

Review

Example: ARMA(2,2)

$$y_t = \frac{5}{6} y_{t-1} - \frac{1}{6} y_{t-2} + w_t - 1.7 w_{t-1} + 0.72 w_{t-2}$$

$$\varphi(z) y_t = \theta(z) w_t$$

where $\varphi(z) = 1 - \frac{5}{6}z + \frac{1}{6}z^2$
 $= (1 - \frac{2}{3}z)(1 - \frac{1}{2}z)$ $|3| > 1$; $|2| > 1$
roots at $z = 3, 2$. So it is causal

$$\theta(z) = 1 - 1.7z + 0.72z^2$$
$$= (1 - 0.9z)(1 - 0.8z)$$

roots at $z = \frac{1}{0.9}, \frac{1}{0.8}$ so it is invertible.

• Get wild decomposition (i.e., MA(∞) form):

given $\varphi(z) y_t = \theta(z) w_t$

we seek $\psi(z)$ such that

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

Substituting, we get

$$\varphi(z) \psi(z) w_t = \theta(z) w_t$$

So, we must have

$$\varphi(z) \psi(z) \equiv \theta(z) \quad \text{for all } z.$$

$$\underbrace{\left(1 - \frac{5}{6}z + \frac{1}{6}z^2\right)}_{\varphi(z)} \underbrace{\left(1 + \gamma_1 z + \gamma_2 z^2 + \gamma_3 z^3 + \dots\right)}_{\psi(z)} = 1 - 1.7z + 0.72z^2$$

$$z^0: \quad \gamma_1 - \frac{5}{6} = -1.7 \quad \gamma_1 = \frac{5}{6} - 1.7 = -0.8667$$

$$z^1: \quad \gamma_2 - \frac{5}{6}\gamma_1 + \frac{1}{6} = 0.72 \quad \gamma_2 = \frac{5}{6}\gamma_1 - \frac{1}{6} + 0.72$$
$$= -0.16889$$

$$\vec{z}^1: \gamma_3 - 5/6 \gamma_2 + 1/6 \gamma_1 = 0 \quad \gamma_3 = 5/6 \gamma_2 - 1/6 \gamma_1 = 0.0037$$

$$\vec{z}^k: \gamma_k - 5/6 \gamma_{k-1} + 1/6 \gamma_{k-2} = 0$$

Note: This is a difference equation of order 2 depending on φ with initial conditions that depend on θ .

We could also find AR(ω) form...
(exercise)

Now, let's compute the ACF.

Recall,
$$\gamma_h = \varphi_1 \gamma_{h-1} + \varphi_2 \gamma_{h-2} + \sigma_w^2 \sum_{i=0}^{\infty} \theta_i \gamma_{i-h}$$

check you notes. May have slipped sign previous's

$$h=0 \quad \gamma_0 = \varphi_1 \gamma_1 + \varphi_2 \gamma_2 + \sigma_w^2 \left[1 + \theta_1 \gamma_1 + \theta_2 \gamma_2 \right]$$

Yule Walker equation

recall $\gamma_1 = \gamma_{-1}$
 $\gamma_2 = \gamma_{-2}$

$$h=1 \quad \gamma_1 = \varphi_1 \gamma_0 + \varphi_2 \gamma_1 + \sigma_w^2 \left[1 \cdot 0 + \theta_1 \gamma_0 + \theta_2 \gamma_1 \right]$$

recall $\gamma_j = 0$ for $j < 0$

$$h=2 \quad \gamma_2 = \varphi_1 \gamma_1 + \varphi_2 \gamma_0 + \sigma_w^2 \left[1 \cdot 0 + \theta_1 \cdot 0 + \theta_2 \cdot 1 \right]$$

since $\theta_0 = 1$ since $\gamma_0 = 1$

$$h > 2 \quad \gamma_h = \varphi_1 \gamma_{h-1} + \varphi_2 \gamma_{h-2}$$

Notice that this is a difference equation of order 2 ~~with~~ that depends only on φ with initial conditions that depend on θ .

get the initial conditions by solving Yule-Walker equations for $\gamma_0, \gamma_1, \dots, \gamma_2$.

Now, let's compute the PACF

To get φ_{hh} , consider the population regression

$$y_t = \varphi_{h1} y_{t-1} + \dots + \varphi_{hh} y_{t-h} + \varepsilon_t$$

$$\mathbf{X} = \begin{bmatrix} y_{t-1} \\ \vdots \\ y_{t-h} \end{bmatrix} \quad \mathbf{Y} = y_t$$

$$\begin{bmatrix} \varphi_{h1} \\ \vdots \\ \varphi_{hh} \end{bmatrix} = \left[E(\mathbf{X}'\mathbf{X}) \right]^{-1} E(\mathbf{X}'\mathbf{Y})$$

recall:

φ_{hh} is the only coefficient I care about.

$$= \begin{bmatrix} \gamma_0 & \gamma_1 & \gamma_2 & \dots & \gamma_{h-1} \\ \gamma_1 & \gamma_0 & \gamma_1 & \dots & \gamma_{h-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \gamma_{h-1} & \gamma_{h-2} & \dots & \gamma_1 & \gamma_0 \end{bmatrix}^{-1} \begin{bmatrix} \gamma_1 \\ \gamma_2 \\ \vdots \\ \gamma_h \end{bmatrix}$$