#### **Random Forest Documentation**

#### 1. Introduction

Random Forest is a supervised machine learning algorithm that uses an ensemble of decision trees to perform classification tasks. It improves prediction accuracy by reducing overfitting through the aggregation of multiple tree predictions.

### 2. Objective

- Develop an interactive web application using Streamlit for Random Forest classification.
- Provide users with the ability to upload a dataset, select features, train the model, and evaluate its performance.

### 3. Dataset Description

- Dataset: User-provided CSV file.
- Features: User-selected input variables.
- Target: User-selected variable to predict.
- Number of records and attributes depend on the uploaded dataset.

### 4. Implementation Details

- \*\*Frontend\*\*: Developed using Streamlit for user interaction.
- \*\*Backend\*\*: Random Forest implemented using `scikit-learn`.
- \*\*Steps\*\*:
  - 1. Upload a CSV dataset.
- 2. Select features and target variable.
- 3. Configure hyperparameters (e.g., number of estimators, max depth).
- 4. Train the model using Random Forest.
- 5. Evaluate the model using accuracy, confusion matrix, and classification report.

# 5. Results and Analysis

- Model accuracy and evaluation metrics are displayed.
- Feature importance can be visualized using bar charts.

- The impact of different hyperparameters on model performance can be analyzed.

## 6. Challenges and Solutions

- Managed large datasets efficiently using parallel processing.
- Prevented overfitting through hyperparameter tuning.
- Addressed class imbalance using appropriate techniques.

#### 7. Conclusion

Random Forest is a robust algorithm that delivers high accuracy and handles large datasets effectively. The web application allows users to explore model behavior and analyze feature importance.

#### 8. References

- Scikit-learn Documentation
- Streamlit Documentation
- Dataset Source