

SERIES TEMPORALES

Escribimos los siguientes modelos:

$$\text{AR}(1): y_t = \phi y_{t-1} + \varepsilon_t$$

$$\text{AR}(2): y_t = \phi_1 y_{t-1} + \phi_2 y_{t-2} + \varepsilon_t$$

$$\text{MA}(1): y_t = \varepsilon_t + \theta_1 \varepsilon_{t-1}$$

$$\text{MA}(2): y_t = \varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2}$$

$$\text{ARMA}(1,1): y_t = \phi y_{t-1} + \varepsilon_t + \theta_1 \varepsilon_{t-1}$$

$$\text{ARMA}(2,2): y_t = \phi_1 y_{t-1} + \phi_2 y_{t-2} + \varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2}$$

Propiedades de las series estacionarias:

$$E(y_t) = \mu, \text{ para todo } t$$

$$\text{VAR}(y_t) = \text{VAR}(y_{t-1}) \text{ para todo } t$$

$$E(\varepsilon_t) = 0, \text{ para todo } t$$

$$\text{VAR}(\varepsilon_t) = \sigma^2 \text{ para todo } t$$

Calculo de valores estadísticos.

AR(1)

Media

$$\mu = E(y_t) = E(\phi y_{t-1} + \varepsilon_t) = \phi E(y_{t-1}) + E(\varepsilon_t) = \phi \mu + 0 =$$

$$\mu = \phi \mu \quad \mu = 0$$

Varianza

$$\text{VAR}(y_t) = \text{VAR}(\phi y_{t-1} + \varepsilon_t) = \phi^2 \text{VAR}(y_{t-1}) + \text{VAR}(\varepsilon_t) = \phi^2 \text{VAR}(y_t) + \sigma^2$$

$$\text{VAR}(y_t) - \phi^2 \text{VAR}(y_{t-1}) = \sigma^2$$

$$(1 - \phi^2) \text{VAR}(y_t) = \sigma^2$$

$$\text{VAR}(y_t) = \frac{\sigma^2}{1 - \phi^2} ; \quad V_0 = \frac{\sigma^2}{1 - \phi^2}$$

Autocovarianzas

$$V_1 = \text{COV}(y_t, y_{t+1}) = \text{COV}(y_t, \phi y_t + \varepsilon_{t+1}) = \phi \text{COV}(y_t, y_t) + \text{COV}(y_t, \varepsilon_{t+1}) = \phi V_0 = \phi \frac{\sigma^2}{1 - \phi^2}$$

$$V_2 = \text{COV}(y_t, y_{t+2}) = \text{COV}(y_t, \phi y_{t+1} + \varepsilon_{t+2}) = \text{COV}(y_t, \phi y_{t+1}) + \text{COV}(y_t, \varepsilon_{t+2}) = \phi V_1 = \phi^2 V_0$$

$$V_k = \phi^k V_0$$

Autocorrelaciones

$$\rho_0 = \frac{V_0}{V_0} = 1$$

$$\rho_1 = \frac{V_1}{V_0} = \frac{\phi V_0}{V_0} = \phi$$

$$\rho_2 = \frac{V_2}{V_0} = \frac{\phi^2 V_0}{V_0} = \phi^2$$

$$\rho_k = \frac{V_k}{V_0} = \phi^k$$

AR(2)

Media

$$\begin{aligned}\mu &= E(y_t) = E(\phi_1 y_{t-1} + \phi_2 y_{t-2} + \varepsilon_t) = \\ &\phi_1 E(y_{t-1}) + \phi_2 E(y_{t-2}) + E(\varepsilon_t) = \phi_1 \mu + \phi_2 \mu + 0 \\ \mu &= (\phi_1 + \phi_2) \mu \quad \mu = 0\end{aligned}$$

