

## U-2 Classification Methods



Support Vector Machine.

- supervised ML algo
- binary classification → goal is to separate data points into 2 classes
- Hyperplane → finds it that best separates.
- Support Vectors → datapoints closest to hyperplane
- Kernel Trick — SVM → handles nonlinear data by using kernel func.

• Multi Class Classifier.

• Sensitive to scale

• strengths → effective in high dimensional space.  
     ↳ no. of features > no. of samples  
 robust against → overfitting → when trained.

- Demerits → expensive for large datasets
- model interpretation → challenging.

Appl<sup>n</sup> → Image classif<sup>n</sup>, text categorization etc



Hyperplane.

decision boundary divides input space into two or more regions.

• directly proportional to no. of features



## \* - optimal separating hyperplanes -

- correctly classifies all data which being farthest away from decision boundary

## \* Kernel functions -

- used to take data as input & transfer into required form of processing data.
- set of mathematical functions used in support vector Machine. providing window to manipulate data
- returns inner product b/w two parts in standard feature dimension
- symmetric • smoothness • complexity • positive definiteness • non-negativity

## \* Kernel Selection -

choosing appropriate kernel kernel from using kernel methods.

- 1) Purpose.  
↓  
transform data in higher-dimensional space; to find complex, non-linear relationship
- 2) common kernels
  - ↳ Linear
  - ↳ Polynomial
  - ↳ Radial Basis Fun<sup>n</sup>
  - ↳ Sigmoid Kernel.
- 3) Kernel Selection
  - 4) Hyperparameter tuning.
  - 5) Domain Knowledge
  - 6) Experimentation
  - 7) Performance Evaluation



## Ensemble methods

- techniques → improve accuracy of results by combining multiple models instead of using single model.
- Ideal for regression & classification reduce bias & variance to boost accuracy of models.

Bagging

Boosting

Stacking

Reduce Variance

Reduce Bias

Improve accuracy

Base Learner Types

Homogeneous

Homogenous

Heterogeneous

Base Learner Training

Parallel

Sequential

Refinement

Aggregation

Max Voting averaging

Weighted Avg

Weighted Avg

~~Decision Tree~~

Eg Algo

Decision Tree

XGBoost



# \* Random Forest

- consists of multiple random decision tree -
- Two types of randomnesses are built into trees
  - 1) each tree is built on random sample from original data.
  - 2) each tree node  $\rightarrow$  subsets of feature are randomly selected to generate best split.
- greater no. of trees leads to higher accuracy & prevents overfitting
- takes less training time when compared to others
  - predicts output  $\rightarrow$  high accuracy  $\rightarrow$  even for large datasets
  - maintains accuracy  $\rightarrow$  when large portion of data is missing
  - performs both classification & regression task.
  - can handle large dataset with high dimensionality
  - used for both classification & regression is not suitable for regression task
- Demerits  $\rightarrow$  • Random Feature Selection

hyperparameter tuning  $\rightarrow$  optimize model performance

## \* Growing of Random Forest

$\rightarrow$  process of creating multiple decision trees

with ensemble -

- combines decision trees  $\rightarrow$  improve accuracy & reduce overfitting

- Steps  $\rightarrow$
- 1) Data preparation
  - 2) Bootstrapping
  - 3) Random Feature selection
  - 4) Decision Tree Building
  - 5) Parallelization
  - 6) Ensemble Creation
  - 7) Prediction Aggregation
  - 8) Out of Bag Error Estimation
  - 9) Feature Importance.
  - 10) Hyperparameter tuning



## \* Random Feature Selection -

Technique  $\rightarrow$  Random Forest

- $\rightarrow$  Prevents overfitting
- $\rightarrow$  Improves robustness
- $\rightarrow$  Increases accuracy

Steps  $\rightarrow$  1) Bootstrap Samples

2) Random Subset of Features

3) Diversity & Decorrelation

4) Preventing Dominance

5) Hyperparameter Control

6) Balancing Bias & Variance

7) Feature Importance