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Different types of data for machine

Content

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Algorithms for types of ML

Example of Supervised ,Unsupervised and reinforcement machine learning

Applications of ML

Different Use Cases for ML models

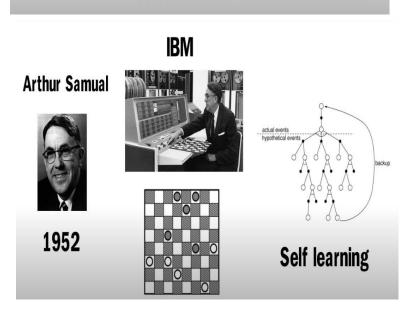
Different Terms uses in the machine learning models

Life Cycle of ML models

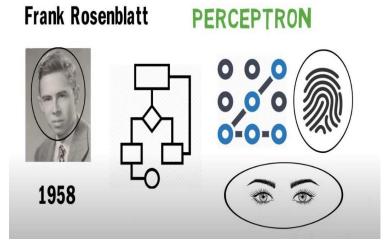
Machine learning (ML) is the study of computer algorithms that improve automatically through experience and by the use of data.

Its self learning process without explicitly involved any users. It learn from past data or experienced.

History







History

1979

Stanford Cart







Stanford students built the Stanford Cart, a remotely-controlled, autonomous cart that could navigate on its own and avoid bumping into objects. It was designed to help study the remote control of a Moon rover.

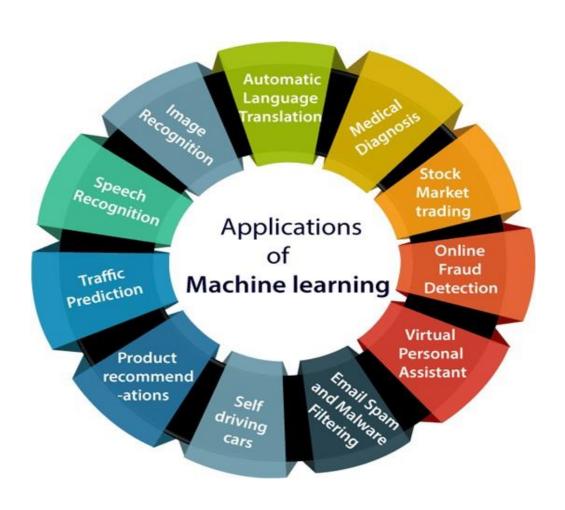


Terrence Sejnowski invented NETtalk, an ML-based computer program that could perform cognitive tasks like a human. With written English text and phonetic transcriptions as input, it learned to "talk" like a baby.

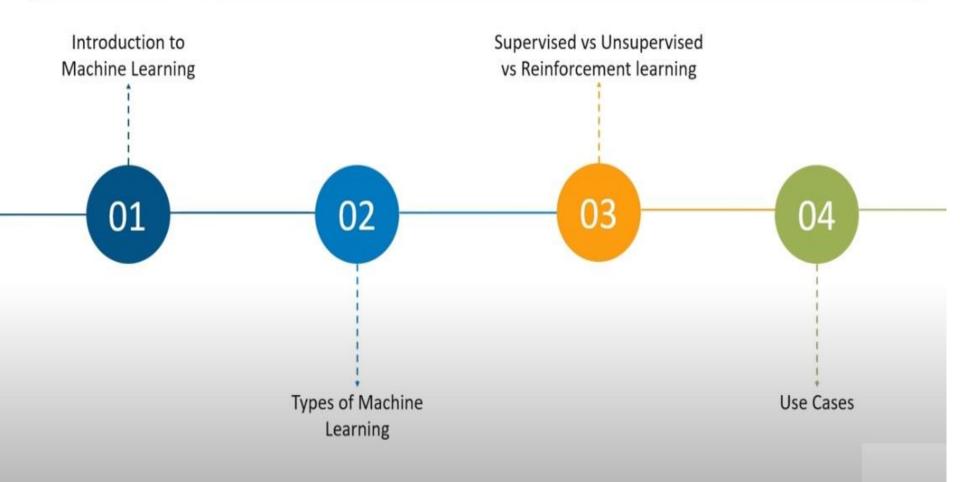


In October 2015, the distributed version of AlphaGo defeated the European Go champion Fan Hui, a 2-dan (out of 9 dan possible) professional, five to zero. This was the first time a computer Go program had beaten a professional human player on a full-sized board without handicap.

