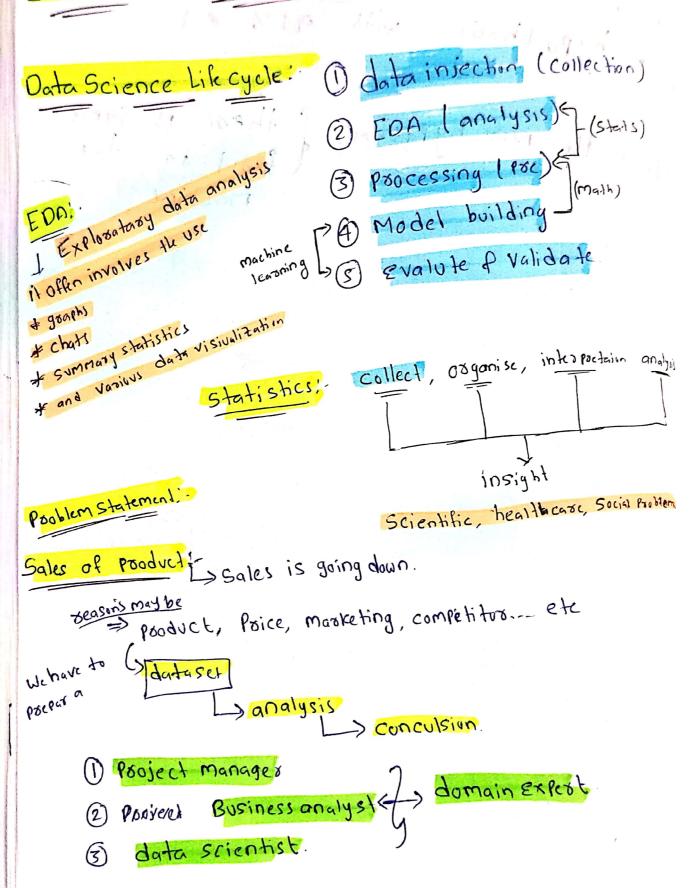
+ Have Krishna +

## EDA & feature Engineering

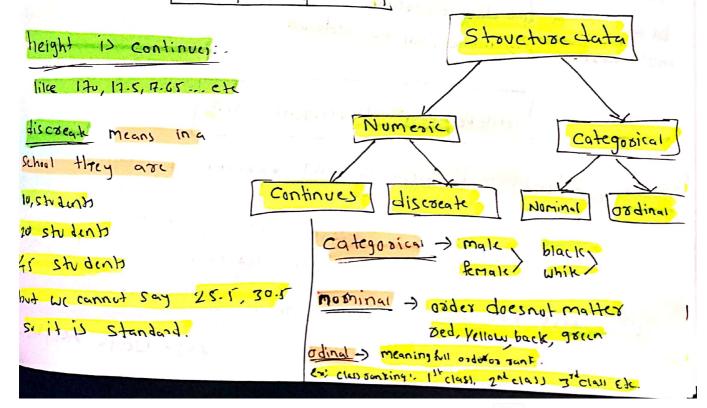


data can be Present in (HDFS) hadoordishibuted file system. (Bigdata tools), ocmok location (521, NOSZI) -> Some Pile formate (csv, tsv, xML, 1500 Excel. etc) websites.

