

Ch 3 ARIMA Models

- Recall last week: Box-Jenkins Methodology.
(See also Sec. 3.1 of book).

Example AR(1)

$$y_t = \varphi y_{t-1} + w_t, \quad w_t \sim WN(\sigma_w^2)$$
$$t = 0, \pm 1, \pm 2, \dots \quad |\varphi| < 1$$

Recall:

$$\begin{aligned} y_t &= \varphi (\varphi y_{t-2} + w_{t-1}) + w_t \\ &= \varphi^2 y_{t-2} + \varphi w_{t-1} + w_t \\ &= \varphi^2 (\varphi y_{t-3} + w_{t-2}) + \varphi w_{t-1} + w_t \\ &= \varphi^3 y_{t-3} + \varphi^2 w_{t-2} + \varphi w_{t-1} + w_t \\ &= \sum_{l=0}^{\infty} \varphi^l w_{t-l} \end{aligned}$$

Forecasting

$$\begin{aligned} \hat{y}_{t+1|t} &= \varphi y_t + E[w_{t+1}] \\ &= \varphi y_t \end{aligned}$$

$$\hat{y}_{t+2|t} = \varphi^2 y_t + E[w_{t+2}] + \varphi E[w_{t+1}]$$

$$\vdots = \varphi^2 y_t$$

$$\hat{y}_{t+k|t} = \varphi^k y_t$$

Assume $h > 0$

$$\gamma(-h) = \gamma(h) = \text{cov}(y_{t+h}, y_t)$$

$$= E \left[\left(\sum_{i=0}^{\infty} \varphi^i w_{t+h-i} \right) \left(\sum_{j=0}^{\infty} \varphi^j w_{t-j} \right) \right]$$

let $k = i - h$

$$= E \left[\left(\sum_{k=-h}^{\infty} \varphi^{k+h} w_{t-k} \right) \left(\sum_{j=0}^{\infty} \varphi^j w_{t-j} \right) \right]$$

$$= E \left[\sum_{k=-h}^{\infty} \sum_{j=0}^{\infty} \varphi^{k+h} \varphi^j w_{t-k} \cdot w_{t-j} \right]$$

These are all zero unless $k = j$

$$= E \left[\sum_{j=0}^{\infty} \varphi^{j+h} \varphi^j w_{t-j}^2 \right]$$

$$= \sum_{j=0}^{\infty} \varphi^{2j+h} \cdot \sigma_w^2$$

$$= \sigma_w^2 \frac{\varphi^h}{1 - \varphi^2}$$

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

Also, $\mu = E[y_t] = 0$.

Example

$$Z_t = \alpha + \varphi Z_{t-1} + w_t$$

↓
constant

$$E[Z_t] = \alpha + \varphi E[Z_{t-1}] + E[w_t]$$

$$\mu = \alpha + \varphi \cdot \mu$$

$$\mu = \frac{\alpha}{1 - \varphi}$$

$\{y_t\}$ & $\{z_t\}$ have the same ACF.

Note that any AR(1) model can be transformed to one that is mean zero by

$$y_t = z_t - \mu$$

We will frequently assume $\alpha = 0$ without loss of generality.

Def: The AR(p) is of form

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

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

Assuming stationarity,

$$\mu = \alpha + \varphi_1 \mu + \dots + \varphi_p \mu$$

$$\mu = \frac{\alpha}{1 - \varphi_1 - \varphi_2 - \dots - \varphi_p}$$

The demarcated model is

$$\begin{aligned} z_t &= y_t - \mu \\ &= \phi_1 z_{t-1} + \dots + \phi_p z_{t-p} + w_t \end{aligned}$$

Recall, we can also write this as

$$z_t - \phi_1 z_{t-1} - \phi_2 z_{t-2} - \dots - \phi_p z_{t-p} = w_t$$

$$(1 - \phi_1 B - \phi_2 B^2 - \dots - \phi_p B^p) z_t = w_t$$

$$\varphi(z) = 1 - \phi_1 z - \dots - \phi_p z^p$$

is referred to as a "lag polynomial"

↓

$$\varphi(B) = 1 - \phi_1 B - \dots - \phi_p B^p$$

is referred to as the autoregressive operator

Explosive AR models

example: AR(1) with $|\varphi| > 1$ is explosive.

