

# Time Series Analysis

## Basics of Time Series Analysis

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Overview

# About This Lesson



# Course Road Map

**Pre-requisite:**  
**Basic Statistical**  
**Modeling &**  
**Inference**

## Basic Time Series Modeling

- Trend
- Seasonality
- Serial Dependence
- Prediction

## Univariate Analysis

Conditional Mean Model:  
**ARIMA**

Conditional Variance Model:  
**GARCH**

Joint Model:  
**ARIMA-GARCH**

## Multivariate Analysis

- Correlation between & within time series
- Multivariate AR model (VAR)

# Course Road Map: Pre-requisite

**Pre-requisite:  
Basic Statistical  
Modeling &  
Inference**

- *Distribution of a Random Variable*
- *Statistical Estimation*
- *Statistical Inference*
- *Regression Analysis*

# Model Estimation in Time Series Analysis

## **Approaches:**

1. Method of Moments (MOM)
  - Durbin-Levinson Algorithm (Linear Prediction)
  - Yule-Walker Algorithm (AR model estimation)
  - Innovation Algorithm (Linear Prediction & ARMA model estimation)
2. Maximum Likelihood Estimation
  - Most common approach in time series model estimation
  - ARMA, GARCH, VAR

## ***Why important?***

Statistical properties and statistical inference of the estimated model are different depending on the estimation approach.

# Statistical Inference in Time Series Analysis

## **Approaches:**

### 1. Confidence & Prediction Intervals

- Evaluate statistical significance of model coefficients
- Provide prediction bands for time series forecasts

### 2. Hypothesis Testing

- Evaluate statistical significance of model coefficients
- Evaluate whether data follows a prescribed distribution (e.g. normal)
- Evaluate model goodness of fit

## ***Why important?***

Statistical inference is used in model interpretation and evaluation.

# Regression Analysis in Time Series Analysis

## General Regression model:

Data:  $\{y_1, (x_{1,1}, \dots, x_{1,p})\}, \dots, \{y_n, (x_{n,1}, \dots, x_{n,p})\}$

Model:  $Y_i = \beta_0 + \beta_1 x_{i,1} + \beta_2 x_{i,2} + \dots + \beta_p x_{i,p} + \varepsilon_i, \quad i = 1, \dots, n$

## Time Series Regression models:

Data:  $\{y_1, t_1\}, \dots, \{y_n, t_n\}$

Model:

- Basic model decomposition:  $Y_t = m_t + s_t + X_t$
- ARMA, GARCH, VAR

## ***Why important?***

Regression modeling is at the basis of all models discussed in this course.

# Course Road Map: Basic Concepts

## Basic Time Series Modeling

- Trend
- Seasonality
- Serial Dependence
- Prediction



- Time Series Definitions
- Basic Decomposition: Trend & Seasonality analysis
- Stationarity: The basic concept for all time series models
- Prediction: Best Linear Predictor



# Basics of Time Series Analysis

## **What will this module cover?**

1. Basic time series modeling
  - Definition of a time series
  - Simple decomposition of a time series into trend & seasonality
  - Approaches for estimating trend and seasonality
2. Fundamental time series concepts
  - Stationarity
  - Linear Process
  - Prediction

## ***Why important?***

The concepts introduced in this module are at the basis of all time series models introduced in this course.

# Data Examples using R Statistical Software

## 1. Data Examples

- Average monthly temperature records starting in 1879 until 2016
- Emergency Department Volume
- Bitcoin Price

## 2. R Statistical Software

- Visual analytics
- Evaluating properties and characteristics of a time series
- Fitting linear regression models for seasonality
- Fitting parametric and non-parametric trends

### ***Why important?***

Fundamentals of time series modeling are best understood by illustrating them using data examples.

# Summary

