

# Time Series Analysis - Taching Session 01

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## 1 Assignment 12

**Exercise:** Let  $x_t$  be a zero-mean, unit-variance stationary process with autocorrelation function  $\rho_h$ . Suppose that  $\mu_t$  is a nonconstant function and that  $\sigma_t$  is a positive-valued nonconstant function. The observed series is formed as  $y_t = \mu_t + \sigma_t x_t$

a) Find the mean and covariance function for the  $y_t$  process.

b) Show that the autocorrelation function for the  $y_t$  process depends only on the time lag. Is the  $y_t$  process stationary?

c) Is it possible to have a time series with a constant mean and with  $\text{corr}(y_t, y_{t+h})$  free of  $t$  but with  $y_t$  not stationary?

### 1.1 Mean and Covariance Function

#### 1.1.1 Mean

$$E[y_t] = E[\mu_t] + E[\sigma_t x_t] = E[\mu_t] + E[\sigma_t] E[x_t]$$

As  $E[x_t]$  is 0, we obtain

$$= E[\mu_t] = \mu_t$$

As  $\mu_t$  is a non-constant function by definition.

### 1.1.2 Covariance Function

$$\gamma(s, t) = \text{cov}(y_s, y_t) = E[(y_s - E[y_s])(y_t - E[y_t])]$$

We substitute  $y_s = \mu_s + \sigma_s x_s$  (by definition) and we already know that  $E[s_t] = \mu_t$ .

$$= E[\mu_s + \sigma_s x_s - \mu_s](\mu_t + \sigma_t x_t - \mu_t) = E[\sigma_s x_s \sigma_t x_t]$$

As  $\sigma$  is a function by definition, there is no randomness and we can put it out of the expectation.

$$= \sigma_s \sigma_t E[x_s x_t]$$

We know that  $E[x_t] = 0$ , so we can easily add it to get a known shape.

$$= \sigma_s \sigma_t E[(x_s - E[x_s])(x_t - E[x_t])]$$

The known shape is the definition of the covariance, so we obtain the following.

$$= \sigma_s \sigma_t \text{cov}(x_s, x_t) = \sigma_s \sigma_t \gamma(x_s, x_t)$$

The ACF is defined by

$$\rho(s, t) = \frac{\gamma(x_s, x_t)}{\sqrt{\gamma(x_s, x_s)\gamma(x_t, x_t)}}$$

Solving for  $\gamma(s, t)$  and then plugging in in the equation above we obtain

$$= \sigma_s \sigma_t \rho(x_s, x_t) \sqrt{\gamma(x_s, x_s)\gamma(x_t, x_t)}$$

We know that  $\gamma(x_s, x_s) = \text{var}(x_s) = 1$  as that is given by definition. Therefor the ACF is finally given by

$$= \sigma_s \sigma_t \rho(x_s, x_t) = \sigma_s \sigma_t \rho_x(s, t)$$

## 2 Assignment 18

**Exercise:** For each of the following ARMA models, find the roots of the AR and MA polynomials, identify the values of p and q for which they are ARMA(p,q) (be careful of parameter redundancy), determine whether they are causal, and determine whether they are invertible. In each case,  $w_t \sim \mathcal{N}(0, 1)$ .

c)  $x_t - 3x_{t-1} = w_t + 2w_{t-1} - 8w_{t-2}$

d)  $x_t - 2x_{t-1} + 2x_{t-2} = w_t - \frac{8}{9}w_{t-1}$

e)  $x_t - 4x_{t-2} = w_t - w_{t-1} + 0.5w_{t-2}$

f)  $x_t - \frac{9}{4}x_{t-1} - \frac{9}{4}x_{t-2} = w_t$

## 2.1 c) Solution

### 2.1.1 Roots of AR and MA Polynomials

Applying the autoregressive operator for  $\phi$  and  $\theta$  gives us:

AR:  $\phi(B) = 1 - 3B$ , thus  $Z_\phi = (1, -3)$

MA:  $\theta(B) = 1 + 2B - 8B^2$ , thus  $Z_\theta = (1, 2, -8)$

For both parts we calculate the zero points. We start with AR.

