

## Project Development Phase Model Performance Test

Date	10 November 2022
Team ID	PNT2022TMID44190
Project Name	Exploratory Analysis of Rainfall Data in India for Agriculture
Maximum Marks	10 Marks

### Model Performance Testing:

S.N o.	Parameter	Values	Screenshot
1.	Metrics	<p><b>Classification Model:</b> <b>Random Forest</b></p> <p><b>Confusion Matrix –</b> [[31372 1448] [ 4726 4691]]</p> <p><b>Accuracy Score–</b> 0.8538248455145963</p> <p><b>Classification Report –</b> Accuracy: 0.8538248455145963 Precision: 0.7641309659553673 Recall: 0.49814165870234683 F1-score: 0.6031113396760092</p>	<p><b>Random forest Confusion matrix</b></p> <pre>conf_matrix = metrics.confusion_matrix(y_test,t1)</pre> <pre>fig,ax = plt.subplots(figsize=(7.5,7.5)) ax.matshow(conf_matrix,alpha=0.3) for i in range(conf_matrix.shape[0]):     for j in range(conf_matrix.shape[1]):         ax.text(x=j, y=i, s=conf_matrix[i,j], va='center', ha='center',size='xx-large') plt.xlabel('Predictions',fontsize=18) plt.ylabel('Actuals',fontsize=18) plt.title('Confusion Matrix',fontsize=18) plt.show()</pre>  <pre>t1 = Rand_forest.predict(X_test_scaled)</pre> <pre>print("Rand_forest:",metrics.accuracy_score(y_test,t1))</pre> <p>Rand_forest: 0.8538248455145963</p>

			<pre>print("***10, "Classification Report", "***10)  print("-"*30) print(classification_report(y_test, t1)) print("-"*30)</pre> <p>***** Classification Report *****</p> <pre>-----               precision    recall  f1-score   support           0           0.87        0.96         0.91        32820          1           0.76        0.50         0.60         9417   accuracy          0.82        0.73         0.85        42237  macro avg          0.82        0.73         0.76        42237  weighted avg       0.85        0.85         0.84        42237  -----</pre>
2.	Tune the Model	Hyperparameter Tuning & Validation Method – RandomizedSearchCV	<h2>Hyperparameter Tuning</h2> <pre>from sklearn.ensemble import RandomForestRegressor rf = RandomForestRegressor(random_state = 42) from pprint import pprint # Look at parameters used by our current forest print('Parameters currently in use:\n') pprint(rf.get_params())</pre> <p>Parameters currently in use:</p> <pre>{'bootstrap': True,  'ccp_alpha': 0.0,  'criterion': 'mse',  'max_depth': None,  'max_features': 'auto',  'max_leaf_nodes': None,  'max_samples': None,  'min_impurity_decrease': 0.0,  'min_impurity_split': None,  'min_samples_leaf': 1,  'min_samples_split': 2,  'min_weight_fraction_leaf': 0.0,  'n_estimators': 100,  'n_jobs': None,  'oob_score': False,  'random_state': 42,  'verbose': 0,  'warm_start': False}</pre>

			<pre> n_estimators = [10,20,30,50] max_features = ['auto', 'sqrt'] max_depth = [int(x) for x in np.linspace(10, 50, num = 8)] min_samples_split = [4, 8, 10] min_samples_leaf = [2, 4, 6] bootstrap = [True, False] # Create the random grid random_grid = {'n_estimators': n_estimators,                'max_features': max_features,                'max_depth': max_depth,                'min_samples_split': min_samples_split,                'min_samples_leaf': min_samples_leaf,                'bootstrap': bootstrap}  from sklearn.model_selection import RandomizedSearchCV rf = RandomForestRegressor() rf_random = RandomizedSearchCV(estimator = rf,param_distributions = random_grid,r  rf_random.fit(X_train_scaled, y_train)  Fitting 5 folds for each of 100 candidates, totalling 500 fits RandomizedSearchCV(cv=5, estimator=RandomForestRegressor(), n_iter=100,                   n_jobs=-1,                   param_distributions={'bootstrap': [True, False],  'max_depth': [10, 15, 21, 27, 32, 38,   44, 50],  'max_features': ['auto', 'sqrt'],  'min_samples_leaf': [2, 4, 6],  'min_samples_split': [4, 8, 10],  'n_estimators': [10, 20, 30, 50]},                   random_state=35, verbose=2)  best_params = rf_random.best_params_  print ('Best Parameters is', best_params)  Best Parameters is {'n_estimators': 50, 'min_samples_split': 10, 'min_samples_l eaf': 6, 'max_features': 'sqrt', 'max_depth': 21, 'bootstrap': False}  print(f'Accuracy =: {round(rf_random.score(X_train_scaled, y_train) * 100, 2)}%')  Accuracy =: 75.07% </pre>
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