Application of M/C Learning

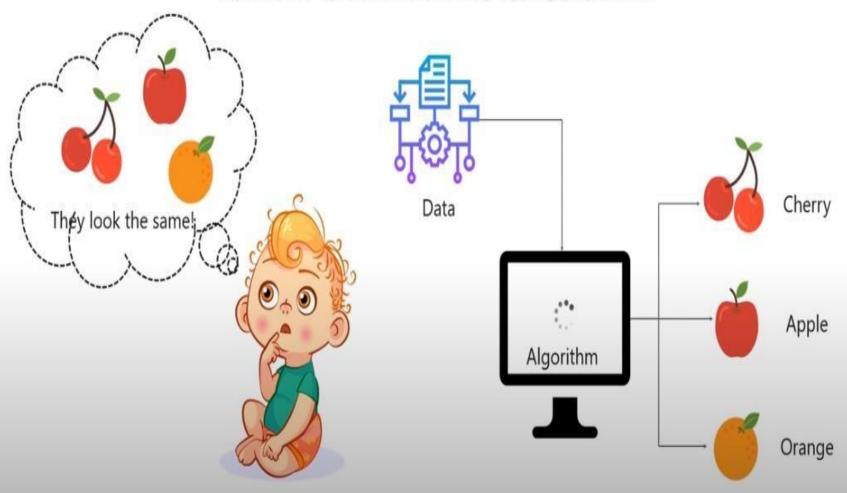


Agenda



What Is Machine Learning?

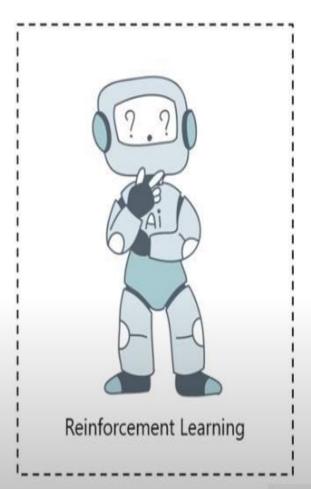
Machine learning is a subset of artificial intelligence (AI) which provides machines the ability to learn automatically & improve from experience without being explicitly programmed.



