Taylor R. Brown PhD

Department of Statistics, University of Virginia

Spring 2020

# Readings for Unit 5

Textbook chapter 2.2 (pages 54 to 62).

#### Last Unit

Linear Regression Overview

#### This Unit

- Detrending with Linear Regression
- ② Differencing for Stationarity
- Backshift Operator

#### Motivation

With time series data, we need to account for the dependence between the values in the series. Stationarity for a time series enables us to measure the dependence, since the dependence structure is regular and does not change over time. When we do not have stationarity, we need methods to reduce the effects of nonstationarity so that stationary properties can be obtained.

# Detrending with Linear Regression

Recall that in time series analysis, we prefer to work with stationary processes, to better estimate autocorrelation and other quantities of interest. If our process has a linear trend, we could use linear regression to remove the trend ("detrend").

#### Detrending with Linear Regression

Consider the following model:

$$x_t = \mu_t + y_t \tag{1}$$

where  $y_t$  is a **zero mean stationary process**, e.g. MA(2), AR(1), white noise, etc., and  $\mu_t$  is a **deterministic trend**, e.g.  $\mu_t = \beta_0 + \beta_1 t$ .

We can view  $x_t$  as having stationary behavior around a trend. A strong trend,  $\mu_t$ , can obscure the behavior of the stationary process,  $y_t$ . Hence, we may want to remove the trend as a first step in exploratory analysis of such time series, especially if the goal is to understand the behavior of  $y_t$ .

## Detrending with Linear Regression

#### The steps involved are

- Obtain an **estimate** of the trend component,  $\hat{\mu_t}$ , e.g. via OLS.
- Work with the residuals  $e_t = x_t \hat{\mu_t}$ .

Recall in this earlier example that a regression of marriages in the Church of England against time should include a quadratic term, so the trend is

$$\mu_t = \beta_1 + \beta_2 t + \beta_3 t^2.$$

Using OLS regression, we obtained

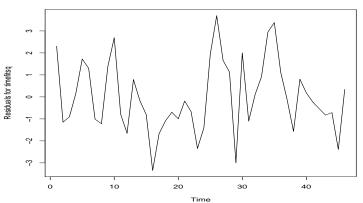
$$\hat{\mu_t} = -15130 + 16.43t - 0.00444t^2.$$

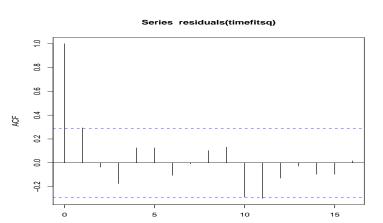
The detrended series is obtained by subtracting  $\hat{\mu_t}$  from the observed values  $x_t$ 

$$e_t = x_t + 15130 - 16.43t + 0.00444t^2$$
.

Next, we observe the plot of  $e_t$  and its ACF.

#### Time Series Plot of Residuals from timefitsq





Lag

### Summary of Detrending

We find a good point estimate of  $\mu_t$  and then look at the residuals

$$e_t = x_t - \hat{\mu}_t.$$

One thing we could do is to "detrend" with linear regression, e.g. use OLS regression to obtain  $\hat{\mu_t} = \hat{\beta_1} + \hat{\beta_2}t$ . With the residuals, based on the structure of their ACF, we have an idea on the model that may reasonably describe the stationary process  $y_t$ .

# Summary of Detrending

**Question:** What is an implication if the residuals after detrending have a pattern / trend?

2 Differencing for Stationarity

## Differencing

Let  $\nabla$  denote the differencing operation. Applying  $\nabla$  to  $x_t$  results in

$$\nabla x_t = x_t - x_{t-1},\tag{2}$$

which is also called the first difference.

Suppose we assume the trend component  $\mu_t$  follows a simple linear regression, then

$$\nabla x_t = \nabla(\beta_0 + \beta_1 t + y_t)$$

$$=$$

$$=$$

Notice that the detrending may give us a more accurate representation of  $y_t$ , whereas differencing completely **removes**  $\beta_0$ and turns  $\beta_1$  into the **mean** of the series  $\{\nabla x_t\}$ .

