### **Assignment: Ensemble Learning and Tree-Based Algorithms**

**Objective**: To understand and implement various tree-based algorithms and ensemble methods, evaluate their performance, and interpret their results.

#### Part 1: Decision Trees

- 1. **Dataset Selection**: Choose a dataset suitable for classification. This could be from UCI Machine Learning Repository, Kaggle, or any other source.
- 2. **Data Exploration**: Conduct a brief exploratory data analysis (EDA) to understand the dataset's features and target variable.
- 3. **Decision Tree Implementation**: Train a decision tree classifier. Visualize the tree and interpret the results.
- 4. **Evaluation**: Use appropriate metrics (e.g., accuracy, precision, recall, F1-score) to evaluate the model's performance on a test set.

### Part 2: Bagging

- 1. **Bootstrap Sampling**: From the dataset used in Part 1, create 10 bootstrap samples.
- 2. **Model Training**: Train a decision tree on each bootstrap sample.
- 3. **Aggregation**: Aggregate predictions from all trees to make the final prediction.
- 4. Evaluation: Compare the bagging ensemble's performance against the single decision tree from Part 1.

### **Part 3: Random Forest**

- 1. **Model Implementation**: Train a Random Forest classifier on the dataset.
- 2. **Feature Importance**: List the top 5 features based on their importance.
- 3. **Evaluation**: Compare the Random Forest's performance against the models from previous parts.

# **Part 4: Gradient Boosting**

- 1. **XGBoost Implementation**: Train an XGBoost classifier. Adjust hyperparameters like **learning\_rate**, **max\_depth**, and **n\_estimators**.
- 2. Feature Importance with XGBoost: Identify and visualize the importance of features in the trained model.
- 3. Evaluation: Compare the XGBoost model's performance against the other models.

# **Part 5: Comparative Analysis**

1. **Performance Metrics**: Tabulate the performance metrics (accuracy, precision, recall, F1-score) of all models side by side.

		accuracy	precision	recall	F1-score
	<b>Decision Trees</b>	0.85	0.85	0.85	0.85
	Bagging	0.85	0.85	0.85	0.85
	<b>Random Forest</b>	0.85	0.85	0.85	0.85
	<b>Gradient Boosting</b>	0.85	0.85	0.85	0.85

2. **Discussion**: Write a brief report discussing the performance of each model, situations where each model might be preferred, and any challenges faced during implementation.

### **Submission Guidelines:**

• Submit the code file (Jupyter notebooks) with detailed comments and visualizations like tree diagrams, feature importance plots, and performance metric graphs and metric TAB.