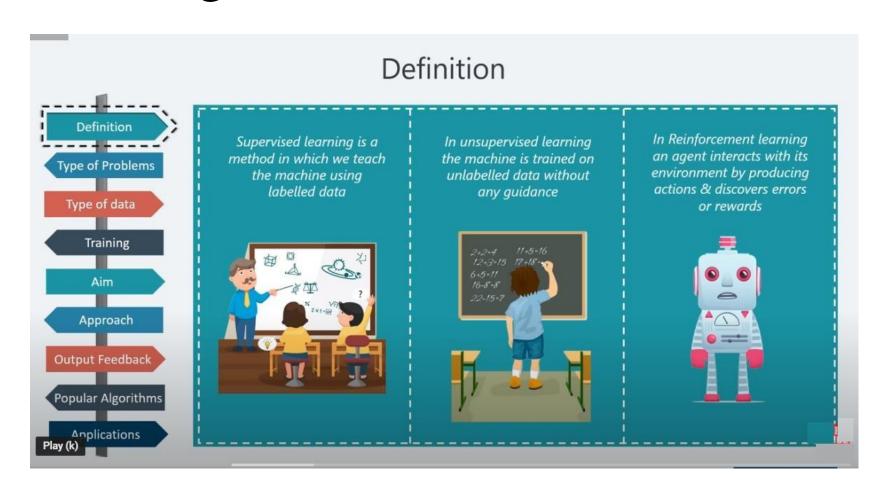
Types Of Machine Learning



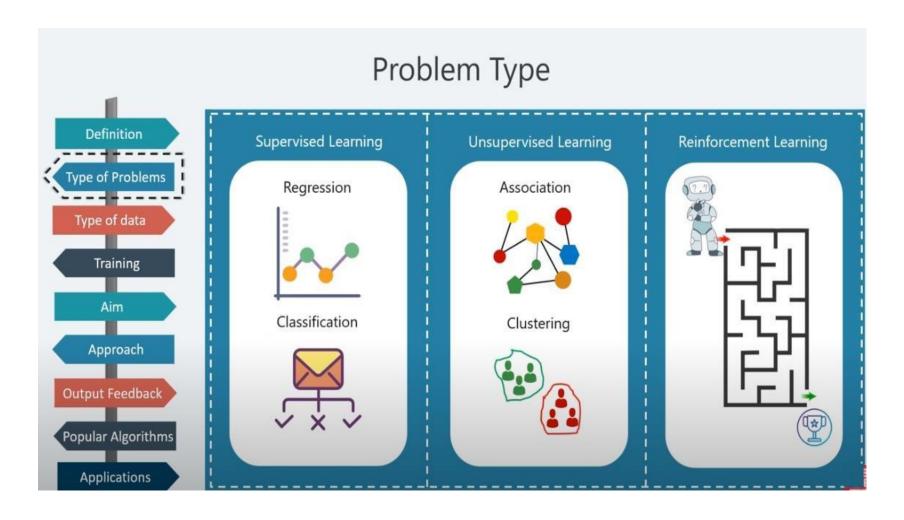




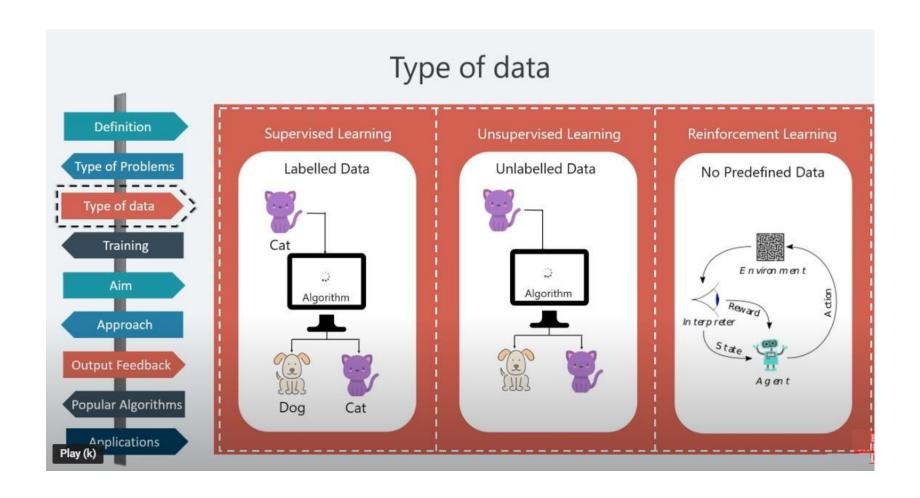
Difference Between Types of M/c Learning