$$p_\phi(z) = 1 - 3z$$

setting  $p_\phi(z) = 0$

$$1 - 3z = 0$$

From that we get  $z_{\theta 1} = \frac{1}{3}$ . Now we do the MA part.

$$p_\theta(z) = 1 + 2z - 8z^2$$

setting  $p_\theta(z) = 0$

$$0 = 1 + 2z - 8z^2$$

First we divide by  $-8$  and swap sides:

$$z^2 - \frac{1}{4}z - \frac{1}{8} = 0$$

We add  $(\frac{1}{4}/2)^2$  on both sides

$$z^2 - \frac{1}{4}z - \frac{1}{8} + \frac{1}{64} - \frac{8}{64} = \frac{1}{64}$$

Next we create the polynomial form and add  $\frac{8}{64}$  on both sides

$$(z - \frac{1}{8})^2 = \frac{9}{64}$$

Taking the square root

$$z - \frac{1}{8} = \pm \frac{3}{8}$$

Thus it follows that  $z_{\theta 1} = \frac{1}{2}$  and  $z_{\theta 2} = -\frac{1}{4}$ .

### 2.1.2 Finding p and q

To check for redundancy, we have to identify any common roots. As  $z_\phi \cap z_\theta = \{\}$ ,  $\phi(B)$  and  $\theta(B)$  share no common roots and no redundancy is given.

Therefore p and q are simply given by the highest order terms, so the model is ARMA( $p = 1, q = 2$ ).

### 2.1.3 Causality and Invertibility

**Causality:** As  $\forall Z_\theta : |Z_\theta| > 1$  is **not** given, the model is not causal.

**Invertibility:** As  $\forall Z_\phi : |Z_\phi| > 1$  is **not** given, the model is not invertible.

## 2.2 d) Solution

### 2.2.1 Roots of AR and MA Polynomials

Applying the autoregressive operator for  $\phi$  and  $\theta$  gives us:

AR:  $\phi(B) = 1 - 2 + 2B^2$ , thus  $Z_\phi = (1, -2, 2)$

MA:  $\theta(B) = 1 - \frac{8}{9}B$ , thus  $Z_\theta = (1, -\frac{8}{9})$

For both parts we calculate the zero points. We start with AR.

$$p_\phi(z) = 1 - 2z + 2z^2$$

Setting  $p_\phi(z) = 0$  and swapping sides

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

Divide by 2 and order

$$z^2 - z + \frac{1}{2} = 0$$

We add  $(1/2)^2$  on both sides

$$z^2 - z + \frac{1}{2} + \frac{1}{4} = \frac{1}{4}$$

Next we create the polynomial form and subtract  $\frac{1}{2}$  on both sides:

$$(z - \frac{1}{2})^2 = -\frac{1}{4}$$

Taking the square root:

$$z - \frac{1}{2} = \pm \frac{1}{2}i$$

From that we get  $z_{\theta 1} = \frac{1}{2} + \frac{1}{2}i$  and  $z_{\theta 2} = \frac{1}{2} - \frac{1}{2}i$ . Now we do the MA part.

$$p_\theta(z) = 1 - \frac{8}{9}z$$

setting  $p_\theta(z) = 0$

$$0 = 1 - \frac{8}{9}z$$

So  $z_{\theta 1} = \frac{8}{9}$

### 2.2.2 Finding p and q

To check for redundancy, we have to identify any common roots. As  $z_\phi \cap z_\theta = \{\}$ ,  $\phi(B)$  and  $\theta(B)$  share no common roots and no redundancy is given.

Therefore p and q are simply given by the highest order terms, so the model is ARMA( $p = 1, q = 2$ ).

### 2.2.3 Causality and Invertibility

**Causality:** As  $\forall Z_\theta : |Z_\theta| > 1$  is **not** given, the model is not causal.

**Invertibility:** As  $\forall Z_\phi : |Z_\phi| > 1$  is **not** given, the model is not invertible.

## 2.3 e) Solution

### 2.3.1 Roots of AR and MA Polynomials

Applying the autoregressive operator for  $\phi$  and  $\theta$  gives us:

AR:  $\phi(B) = 1 - 4B^2$ , thus  $Z_\phi = (1, 0, -4)$

MA:  $\theta(B) = 1 - B + \frac{1}{2}B^2$ , thus  $Z_\theta = (1, -1, -\frac{1}{2})$