Varianza

$$\begin{aligned}V_0 &= \text{VAR}(y_t) = \text{VAR}(\phi_1 y_{t-1} + \phi_2 y_{t-2} + \varepsilon_t) = \phi_1^2 \text{VAR}(y_{t-1}) + \\ &\phi_2^2 \text{VAR}(y_{t-2}) + \text{VAR}(\varepsilon_t) + 2\phi_1^2 \phi_2^2 \text{Cov}(y_{t-1}, y_{t-2}) \\ &= \phi_1^2 V_0 + \phi_2^2 V_0 + 2\phi_1^2 \phi_2^2 V_1 + \sigma^2 \\ (1 - \phi_1^2 - \phi_2^2) V_0 &= \sigma^2 + 2\phi_1^2 \phi_2^2 V_1; \quad V_0 = \frac{\sigma^2 + 2\phi_1^2 \phi_2^2 V_1}{1 - \phi_1^2 - \phi_2^2}\end{aligned}$$

Autocovarianzas

$$\begin{aligned}V_1 &= \text{COV}(y_t, y_{t+1}) = \text{COV}(y_t, \phi_1 y_t + \phi_2 y_{t-1} + \varepsilon_{t+1}) = \\ &\phi_1 \text{COV}(y_t, y_t) + \phi_2 \text{COV}(y_t, y_{t-1}) + \text{COV}(y_t, \varepsilon_{t+1}) = \\ &\phi_1 V_0 + \phi_2 V_1 + 0 \\ V_1 &= \phi_1 V_0 + \phi_2 V_1 \\ (1 - \phi_2) V_1 &= \phi_1 V_0 \\ V_1 &= \frac{\phi_1 V_0}{1 - \phi_2}\end{aligned}$$

$$\begin{aligned}
V_2 &= \text{COV}(y_t, y_{t+2}) = \text{COV}(y_t, \phi_1 y_{t+1} + \phi_2 y_t + \varepsilon_{t+2}) \\
&= \phi_1 \text{COV}(y_t, y_{t+1}) + \phi_2 \text{COV}(y_t, y_t) \\
&\quad + \text{COV}(y_t, \varepsilon_{t+2}) = \phi_1 V_1 + \phi_2 V_0 + 0
\end{aligned}$$

$$\begin{aligned}
V_3 &= \text{COV}(y_t, y_{t+3}) = \text{COV}(y_t, \phi_1 y_{t+2} + \phi_2 y_{t+1} + \varepsilon_{t+3}) \\
&= \phi_1 \text{COV}(y_t, y_{t+2}) + \phi_2 \text{COV}(y_t, y_{t+1}) \\
&\quad + \text{COV}(y_t, \varepsilon_{t+3}) = \phi_1 V_2 + \phi_2 V_1
\end{aligned}$$

$$\begin{aligned}
V_4 &= \text{COV}(y_t, y_{t+4}) = \text{COV}(y_t, \phi_1 y_{t+3} + \phi_2 y_{t+2} + \varepsilon_{t+4}) \\
&= \phi_1 \text{COV}(y_t, y_{t+3}) + \phi_2 \text{COV}(y_t, y_{t+2}) \\
&\quad + \text{COV}(y_t, \varepsilon_{t+4}) = \phi_1 V_3 + \phi_2 V_2
\end{aligned}$$

$$V_k = \phi_1 V_{k-1} + \phi_2 V_{k-2} \quad \text{para todo } k > 1$$

Autocorrelaciones

$$\rho_0 = \frac{V_0}{V_0} = 1$$

$$\rho_1 = \frac{V_1}{V_0}$$

$$\rho_2 = \frac{\phi_1 V_1 + \phi_2 V_0}{V_0}$$

$$\rho_3 = \frac{\phi_1 V_2 + \phi_2 V_1}{V_0}$$

$$\rho_k = \frac{\phi_1 V_{k-1} + \phi_2 V_{k-2}}{V_0}$$

MA(1)