Problem Types



Types of Data



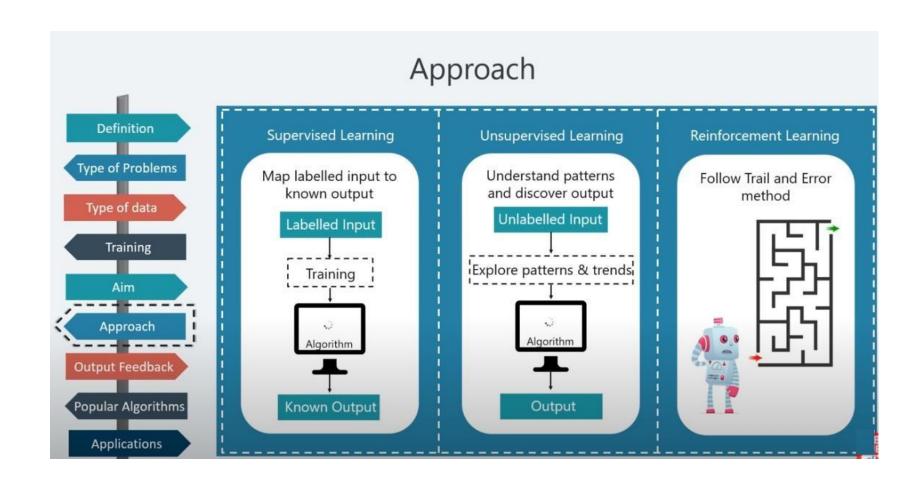
Training



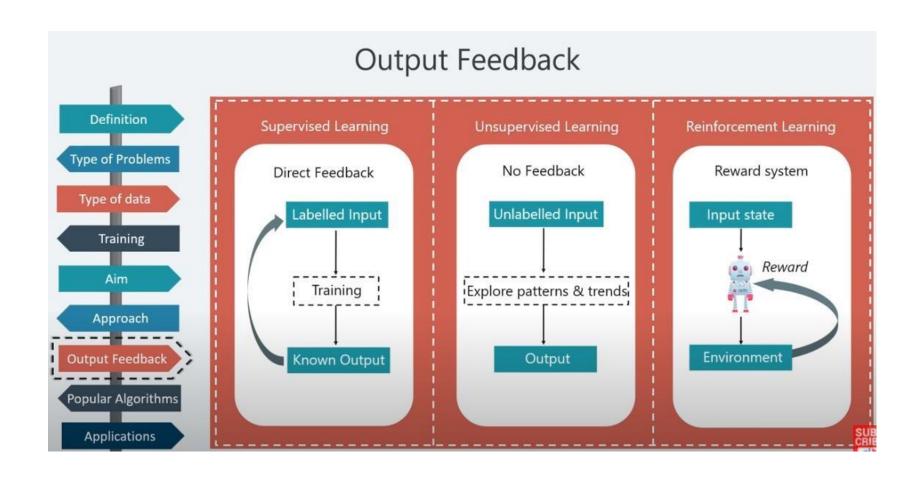
Aim



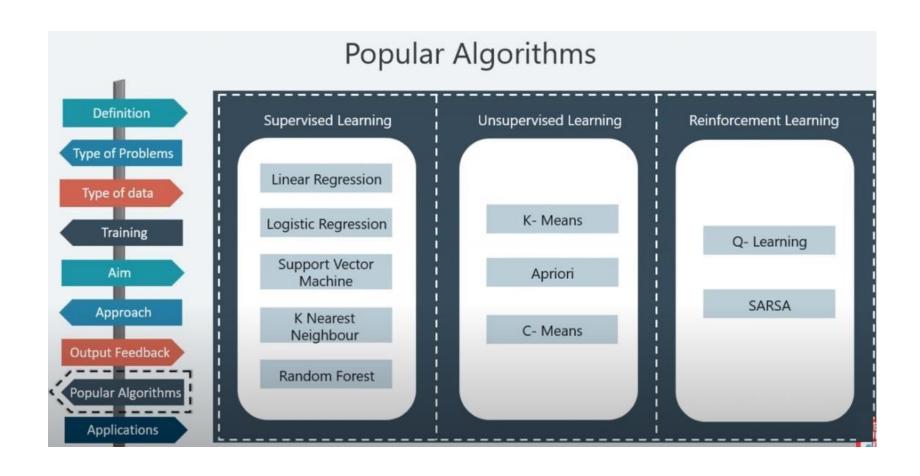
Approach



Approach



Algorithms



Real Life Example of Supervised machine learning

Email Filtering

Image Classification

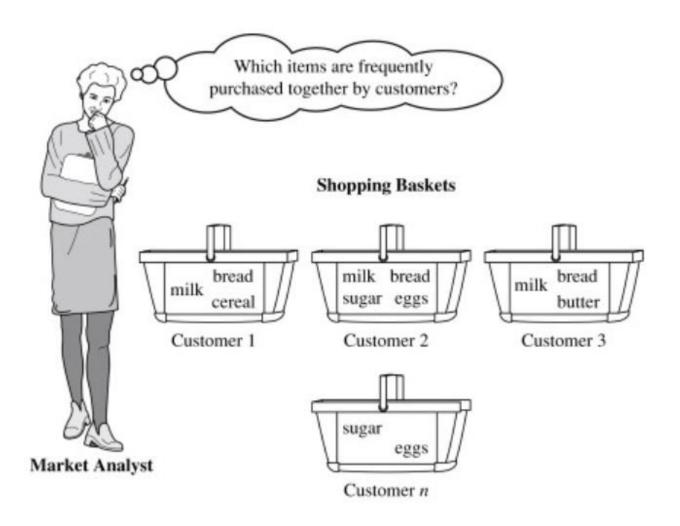
Fraud Detection

Visual Recognition

Different recommendations system:

Amazon, Facebook, Google, Netflix, Youtube, Contains supervised learning techniques.

