

Assignment - 1

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- ① (b) - Small K with noisy data
- ② (b) - Small changes in data lead to different trees
- ③ (c) - Reducing variance
- ④ (c) - All features are considered at each split
- ⑤ (a) - Target variable is categorical
- ⑥ (c) - Sigmoid
- ⑦ (c) - Accuracy
- ⑧ (d) - overfitting
- ⑨ (c) - Because distance calculation depends on the scale
- ⑩ (c) - Logistic Regression
- ⑪ overfitting in Decision Trees using depth as a parameter

Decision tree splits the data again and again to make decisions. If the tree depth is small, the model is simple.

if the tree depth is large the tree is noisy and small details

~~how~~ Bagging of Random forest address this problem differently

Bagging address this problem

Data is sampled randomly with replacement.

Each tree is trained on a slightly different dataset.
Final result is decided by Voting.

Random forest

It also uses different data samples. And also uses random features at each split.

~~Random Forest working in detail, including~~

- Bootstrap Sampling.

Random feature selection
Majority voting

Random forest is an ensemble learning algorithm that builds many decision trees and combines their

Result to make a final prediction.

① Bootstrap Sampling: From the original dataset, multiple new datasets are created. Data is selected randomly with replacement. Each new dataset is called a bootstrap sample.

② Random Feature Selection: When a tree is splitting a node, it does not use all features. Thus, they select a random subset of features.

③ Majority Voting: In the majority voting, it is the final result or prediction. The class with the maximum votes becomes the final output.

Q2 A fraud detection model produced the following results:

Predicted	Predicted Not
Fraud	Fraud
Actual fraud	120
Actual not fraud	50
	30
	800

- Calculate Accuracy
- Calculate Precision
- Calculate Recall
- Calculate F1 Score
- Is this model acceptable for fraud detection? Justify your answer.

	Predicted Fraud	Predicted Not Fraud
Actual Fraud	True Positive = 120	False Negative = 30
Actual Not Fraud	False Positive = 50	True Negative = 800
Total	TP = 120	
	FN = 30	
	FP = 50	
	TN = 800	
Total	1000	

a) Accuracy

$$A = \frac{TP + TN}{Total} = \frac{120 + 800}{1000} = \frac{920}{1000} = 0.92$$

Precision = $\frac{TP}{TP + FP}$

$$(b) P = \frac{TP}{TP + FP} = \frac{120}{120 + 50} = \frac{120}{170} = 0.706 \quad (70.6\%)$$

(c) Recall = $\frac{TP}{TP + FN} = \frac{120}{120 + 30} = \frac{120}{150} = 0.8$
 $= (80\%)$

(d) F1 Score = $\frac{2 \times \text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$

$$F1 = \frac{2 \times 0.706 \times 0.8}{0.706 + 0.8} = 0.75$$

(e) It's Acceptable for fraud detection.
 Because :- Accuracy is high (92%)

Recall is 80% \rightarrow 20% fraud cases are missed.