Recall that a linear process is one that can be written in form:

$$y_t = \mu + \sum_{i=-\infty}^{\infty} \psi_i w_{t-i}$$

$$\text{where } \sum_{i=-\infty}^{\infty} |\psi_i| < \infty.$$

Recall also that we can write the AR(1) model

$$\text{as } y_t = \sum_{i=0}^{\infty} \varphi^i y_{t-i}$$

So, $\{y_t\}$ is a linear process if and only if $|\varphi| < 1$.

Now suppose $|\varphi| > 1$.

$$y_t = \varphi y_{t-1} + w_t$$

$$y_{t-1} = \frac{1}{\varphi} y_t - \frac{1}{\varphi} w_t$$

$$= \frac{1}{\varphi} \left[\frac{1}{\varphi} y_{t+1} - \frac{1}{\varphi} w_{t+1} \right] - \frac{1}{\varphi} w_t$$

$$= \frac{1}{\varphi^2} y_{t+1} - \frac{1}{\varphi^2} w_{t+1} - \frac{1}{\varphi} w_t$$

$$= \vdots$$

$$= - \sum_{j=1}^{\infty} \varphi^{-j} w_{t+j}$$

Since $\left| \sum_{j=1}^{\infty} \varphi^{-j} \right| < \infty$, this shows that this model can be written a linear process.

recall, $y_t = \varphi y_{t-1} + w_t$ is causal if, only if $|\varphi| < 1$.

We are almost always interested in causal models.

Method of matching coefficients

Given the AR(P) model

$$(1) \quad \varphi(B) y_t = w_t$$

we seek to write this in linear process form

$$(2) \quad y_t = \psi(B) w_t = \sum_{j=0}^{\infty} \psi_j w_{t-j}$$

For the AR(1) model this was easy.

In general, do as follows.

So, best choice (2) mit (1)

$$\varphi(B) \psi(B) w_t = w_t$$

$$\text{So, } \varphi(z) \psi(z) = 1$$

$$(1 - \varphi_1 z - \varphi_2 z^2 - \dots - \varphi_p z^p) (1 + \gamma_1 z + \gamma_2 z^2 + \gamma_3 z^3 + \gamma_4 z^4 + \dots) = 1$$

$$1: 1 \cdot 1 = 1 \quad \gamma_0 = 1$$

$$z: \gamma_1 z - \varphi_1 z = 0 \quad \text{so, } \gamma_1 = \varphi_1$$

$$z^2: \gamma_2 - \varphi_1 \gamma_1 - \varphi_2 = 0 \quad \text{so } \gamma_2 = \varphi_1 \gamma_1 + \varphi_2$$

$$z^3: \gamma_3 - \varphi_1 \gamma_2 - \varphi_2 \gamma_1 - \varphi_3 = 0 \quad \text{so, } \gamma_3 = \varphi_1 \gamma_2 + \varphi_2 \gamma_1 + \varphi_3$$

⋮

$$y_t = \mu + \sum_{j=0}^{\infty} \gamma_j w_{t-j}$$

is sometimes referred to as the "Wold" form
or MA(∞).

Exercise: Write

$$y_t = 0.5 y_{t-1} + 0.3 y_{t-2} + w_t$$

in Wold form.

given

$$\varphi(B) y_t = w_t$$

$$\text{where } \varphi(z) = 1 - 0.5z - 0.3z^2$$

We seek

$$y_t = \psi(B) w_t$$

Substituting, we get

$$\varphi(B) \psi(B) w_t = w_t$$

So,

$$\varphi(z) \psi(z) = 1$$

$$\left(\underbrace{(1 - 0.5z - 0.3z^2)}_{\text{}} \left(1 + \psi_1 z + \psi_2 z^2 + \psi_3 z^3 + \dots \right) \right) = 1$$

$$z^0: -0.5 + \psi_1 = 0 \quad \text{so, } \psi_1 = 0.5$$

$$z^1: -0.3 - 0.5\psi_1 + \psi_2 = 0 \quad \text{so, } \psi_2 = 0.5\psi_1 + 0.3$$
$$= 0.5 \cdot 0.5 + 0.3$$
$$= 0.55$$

$$z^2: -0.3\psi_1 - 0.5\psi_2 + \psi_3 = 0 \quad \text{so, } \psi_3 = 0.5\psi_2 + 0.3\psi_1$$
$$= 0.425$$

⋮

$$z^k: -0.3\psi_{k-2} - 0.5\psi_{k-1} + \psi_k = 0$$
$$\psi_k = 0.5\psi_{k-1} + 0.3\psi_{k-2}$$