Real Life Example of Unsupervised machine learning



Real Life Example of Reinforcement machine learning

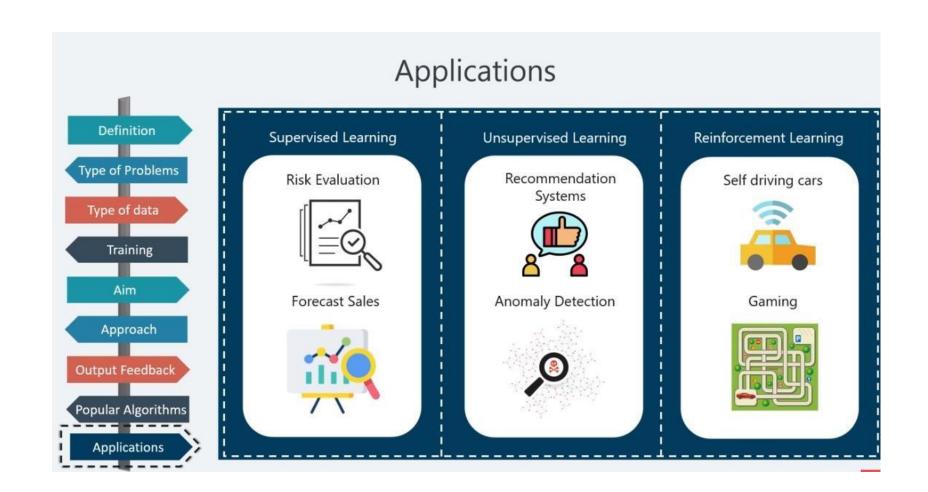
Robots equipped with visual sensors from to learn their surrounding environment

Traffic analysis and real-time road processing by video segmentation and frame-by-frame image processing

CCTV cameras for traffic and crowd analytics

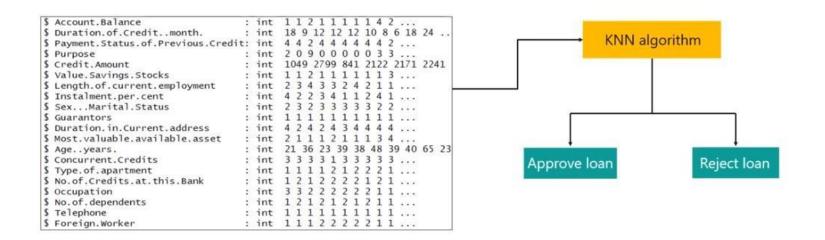
Scanners to understand and interpret text

Applications



Use Case 1

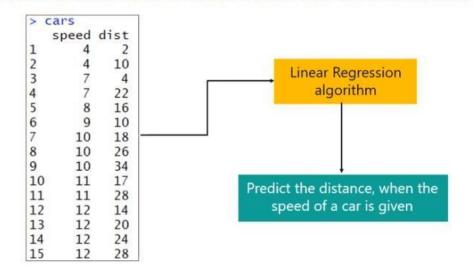
Problem Statement: Study a bank credit dataset and make a decision about whether to approve the loan of an applicant based on his profile



https://colab.research.google.com/drive/1cuKljFSMx3eD m3_LTyh-OjlUs4nlQYdf#scrollTo=x2RUPvunWH0E

Use Case 2

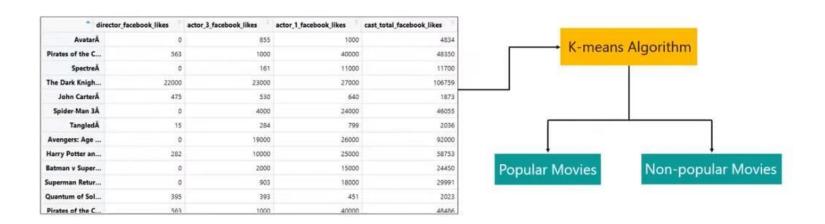
Problem Statement: To establish a mathematical equation for distance as a function of speed, so you can use it to predict distance when only the speed of the car is known.



https://colab.research.google.com/drive/1cuKljFSMx3eDm3 LTyh
-OjlUs4nlQYdf?usp=sharing

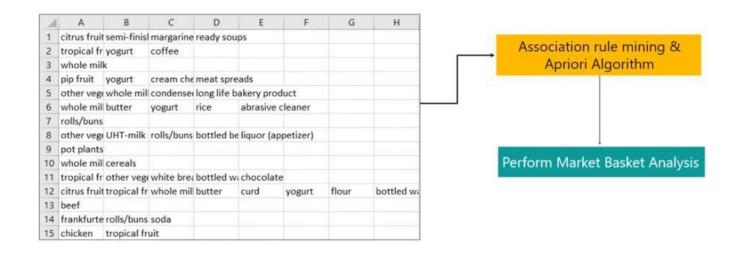
Use Case 3

Problem Statement: To cluster a set of movies as either good or average based on their social media out reach