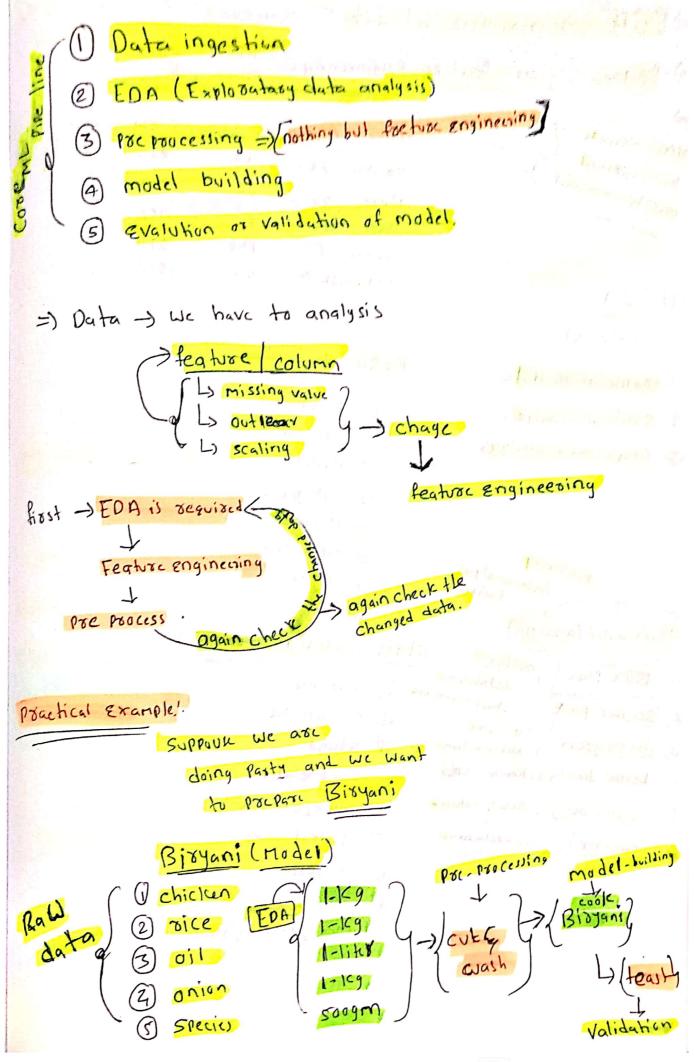
types of data: L) Batch data, Streaming data mini batch historical data data (Persodic) little more forg)

- 1) Structure data :- table formate
- 2) unstaucture data! · video, image, voice, sound, text
- 3 Semi stoucture data! XML, ison L) deep learning

Stoucture data: (Kg) (cm) weight height BMI 70 170 22 24 180 80 90 140 26 100 200 30 160



| Ext data mis  |          | Slawer   |          | -            | > bi-var   | iant    | multivarial        |  |  |  |
|---|----------|--|----------|--------------|------------|---------|--------------------|--|--|--|
|   |          | 0 0 0  | 0 5 0 1  | 1            |            |         |                    |  |  |  |
| N   | ame      | Age  | weight   | 1            | Sex        | height  | Education          |  |  |  |
| N   | avern    | 25   | 70       | Section 1    | male       | 1 }()   | 06                 |  |  |  |
| S   | sikanth  | 24   | Go       |              | male       | 160     | P.C.               |  |  |  |
| V   | 0000     | 35   | 50       |              | Male       | 150     | J'h                |  |  |  |
| S   | 501      | 20   | 70       | 1            | Male       | 170     | 649                |  |  |  |
| 8   | aghava   | 21   | 76       | TOTAL STREET | Male       | 176     | Pa                 |  |  |  |
| F   | oiya     | 32   | 8,0      | 1            | Male       | 180     | PG                 |  |  |  |
| Catagorical Numarical Numerical Catagorical Numerical Catagorical |          |  |          |              |            |         |                    |  |  |  |
| Cat   | agodical | Numadia  | al Numer | 010          | of caradom | 4 1001  | enca Catagonia     |  |  |  |
| I   | •        | continu  | es conti | nic          | nomir      | nal Con | hinisves (ordinal) |  |  |  |
| indicated)  |          |  |          |              |            |         |                    |  |  |  |
| types of datal we have to check)                                  |          |  |          |              |            |         |                    |  |  |  |
| types of deferme have to cheers  PG-1 9 1cv11  Ph1-2              |          |  |          |              |            |         |                    |  |  |  |
| Phi-2   |          |  |          |              |            |         |                    |  |  |  |
| bi - variate as double column                                     |          |  |          |              |            |         |                    |  |  |  |
| multivariate >> more than two column.                             |          |  |          |              |            |         |                    |  |  |  |
| 1710111100  |          | The state of the s |          |              |            |         |                    |  |  |  |
| independent Dependent varian                                      |          |  |          |              |            |         |                    |  |  |  |
| independent - variables- if any variable or value are             |          |  |          |              |            |         |                    |  |  |  |
| independent which means not                                       |          |  |          |              |            |         |                    |  |  |  |
| dependent on any value or variable                                |          |  |          |              |            |         |                    |  |  |  |
| could independent Ext Age, ser                                    |          |  |          |              |            |         |                    |  |  |  |
| defendent variables opposite of independent varable               |          |  |          |              |            |         |                    |  |  |  |
| ex1- weigh, height  |          |  |          |              |            |         |                    |  |  |  |



