

EDA (Exploratory Data Analysis) & Feature Engineering

EDA - EDA is used by data scientists to analyze and investigate data sets and summarize their main characteristics, often employing data visualization methods.

Data Science Life cycle -

- 1) Data Ingestion — Project
- 2) EDA (Analysis)
- 3) Processing (Pre)
- 4) Model
- 5) Evaluate and validate

(big data tools, remote location)
 (SQL, nosql), Some file format
 CSV, tsv, XML, json, website
 → HDFS, NOSQL, Kafka, spark

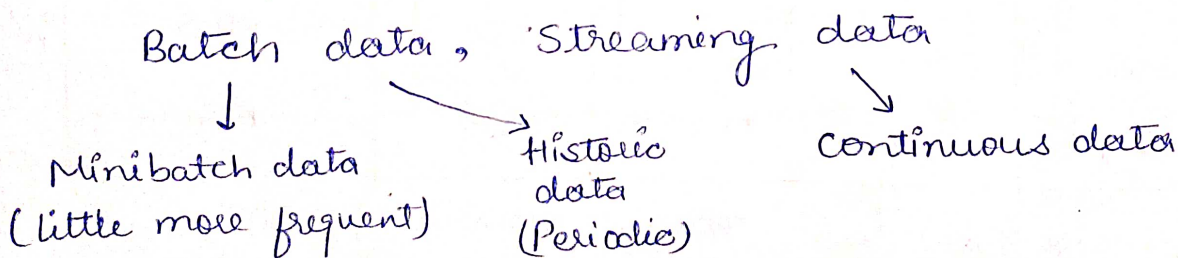
Statistics -

Collect, Organise, Interpretation, Analysis

Insight

(Scientific, healthcare, Social problem)

Data Types -

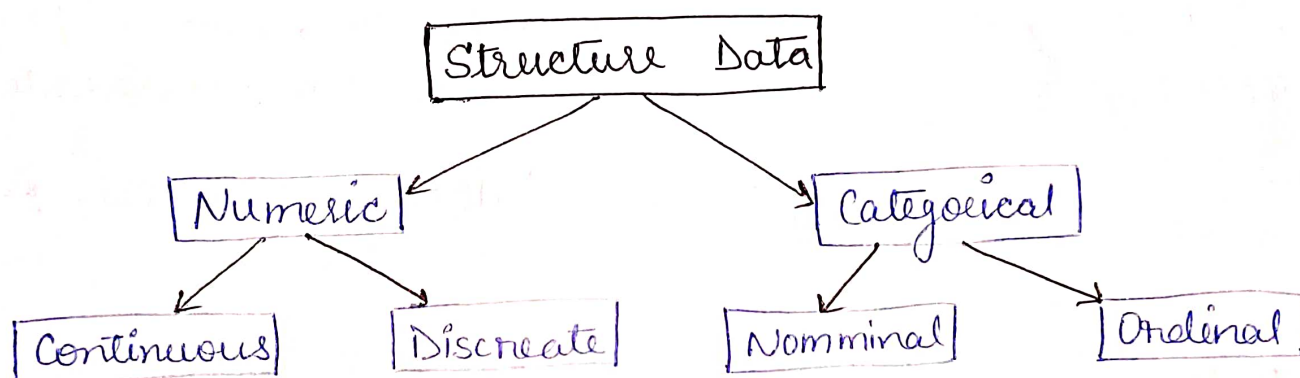


- | | |
|------------------------|-----------------------|
| 1) Structure data | — Table |
| 2) Unstructure data | — Video, Images, text |
| 3) Semi structure data | — XML, Json |

EDA + FE →

Structure Data -

| Weight | Height | BMI |
|--------|--------|-----|
| 70 | 170 | 22 |
| 80 | 180 | 24 |
| 90 | 190 | 26 |
| 100 | 200 | 30 |
| 60 | 160 | 21 |



STUDENT PERFORMANCE -

| FEATURE | | | | | |
|---------|-----|--------|--------|--------|-----------|
| Name | Age | Height | Sex | Weight | Education |
| Sunny | 25 | 170 | Male | 70 | UG |
| Anujit | 30 | 180 | Male | 80 | PG |
| Prityam | 35 | 160 | Male | 60 | UG |
| Pritya | 20 | 150 | Female | 55 | Phd |
| Aditi | 27 | 145 | Female | 58 | PG |

Categorical → Nominal Numerical → Continuous Numerical → Continuous Categorical → Nominal Numerical → Continuous Categorical → Ordinal

EDA - TYPE OF DATA

Univariate - Single Column
 Bivariate - Two Column
 Multivariate - More than two column

Independent / Dependent -

Age, height, Sex - Dependent [Weight]
 Age, height - Independent

Core ML Pipeline

- 1) Data ingestion
- 2) EDA
- 3) Preprocessing → FE
- 4) Model building
- 5) Evaluation & validation

Data - Analysis

- 1) Missing value
- 2) Outliers
- 3) Scaling

} Feature/Column

First EDA is required on FE ~~or~~ PP? - EDA → PP/FE

| Name | Age | Education | Salary | Exp |
|---------|-----|-----------|--------|-----|
| Sunny | 25 | UG | 25K | 2 |
| Deepak | 30 | PG | 30K | 3 |
| Rushi | 40 | UG | 40K | 5 |
| Aman | 50 | Phd | 50K | 10 |
| Shalini | 20 | UG | 35K | 1 |

Steps -

EDA (Analysis)

- Profile of the data
- Statistical analysis
- Graph based analysis

Profile of the data

Stats based Interpretation

| | | |
|---------------------|----------------------|--------------------|
| Row | Graph based analysis | Variance |
| Column | | Covariance |
| No of missing value | | Standard deviation |
| Categorical | | Correlation |
| Numerical | | Chi-Square |
| Duplicate | | T-test |
| Data type | | Z-test |
| RAM | | Anova test |
| Data Size | | mean/median/mode |

★ Based on EDA can we do processing of the data?
Yes

STEPS FOR FEATURE ENGINEERING

- 1.) Missing value handle
- 2.) Outliers handle
- 3.) Scaling of data
- 4.) Transformation (log, Boxcox, Sqrt, Cube)
- 5.) encoding
- 6.) Imbalance data
- 7.) Feature selection
- 8.) Dimension reduction (PCA, ESNE)
- 9.) Duplicate value/Column
- 10.) Split / merge / drop / add

25 Sep 22 / Sunday

WAY OF PERFORMING FEATURE ENGINEERING —

1.) Missing Value handle

- 1) Random no filling
- 2) Forward filling / backward filling
- 3) Statistical approach - mean, median, mode
- 4) end of the distribution
- 5) drop the row
- 6) Knn - imputer
- 7) Can we take that ML algorithm which missing value.
- 8) Can create own ML model and predict missing value.

2.) Outlier

detect

Z-Score
IQR
box-plot
Scatter plot
Violin plot

handling

drop
median
replace / trimming

3.) Transformation / Sc

box-cox
power transformation
log
square
Cube
Yeo johnson

4.) Scaling

Standardization
Min/Max
Unit scaling

5.) Encoding

One hot
label encoding
binary encoding
target guided encoding
hash encoding

6.) Imbalanced

collect more data
undersampling
oversampling
cluster-based oversampling