

### 3.1

For  $MA(1)$ , the autocorrelation function is given as:

$$\begin{aligned}\gamma(h) &= Cov[x_{t+h}, x_t] \\ \gamma(h) &= Cov[w_{t+h} + \theta w_{t+h-1}, w_t + \theta w_{t-1}]\end{aligned}$$

When  $h = 0$ :

$$\begin{aligned}\gamma(0) &= Cov[w_t + \theta w_{t-1}, w_t + \theta w_{t-1}] \\ \gamma(0) &= Var[w_t + \theta w_{t-1}] \\ \gamma(0) &= (1 + \theta^2)\sigma_w^2\end{aligned}$$

When  $h = 1$ :

$$\begin{aligned}\gamma(1) &= Cov[w_{t+1} + \theta w_t, w_t + \theta w_{t-1}] \\ \gamma(1) &= E[(w_{t+1} + \theta w_t)(w_t + \theta w_{t-1})] - E[w_{t+1} + \theta w_t]E[w_t + \theta w_{t-1}] \\ \gamma(1) &= E[(w_{t+1} + \theta w_t)(w_t + \theta w_{t-1})] - 0 \\ \gamma(1) &= E[\theta w_t w_t] \\ \gamma(1) &= \theta \sigma_w^2\end{aligned}$$

Solve for  $\rho_x(1)$ :

$$\begin{aligned}\rho_x(1) &= \frac{\gamma(1)}{\gamma(0)} \\ \rho_x(1) &= \frac{\theta \sigma_w^2}{(1 + \theta^2)\sigma_w^2} \\ \rho_x(1) &= \frac{\theta}{1 + \theta^2}\end{aligned}$$

Find which  $\theta$  results in minimum and maximum values of  $\rho_x(1)$ :

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

Set  $\frac{\delta \rho_x(1)}{\delta \theta} = 0$ , then the maximum and minimum values are at:

$$\theta = \pm 1$$

At  $\theta = 1$ :

$$\rho_x(1) = \frac{1}{1 + 1} = \frac{1}{2}$$

At  $\theta = -1$ :

$$\rho_x(1) = \frac{-1}{1 + 1} = -\frac{1}{2}$$

We know that  $\theta = \pm 1$  are global maximums and minimums since both limits to infinity are within the range of the minimum and maximum:

$$\begin{aligned}\lim_{\theta \rightarrow -\infty} \frac{\theta}{1 + \theta^2} &= \lim_{\theta \rightarrow -\infty} \frac{1}{2\theta} = 0 \\ \lim_{\theta \rightarrow \infty} \frac{\theta}{1 + \theta^2} &= \lim_{\theta \rightarrow -\infty} \frac{1}{2\theta} = 0\end{aligned}$$

Since  $-\frac{1}{2} \leq \rho_x(1) \leq \frac{1}{2}$ :

$$|\rho_x(1)| \leq \frac{1}{2}$$

### 3.4

(a)

Rewriting the model with backshift operator:

$$\begin{aligned}x_t &= .8Bx_t - .15B^2x_t + w_t - .3Bw_t \\x_t - .8Bx_t + .15B^2x_t &= w_t - .3Bw_t \\(1 - .3B)(1 - .5B)x_t &= (1 - .3B)w_t\end{aligned}$$

We can reduce the model to:

$$(1 - .5B)x_t = w_t$$

This is an  $AR(1)$  model.

AR polynomial:

$$\phi(B) = 1 - .5B$$

MA polynomial:

$$\theta(B) = 1$$

Check the roots of AR polynomial:

$$\begin{aligned}1 - .5z &= 0 \\z &= 2\end{aligned}$$

Since the root is outside of the unit circle, this process is causal.

There is no MA component, so this process is trivially invertible.

The causal representation is:

$$\begin{aligned}x_t &= .5x_{t-1} + w_t \\x_t &= .5(.5x_{t-2} + w_{t-1}) + w_t \\&\dots \\x_t &= \sum_{j=0}^{\infty} .5^j w_{t-j}\end{aligned}$$

The invertible representation is:

$$w_t = x_t - .5x_{t-1}$$

(b)

Rewriting with backshift operator:

$$\begin{aligned}x_t &= Bx_t - .5B^2x_t + w_t - Bw_t \\x_t - Bx_t + .5B^2x_t &= w_t - Bw_t \\(1 - B + .5B^2)x_t &= (1 - B)w_t\end{aligned}$$

Since we can't factor  $1 - B + .5B^2$ , our model is already in simplest form with no redundant parameters.

This is an  $ARMA(2, 1)$  model.

AR polynomial:

$$\phi(B) = 1 - B + .5B^2$$

MA polynomial:

$$\theta(B) = 1 - B$$

Check the roots of AR polynomial:

$$1 - z + .5z^2 = 0$$

Using the quadratic formula:

$$\begin{aligned} z &= \frac{1 \pm \sqrt{(-1)^2 - 4 * .5 * 1}}{2 * .5} \\ z &= 1 \pm \sqrt{-1} \\ z &= 1 \pm i \end{aligned}$$

Since the complex roots are outside of the unit circle, this process is causal.

Check the roots of MA polynomial:

$$\begin{aligned} 1 - z &= 0 \\ z &= 1 \end{aligned}$$

Since the root is on the unit circle, this process is NOT invertible.

Since this process is causal, this can be written as:

$$x_t = \sum_{j=0}^{\infty} \psi_j w_{t-j}$$

To find the causal representation, derive the Coefficients. Let:

$$\psi(B) = \sum_{j=0}^{\infty} \psi_j B^j$$

Since:

$$x_t = \frac{\theta(B)}{\phi(B)} w_t$$

We can use the identity:

$$\phi(B)\psi(B) = \theta(B)$$

In time-domain convolution, this gives the recursion:

$$\sum_{k=0}^2 \phi_k \psi_{j-k} = \theta_j$$

Where:

- $\phi_0 = 1, \phi_1 = -1, \phi_2 = 0.5$
- $\theta_0 = 1, \theta_1 = -1, \theta_j = 0$  for  $j \geq 2$

We initialize:

- $\psi_{-1} = 0$
- $\psi_0 = 1$

Then recursively compute:

- $\psi_1 = \psi_0 - 0.5\psi_{-1} + \theta_1 = 1 - 0 - 1 = 0$
- $\psi_2 = \psi_1 - 0.5\psi_0 + 0 = 0 - 0.5 = -0.5$
- $\psi_3 = \psi_2 - 0.5\psi_1 = -0.5 - 0 = -0.5$
- $\psi_4 = \psi_3 - 0.5\psi_2 = -0.5 + 0.25 = -0.25$
- etc.

The coefficients follow the second-order linear recurrence:

$$\begin{cases} \psi_0 = 1 \\ \psi_1 = 0 \\ \psi_j = \psi_{j-1} - 0.5\psi_{j-2}, \quad j \geq 2 \end{cases}$$

The causal representation of the process is:

$$x_t = \sum_{j=0}^{\infty} \psi_j w_{t-j}$$

Where:

$$\begin{cases} \psi_0 = 1 \\ \psi_1 = 0 \\ \psi_j = \psi_{j-1} - 0.5\psi_{j-2}, \quad j \geq 2 \end{cases}$$