Exploratory Data Analysis (EDA)

Session-1

In the world of data & data science, EDA plays a prominent role. It is the first step in any data science use-case after data gathering & cleaning.

Let's take a look at the life cycle of Data Science Project:

Life Cycle of a Data Science (DS) Use-Case begins with:

1) Requirement Gathering: At this stage, the problem statement is understood based on which team is formed. Roles are defined & responsibilities are distributed. After this, the team begins to think what data do they require & how would they get that.

2) EDA: At this stage, the data is gathered through various sources like, third party APIs, Web Scrapers & other data vendors. Sometimes, the organization for which we're working gives us the data. Data is stored in various SQL & NoSQL Databases. Statistical Analysis done over the data.

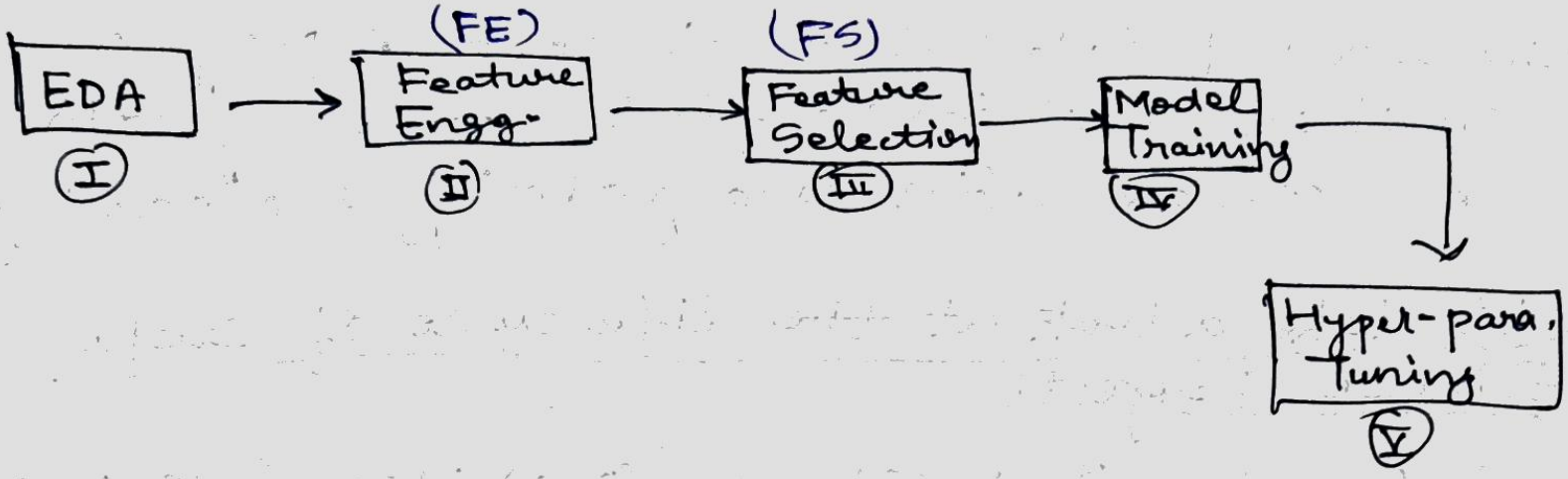
3) Feature Enggineering: It is the next step in which using domain knowledge we extract features from raw data.

4) Feature Selection: Based on knowledge & experience, we select the features & discard others that are not going to play a prominent role in our ML Algo.

5) Model Creation: A suitable ML model is created based on the requirement & data.

6) Hyper-parameter Tuning: This helps in increasing the over-all efficiency of the model by tuning the parameters of Mathematical equations of ML Model. It's a never ending process.

Diagrammatic Representation of Life Cycle of DS Project:



Since, these sessions will primarily focus on EDA.
∴ We're not going to worry about the ongoing steps but we'll cover them gradually!!

EDA or Exploratory Data Analysis or Statistical Data Analysis requires Maths & Statistics, as the name suggests.

∴ Statistics: It is the science of collecting, organizing, analyzing, presenting/visualizing & drawing conclusions from data.

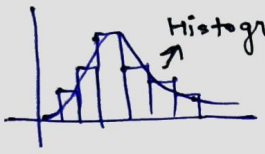
Data: "Facts or pieces of Information"

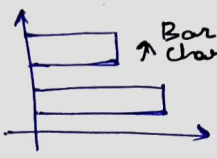
E.g. Ages of Students, weights of people.

