

Time Series Components

Introduction

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What is a Time Series?

- A time series is a sequence of numerical data points in successive order, usually occurring in uniform intervals.
- Time series analysis comprises methods for analyzing time series data in order to extract meaningful statistics and other characteristics of the data.
- Time series forecasting is the use of a model to predict future values based on previously observed values.

Courtesy: Wikipedia



- Trend: Indicates a long term increase or decrease in the data. It may be linear or non-linear.
- Seasonal: Seasonality is a pattern observed with regular intervals of time. e.g. Sale of woolen clothes increases in winter and is relatively low in other seasons.
- Cyclic: Data exhibits rise and fall not in regular time intervals. e.g. Recession and Boom
- Random: This is an error component. Also called irregular component.



Classical Decomposition

- There are two types of classical decompositions:
 - Additive
 - Multiplicative
- We assume here that the seasonal component is constant from year to year.
- Suppose that we have m seasonal periods.
 Then there are m seasonal values which are called seasonal indices.



Notations

- y_t : Value in time series at time t
- \widehat{T}_t : Trend-cycle component (Moving Average) calculated for time t
- \widehat{S}_t : Seasonal Index for time t



Additive Decomposition

- 1. If m is even number, then centered MA is calculated otherwise non-centered MA is calculated.
- 2. Calculate the de-trended series, y_t - \widehat{T}_t
- 3. For estimating the seasonal component for each month, a simple average is calculated for detrended values for that particular month. It is denoted by \widehat{S}_t .
- 4. The random component is calculated by subtracting seasonal and trend-cycle components. $\widehat{E_t} = \widehat{y_t} \widehat{T_t} \widehat{S_t}$



Example

```
In [55]: from statsmodels.tsa.seasonal import seasonal_decompose
     ...: from matplotlib import pyplot
...: series = df['Milk']
     ...: result = seasonal_decompose(series, model='additive',freq=12)
In [56]: result.plot()
     ...: pyplot.show()
 Observed
   800
   600
   800
   700
 Seasonal
 Residual
                        50
                                75
                                               125
                                       100
                                                       150
                                                               175
```

Multiplicative Decomposition

- 1. If m is even number, then centered MA is calculated otherwise non-centered MA is calculated.
- 2. Calculate the de-trended series, y_t/\widehat{T}_t
- 3. For estimating the seasonal component for each month, a simple average is calculated for detrended values for that particular month. It is denoted by \widehat{S}_t .
- 4. The random component is calculated by subtracting seasonal and trend-cycle components. $\widehat{E_t} = \widehat{y_t}/(\widehat{T_t}\widehat{S_t})$



Example

```
In [57]: result = seasonal_decompose(series, model='multiplicative',freq=12)
In [58]: result.plot()
    ...: pyplot.show()
  Observed
     800
     600
     800
     700
   Seasonal
     1.0
   1.025
   1.000
   0.975
                                            100
                                     75
                                                     125
                            50
                                                             150
                                                                      175
```