

Standard assumptions for ARMA(p,q) model

$$\phi(B)y_t = \theta(B)\omega_t$$

(1) causal: $\{y_t\}$ can be written in form

$$y_t = \sum_{j=0}^{\infty} \psi_j \omega_{t-j} = \psi(B)\omega_t$$

where $\psi_0 = 1$; $\sum_{j=0}^{\infty} |\psi_j| < \infty$

(2) invertible: $\{y_t\}$ can be written in form

$$\omega_t = \sum_{j=0}^{\infty} \pi_j y_{t-j} = \pi(B)y_t$$

where $\pi_0 = 1$; $\sum_{j=0}^{\infty} |\pi_j| < \infty$

(3) No parameter redundancy

$\phi(z)$ & $\theta(z)$ have no common factors.

Example: consider the model

$$(*) \quad y_t = \omega_t, \quad \omega_t \sim \text{WN}(\sigma^2 \omega)$$

Now suppose we try fitting an ARMA(1,1) model to data generated from (*).

$$(**) \quad (1 - \phi_1 B)y_t = (1 + \theta_1 B)\omega_t$$

recall that (**) & (*) are identical

$\psi_1 = -\theta_1$. In practice I will typically find that the hypothesis $\hat{\psi} = -\hat{\theta}$ is not rejected.

Theorem: An ARMA(p,q) model is causal iff all roots of $\varphi(z)$ lie outside the unit circle (in the complex plane).

Theorem: An ARMA(p,q) model is invertible iff all roots of $\theta(z)$ lie outside the unit circle (in the complex plane).

Example. $y_t = \varphi_1 y_{t-1} + w_t$

$$\varphi(z) = 1 - \varphi_1 z$$

$$1 - \varphi_1 z = 0 \quad \text{iff} \quad z = \frac{1}{\varphi_1}$$

This is outside the unit circle iff $|\varphi_1| < 1$.

Example: $y_t = 2y_{t-1} + 3y_{t-2} + w_t$

$$\varphi(z) = 1 - 2z - 3z^2 = 0$$

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$$(1 - 3z)(1 + z)$$

has roots $z = 1/3, -1$

Neither are outside unit circle, so not causal.

Example : $y_t = w_t - \frac{1}{4} w_{t-2}$

$$\begin{aligned} \theta(z) &= 1 - \frac{1}{4} z^2 \\ &= (1 - \frac{1}{2} z)(1 + \frac{1}{2} z) \end{aligned}$$

has roots

$$z = \pm 2$$

Both roots are outside unit circle, so invertible.

Example : $y_t = 0.4 y_{t-1} + 0.45 y_{t-2} + w_t + w_{t-1} + 0.25 w_{t-2}$

$$\phi(z) = 1 - 0.4z - 0.45z^2 = \cancel{(1 + 0.5z)}(1 - 0.9z)$$

$$\theta(z) = 1 + z + 0.25z^2 = \cancel{(1 + 0.5z)^2}$$

There is parameter redundancy. After eliminating the redundancy, the model is

$$y_t - 0.9 y_{t-1} = w_t + 0.5 w_{t-1}$$

is both invertible & causal.

We learnt previously how to write an ARMA (p,q) model in AR(∞) or MA(∞) using coefficient matching.

Can also do using R functions

ARMA to AR ()

ARMA to MA ()

3.3 Difference equations

A homogeneous difference equation of order p in of form:

$$u_n = \alpha_1 u_{n-1} + \alpha_2 u_{n-2} + \dots + \alpha_p u_{n-p}$$

Using lag operator notation, we can write this as

$$\alpha(B) u_n = 0$$

where $\alpha(z) = 1 - \alpha_1 z - \alpha_2 z^2 - \dots - \alpha_p z^p$

is referred to as the characteristic polynomial.

Given initial conditions u_0, u_1, \dots, u_{p-1} , we can easily solve for $u_p, u_{p+1}, u_{p+2}, \dots$

Theorem

It turns out that $\{u_i\}$ has the property

$$\lim_{i \rightarrow \infty} u_i = 0$$

iff the roots of the characteristic polynomial are all outside the unit circle.

Remark: If we write an ARMA(p,q) model in AR(∞) form, the coefficients $\pi_1, \pi_2, \pi_3, \dots$ form a difference equation of order q with characteristic polynomial $\theta(z)$ & initial conditions are determined by $\varphi_1, \varphi_2, \dots, \varphi_p$.

