

Regression:

Linear- (one variable input, one variable output)

Multiple-(multiple variable input , one variable output)

Evaluation:

- Mean_absolute_error
- Mean_squared_error
- Root_mean_squared_error

Applications:

- Weather prediction
- House price prediction

Classification:

Evaluation parameters:

- Recall/TPR/Sensitivity= $TP/(TP+FN)$
- Specificity= $TN/(TN+FP)$
- FPR= $FP/(TN+FP)=1-\text{Specificity}$
- Confusion matrix
- Precision= $TP/(TP+FP)$
- Accuracy- no of samples correctly classified= $(TP+TN)/(TP+TN+FP+FN)$
- Classification report

Applications:

- Medical disease diagnosis
- Covid positive or negative based on symptoms

AUC ROC curve:

(Area under the curve) and (Receiver operating characteristics)

If 1-> clearly able to distinguish

If 0.7 -> (existence of FP, FN)

If 0.5-> Not able to distinguish

If 0-> It is predicting opposite labels

SVM:

- Support vectors- These are the points of one class that are nearer to the points of other class and thus help in deciding the margin between classes
- Can be used for both classification and regression.
- Mostly used for binary classification

Naive bayes:

Advantages:

- Easy to determine the class using the mathematical formula given the frequency distribution of various classes as per attributes

Disadvantages:

- Class conditional Independence

Applications:

- Spam/ham same
- Text classification
- Sentimental Analysis (Ex: positive/negative feedback)
- Recommender systems

KNN:

- Lazy learner
- Even if k is small the distance between all the points of the dataset and the given point should be calculated then sorted to find the k nearest neighbours
- Consumes a lot of time during testing phase, training phase just stores data

Apriori:

Quantification is done by

- a) Support
- b) Confidence
- c) Lift

How to improve apriori

FP growth:(method)

K Means:

- Mutually exclusive clusters
- Measure used is mean (or) median for cluster centroids
- Voronoi cells
- Iterative relocation technique : The data points are assigned to clusters iteratively to improve the quality of clusters and minimize squared error within the cluster and maximize the separation between clusters
- Expectation-Maximization
- E step is to assign data point to cluster
- M step is to calculate the centroid

- Normalise it to stop one attribute from overweighting other (units are different)

Disadvantages:

- No of clusters should be known in advance (else trial and error)
- Sensitive to noise (different clusters for pure dataset and the one with noise)
- Different initialization values -> different clusters (because it settles in local optimum rather than global optimum)
- Spherical shape
- Gives focus to bigger clusters to minimize the variation

Applications:

- Market / customer segmentation
- Handwriting recognition
- Document clustering
- Image recognition

Evaluation Method:

a) Elbow method

- Choosing right k using SSE (sum of squared errors) where it forms an elbow and flattens out

b) Silhouette Analysis

- Degree of separation between clusters

a = intra cluster distance

b = inter cluster distance

Coefficient = $(b - a) / \max(a, b)$

If 0, diff clusters are near to each other

If 1, well apart and clearly distinguishable

If -1, overlapped