For both parts we calculate the zero points. We start with AR.

$$p_\phi(z) = 1 - 4Z^2$$

setting  $p_\phi(z) = 0$

$$1 - 4Z^2 = 0$$

dividing by 4 and putting  $Z$  on one side

$$Z^2 = \frac{1}{4}$$

From that we get  $z_{\theta 1} = \frac{1}{2}$  and  $z_{\theta 2} = -\frac{1}{2}$ . Now we do the MA part.

$$p_\theta(z) = 1 - z + \frac{1}{2}z^2$$

setting  $p_\theta(z) = 0$

$$0 = 1 - z + \frac{1}{2}z^2$$

First we divide by  $\frac{1}{2}$  and swap sides:

$$z^2 - 2z + 2 = 0$$

We add  $(\frac{2}{2}/2)^2$  on both sides

$$z^2 - 2z + 1 + 2 = 1$$

Next we create the polynomial form and subtract  $-2$  on both sides

$$(z - 1)^2 = -1$$

Taking the square root

$$z - 1 = \pm i$$

Thus it follows that  $z_{\theta 1} = 1 + 1i$  and  $z_{\theta 2} = 1 - 1i$ .

### 2.3.2 Finding p and q

To check for redundancy, we have to identify any common roots. As  $z_\phi \cap z_\theta = \{\}$ ,  $\phi(B)$  and  $\theta(B)$  share no common roots and no redundancy is given.

Therefore p and q are simply given by the highest order terms, so the model is ARMA( $p = 2, q = 2$ ).

### 2.3.3 Causality and Invertibility

**Causality:** As  $\forall Z_\theta : |Z_\theta| > 1$  is **not** given, the model is not causal.

**Invertibility:** As  $\forall Z_\phi : |Z_\phi| > 1$  is given, the model is invertible. The length of all elements of the set  $Z_\phi$  is given by  $|Z_\phi| = \sqrt{1^2 + (\pm 1)^2} \approx 1.414214$ .

## 2.4 f) Solution

### 2.4.1 Roots of AR and MA Polynomials

Applying the autoregressive operator for  $\phi$  and  $\theta$  gives us:

AR:  $\phi(B) = 1 - \frac{9}{4}B - \frac{9}{4}B^2$ , thus  $Z_\phi = (1, -\frac{9}{4}, -\frac{9}{4})$

MA:  $\theta(B) = 1$ , thus  $Z_\theta = (1)$

For both parts we calculate the zero points. We start with AR.

$$p_\phi(z) = 1 - \frac{9}{4}z - \frac{9}{4}z^2$$

setting  $p_\phi(z) = 0$

$$1 - \frac{9}{4}z - \frac{9}{4}z^2 = 0$$

dividing by  $\frac{9}{4}$  and ordering

$$z^2 + z - \frac{4}{9} = 0$$

Adding  $(\frac{1}{2})^2$  on both sides

$$z^2 + z + \frac{1}{4} - \frac{4}{9} = \frac{1}{4}$$

Forming the polynomial and adding  $\frac{4}{9}$

$$(z + \frac{1}{2})^2 = \frac{25}{36}$$

Taking the square root

$$z + \frac{1}{2} = \pm \frac{5}{6}$$

From that we get  $z_{\theta 1} = \frac{1}{3}$  and  $z_{\theta 2} = -\frac{4}{3}$ . Now we do the MA part.

$$p_{\theta}(z) = 1$$

As setting this to 0 is an inequality, this ARMA model does not have an MA part.

#### 2.4.2 Finding p and q

To check for redundancy, we have to identify any common roots. As  $\theta(B)$  is not given, it does not really make sense to ask this question. There are no common roots.

Therefore p and q are simply given by the highest order terms, so the model is ARMA( $p = 2, q = 0$ ) which can also be expressed as AR(2).

#### 2.4.3 Causality and Invertibility

**Causality:** As  $\forall Z_{\theta} : |Z_{\theta}| > 1$  is **not** given, the model is not causal.

**Invertibility:** As the model does not have an MA part, it does not make sense to talk about invertibility as an MA( $\infty$ ) representation cannot exist for something that is non-existent. Also a (weak) stationary property does not make sense in this case.