Analogously, if we try to write the model in MA(∞) form, the coefficients ψ_1, ψ_2, \dots form a difference equation of order p with characteristic polynomial $\varphi(z)$ & initial conditions determined by $\theta_1, \theta_2, \dots, \theta_q$.

Yule-Walker equations

Example: ACF of AR(2)

$$y_t = \phi_1 y_{t-1} + \phi_2 y_{t-2} + w_t$$

Assume causal.

Multiplying through by y_{t-h} & taking $E[\cdot]$,
we get

$$(*) \quad E[y_{t+h} y_t] = \phi_1 E[y_{t+h} y_{t-1}] + \phi_2 E[y_{t+h} y_{t-2}] + E[y_{t+h} w_t]$$

$$\text{Note that } E[y_{t+h} w_t] = \begin{cases} \sigma_w^2 & h=0 \\ 0 & h=1, 2, 3, \dots \end{cases}$$

So (*) implies

$$\left. \begin{aligned} (h=0) \quad \gamma_0 &= \phi_1 \gamma_1 + \phi_2 \gamma_2 + \sigma_w^2 \\ (h=1) \quad \gamma_1 &= \phi_1 \gamma_0 + \phi_2 \gamma_1 \\ (h=2) \quad \gamma_2 &= \phi_1 \gamma_1 + \phi_2 \gamma_0 \end{aligned} \right\} \begin{array}{l} \text{Yule-Walker} \\ \text{equations.} \\ \text{If we know } \sigma_w^2, \\ \text{3 eq's in 3 unknowns.} \end{array}$$

Can solve for $\gamma_0, \gamma_1, \gamma_2$

$$(h=3) \quad \gamma_3 = \phi_1 \gamma_2 + \phi_2 \gamma_1$$

$$(\cancel{h=4}) \therefore \gamma_h = \phi_1 \gamma_{h-1} + \phi_2 \gamma_{h-2} \quad h=3, 4, 5, \dots$$

~~Recall, AR(1)~~

$$y_t = \phi y_{t-1} + \omega_t$$

$$\delta_h = \phi^h \delta_0$$

$$1 - \phi^2$$

In practice it is easier to divide the Yule-Walker equations by δ_0 .

$$P_0 = 1$$

$$P_1 = \phi_1 P_0 + \phi_2 P_1$$

$$= \phi_1 / (1 - \phi_2)$$

$$P_2 = \phi_1 P_1 + \phi_2 P_0$$

$$P_h = \phi_1 P_{h-1} + \phi_2 P_{h-2}, \quad h = 2, 3, 4, \dots$$

So, $\delta_h = P_h \cdot \delta_0$

where $\delta_0 = \frac{\sigma_\omega^2}{1 - \phi_1 P_1 - \phi_2 P_2}$

ACF of AR(p)

$$y_t = \varphi_1 y_{t-1} + \dots + \varphi_p y_{t-p} + w_t$$

Assume causal.

$$E[y_{t+h} y_t] = \varphi_1 E[y_{t+h} y_{t+1}] + \dots + \varphi_p E[y_{t+h} y_{t-p}] + E[y_{t+h} w_t]$$

$$\text{Note that } E[y_{t+h} w_t] = \begin{cases} \sigma_w^2 & h=0 \\ 0 & h>0 \end{cases}$$

$$\gamma_{-h} = \gamma_h = \varphi_1 \gamma_{h-1} + \varphi_2 \gamma_{h-2} + \dots + \varphi_p \gamma_{h-p} \quad h=1, 2, 3, \dots$$

is a difference equation with initial conditions

$$\gamma_0 = \varphi_1 \gamma_1 + \dots + \varphi_p \gamma_p + \sigma_w^2$$

$$\gamma_1 = \varphi_1 \gamma_0 + \dots + \varphi_p \gamma_{p-1}$$

\vdots

$$\gamma_p = \varphi_1 \gamma_{p-1} + \dots + \varphi_p \gamma_0$$

Yule-Walker equations

Given σ_w^2 , this is p equations in p unknowns, so I can solve for

$$\gamma_0, \gamma_1, \dots, \gamma_{p-1}$$

For next time read about PACF.