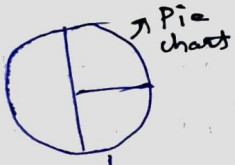
Types of Statistics

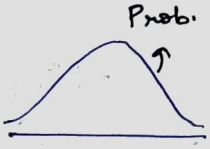
Descriptive Stats → (EDA + FE)

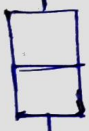
→ Consists of organizing & summarizing the data.

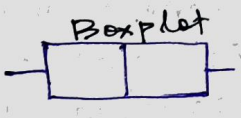




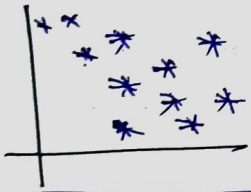








Candlestick Chart



→ Scatter Plot

Inferential Stats

→ consists of collecting sample data & making conclusions about the population data using some Statistical experiments.

↳ Hypothesis Testing

- → Z-test
- → t-test
- → χ^2 (chi-Square)-Test
- → F-test (ANOVA)

→ P-Value

e.g. Exit Polls of News channels after Elections. Surveys a handful people & predicts the results for whole state/country.

Sample Data ➡ Population Data

E.g. Let's say there are 20 classrooms in a University & you've collected the ages & weights of students in one classroom:

Ages: {21, 20, 18, 34, 17, 22, 24, 25, 26, 23, 22}

Weights: {60, 65, 56, 70, ... , 67}

Based on the classification that we've done above of statistics, we can ask the following questions:

Descriptive Stats: What is the average age of students in the classroom?

→ Relationship b/w Age & weight?

Inferential Stats: Is the average age of the students in the classroom less than | greater than | equal to the average age of students in the university?

→ Sampling Techniques:

1) Simple Random Sampling:
2) ~~Stat~~ Stratified Sampling:
3) Systematic Sampling
4) Convenience Sampling

1) Simple Random Sampling: This sampling is simplest of all samplings. In this, we choose a fixed number of items from the population. The probability of every item getting selected is equal because of which the sampling becomes unbiased.
E.g. Drawing a lottery winner.

2) Stratified Sampling: The word ~~means~~ 'strata' means to group/layers. We group the data based on some categorical feature & then sample data from them as we're doing previously.

E.g. In Exit polls: Making Groups

Less than 18	More than 18
(Not useful)	(Useful)
∵ Didn't Vote	∵ They vote

3) Systematic Sampling: We draw every n^{th} item from the population.

E.g. Banks choosing that they'll call every 3rd customer for Loan/Credit Card.

→ Door-to-Door Salesman decides to visit every 2nd house that comes in the way.

4) Convenience Sampling: It is used when we want correct, true & quality data.
We draw only those people that are ready to comply with us.

E.g. Youtube, these days runs a survey for its improvement. Survey contains some questions regarding our personal experience with youtube but. But, it does not forces anyone to fill that survey.
We can easily skip that.

Variable: A variable is a property that can take any values.

↳ (Vary - Able) → Able to Vary

Types of Variables

Quantitative Variables

→ can be measured numerically

→ e.g. Age, weight, height, rainfall, temp, distance

→ They can contain any number of values.

Age / Weight

Qualitative Variables

→ Categorical Variables

~~(Based on sex~~

→ e.g. Gender, Marital Status

→ They can only have a fixed no. of values.

Gender

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* Quantitative Variables can be classified further:

e.g. Age: 15, 16, 17, 18, ... (Whole Numbers) { There can't be any other no b/w 15 & 16

Weight: 35, 36, 36.5, 37, ... (Real Numbers) { There can be another value b/w 36 & 36.5 like 36.4 }

Quantitative Variables

Discrete

→ The whole number data that we discussed previously, comes under discrete variables

→ E.g. Pincode (Fixed, Whole Numbers)

Continuous

→ The Real Numbers data comes under Continuous Variables.

→ E.g. Height, Rainfall (Not Fixed, Real Number)

e.g. Let's classify variables to the data:

→ Marital Status / Gender : Categorical / Qualitative

→ River Length: Continuous Variable

→ Movie Duration : Continuous Variable

→ Pincode : Discrete (100110, 100111, 100112, ... etc.)

→ IQ : Discrete (100, 110, 120, ... etc.)