Media

$$\mu = E(y_t) = E(\varepsilon_t + \theta\varepsilon_{t-1}) = E(\varepsilon_t) + \theta E(\varepsilon_{t-1}) = 0$$

Varianza

$$\begin{aligned} \text{VAR}(y_t) &= \text{VAR}(\varepsilon_t + \theta\varepsilon_{t-1}) = \text{VAR}(\varepsilon_t) + \theta^2 \text{VAR}(\varepsilon_{t-1}) = \\ &\sigma^2 + \theta^2 \sigma^2 = V_0 = (1 + \theta^2) \sigma^2 \end{aligned}$$

Autocovarianzas

$$V_1 = \text{COV}(y_t, y_{t+1}) = \text{COV}(\varepsilon_t + \theta\varepsilon_{t-1}, \varepsilon_{t+1} + \theta\varepsilon_t) = \theta \text{cov}(\varepsilon_t, \varepsilon_t) = \theta\sigma^2$$

$$V_2 = \text{COV}(y_t, y_{t+2}) = \text{COV}(\varepsilon_t + \theta\varepsilon_{t-1}, \varepsilon_{t+2} + \theta\varepsilon_{t+1}) = 0$$

$$V_k = 0 \text{ Para todo valor de } k > 1$$

Autocorrelaciones

$$\rho_0 = \frac{V_0}{V_0} = 1$$

$$\rho_1 = \frac{V_1}{V_0} = \frac{\theta\sigma^2}{(1 + \theta^2)\sigma^2} = \frac{\theta}{1 + \theta^2}$$

$$\rho_2 = \frac{V_2}{V_0} = 0$$

$\rho_k = 0$ para todo $k > 1$

MA(2)

Media

$$\mu = E(y_t) = E(\varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2})$$

Varianza

$$\begin{aligned} \text{VAR}(y_t) &= \text{VAR}(\varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2}) = \\ \text{VAR}(\varepsilon_t) &+ \theta_1^2 \text{VAR}(\varepsilon_{t-1}) + \theta_2^2 \text{VAR}(\varepsilon_{t-2}) = \\ \sigma^2 &+ \theta_1^2 \sigma^2 + \theta_2^2 \sigma^2 = (1 + \theta_1^2 + \theta_2^2) \sigma^2 \end{aligned}$$

Autocovarianzas

$$\begin{aligned} V_1 &= \text{COV}(y_t, y_{t+1}) = \text{COV}(\varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2}, \varepsilon_{t+1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) \\ &= \theta_1 \text{cov}(\varepsilon_t, \varepsilon_t) + \theta_1 \theta_2 \text{ov}(\varepsilon_{t-1}, \varepsilon_{t-1}) = \theta_1 \sigma^2 + \theta_1 \theta_2 \sigma^2 \\ &= (\theta_1 + \theta_1 \theta_2) \sigma^2 \end{aligned}$$

$$\begin{aligned} V_2 &= \text{COV}(y_t, y_{t+2}) = \text{COV}(\varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2}, \varepsilon_{t+2} + \theta_1 \varepsilon_{t+1} + \theta_2 \varepsilon_t) \\ &= \theta_2 \text{ov}(\varepsilon_t, \varepsilon_t) = \theta_2 \sigma^2 \end{aligned}$$

$$\begin{aligned} V_3 &= \text{COV}(y_t, y_{t+3}) = \\ \text{COV}(\varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2}, \varepsilon_{t+3} + \theta_1 \varepsilon_{t+2} + \theta_2 \varepsilon_{t+1}) &= 0 \end{aligned}$$

