Data Science and its own importance

Introduction: Data science is an inter-disciplinary field that uses scientific methods, processes, algorithms and systems to extract knowledge and insights from many structural and unstructured data. Data science is related to data mining, machine learning and big data.

What is Data?

- Data is often viewed as the lowest level of abstraction from which information and knowledge are derived.
- Data can be numbers, words, measurements, observations or even just descriptions of things.
 Also, data is a representation of a fact, figure and idea.
- Data on its own carries no meaning. If data to be an information, it must be interpreted and take on a meaning.

Scientific Methods Real-world Applications Data Science Data Consulting Data Technology Data Research

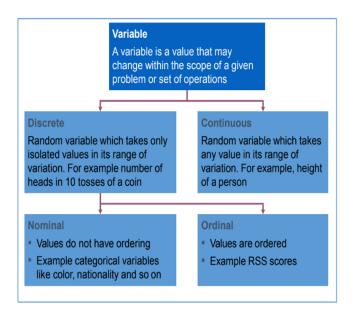
Types of Data/Variables:

Catagorical data is the data that is non numeric.

e.g.. Favourite color, Place of Birth, Types of Car

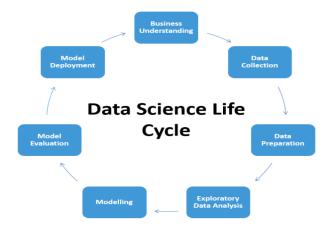
Quantitative Data is numeric. There are 2 types of quantitative data.

- Discreate data can only take specific values;
 e.g. shoe size,number of brothers, number ofcars in a car park.
- Continuous data can take any numerical value;
 e.g. height, mass,length.

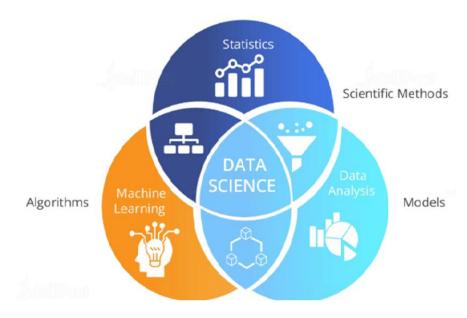


Data Science Steps:

- Data Gathering: Unless you're at a company with great data governance you're likely going to have some trouble accessing the data you want. Whether that's because your company has neglected to put the necessary systems in place to gather data, or the data that they are collecting is fragmented and scattered across the organization, you'll have to first spend some time gathering whatever data you'll need to do your job.
- Data Preparation: Once you have access to data, you'll need to spend some time cleaning and formatting it. This is where Data Science can often become more of an art, then a science. Unlike datasets you'll find in competitions, the real world has very messy data sets. Missing values, error in data collection, data formatting, normalization, outliers - these are all issues that you'll have to learn to deal with.
- Exploration: Before diving into building any models, you'll want to explore the
 data to try to glean some insights. Clustering algorithms, scatterplots, bar
 graphs, Chernoff faces are all interesting ways of visualizing data that will lead to
 a better understanding of the structure of your data and aid you in your model
 building step.
- Model Building: With your data cleaned and formatted, you'll have an
 opportunity to explore a variety of models to see which one works best. Random
 Forests, SVM's, Bayesian Predictors Neural Networks, Deep Learning, KNearest Neighbours all models you should familiarize yourself with. There is no
 one model fits all, and so you again will need to develop intuition on which model
 suits your particular problem.
- Model Validation: Prediction accuracy is a common benchmark for whether
 your model is performing well, however often times there are other evaluation
 metrics to consider. False positives and false negatives are important to think
 about from the perspective of the problem you're working on. If you're predicting
 disease, you'll care more about minimizing false negative, since it may result in a
 persons death -whereas a false positive will only lead to additional testing.
- Model Deployment: Finally you'll deploy your model into the wild, as you
 gather more data and feedback on how its doing you'll be able to tweak and
 improve it as time goes on.

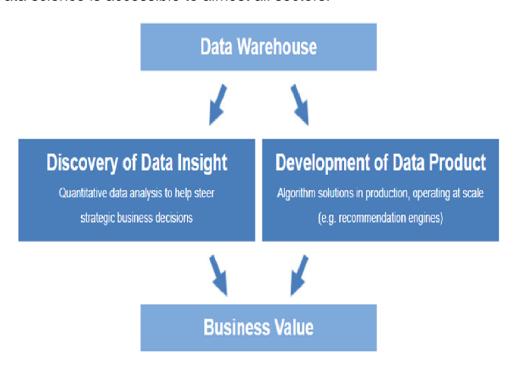


Key Elements of Data Science:



Data Science & its importance:

- Data science helps brands to understand their customers in a much enhanced and empowered manner.
- It allows brands to communicate their story in such a engaging and powerful manner.
- Big Data is a new field that is constantly growing and evolving.
- Its findings and results can be applied to almost any sector like travel, healthcare and education among others.
- Data science is accessible to almost all sectors.



Python Libraries:

Numpy, Pandas, Seaborn, Bokeh, SciPy, Scrapy, Keras, SciKit-Learn, PyTorch, TensorFlow, XGBoost, Matplotlib, Plotly, Pydot, StatsModel.

Open Source IDE for Programming:

- ▶ Google CoLab
- ► Anaconda (Jupyter Notebook)
- Pycharm

Data Repositories:

- Kaggle
- ▶ GitHub
- UCI Machine Learning