

→ Definición de un ruido blanco.

→ Procesos estacionarios

→ Ejemplos de procesos estacionarios

a) $\{X_t\}_{t=1}^{\infty}$, donde $X_t \sim N(0,1)$ para $t=1,2,\dots$

$$\mathbb{E}[X_t] = 0, \text{Var}(X_t) = 1 \quad \forall t$$

$$\gamma(t,s) = \mathbb{E}[X_t X_s] = \begin{cases} 1 & \text{si } t=s \\ 0 & \text{si } t \neq s \end{cases}$$

$$\rho(t,s) = \frac{\gamma(t,s)}{\gamma(t,t)} = \begin{cases} 1 & \text{si } t=s \\ 0 & \text{si } t \neq s \end{cases}$$

b) Sea $\{S_t\}_{t=1}^{\infty}$ una caminata aleatoria,
es decir, $S_t = X_1 + X_2 + X_3 + \dots + X_t$, o

$$S_t = \sum_{i=1}^t X_i, \text{ donde } X_i \sim WN(0, \sigma^2)$$

$$\mathbb{E}[S_t] = 0 = \mathbb{E}\left[\sum_{i=1}^t X_i\right] = \sum_{i=1}^t \mathbb{E}[X_i]$$

$$\text{Var}(S_t) = \text{Var}\left(\sum_{i=1}^t X_i\right) = \sum_{i=1}^t \text{Var}(X_i) = \sum_{i=1}^t \sigma^2$$

$$= t \sigma^2$$

Depende explícitamente de t .

$$\gamma(t+h, t) = \text{Cov}(S_{t+h}^{\uparrow}, S_t)$$

$$= \text{Cov}(S_t + \sum_{i=1}^h X_{t+i}, S_\tau)$$

$$= \cancel{\text{Cov}}(S_t, S_t) + \text{Cov}\left(\sum_{i=1}^h \cancel{X_{t+i}}, S_t\right)$$

$$= t \sigma^2$$

De aquí, concluimos que S_t no es un proceso estacionario.

→ Consideremos un proceso de medias móviles de orden 1, MAC(1)

$X_t = W_t + \Theta W_{t-1}$, donde $W_t \sim WN(0, \sigma^2)$
 \downarrow \downarrow \downarrow \nwarrow $t = \pm 1, \pm 2, \dots, t \in \mathbb{Z}$
 ¿Es estacionario? $\Theta \in \mathbb{R}$ $X_t = W_{t-1} + \Theta W_{t-2}$

$$\mu_t = E[X_t] = E[w_t] + 0 E[w_{t-1}] = 0$$

$$\begin{aligned} \text{Var}(X_t) &= \text{Var}(W_t) + \Theta^2 \text{Var}(W_{t-1}) \\ &= \sigma^2 + \Theta^2 \sigma^2 = \sigma^2 (1 + \Theta^2) \end{aligned}$$

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

* Recordemos que $X_t = W_t + \Theta W_{t-1}$

- Cuando $h=0 \Rightarrow \text{Cov}(X_t, X_{t+0}) = \text{Var}(X_t) = \sigma^2(1+\theta^2)$

• Cuando $h = \pm 1$

$$\begin{aligned} \text{Cov}(X_t, X_{t+1}) &= \text{Cov}(W_t + \Theta W_{t-1}, W_{t+1} + \Theta W_t) \\ &= \text{Cov}(W_t, W_{t+1}) + \text{Cov}(W_t, \Theta W_t) + \text{Cov}(\Theta W_{t-1}, W_{t+1}) + \text{Cov}(\Theta W_{t-1}, \Theta W_t) \\ &= \Theta \text{Var}(W_t) = \boxed{\Theta \sigma^2} \end{aligned}$$

$$\text{Cov}(X_t, X_{t-1}) \xrightarrow[t=S+1 \rightarrow t-1=S]{\text{renombrar}} \text{Cov}(X_{S+1}, X_S) = \boxed{\Theta \sigma^2}$$