$V_k = 0$ Para todo valor de $k > 2$

Autocorrelaciones

$$\rho_0 = \frac{V_0}{V_0} = 1$$

$$\rho_1 = \frac{V_1}{V_0} = \frac{(\theta_1 + \theta_1\theta_2) \sigma^2}{(1 + \theta_1^2 + \theta_2^2) \sigma^2} = \frac{\theta_1 + \theta_1\theta_2}{1 + \theta_1^2 + \theta_2^2}$$

$$\rho_2 = \frac{V_2}{V_0} = \frac{\theta_2 \sigma^2}{(1 + \theta_1^2 + \theta_2^2) \sigma^2} = \frac{\theta_2}{1 + \theta_1^2 + \theta_2^2}$$

$$\rho_3 = 0$$

Condiciones estacionaridad

MA - Siempre es estacionario

AR - Si las raíces, en valor absoluto, son menores que 1

ARMA - Si la parte AR es estacionaria

De manera concreta:

$$\text{AR}(1) \quad |\phi| < 1$$

$$\text{AR}(2) \quad \phi_1 + \phi_2 < 1$$

$$\phi_2 - \phi_1 < 1$$

$$\phi_2 > -1$$

Para valores mayores de 2 es necesario calcular las raíces con un ordenador.

Selección de modelos

Existen tres opciones para determinar cuál es el mejor modelo

- $AIC = -2\log - \text{lik} + 2k$ $k = \text{números de parámetros}$
- $BIC = -2\log - \text{lik} + k * \ln(n)$ $n = \text{números de observaciones}$

En ambos el mejor es el menor.

- $LM = 2(\log - \text{Lik}_{MNR} - \text{Log} - \text{lik}_{MR}) \sim \chi_k^2$
siendo k el número de restricciones

Se compara con el valor en tablas de la χ^2

Tienen que estar anidados es decir ser una modificación de otro.

No se puede sustraer de ambas partes a la vez (AR y MA)

Ejemplo:

Elegir el mejor modelo

$N=1000$

Modelo	Log-lik
AR (2)	50,7
MA (2)	29,3
ARMA(1,1)	50,2
ARMA(2,2)	55,3

AIC

$$AR(2) = -2 * 50,7 + 2 * 2 = -97,4$$

$$MA(2) = -2 * 29,3 + 2 * 2 = -54,6$$

$$ARMA(1,1) = -2 * 50,2 + 2 * 2 = -96,4$$

$$ARMA(2,2) = -2 * 55,3 + 2 * 4 = -102,4$$

BIC

$$AR(2) = -2 * 50,7 + 2\ln(1000) = -87,58$$

$$MA(2) = -2 * 29,3 + 2\ln(1000) = -44,78$$

$$ARMA(1,1) = -2 * 50,2 + 2\ln(1000) = -86,58$$

$$ARMA(2,2) = -2 * 55,3 + 2\ln(1000) = -82,77$$

LM

AR(2) versus MA(2) (NO)

AR(2) versus ARMA(1,1) (No)

AR(2) versus ARMA(2,2) (SI)

$$H_0: \theta_1 = \theta_2 = 0$$

$$H_1: \text{Algún } \theta_i \neq 0$$

$$LM = 2 * (55,3 - 50,7) = 9 \sim \chi_2^2 = 5,99$$

Rechazo HO elijo ARMA (2,2)

MA(2) versus ARMA(1,1) (NO)

MA(2) versus ARMA(2,2) (SI)

$$H_0: \phi_1 = \phi_2 = 0$$

$$H_1: \text{Algún } \phi_i \neq 0$$

$LM = 2 * (55,2 - 29,3) = 51,8 \sim x_2^2 =$
 5,99 rechazo H_0 elijo ARMA (2,2)

ARMA(1,1) versus ARMA(2,2) (NO)

Respuesta final \leftrightarrow ARMA (2,2)

Predicciones

ARMA(2,2)

$$E(Y_{T+k|T}) = E(\phi_1 Y_{T+k-1|T} + \phi_2 Y_{T+k-2|T} + \varepsilon_{t+k|T} + \theta_1 \varepsilon_{t+k-1|T} + \theta_2 \varepsilon_{t+k-2|T})$$

$$\tilde{y}_{T+k|T} = \phi_1 \tilde{y}_{T+k-1|T} + \phi_2 \tilde{y}_{T+k-2|T} + \tilde{\varepsilon}_{T+k|T} + \theta_1 \tilde{\varepsilon}_{T+k-1|T} + \theta_2 \tilde{\varepsilon}_{T+k-2|T}$$

