

## Chapter 4 Non-stationary TS Models

### 4.1. Nonstationarity in Mean

#### 4.1.1. Deterministic Trend Models

Let  $\{x_t\}$  be a sequence of stationary time series.

$Z_t$  is called a deterministic trend model, if

$$Z_t = \alpha_0 + \alpha_1 t + x_t, \quad \alpha_1 \neq 0.$$

$Z_t$  is not stationary.

**Feature:** If  $Z_t$  is a deterministic trend model, then after transformed:

$$\dot{Z}_t = Z_t - \alpha_0 - \alpha_1 t.$$

Then  $\dot{Z}_t = x_t$  and hence  $\dot{Z}_t$  is stationary.

Other Deterministic Trend Models:

$$Z_t = \alpha_0 + \alpha_1 t + \alpha_2 t^2 + x_t,$$

$$Z_t = \gamma_0 + \gamma_1 \cos(t\omega + \theta) + x_t,$$

$$Z_t = \gamma_0 + \sum_{j=1}^m (\alpha_j \cos t\omega_j + \beta \sin(t\omega_j)) + x_t.$$

### 4.1.2. Stochastic Trend Models

Let  $\{x_t\}$  be a sequence of stationary time series.

$Z_t$  is called a Stochastic Trend Model, if

$$Z_t = Z_{t-1} + x_t, \quad \text{or} \quad (1 - B)Z_t = x_t$$

$Z_t$  is not stationary.

In particular, when  $x_t = a_t$ ,

$Z_t = Z_{t-1} + a_t$  is a random walk.

**Feature:** If  $Z_t$  is a Stochastic Trend Model, then after differenced:

$$\dot{Z}_t = Z_t - Z_{t-1}$$

$\dot{Z}_t$  is stationary.

General Stochastic Trend Models:

$$(1 - B)^d Z_t = x_t, \quad d \geq 1.$$

Let  $\dot{Z}_t = (1 - B)^d Z_t$ . Then  $\dot{Z}_t$  is stationary.

## 4.2. Autoregressive Integrated Moving-average Model

### 4.2.1. The General ARIMA Model

Let  $Z_t$  is a General Stochastic Trend Model:

$$(1 - B)^d Z_t = x_t, \quad d \geq 1.$$

If  $x_t$  is a weakly stationary and invertible ARMA model,

$$\phi_p(B)x_t = \theta_q(B)a_t,$$

where  $\phi_p(B)$  and  $\theta_q(B)$  have no common roots. Then:

$$\phi_p(B)(1 - B)^d Z_t = \theta_q(B)a_t,$$

$Z_t$  is called ARIMA( $p, d, q$ ) model.(?)

If  $x_t$  is the following ARMA model,

$$\phi_p(B)x_t = \theta_0 + \theta_q(B)a_t,$$

then,

$$\phi_p(B)(1 - B)^d Z_t = \theta_0 + \theta_q(B)a_t,$$

$Z_t$  is called ARIMA( $p, d, q$ ) model.

### 4.2.2. The Random Walk model

ARIMA(0, 1, 0) model:

$$(1 - B)Z_t = a_t \text{ or}$$

$$Z_t = Z_{t-1} + a_t \text{ (random walk)}$$

$$Z_t = \theta_0 + Z_{t-1} + a_t$$

— called the random walk with drift.

**Example 4.1:** Simulated 100 values from

$$(1 - B)Z_t = a_t,$$

and

$$(1 - B)Z_t = 4 + a_t,$$

Show the sample ACF and PACF.

### 4.2.3. The ARIMA(0, 1, 1) or IMA(1, 1) Model

ARIMA(0, 1, 1) model:

$$(1 - B)Z_t = (1 - \theta B)a_t.$$

or

$$Z_t = Z_{t-1} - \theta a_{t-1} + a_t,$$

where  $|\theta| < 1$ .

### Expansion:

$$a_t = Z_t - \alpha \sum_{j=1}^{\infty} (1 - \alpha)^{j-1} Z_{t-j}.$$

or

$$Z_t = \alpha \sum_{j=1}^{\infty} (1 - \alpha)^{j-1} Z_{t-j} + a_t,$$

where  $\alpha = 1 - \theta$ .

**Example 4.2:** Simulated 100 values from three models:

ARIMA(1, 1, 0) model:

$$(1 - 0.8B)(1 - B)Z_t = a_t,$$

ARIMA(0, 1, 1) model:

$$(1 - B)Z_t = (1 - 0.75B)a_t,$$

ARIMA(1, 1, 1) model:

$$(1 - 0.9B)(1 - B)Z_t = (1 - 0.5B)a_t,$$

a. Show the sample ACF and PACF.

b. Let  $W_t = (1 - B)Z_t$ . Show the sample ACF and PACF of  $W_t$ .