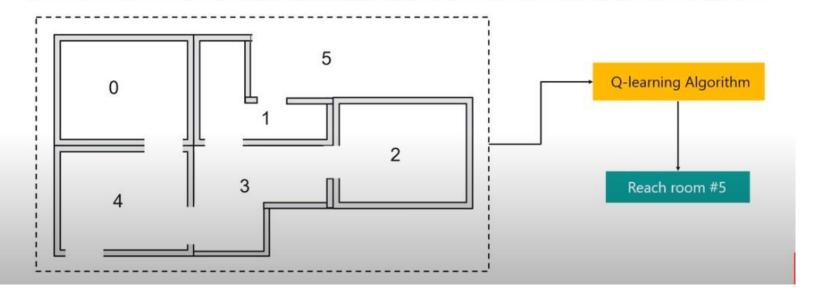
Use Case 4

Problem Statement: To perform Market Basket Analysis by finding association between items bought at the grocery store



Use Case 5

Problem Statement: Place an agent in any one of the rooms (0,1,2,3,4) and the goal is to reach outside the building (room 5)



Terms.....

Labels

Label is a value or thing we are trying to predict.

The label could be future price of a product, it can be whether the email needs to be routed to SPAM or INBOX

If we take example of following equation:

$$Y = Mx + C$$

So Y is the label in this case.

Terms.....

Features

A feature is an input variable - the x variable in simple linear regression

$$Y = Mx + C$$

A simple machine learning project might have just one feature.

while a more complex machine learning project could use hundreds of features like:

In the modeling and prediction of what would be the future price of product, the features could include the following:

- · Pack size of the product
- · Month of the year
- Competitors price of similar product

Examples

An "example" is a particular instance of data.

Examples are of two categories:

<u>Labeled</u>

This includes both features and Label

Unlabeled

This includes only features

Labelled Examples

Here first 3 columns are features and 4th column is the label Labelled examples are use to train and test the models to be used for predictions.

month of the year		competitor price	Products retail price
1	small	10	9.5
2	small	11	10.5
3	small	10.5	10
4	small	9.5	9

Unlabelled Examples

It contains only features and no label

The model trained using labelled examples is then used to predict the labels on unlabelled examples

month of the year		competitor price	
1	small	10	
2	small	11	
3	small	10.5	
4	small	9.5	

Model

It defines the relationship between features and label.

This relationship is derived by trying to fit various readily available algorithms or writing an custom algorithm.

Two key terms related to models are:

- <u>Training</u> This is the process where we feed the labelled data to the model and make it learn the relationship between features and label
- <u>Prediction</u> This is the process we feed unlabelled data to the trained model and obtain the values of labels

Regression vs Classification

A regression model predicts continuous values.

For example, regression models make predictions that answer questions like the following

- What is the value of a house in California?
- What is the probability that a user will click on this ad?

A classification model predicts discrete values.

For example, classification models make predictions that answer questions like the following:

- · Is a given email message spam or not spam?
- Is this an image of a dog, a cat, or a hamster?

- Analytics: Descriptive, Predictive,
 Prescriptive.
- Visualization : Data in to Graphs

- DataSet
- DataFrame
- Data
- RawData

Storage of data

- Outliers
- Missing Values/Imputation
- Feature Selection/Dimensionality Reduction
- Imbalance Data: Oversampling and Undersampling
- Time Series Data
- Feature Engineering

Outliers = odds

Detecting Outliers:

BoxPlot

Quartliles

Scatter Plot

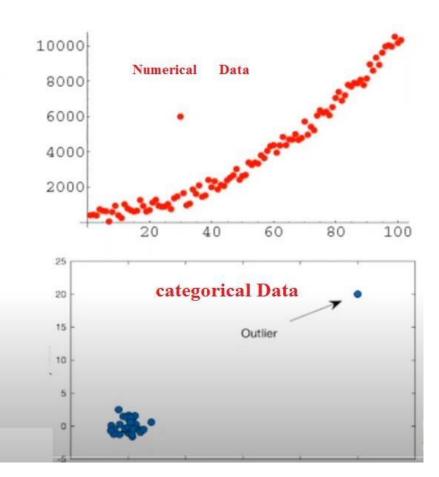
Z-score

Treating Outliers

Capping

Deletion

Replacing by mean, median and mode



Input and Output

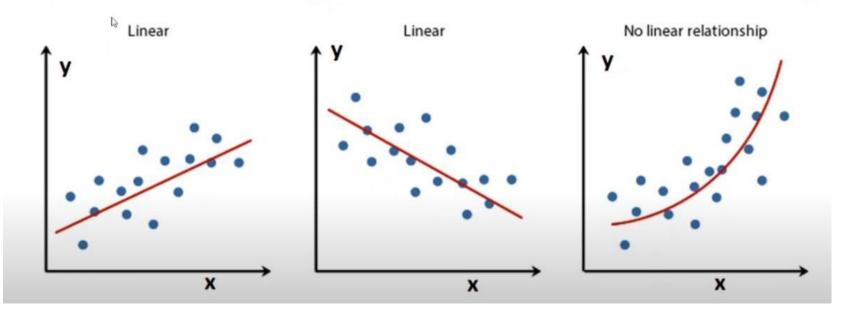
- Input Variables | Features | Columns |
 Dimensions | Characteristics | Independent
 Variables | X | Multiple
- Output variables | Outcome | Result | Target
 | Y | Dependent Variables | Y Predicted |
 | Single

