



# Time Series

☰ Status	Done
🔗 Files	
🔗 Link	
☰ Type	All Around
☰ Course	Data Analysis

## Weak stationarity of a series

Stationarity statistical tests

Correlogram

! This makes quite a small portion of the exam, if any, maybe one or two multiple choice questions, so not really worth revising, but its very interesting for real world analysis of time series

- Temporal = related to time

## Weak stationarity of a series

- **Weak stationarity of a (time) series** is the problem that arrises when we try to perform summary/inferential/descriptive statistics of a (time) series
  - We usually assume that there is no correlation between the variables, but if there is a correlation then the usual approaches are not correct
  - We need to remove this correlation before continuing so the time series “doesn’t change its probabilistic character” over time

- Stationary → does not have these seasonal effects  
Non Stationary → has these seasonal effects
- Basically its like you wanna study the time series but you wanna look at overall trends and generalise the daily/seasonal/whatever fluctuations that repeat every x period because these are distractions and they are not important in the overall trend
- **Formally**, a time series  $X_t$  is said to be weakly stationary, if :

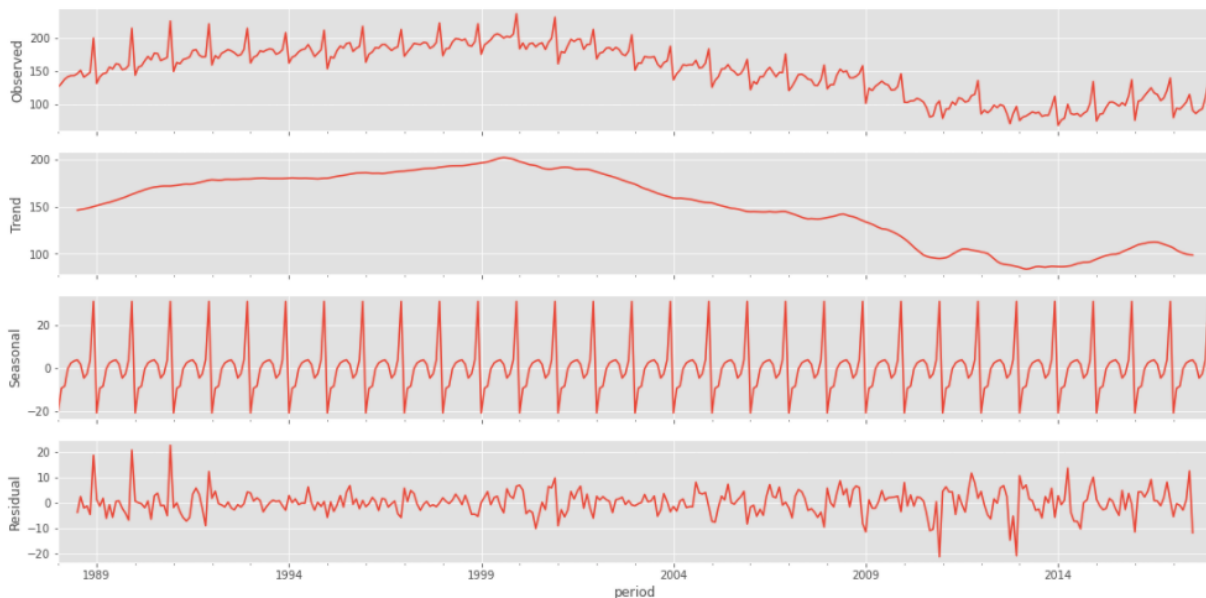
- $E[X_t] = \mu$ : all  $X$  have identical expected mean
- $Var(X_t) = \sigma^2$ : all  $X_t$  have identical variance
- $Cov(X_t, X_{t+h}) = \gamma_h$ : autocovariance depends only on lag  $h$

- How can we deal with this? We can decompose a time series as follows

$$x_t = m_t + s_t + r_t$$

= trend + seasonal effect + stationary remainder

- Example, we can see some repeating patterns in this graph with the raw series



## Stationarity statistical tests

- Augmented Dickey-Fuller (ADF) unit root test
- Kwiatkowski-Phillips-Schmidt-Shin (KPSS) test for stationarity

## Correlogram

- Correlogram is a visual way to show serial correlation in data that changes over time
- The intuition behind it is simple, if its a purely random series,  $x_n$  can not help in predicting  $x_{n+1}$
- We can put this in a value called **Sample Autocorrelation Coefficient**
  - The closer to 0 is it, the more random the source of the data is, especially if n is large

$$r_1 = \frac{\sum_{t=1}^{n-1} (x_t - \bar{x})(x_{t+1} - \bar{x})}{\sum_{t=1}^n (x_t - \bar{x})^2}$$