



Model Optimization and Tuning Phase Template

Date	14 July 2024
Team ID	SWTID1720174920
Project Title	Human Resource Management: Predicting Employee Promotions Using Machine Learning
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
DecisionTree	<pre>dt = DecisionTreeClassifier(criterion='entropy',max_depth=5,min_samples_split=10,</pre>	accuracy_score(y_test,y_pred_dt) 0.7149980055843638
Classifier	y_proj_descriptedectx_costy	
RandomForest	<pre>rf=RandomForestClassifier(n_estimators=100, max_depth=5, random_state=42) rf.fit(x_train,y_train) y_pred_rf=rf.predict(x_test)</pre>	accuracy_score(y_test,y_pred_rf) 0.7957718388512166
Classifier		
Kneighbours	<pre>knn=KNeighborsClassifier(n_neighbors=3, weights='uniform',algorithm='auto',leaf_size=10) knn.fit(x_train,y_train) y_pred_kn=knn.predict(x_test)</pre>	accuracy_score(y_test,y_pred_kn) 0.9032708416433984
Classifier		





Gradient Boosti

ng Classifier

 $$xg$-GradientBoostingClassifier(n_estimators=100, learning_rate=0.1, max_depth=3, random_state=42) xg. fit(x_train,y_train) $y_pred_xg=xg$.predict(x_test)$

accuracy_score(y_test,y_pred_xg)

0.864379736737136

Performance Metrics Comparison Report (2 Marks):

Model		Base	line M	etric		Optimized Metric					
DecisionTree Classifier	accuracy macro avg weighted avg confusion_matrix array([[9289, 7]	0.95 0.93 0.94 0.94	recall 0.93 0.95 0.94 0.94 pred)	f1-score 0.94 0.94 0.94 0.94	support 10035 10021 20056 20056 20056	y_pred_dt=decisionTree(x_train,x_test,y_train,y_test) DecisionTreeClassifier Confusion matrix [[6474 3561] [2155 7866]] Classification report					
RandomForest Classifier	accuracy macro avg weighted avg confusion_matrix(array([[9498, 53	0.95 0.95 0.95 0.95 0.95	0.95 0.95 0.95 0.95 red)	f1-score 0.95 0.95 0.95 0.95 0.95	support 10035 10021 20056 20056 20056	y_pred_rf=randomForest(x_train,x_test,y_train,y_test) RandomForestClassifier Confusion matrix [[7317 2718] [1378 8643]] Classification report					





						y pred kn=KNN(x	train.×	test.v tr	ain.v test)
	classification_r			,,	2111,9_0050	,				
	р	recision	recall	f1-score	support	Confusion matrix [[8451 1584]				
Vnoighbourg	0	0.96	0.82	0.89	10035	[356 9665]] Classification	report			
Kneighbours	1	0.84	0.97	0.90	10021	р	recision	recall	f1-score	support
Classifier	accuracy	0.00	0.00	0.90	20056	0	0.96	0.84	0.90	10035
Classifier	macro avg weighted avg	0.90 0.90	0.90 0.90	0.89 0.89	20056 20056	1	0.86	0.96	0.91	10021
I	confusion_matrix(y_te	est,y_pred)				accuracy			0.90	20056
1	array([[8242, 1793],					macro avg	0.91	0.90	0.90	20056
	[308, 9713]],	dtype=int64	.)			weighted avg	0.91	0.90	0.90	20056
	classification_re	<pre>y_pred_xg=xgboost(x_train,x_test,y_train,y_test)</pre>								
	precision recall f1-score support				support	Confusion matrix [[8409 1626]				
GradientBoosti	0	0.88	0.84	0.86	10035	[1094 8927]]				
	1	0.85	0.89	0.87	10021	Classification r	•			
ng						pr	ecision	recall	f1-score	support
8	accuracy			0.86	20056	0	0.88	0.84	0.86	10035
	macro avg	0.87	0.86	0.86	20056	1	0.85	0.89	0.87	10033
Classifier	weighted avg	0.87	0.86	0.86	20056	_	0.03	0.03	0.07	10021
	confusion_matrix(y_	test,y_pred)			accuracy			0.86	20056
	(550400 4505)					macro avg weighted avg	0.87 0.87	0.86 0.86	0.86 0.86	20056 20056
	array([[8409, 1626], [1094, 8927]], dtype=int64)					weighted avg	0.07	0.00	0.00	20036

Final Model Selection Justification (2 Marks):

Final Model	Reasoning
	RandomForest Classifier is selected due to its robust performance across various datasets without requiring extensive hyperparameter tuning. It effectively reduces overfitting by aggregating multiple decision trees, each trained on a random subset of features and
RandomForest Classifier	observations. This ensemble approach improves generalization and handles outliers and noisy data well.