Normalization

Normalization = Scaling

Normalization means to scale a variable to have a values between 0 and 1. Goal of Normalization is to change the values of numerical columns in the dataset to a common scale, without distorting differences in the ranges of values.

Linear Relationship and Non-Linear Relationship

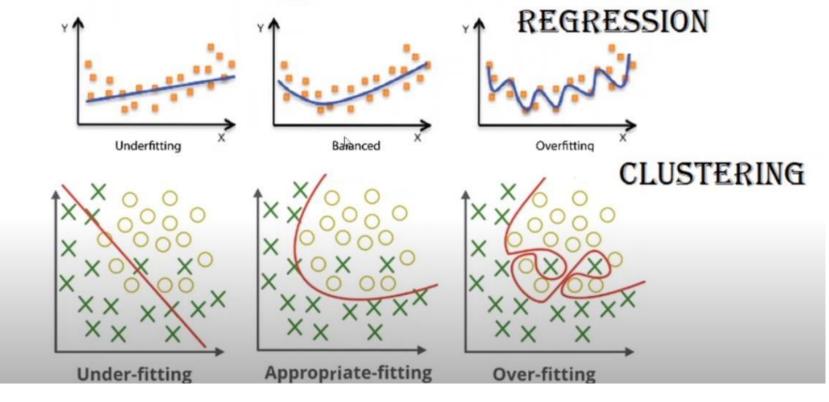


Train and Test Splitting

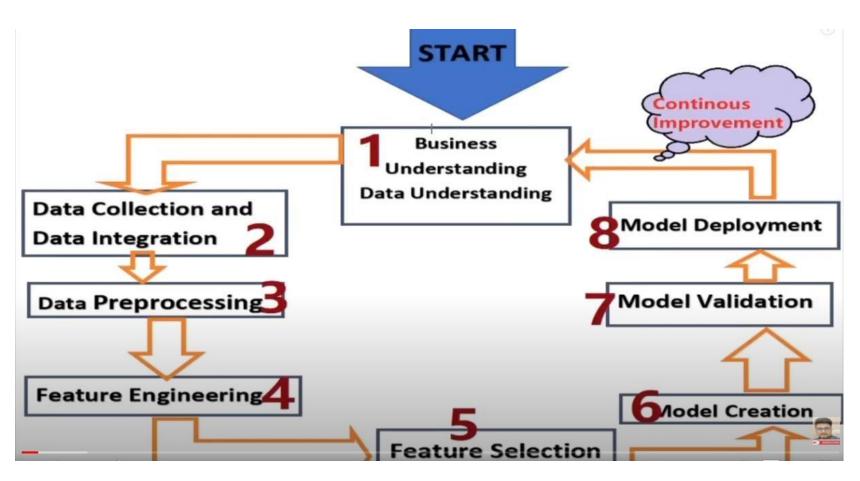
- X_train, Y_train
- X_test, Y_test

Random State and Sample Size

Underfitting and Overfitting



Life Cycle



1) BUSINESS UNDERSTANDING AND DATA UNDERSTANDING

- Nature of business
- Requirements
- · What we need?
- Goal (Prediction , Descriptive and Prescriptive)
- Final Deployment (On-premises or Clodu) (Online or Offline Learning)
- UAT
- Timelines/Deadlines
- Read all Documents/SOPS

2) DATA COLLECTION AND INTEGRATION

- Type of Data (structured/unstructured or offline/online)
- Source of Data (on premises or Cloud)
- Size of data
- Format of data
- Data transport
- Data Ingestion/Data Collection from all sources
- Data Integration (BODS ETL)

3) DATA PREPROCESSING (cd mouz)

- Unstructured in to Structured Format of data.
- Cleaning: (Noise, Special Chars, Lower),
- Removing Duplicate,
- Removing Missing,
- · Removing Outliers,
- Unidecode,
- Removing Zeroes values(imputation,
- Rounding,
- Formatting and Repairing(MCORDF)
- Organize the data: version conrol of data, auditing, maintaining data, Court Reviews conferences.

