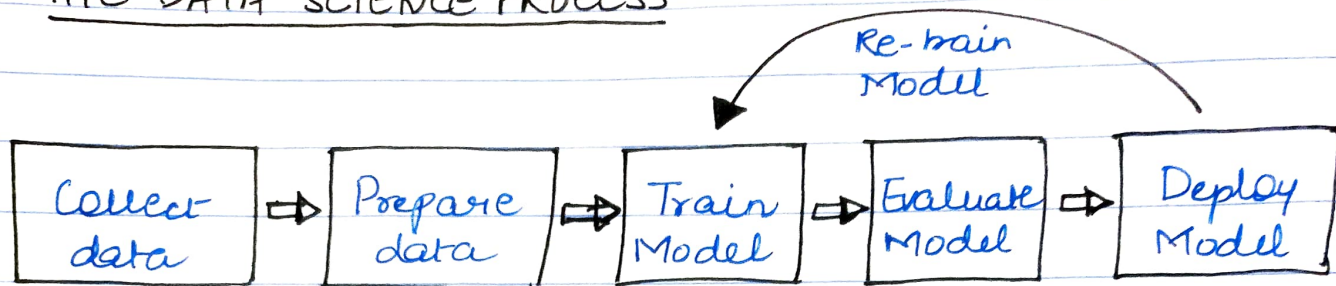


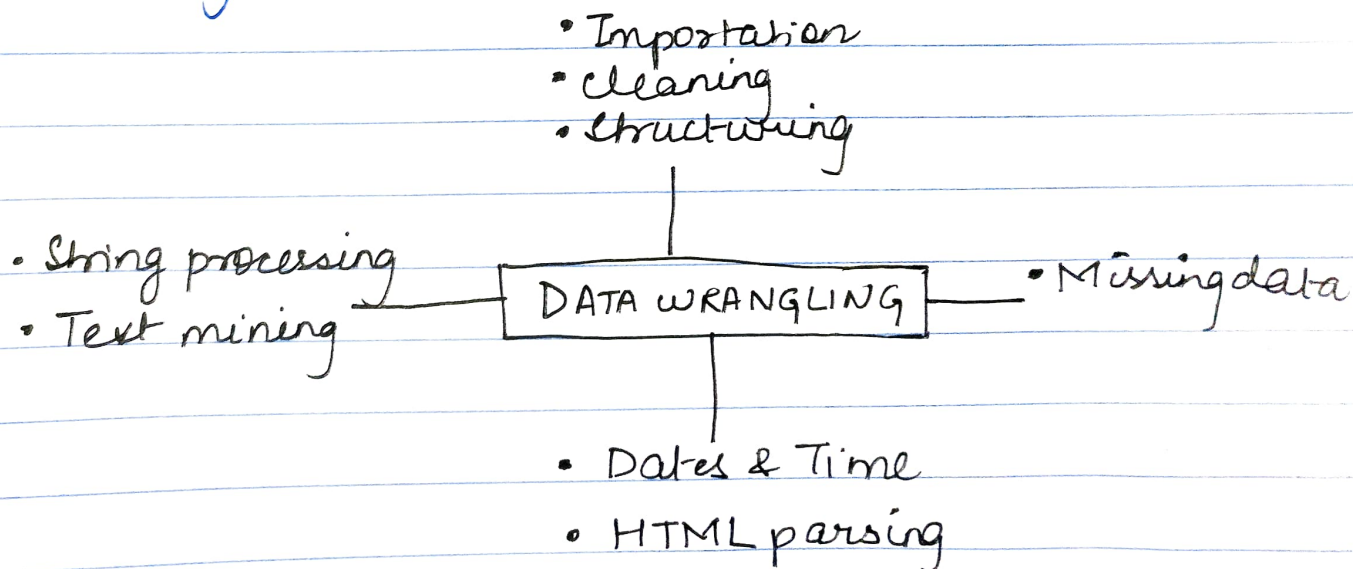
9/07/20

5. THE DATA SCIENCE PROCESS



II) Prepare data \Rightarrow Performing data wrangling
(referred to as data munging - is a process of transforming & mapping data from one "raw" data form into another format with the intent of making it more appropriate & valuable for a variety of downstream purposes eg analytics. \neq Involves processes :

1) cleaning, 2) structuring, 3) enriching raw data into desired format for better decision making in less time.)



\therefore in data preparation \rightarrow we identify or create the features needed for the model.

- III) Train Model -
- 1) select an algorithm
 - 2) prepare our training, testing & validation datasets
 - 3) Iteratively evaluate the model to identify the best-performing version & sanity check the outcome.

- IV) Evaluate Model - Run the model through a final exam using data from our validation data-set & see how it performs.

- V) Deploy Model - Pack into the model in-dependencies up for the deployment, use it within our web-service or within an API in an application & measure the ongoing performance of the model.

- VI) Retrain Model (Last Step) - it is an iterative step for models in production.

Developer's perspective

- 1) collect data - write code
 - 2) Prepare data - write queries & code
 - 3) Train model - write code, do some math \Rightarrow feature vectorization, feature scaling & tuning the ML algo.
 - 4) Evaluate model - computing evaluation metrics or evaluation graphs on the ^{test} data sets.
 - 5) Deploy Model - DevOps \Rightarrow involve training, evaluation & deployment scripts in respective build & release pipelines.
 \rightarrow means you can access any version of the product.
- * make sure all models & deployments are versioned & artefacts are archived *

6. COMMON TYPES OF DATA

- **Numerical** - integers or floats
eg identifiers of items, different properties like sales amount, house prices.

★ All data in ML eventually ends up being numerical data, whether its numerical in its original form or processed from other more complex structured forms like image, speech, or text.

- **Time-series** - series of numerical values that can be ordered, typically data collected over equally spaced points in time, but it also can be data that is ordered based on a non-date-time column. (// numerical data pts. across pts in time)
eg → real-time stock performances, energy demand forecasting, ~~speech data~~

- **Categorical** - includes discrete & limited set of values. eg gender, ethnicity, location ID
less ~~low~~ imp. ↳ high imp.

- **Text** - words, sentences eg newspaper text.

- **Image** - transform into appropriate numeric form

$$\begin{aligned}\mu &= \text{mean} \\ \sigma &= \text{S.D.} \\ x_{\max} - x_{\min} &= \text{range}\end{aligned}$$

TABULAR DATA - most common in ML.

- Row - an item | single observation.
- Column - property
- Cell - individual data point | single value in a row or column
- Column values can be continuous or discrete (categorical)

Discrete \rightarrow eg make (Brand), color

Continuous \rightarrow Price, quantity

\hookrightarrow scaling of input data is done.

* Vectors

In ML, we ultimately always work with numbers or specifically vectors.

A vector is simply an array of numbers, eg (1, 2, 3) - or a nested array that contains other arrays of numbers \rightarrow (1, 2; (1, 2, 3)).

* All non-numerical data types (eg images, text, & categories) must eventually be represented as numbers.

SCALING DATA - 2 methods:

Standardization

Rescales data to have mean = 0 & $\sigma = 1$ (S.D.)

$$\frac{x - \mu}{\sigma}$$

Normalization

Rescales data into range [0, 1].

$$\frac{x - x_{\min}}{x_{\max} - x_{\min}}$$