However, differencing changes  $y_t$  and often introduces additional dependency. Consider regression with MA(1).

$$x_t = \beta_0 + \beta_1 t + w_t + \theta_1 w_{t-1}$$

Differencing the above expression, we have

$$\nabla x_t = \nabla(\beta_0) + \nabla(\beta_1 t) + \nabla w_t + \theta_1 \nabla w_{t-1}$$

$$=$$

$$=$$

which becomes a mean value  $\beta_1$  plus MA(2).  $\{\nabla x_t\}$  is stationary.

# Differencing

**Question:** Suppose instead of a trend,  $\mu_t$  follows a random walk with drift process, derive  $\nabla x_t$ .

- An advantage of differencing over detrending is that fewer parameters are estimated after the differencing operation.
- A disadvantage of differencing is that it often makes an estimate of the stationary process y<sub>t</sub> more difficult.



# Differencing Vs Detrending

- If estimating y<sub>t</sub> is the goal, then **detrending** may be more appropriate.
- If the goal is to make the data stationary without thinking about what  $\mu_t$  is, then **differencing** may be more appropriate.

However, it's dangerous to generalize. You should consider several models, and derive the implications of each tool.

- Backshift Operator

The backshift operator is defined as

$$Bx_t = x_{t-1},$$

which can be extended to powers

$$B^k x_t = x_{t-k}. (3)$$

Thus, the first difference (2) can be written as

**Question:** Denote the second difference,  $\nabla^2 x_t$  in terms of the backshift operator.

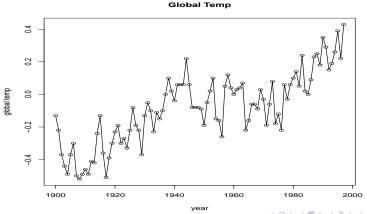
In general, for all positive integer d,

$$\nabla^d = (1 - B)^d. \tag{4}$$

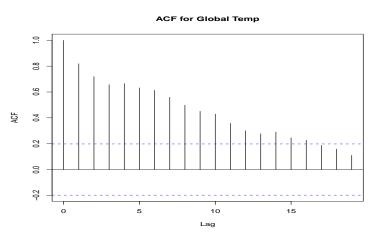
We showed earlier that the first difference eliminates a linear trend. This is an example of a **linear filter** applied to eliminate a linear trend. Next, we show that the second difference eliminates a quadratic trend.

- Detrending with Linear Regression
- 2 Differencing for Stationarity
- Backshift Operator
- Worked Example

Let's look at average global temperatures from 1900 to 1997. The data are measured in deviation in degrees centigrade from the 1961-1990 average. You can find the data (global.txt) in Collab.



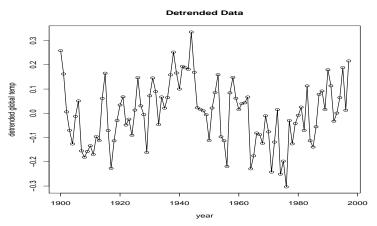




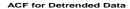
Question: What does this indicate about stationarity?

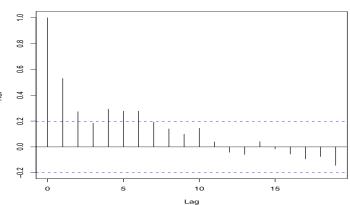


Next, we look at detrended data. These are the residuals obtained from regressing  $x_t$  on t

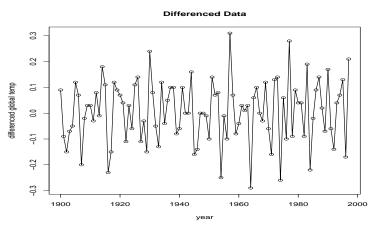




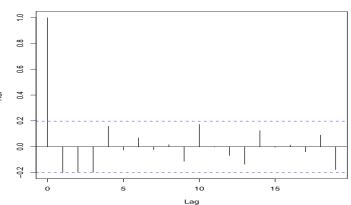




Finally, we look at differenced data.



#### ACF for Differenced Data



**Question:** Based on the plots and ACFs, what can we say about the behavior of global temperature?

**Question:** How would you estimate the parameters of the process you are considering for the behavior of global temperature?