



Model Development Phase Template

Date	15 April 2024	
Team ID	Team-738164	
Project Title	Rainfall Prediction Using Machine Learning	
Maximum Marks	4 Marks	

Initial Model Training Code, Model Validation and Evaluation Report

Initial Model Training Code:

Logistic Regression

Baseline

```
In []: logreg = LogisticRegression(solver='liblinear', random_state=42)
logreg.fit(x_train, y_train)
    y_pred = logreg.predict(X_test)
    y_pred
Out[76]: array([0, 0, 0, ..., 0, 0, 0])
```

```
In []: from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay

def conf_matrix(model, X_test, y_test, cmap='Blues'):

    # Calculate confusion matrix
    y_pred = model.predict(X_test)
    cm = confusion_matrix(y_test, y_pred)

# Plot confusion matrix using ConfusionMatrixDisplay
    disp = ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=model.classes_)
    disp.plot(cmap='Blues')
    plt.grid()
    plt.show()

def roc_curve_custom(model, X_test, y_test):

# Generate predicted probabilities
    y_proba = model.predict_proba(X_test)

# Plot ROC curve using RocCurveDisplay
    disp = RocCurveDisplay.from_estimator(model, X_test, y_test)
    plt.plot([0, 1], [0, 1], color='black', linestyle='--')
    plt.show()
```





```
def evaluate(model, X_train=X_train, X_test=X_test, y_train=y_train, y_test=y_test, y_pred=y_pred):
    # Confusion Matrix'
    print('Confusion Matrix')
    print('\-'*53)
    conf_matrix(model, X_test, y_test)
    print('\n')

# Classification Report
    print('Classification Report')
    print('-**53)
    print(classification_report(y_test, y_pred))
    print('\n')

# ROC Curve
    print('NOC Curve')
    print('***93)
    roc_curve_custom(model, X_test, y_test)
    print('\n')

# Checking model fitness
    print('\n')

# Checking model fitness')
    print('\dots'*is3)
    print('\do
```

Random Forest

Baseline

```
In [ ]: rf = RandomForestClassifier(random_state=42)
    rf.fit(X_train, y_train)
        y_pred_rf = rf.predict(X_test)
        y_pred_rf

Out[102]: array([0, 0, 0, ..., 0, 0, 0])
In [ ]: evaluate(rf, y_pred=y_pred_rf)
```

Decision Tree

Baseline

XGBoost

Baseline

```
In [ ]: xgb = XGBClassifier(random_state=42)
    xgb.fit(X_train, y_train)
    y_pred_xgb = xgb.predict(X_test)
    y_pred_xgb

Out[111]: array([0, 0, 0, ..., 0, 0, 0])
In [ ]: evaluate(xgb, y_pred=y_pred_xgb)
```





Model Validation and Evaluation Report:

Model	Classification Report	Accuracy	Confusion Matrix
Logistic Regression (Baseline)	Classification Report precision recall f1-score support 0 0.52 0.80 0.86 2763 1 0.52 0.77 0.62 7926 accuracy 0.79 0.78 0.74 35549 macro avg 0.72 0.78 0.74 35549 weighted avg 0.83 0.79 0.80 35549	79%	Confusion Natrix - 20000 - 17500 - 15000 - 15000 - 12500 - 10000 - 10000 - 7500 - 5000 - 2500 - 2500 - 10000 - 2500
Random Forest (Baseline)	Classification Report precision recall fi-score support 0 0.87 0.95 0.91 27623 1 0.75 0.51 0.61 7926 accuracy 0.81 0.73 0.76 35549 macro avg 0.81 0.73 0.76 35549 weighted avg 0.85 0.85 0.84 35549	85%	Confusion Matrix - 25000 - 260290 - 1327 - 20000 - 15000 - 10000 - 10000 - 5000 0 Predicted label
Decision Tree (Baseline)	Classification Report precision recall f1-score support 0 0.87 0.86 0.86 27623 1 0.52 0.54 0.53 7926 accuracy 0.99 35549 macro avg 0.69 0.70 0.89 35549 weighted avg 0.79 0.79 0.79 35549	79%	Confusion Natrix - 22500 - 20000 - 17500 - 15000 - 12500 - 10000 - 10000 - 7500 - 5000 - 5000
XGBoost (Baseline)	Classification Report Precision recall f1-score support	86%	0 26090 1533 - 20000 - 15000 - 15000 - 10000 - 10000 - 5000 - 5000