

# Q/A.

## 1. Linear Regression

Over View of Machine Learning

It is used to predict real values.  
(or) predicting output variable is "Continuous Data".  
a relationship is established between independent and dependent variable by fitting a best fit line.  
(or) Assuming relation between  $x$  &  $y$  is "Linear".  
The best fit line is known as "regression line".

Line represented as

$$y = mx + b$$

Annotations:  
-  $y$ : dependent variable  
-  $x$ : independent variable  
-  $m$ : slope  
-  $b$ : intercept

## 2. Logistic Regression

It is a classification algorithm used to estimate "Discrete Data values". based on given set of independent variables. its output value lies between [0 and 1] or [Yes/No]. Uses probability.

The Sigmoid function.

$$\frac{e^y}{1 + e^y}$$



### 3. Decision tree

It is Supervised learning algorithm Used for classification problems which works for both Categorical and Continuous dependent variables.  
In this algorithm, "population" is split" into two (or) more homogenous sets according to our

Given Conditions.

It uses different Techniques Like

- \* Gini impurity
- \* Information Gain
- \* Chi-Square Entropy

### 4. Random Forest

A Collection of Decision trees is called as "Random Forest." To classify a new object based on its attributes, Each Tree is classified and Take majority of votes for that class and classifies them. (Bagging Technique)  
applicable for both Classification & Regression



## 5. K-Nearest Neighbors

K-Nearest Neighbors is a simple algorithm that stores all available cases and classifies new cases by a majority vote of its "K-Neighbors". It is measured by distance function

\* Manhattan Distance

$$|x_2 - x_1| + |y_2 - y_1|$$

\* Euclidean Distance

$$\sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

## 6. Naive Bayes Theorem (algorithm)

Conditional Probability is based on "Bayes Theorem". It is supervised learning algorithm. Which is used for solving classification problems. Bayes theorem is used to determine probability of a hypothesis with prior knowledge and likelihood.

$$P(A/B) = \frac{P(B/A) * P(A)}{P(B)}$$

\* Naive :- it assumes that each input variable is independent.



## 7. SVM (Support Vector Machine)

it is Supervised Machine Learning algorithm.  
Used for both classification & Regression. Main  
Objective of SVM to find a hyperplane in a "N-dimen  
sional space that distinctly classifies data points.

(or)

SVM to find best line in two dimensions (or)  
The best hyperplane in more than two dimension in  
Order to help us separate our space into classes.

\* Linear Separable

\* Non-Linear Separable (or) kernel SVM

## 8. Boosting Techniques

Ensemble Learning (or) boosting Techniques  
it is Combination of "Weak Learners". that implies  
Combination of estimators (models) with an  
applied coefficient could act as effective Ensemble  
Estimator.

Boosting  
Formula :

$$F_T(x) = \sum_{t=1}^T f_t(x) = \alpha_t h(x)$$

Summation  
of  $t=1$  to " $T$ "

learning  
rate

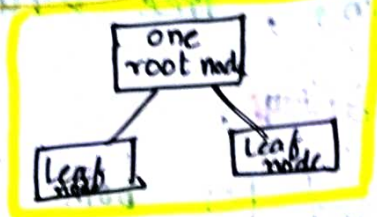
Each  
model



## 9. AdaBoost Algorithm

AdaBoost (adaptive Boosting) Ensemble method. it assigns weights to observations which are incorrectly predicted. Next model works to predict these values correctly which are wrongly predicted before. Each "Weak Learner" is called as "Stump".

• Depth = 1



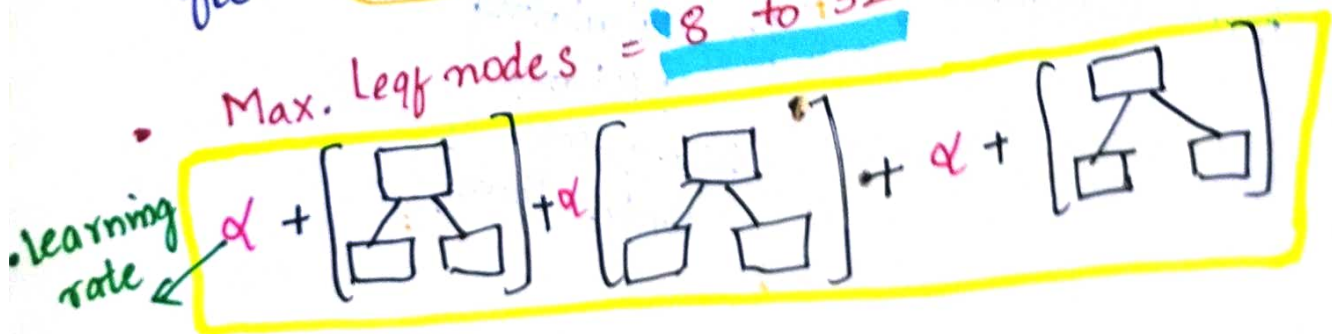
consists of

- 1 root nodes
- 2 leaf nodes

## 10. Gradient Boost

Gradient boost is used to predict both continuous and categorical variables. it is same as boosting techniques. Here we calculate the "Loss function" (single training error) and minimize in a sequential order. The sum of error is minimized and trees are adjusted after multiple times.

Max. Leaf nodes = 8 to 32





## 11. XGBoost

XGBoost also called as "Regularized Boosting". Like  $L_1$  &  $L_2$  regularization, it is "10" times faster than Gradient boost. it implement "parallel processing". it reduces The overfitting, Handle Missing Data, Inbuilt "Cross validation", Tree pruning using depth approach, High flexibility. it can be used for both "classification" and Regression problems.

## 12. K-Means algorithm

it is "Unsupervised" learning algorithm. (or) Unlabel data. K-means algorithm "identifies 'k' no. of Centroids", and then allocates every data point to nearest cluster, while keeping Centroids small as possible. it uses "Distance formula" same as K-nearest Neighbours.

Based on Assumptions.

↓  
For calculating Distance b/w points & cluster them.



### 13. Hierarchical clustering

it combines

Data points which are very

close by distance. That Groups Similar objects into Groups called clusters. This clustering involves creating clusters that are predetermined Ordering

from Top to bottom.

These are two of hierarchical clustering



- \* Agglomerative (From Top to bottom)
- \* Divisive (From bottom to Top)

### 14. DB-Scan clustering

DB-scan (Density Based spatial Clustering of

application with noise)

DB-scan clustering locates regions of "high density"

and Separate Outliers.

DB Scan is to find the neighborhoods of data points Exceeds certain Density threshold.

\* two parameters.

\*  $R$  (radius of neighborhood)

\*  $m$  (min. no. of neighbours)



## 15. Gradient descent algorithm

Gradient Descent is an "Optimization" algorithm for finding a Local minima (minimum point) of a differentiable function. (or) it Gradually reducing Weights and making Errors as Minimum value.

(or) Ex: By applying different " $\alpha$ " values and identifying the perfect " $\alpha$ " value, till we get error as "0".

Zero is called gradient "descent" minimize Cost Function

Update rule  
formula

$$y = m \times b$$

## 6. Stochastic Gradient Descent

it has both \* "Local Minima" and "Global minima".

Entire data We divide into parts (or) batches.

Each part we are identifying The "local minima" and from that Over all identify The "Global minima" is called "stochastic Gradient Descent".

Ex:- State ranker (local Minima)  
All india ranker (Global Minima)