K=1

$$\tilde{y}_{T+1|T} = \phi_1 \tilde{y}_{T|T} + \phi_2 \tilde{y}_{T-1|T} + \tilde{\varepsilon}_{T+1|T} + \theta_1 \tilde{\varepsilon}_{T|T} + \theta_2 \tilde{\varepsilon}_{T-1|T}$$

$$\tilde{y}_{T+1|T} = \phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}$$

K=2

$$\tilde{y}_{T+2|T} = \phi_1 \tilde{y}_{T+1|T} + \phi_2 \tilde{y}_{T|T} + \tilde{\varepsilon}_{T+2|T} + \theta_1 \tilde{\varepsilon}_{T+1|T} + \theta_2 \tilde{\varepsilon}_{T|T}$$

$$\tilde{y}_{T+2|T} = \phi_1 \tilde{y}_{T+1|T} + \phi_2 y_t + \theta_2 \varepsilon_t$$

$$\tilde{y}_{T+2|T} = \phi_1 (\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) + \phi_2 y_t + \theta_2 \varepsilon_t$$

K=3

$$\tilde{y}_{T+3|T} = \phi_1 \tilde{y}_{T+2|T} + \phi_2 \tilde{y}_{T+1|T} + \tilde{\varepsilon}_{T+3|T} + \theta_1 \tilde{\varepsilon}_{T+2|T} + \theta_2 \tilde{\varepsilon}_{T+1|T}$$

$$\tilde{y}_{T+2|T} = \phi_1 \tilde{y}_{T+1|T} + \phi_2 \tilde{y}_{T|T}$$

$$\begin{aligned} \tilde{y}_{T+3|T} &= \phi_1^2 (\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) + \phi_1 (\phi_2 y_t + \theta_2 \varepsilon_t) + \\ &\phi_2 (\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) \end{aligned}$$

Error Cuadrático Medio

$$ECM = E(\tilde{y}_{T+k|T} - y_T)^2$$

K=1

$$ECM_1 = E(\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1})$$

$$-(\phi_1 y_t + \phi_2 y_{t-1} + \varepsilon_{t+1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-2})^2$$

$$= E(\varepsilon_t)^2 = \sigma^2$$

K=2

$$ECM_2 = E(\phi_1 (\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) + \phi_2 y_t + \theta_2 \varepsilon_t$$

$$-(\phi_1 y_{t+1} + \phi_2 y_t + \varepsilon_{t+2} + \theta_1 \varepsilon_{t+1} + \theta_2 \varepsilon_t))^2 =$$

$$= E(\phi_1 (\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) - (\phi_1 y_{t+1} + \varepsilon_{t+2} + \theta_1 \varepsilon_{t+1}))^2$$

K=3

$$ECM_3 = E(\phi_1^2(\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) + \phi_1(\phi_2 y_t + \theta_2 \varepsilon_t) +$$

$$\phi_2(\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) \\ - (\phi_1 y_{t+2} + \phi_2 y_{t+1} + \varepsilon_{t+3} + \theta_1 \varepsilon_{t+3} + \theta_2 \varepsilon_{t+1}))^2$$

$$ECM_3 = E(\phi_1^2(\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) + \phi_1(\phi_2 y_t + \theta_2 \varepsilon_t)$$

$$+ \phi_2(\phi_1 y_t + \phi_2 y_{t-1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) \\ - (\phi_1 y_{t+2} + \phi_2 y_{t+1} + \varepsilon_{t+3} + \theta_1 \varepsilon_{t+3} + \theta_2 \varepsilon_{t+1}))^2$$