| - = EDA -) analysis a  | + data         | per from             |           | 79  |
|--|----------------|----------------------|-----------|-----|
| - = EDA - S and 1  |                | as many              | times     |     |
| =) Pre roocessing (00) feature   | Engineering.   | al we w              | 124       |     |
| 2) 100 1000  |                |                      |           |     |
| =) (== afe   | Name Age       | education            | Salary    | Emp |
| Uni-variate  | Koishar 25     | υA                   | 9016      | 2   |
| bi- variant  multi- variant  | Rama 30        | UA                   | 8010      | 3   |
| mulh   | Have 40        | PA                   | 961       | 5   |
|  | gavinda 50     | PHE                  | 8716      | 10  |
|  | jagannadh 20   | 04                   | 1001      | C   |
| () EDA   | s repland      |                      |           |     |
| (analysis)   | Profile of the | data                 |           |     |
| 1) Profile of the data   | Pookie -       |                      |           |     |
|  | (i) ROW        | 77.1                 |           |     |
| 2) Statical analysis   | (2) COlum      | ń                    |           |     |
| 3 Goaph based analysis   | 3 missing      |                      | ,         |     |
| PART E DAINEERING  | 0              |                      | SEL A. C. |     |
|  | ( numan        |                      |           |     |
|  | @ durlica      | 43 18 25 3 5 73 . 45 |           |     |
| t data analysis  | a ptyl         |                      |           |     |
|  | @ RAM          |                      | 90 32     |     |
| graph based (analysis)   | tats based(    | anterpration         | ()        |     |
| U a law outlier,   | <u></u>        |                      |           |     |
| Olistaino III  | Vazianu        |                      |           |     |
| autlican   | (b) (05 vass   | na                   |           |     |
| (3) his to gram distribution   | 3 stand        | 3-4 7 (2-74)         | 1         |     |
| (G) Kernal density estimation (KDE).   | @ Co-bool      | ation                |           |     |
| ( count but -) Rows, column  | @ chiseu       | oc - test-           |           |     |
| (6) hear mar - wordahim  | @ t-test       |                      | la Ca     |     |
| Thomas the state of the state o | P Z- test      | 10.11                |           |     |
|  | ® anoug        | test                 |           |     |
|  | (9) Mean/      | median/ mode         | 14        |     |
|  | Pt maj         | <u>_1</u>            |           |     |
|  |                |                      |           |     |

3) Based on a EDA we can do a processing of the data. 1) missing value handle

@ outlier handle

3) scaling of data

a) toansformation (109, Box, Square, cube)

(5) En-coding

@ Imbalance data

1) feature selection

Dimention reduction (PCA, ESNE) (8)

M. no.06

```
- missing valve
missing null value
                         pre Processing
  EDA
                         handling
outlier
                   -> Encoding
Catog (man, woman)
                   -> Scale (with a certain range)
skewed range
```

(5) -> Count of feature -> handle imbalance -) feature selection -) dimension deduction.

Encoding: Converting the catagorical valves to numerical value is called En-coding.