

## DAY 6

### Time Series

#### What is a Time Series:

- A dependent variable that is observed at different time intervals usually equal intervals
- Examples
  - U.S.'s GDP over time
  - Customers Reviews during March 2019
  - GPAs of all sophomore students during the summer of '18.
  - Sales and revenue
  - Height of a boy over time.
- Time Series Plots
  - Possible Patterns
    - Horizontal Patterns
      - Where the data seems to fluctuate randomly around a mean
      - An adult's height. (Constant but only may change due to reading error / posture)
    - Trend Pattern
      - When random fluctuation occur, but has a constant trend about it.
      - Eg. Temperature at March (It keep increasing)
      - Types of Trends
        - Uptrend
        - Downtrend
        - Stationary Trend
    - Seasonal Pattern
      - Deals with **seasonality** - short term periodic relationship
      - Identifying a common pattern in the data
      - Seasonal Patterns can exist without trends
        - Eg. Tides, and how far they recede throughout a day.
      - Eg. Low-Med-High is a constant shape shift of the data throughout the months.
      - Common with monthly data
      - Employment graph, Patio Furniture Sales graph
    - Cyclical Patterns
      - Has high runs and low runs (ups and downs)
      - Eg. Employment, Gasoline Prices
    - Irregular Patterns
      - Unobservable, but time dependant
      - E.g. Sales of Medicine due to floods
  - How to Plot
    - Take a variable that you care about
    - Take a series of Y's that are influence by time
  - Terms used in Plotting

- **Seasonality**
  - Short term periodic relationship
- **Cycles**
  - Long term periodic relationship
- **Irregular**
  - Unobservable, but time dependant
  - Eg. Oil spills

#### **What can one achieve with Time-Series:**

- Business Forecasting
- Understanding Past Behavior
- Plan the future
- Evaluating a current accomplishment.

#### **When do we not use Time-Series:**

- When values are constant.
  - Number of coffee sales are constant across months
- Values are in the form of functions
  - No point as you can use a function to calculate.

#### **What is Stationarity:**

- Dataset has a constant mean, one that doesn't vary across time
- Dataset has a constant variance, one that doesn't vary across time
- Autocovariance doesn't depend on time
  - Correlation between values of the different time frames shouldn't be there.
  - E.g.  $t-1$ 's values and  $t$ 's values don't have any correlation between them
- When a dataset has these 3 points it is said to be stationary, and then can time-series can be used upon it.

#### **Checking Stationarity in Python:**

- Python has two popular tools to check the stationarity of a dataset
- **Rolling Statistics**
  - Plot the moving average, variance and check whether that is moving or not.
  - A moving average is one, that is taken in a specific time window only
  - Visual Technique
    - Not deployable on production, but it is useful for Proof of Concept
- **Augmented Dickey Fuller Test**
  - Statistical Test to check for stationarity
  - Data Science related
  - Null Hypothesis is that the Time Series is non-stationary
  - Result consists of Test Statistics and Critical Values

#### **Models to work with Time Series Data**

##### **Arima**

- The ARIMA model shows great promise
- ARIMA = AR + I + MA
  - **AR** - Auto Regression
    - Describes co-relation between previous time period and current time period
  - **I** - Integration
  - **MA** - Moving Average
    - The noise in the time series is averaged out and smoothened out and calculated.
    - The crest / troughs of noise are smoothened out to bring an average forecast
- **Parameters of ARIMA (to predict)**
  - P
    - Autoregressive lags
    - Prediction done using PACR graph (Partial Autocorellation)
  - Q
    - Moving average
    - Prediction done using ACF Plot (Auto Correlation Plot)
  - D
    - Order of differentiation
      - If the order of integration is 1, then  $D = 1$
      - To make data stationary, we use differentiation