/4) FEATURE ENGINEERING (td soieng)

- Data Transformation :Normalization (skewing (z-score)), Logarithm log(x) (heteroscedasticity), 1/x, sqrt(x), exp(x)
- Discretization: Binning, Equal frequency discretization, Equal length discretization,
- Scaling: Standardization, Min-Max Scaling, Mean Scaling, Max Absolute Scaling, Unit norm-Scaling
- · One hot encoding, dummy variables, Rare variables.
- Imbalance dataset: SMOTE, SMOTEtomek, SMOTEEN
- Extracting features from text: Bag of words, Tfidf, n-grams, Word2vec, topic extraction
- create some new features by using domain knowledge from domain expertise or by using internet and google.
- Performing Statistical and Graphical Data Analysis (EDA): story telling

5) Feature Selection

- Convert many features in to important features
- Correlation, Heat Matrix (covariance)
- Multi colinearity (RIDGE, Combining)
- Backward Elimination
- PCA
- NMF

Sklearn library and scikit library

- ICA and FastICA
- SVD

6) Model Creation

- Basis of EDA, we consider Algorithms.
- Free Lunch Theorem no model is work well for every problem meaning of Free lunch theorem
- We consider at least 5 to 6 models.
- Regression: Linear, Lasso, SVR, Random Forest Classfier, Adaboost Regressor,
 XG Boost Regressor
- Classification: Logistics, SVC, NB, Random Forest Classifier, Adaboost Classifier, , XG Boost Regressor, Light GBM and CatGBM
- Anomaly detection: Isolation Forest, logistics and Local outlier Factor.

7) Model Validation and Model Selection

- · We will apply cross validation on all models
- Time and effort and resources if need to consider.
- Tradeoff between Bais/variance and True Positive/True Negative.
- Best Model will be selected for further tuning. (confusion matrix in case of Imbalance Dataset)
- Hyper Parameter optimisation is always important (Grid Search and Random Search)
- Automated code should be written that which model out of these 4-5-6 models.

8) Model Deployment

- Will generate the pickle file for ML Models
- Develop Front end API using Flask or Django framework
- On premises or Cloud
- · Storing the prediction in the Storage.
- Setting up the logging and monitoring frameworks to generate reports and dashboards based on the client requirements and to do continuous monitoring the output to find out whether the model is preforming well or not.

Continuous Improvement

- · Internal Evaluation
- External Evaluation
- UAT done by customer to check how model is responding as per their expectations

Scikit-learn

 Scikit-learn (Sklearn) is the most useful and robust library for machine learning in Python. It provides a selection of efficient tools for machine learning and statistical modeling including classification, regression, clustering and dimensionality reduction via a consistence interface in Python. This library, which is largely written in Python, is built upon NumPy, SciPy and Matplotlib.

Features

Rather than focusing on loading, manipulating and summarising data, Scikit-learn library is focused on modeling the data. Some of the most popular groups of models provided by Sklearn are as follows –

- Supervised Learning algorithms Almost all the popular supervised learning algorithms, like Linear Regression, Support Vector Machine (SVM), Decision Tree etc., are the part of scikitlearn.
- Unsupervised Learning algorithms —On the other hand, it also has all the popular unsupervised learning algorithms from clustering, factor analysis, PCA (Principal Component Analysis) to unsupervised neural networks.
- Clustering –This model is used for grouping unlabeled data.
- Cross Validation —It is used to check the accuracy of supervised models on unseen data.

Continue

- Dimensionality Reduction It is used for reducing the number of attributes in data which can be further used for summarisation, visualisation and feature selection.
- Ensemble methods –As name suggest, it is used for combining the predictions of multiple supervised models.
- Feature extraction It is used to extract the features from data to define the attributes in image and text data.
- Feature selection It is used to identify useful attributes to create supervised models.

Dataset Loading

- A collection of data is called dataset. It is having the following two components –
- **Features** –The variables of data are called its features. They are also known as predictors, inputs or attributes.
- Feature matrix –It is the collection of features, in case there are more than one.
- Feature Names -It is the list of all the names of the features.
- Response It is the output variable that basically depends upon the feature variables. They are also known as target, label or output.
- Response Vector –It is used to represent response column. Generally, we have just one response column.
- Target Names –It represent the possible values taken by a response vector.
- Scikit-learn have few example datasets like iris and digits for classification and the Boston house prices for regression.