S: $|h| > 1 \leadsto \text{Cov}(X_{t+h}, X_t)$

$$\begin{aligned} &= \text{Cov}(W_{t+h} + \Theta W_{t+h-1}, W_t + \Theta W_{t-1}) \\ &= \text{Cov}(W_{t+h}, W_t) + \Theta \text{Cov}(W_{t+h-1}, W_t) + \Theta \text{Cov}(W_{t+h}, W_{t-1}) + \Theta^2 \text{Cov}(W_{t+h-1}, W_{t-1}) = 0 \end{aligned}$$

$$\gamma_h = \begin{cases} \sigma^2(1+\Theta^2) & \text{s: } h=0 \\ \sigma^2\Theta & \text{s: } h=1 \text{ o } h=-1 \\ 0 & \text{s: } |h| > 1 \end{cases}$$

$$\begin{aligned} \text{Cov}(X_t, X_s) &= \mathbb{E}[(X_t - \mu_t)(X_s - \mu_s)] \\ &= \mathbb{E}[X_t X_s] - \cancel{\mu_t} \cdot \cancel{\mu_s} \\ &= \mathbb{E}[X_t X_s] \end{aligned}$$

$$\text{Cov}(\alpha X + \beta Y, Z), \quad X, Y, Z \text{ v.a.'s}$$

$$\alpha, \beta \in \mathbb{R}$$

$$= \alpha \text{Cov}(X, Z) + \beta \text{Cov}(Y, Z)$$

$$\rightarrow \mathbb{E}[(\alpha X + \beta Y)Z] - \mathbb{E}[\alpha X + \beta Y] \mathbb{E}[Z]$$

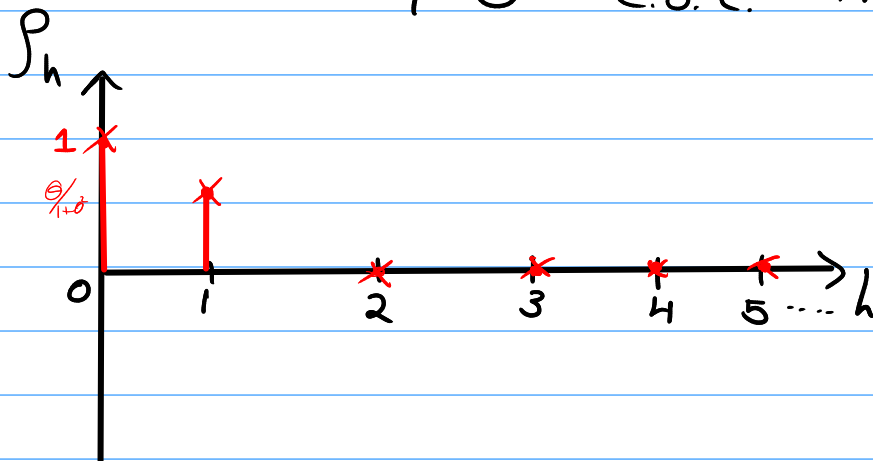
$$\alpha \mathbb{E}[XZ] + \beta \mathbb{E}[YZ] - \alpha \mathbb{E}[X] \mathbb{E}[Z] - \beta \mathbb{E}[Y] \mathbb{E}[Z]$$

$$= \alpha [\underbrace{\mathbb{E}[XZ] - \mathbb{E}[X] \mathbb{E}[Z]}] + \beta [\underbrace{\mathbb{E}[YZ] - \mathbb{E}[Y] \mathbb{E}[Z]}]$$

$$\alpha \cdot \text{Cov}(X, Z) + \beta \cdot \text{Cov}(Y, Z)$$

→ De aquí, concluimos que X_t es estacionario.

$$\rho_h = \frac{\gamma_h}{\gamma_0} = \begin{cases} 1 & h=0 \\ \theta/(1+\theta^2) & h=\pm 1 \\ 0 & \text{e.o.c. } |h|>1 \end{cases}$$



→ Proceso Autorregresivo de orden 1
(AR(1))

$$X_t = \psi X_{t-1} + W_t, \quad \psi \in \mathbb{R}, \quad W_t \sim WN(0, \sigma^2)$$

$$\downarrow$$

$$\text{Si } \psi = 0.5 \Rightarrow X_t = 0.5 X_{t-1} + W_t$$

$$\mathbb{E}[X_t] = 0 \quad \text{y} \quad \mathbb{E}[X_t] = \psi \mathbb{E}[X_{t-1}] + 0$$

$$\mathbb{E}[X_t] = \psi \mathbb{E}[X_{t-1}] \leadsto \mathbb{E}[X_{t-1}] = \psi \mathbb{E}[X_{t-2}]$$

$$\text{Cov}(X_{t+h}, X_t) = \mathbb{E}[X_{t+h} X_t] - \cancel{\mathbb{E}[X_{t+h}]}^0 \cancel{\mathbb{E}[X_t]}^0$$

$$\text{Cov}(\psi X_{t+h-1} + V_t, X_t) = \psi \text{Cov}(X_{t+h-1}, X_t) + \text{Cov}(\cancel{V_t}^0, X_t)$$

