all roots  $|a| \neq 1 \Rightarrow$  has

stationary

(p3)  $y_t = 1.4 y_{t-1} + \varepsilon_t$

solutions

$$y_t = \beta y_{t-1} + \varepsilon_t, \quad 0 < \beta < 1, \quad \varepsilon_t - WN$$

$$E(y_0) = 0$$

$$\text{Var}(y_0) = \frac{1}{1-\beta^2} \cdot \sigma_\varepsilon^2$$

$$E(y_t) = E(\beta^t y_0 + \beta^{t-1} \varepsilon_1 + \dots + \varepsilon_t) =$$

$$= \beta^t E(y_0) = 0$$

$$\text{Var}(y_t) = \beta^{2t} \text{Var}(y_0) + \beta^{2t-2} \sigma_\varepsilon^2 + \dots + \sigma_\varepsilon^2 =$$

$$= \beta^{\textcircled{2t}} \frac{1}{1 - \beta^2} \cdot \sigma_\varepsilon^2 + \frac{1 - \beta^{\textcircled{2t}}}{1 - \beta^2} \cdot \sigma_\varepsilon^2 =$$

$$= \frac{1}{1 - \beta^2} \cdot \sigma_\varepsilon^2 \quad \frac{b_0}{1 - q} = \sum_{q=0}^t \beta^q \cdot \sigma_\varepsilon^2$$

$$\text{Cov}(y_t, y_{t+s}) = \text{Cov}(y_t, \beta^s y_t + \beta^{s-1} \varepsilon_{t+1} + \dots + \varepsilon_{t+s})$$

$$= \beta^s \text{Var}(y_t) + 0$$

④ For sample large enough

$y_t$  - weakly persistent / dependent

$$\text{Cov}(y_t, y_{t+s}) \rightarrow 0, \quad s \rightarrow \infty$$

(p4)

Non-stationary:

$$y_0 = 0$$

Random Walk:

$$y_t = y_{t-1} + \varepsilon_t$$

↳ DS-process

$$E(y_t) = E(\varepsilon_0) = 0$$

DS:

$$\Delta y_t = \varepsilon_t$$

$$\text{Var}(y_t) = t \cdot \sigma_\varepsilon^2$$

$$\forall s > 0 \quad \text{Cov}(y_t, y_{t+s}) = t \cdot \sigma_\varepsilon^2$$

$$\forall s > 0 \quad \text{Cov}(y_t, y_{t-s}) = (t-s) \sigma_\varepsilon^2$$

TS-process:

$$y_t - \hat{\alpha} - \hat{\beta} t \Rightarrow \text{stat.}$$

$$y_t = \underbrace{\alpha + \beta t}_{\text{TS}} + \underbrace{\varepsilon_t}_{\text{DS}}$$

$$\Delta y_t = y_t - y_{t-1} = \beta + \varepsilon_t - \varepsilon_{t-1}$$

(p5)

Spurious Regressions:

TS-process:

