# Forecasting Inflation Rates





# Seasonal Models



# Seasonality: A recurring pattern over a fixed time interval

- Frequency can be of different length
- Monthly data = Frequency of 12

Dataset: Monthly US inflation rates

Modeling seasonal time series data

#### Standard models:

- Seasonal decomposition
- Seasonal ARIMA
- Holt-Winters exponential smoothing

Visualizing seasonal data



# Seasonality in Time Series





Seasonal time series

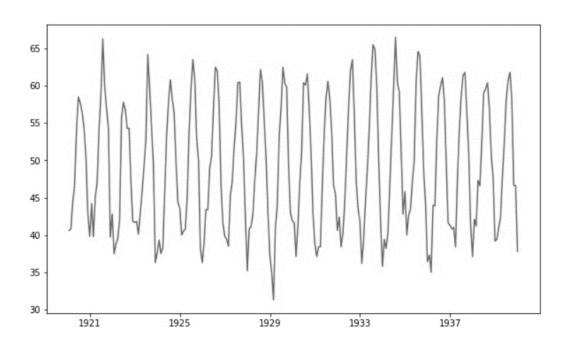
Adjusted analytical approach

Frequency to frame the recurring pattern

- Function: ts(frequency = )



# Temperature Measurements



54 - 52 - 50 - 48 - 46 - 44 - 42 - 1929 1934 1939

**Time Series with Seasonality** 

Recurring pattern over a given time interval

**Time Series without Seasonality** 

The recurring pattern disappears when seasonality is extracted



## Multiple Seasonality





#### **Daily Recurring Patterns**

More orders placed in the evening than in the morning

#### **Weekly Recurring Patterns**

More orders placed in the weekend than during the week



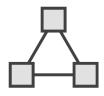
### Standard Models for Seasonal Datasets



Seasonal ARIMA (SARIMA)



Holt-Winters exponential smoothing



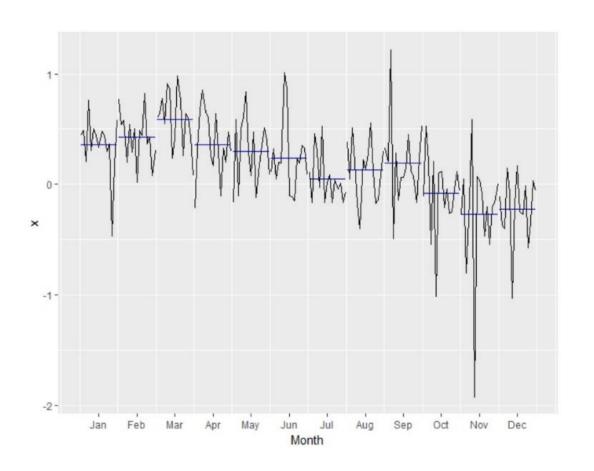
Seasonal decomposition: Trend, seasonality, and residuals

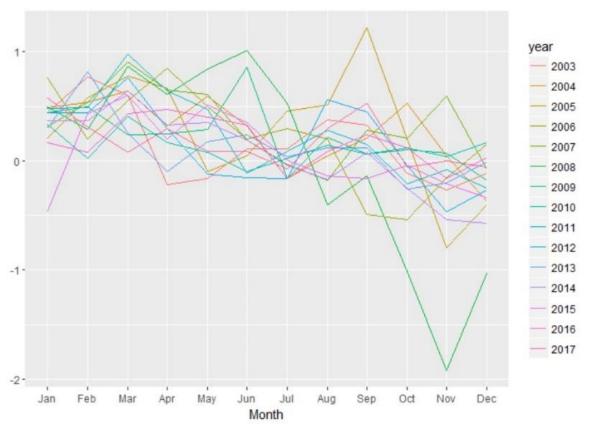


# Visualizing time series data helps to identify patterns like trend or seasonality



# Visualizing Seasonal Time Series





**Month plot** 

**Season plot** 



# The US Inflation Dataset





#### Inflation rates

A measure of change in purchasing power

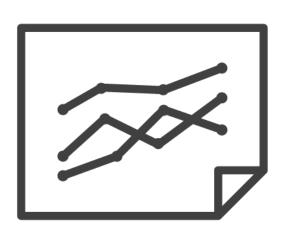
#### Affects investment opportunities

- Stocks, property, precious metals, oil

#### The era of cheap money

- Currency depreciation





#### Monthly US inflation rates (2003-2017)

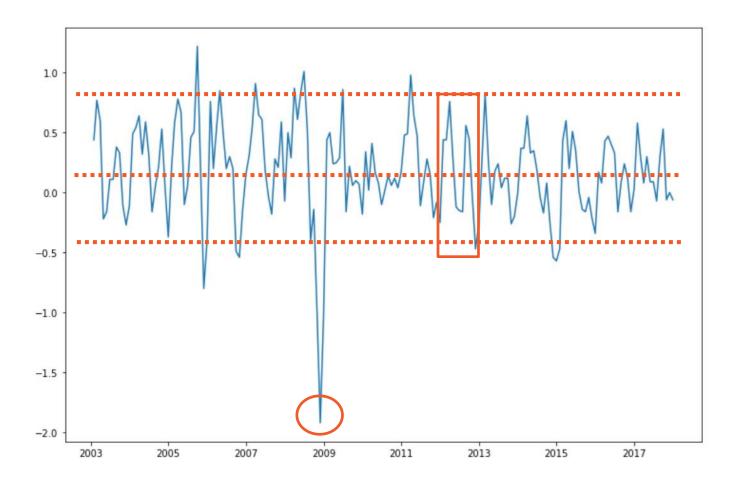
Source: statbureau.org

#### Month-on-month inflation rate

- More intuitive than the year-on-year change
- The difference between two consecutive months
- Monthly rates can be negative



Financial crash in 2008
Seasonal dataset
12 observations/cycle
Negative values
Constant variance





# Importing the Data into R





#### Pasting the data into R

Avoid row IDs and headers

- Time stamp will be generated in R

Chronologically ordered vector

Pre-format the data in Excel



# Tools and functions for time series analysis require the data to be in 'ts' format



```
mydata = scan()
ts(mydata, start = c(2003, 3), frequency = 12)
```

# Specifying an Offset in the Timestamp Use the 'start' argument with two integers

- Start year
- Start month



# Seasonal Decomposition



# Seasonal Decomposition

Dividing the data into trend, seasonality, and remainder

Additive and multiplicative methods

Simple, easy to use approach

Possible drawbacks



## Drawbacks of Seasonal Decomposition



First observations are NA



Slow to catch changes



Constant seasonal component



#### Alternative Methods

X 11

**SEATS** 

STL

Model values for all observations

Adjusted seasonal component



# STL Decomposition

Seasonal and trend decomposition with loess

Robust against outliers

Additive model is preferred

Seasonal and trend cycles may adjust

Argument s.window

- Number of required seasonal cycles
- $x \ge 7; x \nmid 2$



### Forecasting with STL Decomposition

Feeding an 'STL' object into the forecast() function

Feeding a 'ts' object into the stlf() function



## Seasonal ARIMA Model



#### ARIMA Parameter Selection

#### **Manual Method**

Parameter identification via differencing and data visualization (ACF, PACF)

#### **Automated Method**

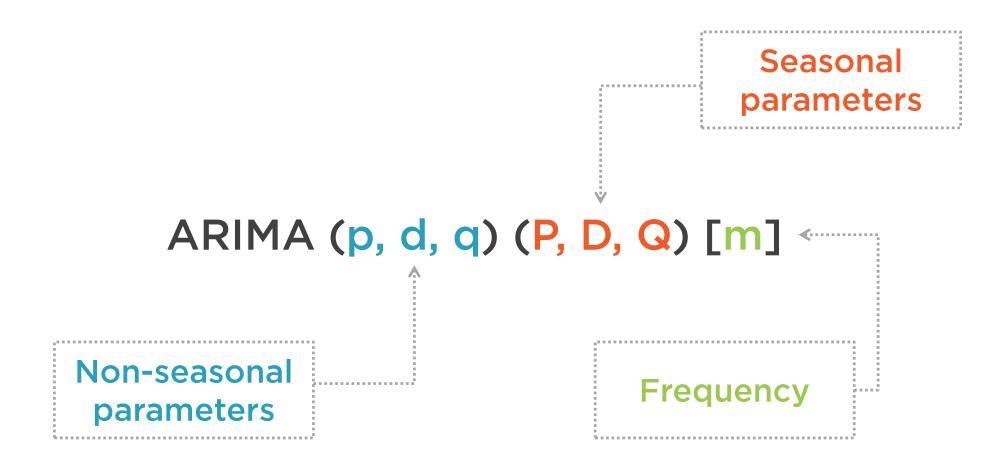
R estimates the parameters automatically by using the auto.arima() function



# Seasonal ARIMA models have two sets of parameters



#### Seasonal ARIMA Model Parameters





# Only models of the same class can be compared with the information criterion



### Comparison of Models Improves the Analysis

Seasonal ARIMA model Exponential smoothing model

**ETS** model



# **Exponential Smoothing Model**



# Exponential Smoothing

#### Two methods with the 'forecast' library

- Function: ets()
- Function: hw() Holt-Winters exponential smoothing

Comparing the model and forecast to previous models

Selects a seasonal model automatically



# Month Plot

Extracts patterns by plotting the seasons (months, days) of a cycle (years, weeks, or other given frequency) in chronological order.

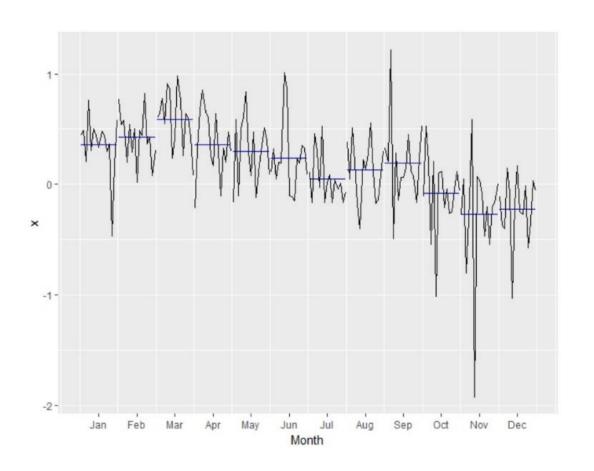


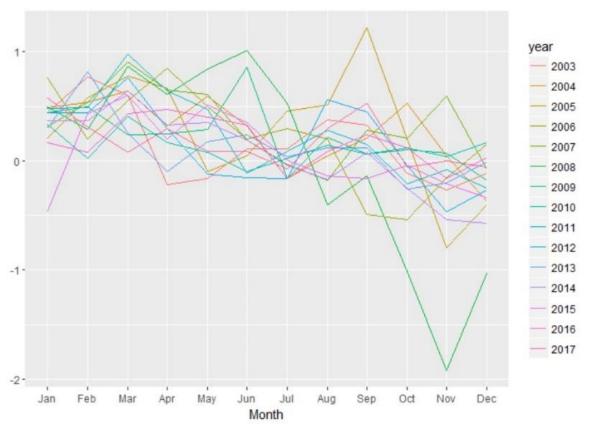
# Season Plot

Extracts and emphasizes patterns by plotting seasonal cycles (years, weeks, or other given frequency) over one another.



# Visualizing Seasonal Time Series





**Month plot** 

**Season plot** 



# Seasonal Models



#### Working with a seasonal time series

- Identifying and capturing recurring patterns
- The frequency influences the availability of models

# Getting a first impression with seasonal decomposition

- Functions: decompose(), stl()

#### Seasonal ARIMA model (SARIMA)

Function: auto.arima()

#### Holt-Winters exponential smoothing

- Functions: ets(), hw()

