



## **Machine Learning Online**

**Assignment - 3: Naive Bayes, Decision Trees, and Ensemble Learning** 

## Part-I: Naive Bayes Classifier

## **Objective:**

To understand the probabilistic foundation of Naive Bayes classifiers and apply them to text and numerical datasets.

## **Assignment Tasks**

## **Task 1: Theory Questions**

Answer in 2–4 sentences:

- 1. What is the core assumption of Naive Bayes?
- 2. Differentiate between GaussianNB, MultinomialNB, and BernoulliNB.
- 3. Why is Naive Bayes considered suitable for high-dimensional data?

## Task 2: Spam Detection using MultinomialNB

- Load a text dataset (e.g., SMS Spam Collection or any public text dataset).
- Preprocess using CountVectorizer or TfidfVectorizer.
- Train a MultinomialNB classifier.
- Evaluate:
  - Accuracy
  - o Precision
  - o Recall
  - Confusion Matrix

#### Task 3: GaussianNB with Iris or Wine Dataset

- Train a GaussianNB classifier on a numeric dataset.
- Split data into train/test sets.
- Evaluate model performance.
- Compare with Logistic Regression or Decision Tree briefly.

#### **Part-II: Decision Trees**

## **Objective:**

To implement Decision Tree classifiers and understand their structure, splits, and overfitting characteristics.

## **Assignment Tasks**

## **Task 4: Conceptual Questions**

Answer briefly:

- 1. What is entropy and information gain?
- 2. Explain the difference between Gini Index and Entropy.
- 3. How can a decision tree overfit? How can this be avoided?

#### **Task 5: Decision Tree on Titanic Dataset**

- Load Titanic dataset (or use a similar dataset).
- Preprocess (handle missing values, encode categorical variables).
- Train a **DecisionTreeClassifier**.
- Visualize the decision tree using plot\_tree.
- Evaluate the model using accuracy and confusion matrix.

## Task 6: Model Tuning

- Use parameters like:
  - o max\_depth
  - o min\_samples\_split
- Show how performance changes.
- Plot training vs testing accuracy to visualize overfitting.

## Part-III: Ensemble Learning - Bagging, Boosting, Random Forest

## **Objective:**

To understand and implement ensemble techniques like Random Forest, AdaBoost, and Gradient Boosting for improving classification performance.

## **Assignment Tasks**

## **Task 7: Conceptual Questions**

#### Answer:

- 1. What is the difference between Bagging and Boosting?
- 2. How does Random Forest reduce variance?
- 3. What is the weakness of boosting-based methods?

#### **Task 8: Random Forest vs Decision Tree**

- Train a **RandomForestClassifier** on the same dataset used for Task 5.
- Compare accuracy, precision, and recall with the standalone decision tree.
- Plot feature importances.

## Task 9: AdaBoost or Gradient Boosting

- Train an AdaBoostClassifier or GradientBoostingClassifier.
- Use a suitable dataset.
- Compare it with Random Forest and Decision Tree in terms of:
  - Accuracy
  - o F1-score
  - o Training time (optional)

#### **Submission Guidelines**

- Submit:
  - Jupyter Notebook
  - o All CSV datasets used or links to source
  - o README with execution steps
- Use proper comments and section headers
- Upload it on Github.



# Coding Blocks | Online