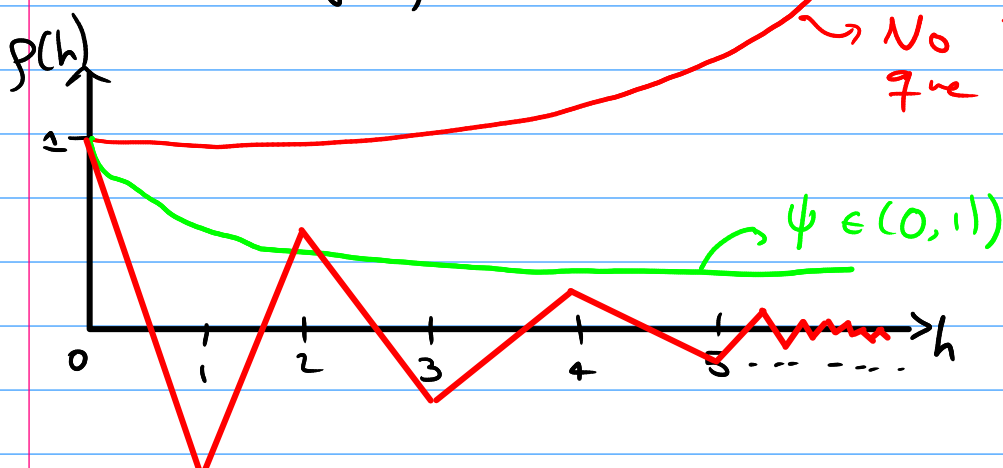
$$= \psi \gamma(h-1) = \psi [\text{Cov}(X_{t+h-1}, X_t)]$$

$$= \psi [\text{Cov}(\psi X_{t+h-2} + W_t, X_t)] =$$

$$= \psi^2 \text{Cov}(X_{t+h-2}, X_t) + \psi \text{Cov}(\cancel{W_t}^0, X_t)$$

$$= \psi^2 \gamma(h-2) = \dots = \psi^h \gamma(0)$$

$$\rho(h) = \frac{\gamma(h)}{\gamma(0)} = \psi^h \quad \rightarrow \quad \left\{ \begin{array}{l} |\psi| > 1 \quad \text{ó} \quad |\psi| < 1 \end{array} \right.$$



X_t es estacionaria

$$\rho(h) = \psi^h, \text{ para } |\psi| < 1 \quad \nearrow$$

¿Cómo calculamos $\text{Var}(X_t)$?

$$\begin{aligned}\text{Var}(X_t) &= \text{Cov}(X_t, X_t) = \\ &= \text{Cov}(\underbrace{\psi X_{t-1}} + \underbrace{W_t}, \underbrace{\psi X_{t-1}} + \underbrace{W_t}) \\ &= \text{Cov}(\psi X_{t-1}, \psi X_{t-1}) + \text{Cov}(W_t, \psi X_{t-1}) \\ &\quad + \text{Cov}(\psi X_{t-1}, W_t) + \text{Cov}(W_t, W_t) \\ &= \psi^2 \text{Cov}(X_{t-1}, X_{t-1}) + 0 \\ &\quad + 0 + \sigma^2\end{aligned}$$

$$\Rightarrow \text{Var}(X_t) = \psi^2 \text{Var}(X_{t-1}) + \sigma^2$$

$$\gamma(0) = \psi^2 \gamma(0) + \sigma^2 \Leftrightarrow \gamma(0) = \frac{\sigma^2}{1 - \psi^2}$$

→ Supongamos que tenemos x_1, x_2, \dots, x_n n observaciones de una serie de tiempo.

La media muestral se define como:

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

La función muestral de autocovarianza se define como:

$$\hat{\gamma}(h) = \frac{1}{n} \sum_{t=1}^{n-h} (x_{t+h} - \bar{x})(x_t - \bar{x})$$

$$\hat{\rho}(h) = \frac{\hat{\gamma}(h)}{\hat{\gamma}(0)}$$