

$$\gamma(h) = \text{cov}(X_t, X_{t+h}), \quad \{X_t\}_{t \in \mathbb{Z}}$$

$$\rho(h) = \frac{\gamma(h)}{\gamma(0)}, \quad \text{ACF}$$

PACF

$$\{X_1, X_2, \dots, X_n\} \text{ nuestra}$$
$$\hat{\gamma}(h) = \frac{\hat{\gamma}(h)}{\hat{\gamma}(0)} \quad \text{— version}$$

empirica

- Corroborar de manera grafica si un ST es estacionaria

$$\{X_t\}_{t \in \mathbb{Z}}, \quad \boxed{X_t = \varepsilon_t}$$

White Noise

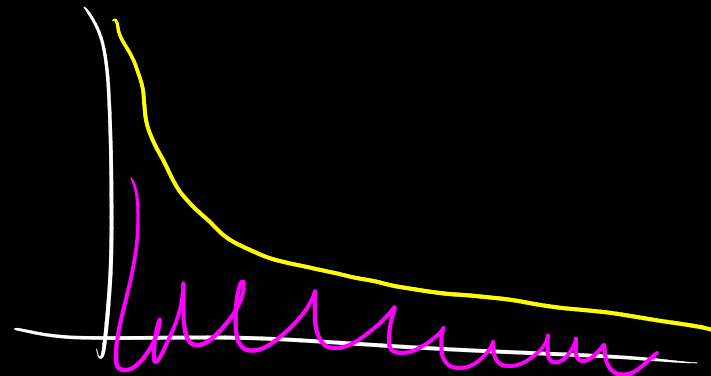
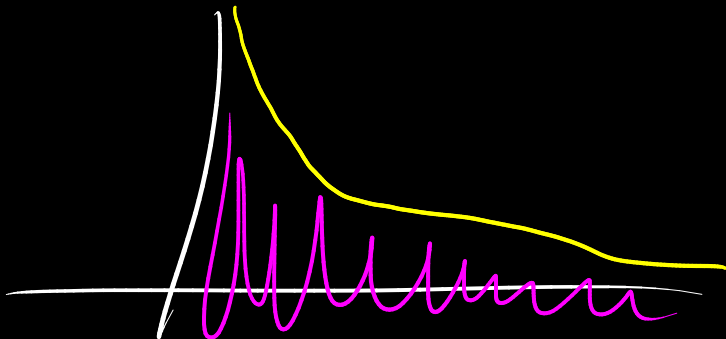
$$\varepsilon_t \sim \text{WN}(0,1)$$

↑

- Normal

- T

ACF (Para ST Estacionarias) PACF



ARIMA (p, d, q)

AR(p): Autoregressive
Autoregresivos

I(d): Integrated (Integrado)

MA(q): - Moving Average
- Medias Móviles

AR(p), $\{z_t\}, t \in \mathbb{Z}$, $\varepsilon_t \sim \mathcal{N}(0,1)$

$$z_t = a_1 z_{t-1} + a_2 z_{t-2} + \dots + a_p z_{t-p} + \varepsilon_t$$

$$= \sum_{i=1}^p a_i z_{t-i} + \varepsilon_t$$

$$\begin{array}{c} |\phi| < 1 \\ z_t = \phi z_{t-1} + \varepsilon_t \\ \hline \text{AR}(1) \\ \uparrow \phi = 1 \\ \text{Camminante} \\ \text{aleatorio} \end{array}$$

MA(q),

$$z_t = b_1 \varepsilon_{t-1} + b_2 \varepsilon_{t-2} + \dots + b_q \varepsilon_{t-q} + \varepsilon_t$$

$$= \sum_{i=1}^q b_i \varepsilon_{t-i} + \varepsilon_t$$

Usando el operador $Lag(\underline{L})$, $\{x_t\}_{t \in \mathbb{Z}}$

$$L x_t = x_{t-1},$$

$$L^2 x_t = L(L x_t) = L x_{t-1} = x_{t-2}$$

$$L^j x_t = x_{t-j}$$

$L(H)$
(Análisis Funcional)

$$AR(p): z_t = \sum_{i=1}^p a_i L^i z_t + \varepsilon_t$$

$$MA(q): z_t = \sum_{i=1}^q b_i L^i \varepsilon_t + \varepsilon_t$$

$$AR(P) : z_t - \sum_{i=1}^p a_i L^i z_t = \underbrace{\left(1 - \sum_{i=1}^p a_i L^i\right)}_{P(L)} (z_t) = \varepsilon_t$$

$$\underline{P(L) z_t = \varepsilon_t}$$

$$\underline{P(L)}$$

$$MA(Q) : z_t = \sum_{i=1}^q b_i L^i \varepsilon_t + \varepsilon_t = \underbrace{\left(\sum_{i=1}^q b_i L^i + 1\right)}_{Q(L)} \varepsilon_t$$

$$\underline{z_t = Q(L) \varepsilon_t}$$

$$\underline{P(x) = \left(1 - \sum_{i=1}^p a_i x^i\right)}$$

$$Q(x) = \left(\sum_{i=1}^q b_i x^i + 1\right)$$

$$P(L)$$

$$Q(L)$$

Polinomios
característicos
de
los Modelos

$$ARMA(P, q): P(L)z_t = Q(L)\varepsilon_t$$

$ARIMA(P, d, q):$

(Cantidad de Raíces Unitarias) $\rightarrow d$

$$(1-L)z_t = w_t$$

$$z_t = z_{t-1} + w_t$$

$$P(L)(1-L)^d z_t = Q(L)\varepsilon_t$$

$AR(P)$ $Integrada(d)$ $MA(q)$

Caminata aleatoria

Movimiento Browniano
 Proceso de Wiener

Proceso	ACF	PACF
AR(p)	Decae exp.	Se corta después de $h=p$
MA(q)	Se corta en $h=q$	Decae exp.
ARMA(p,q)	Decae exp	Deca exp.
ARIMA	<u>Decae exp</u>	<u>Deca exp.</u>

"Decae exp" —

Esta oculto un proceso
 $AR(\infty)$ o $MA(\infty)$

— OLS }
 — ML }

ⓈARIMAⓧ
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