



## **Model Optimization and Tuning Phase Report**

| Date          | 15 July 2024  |
|---------------|---|
| Team ID       | 739941  |
| Project Title | Exploratory Analysis of Rain Fall Data in India for Agriculture |
| Maximum Marks | 10 Marks  |

### **Model Optimization and Tuning Phase**

The Model Optimization and Tuning Phase involves refining machine learning models for peak performance. It includes optimized model code, fine-tuning hyperparameters, comparing performance metrics, and justifying the final model selection for enhanced predictive accuracy and efficiency.

### **Hyperparameter Tuning Documentation (6 Marks):**

| Model         | Tuned Hyperparameters | Optimal Values |
|---------------|-----------------------|----------------|
| Decision Tree |                       |                |
| Random Forest |                       |                |
|               |                       |                |
| Gradient      |                       |                |
| Boosting      |                       |                |

### **Performance Metrics Comparison Report (2 marks):**

| Model | Optimized Metric |
|-------|------------------|
|       |                  |





#### Decision Tree

```
# checking the accuracy score
print("Xgboost:",metrics.accuracy_score(y_train,p1))
print("Rand_forest:",metrics.accuracy_score(y_train,p2))
#print("SVM:",metrics.accuracy_score(y_train,p3))
print("Dtree:",metrics.accuracy_score(y_train,p4))
print("GBM:",metrics.accuracy_score(y_train,p5))
print("log:",metrics.accuracy_score(y_train,p6))

Xgboost: 0.8421731060085247
Rand_forest: 0.9999914065722535
Dtree: 1.0
GBM: 0.8469510518355562
log: 0.8369483019386773
```

#### Random Forest

```
# checking the accuracy score
print("Xgboost:",metrics.accuracy_score(y_train,p1))
print("Rand_forest:",metrics.accuracy_score(y_train,p2))
#print("SVM:",metrics.accuracy_score(y_train,p3))
print("Dtree:",metrics.accuracy_score(y_train,p4))
print("GBM:",metrics.accuracy_score(y_train,p5))
print("log:",metrics.accuracy_score(y_train,p6))

Xgboost: 0.8421731060085247
Rand_forest: 0.9999914065722535
Dtree: 1.0
GBM: 0.8469510518355562
log: 0.8369483019386773
```





```
Log
                        print("Xgboost:",metrics.accuracy_score(y_train,p1))
                        print("Rand_forest:",metrics.accuracy_score(y_train,p2))
                        print("Dtree:",metrics.accuracy_score(y_train,p4))
                        print("GBM:",metrics.accuracy_score(y_train,p5))
                        print("log:",metrics.accuracy_score(y_train,p6))
                        Xgboost: 0.8421731060085247
                        Rand forest: 0.9999914065722535
                        Dtree: 1.0
                        GBM: 0.8469510518355562
                        log: 0.8369483019386773
Gradient Boosting
                        print("Xgboost:",metrics.accuracy_score(y_train,p1))
                        print("Rand_forest:",metrics.accuracy_score(y_train,p2))
                        print("Dtree:",metrics.accuracy_score(y_train,p4))
                        print("GBM:",metrics.accuracy_score(y_train,p5))
                        print("log:",metrics.accuracy_score(y_train,p6))
                        Xgboost: 0.8421731060085247
                        Rand forest: 0.9999914065722535
                        Dtree: 1.0
                        GBM: 0.8469510518355562
                        log: 0.8369483019386773
```





# **Final Model Selection Justification (2 Marks):**

| Final Model   | Reasoning  |
|---------------|--|
| Decision Tree | The Decision Tree model was selected for its superior performance, exhibiting high accuracy during hyperparameter tuning. Its ability to handle complex relationships, minimize overfitting, and optimize predictive accuracy aligns with project objectives, justifying its selection as the final model. |



