



Model Optimization and Tuning Phase Report

Model	Tuned Hyperparameters	Optimal Values
LOGISTICS REGRESSIO N	-	-





Random Forest	-		-
Date		06-06-2	024
Team ID		740055	
Project Title			TION OF PHISHING TE FROM URLS

Model Optimization and Tuning Phase

Maximum Marks

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.

10 Marks

Hyperparameter Tuning Documentation (6 Marks):





KNN	<pre>knn_classifier = KNeighborsClassifier() # Define the hyperparameters and their possible values for tuning param_grid = { 'n_neighbors': [3, 5, 7, 9], 'weights': ['uniform', 'distance'], 'p': [1, 2] }</pre>	<pre># Evaluate the performance of the tuned model accuracy = accuracy_score(y_test, y_pred) print(f'Optimal Hyperparameters: (best_params)') print(f'Accuracy on Test Set: {accuracy}') Optimal Hyperparameters: {'n_neighbors': 9, 'p': 1, 'weights': 'distance'} Accuracy on Test Set: 0.7218934911242604</pre>
Gradient Boosting	<pre># Define the Gradient Boosting classifier gb_classifier = GradientBoostingClassifier() # Define the hyperparameters and their possible values for tuning param_grid = { 'n_estimators': [50, 100, 200], 'learning_rate': [0.01, 0.1, 0.2], 'max_depth': [3, 4, 5], 'min_samples_split': [2, 5, 10], 'min_samples_leaf': [1, 2, 4], 'subsample': [0.8, 1.0] }</pre>	# Evaluate the performance of the toned model accuracy - accuracy (mode) (sets, /pret) print(f "total approxementers: (burstage)") print(f"eccuracy on lest Sets (burstage)") (Setsial Appropriates: ("lauracing rad" i &.1, "ans (depth": 5, "ain_samples_lesd": 2, 'bin_samples_point': 5, 'n_estimaters': 188, 'sambsemple': 8.5) Accuracy on lest Sets A. Tribbreon(1882))

Performance Metrics Comparison Report (2 Marks):

Model	Optimized Metric





<pre>print(classification_repo</pre>	rt(y_test,y_p	ored))		
	precision	recall	f1-score	support
Decision Tree Loan will be Approved Loan will not be Approved		0.68 0.73	0.68 0.74	75 94
accuracy macro avg weighted avg	0.71	0.71 0.71	0.71 0.71 0.71	169 169 169
confusion_matrix(y_test,y_ array([[51, 24],	_pred)			

	<pre>print(classification_repor</pre>	t(y_test,y_p	ored))		
		precision	recall	f1-score	support
Random Forest	Loan will be Approved	0.71	0.83	0.77	75
Tunia on Toron	Loan will not be Approved	0.84	0.73	0.78	94
	accuracy			0.78	169
	macro avg	0.78	0.78	0.77	169
	weighted avg	0.78	0.78	0.78	169
	confusion_matrix(y_test,y_	pred)			
	array([[62, 13], [25, 69]])				





	<pre>print(classification_repor</pre>	t(y_test,y_p	red))				
		precision	recall	f1-score	support		
KNN	Loan will be Approved	0.73	0.59	0.65	75		
	Loan will not be Approved	0.72	0.83	0.77	94		
	accuracy			0.72	169		
	macro avg weighted avg	0.72 0.72	0.71 0.72	0.71 0.72	169 169		
	weighted dvg		0.72	0172	103		
	confusion_matrix(y_test,y_	pred)					
	array([[44, 31],						
	[16, 78]])						
			18.8				
	<pre>print(classification_report(y_test,y_pred))</pre>						
		precision	recall	f1-score	support		
Gradient Boosting	process configuration are considered						
	Loan will be Approved	0.73	0.85		75		
	Loan will be Approved Loan will not be Approved	0.73 0.86	0.85 0.74		75 94		
	Loan will not be Approved accuracy	0.86	0.74	0.80 0.79	94 169		
	Loan will not be Approved accuracy macro avg			0.80 0.79 0.79	94		
	Loan will not be Approved accuracy	0.86 0.80	0.74	0.80 0.79 0.79	94 169 169		
	Loan will not be Approved accuracy macro avg	0.86 0.80 0.80	0.74	0.80 0.79 0.79	94 169 169		
	Loan will not be Approved accuracy macro avg weighted avg confusion_matrix(y_test,y) array([[64, 11],	0.86 0.80 0.80	0.74	0.80 0.79 0.79	94 169 169		
	Loan will not be Approved accuracy macro avg weighted avg confusion_matrix(y_test,y	0.86 0.80 0.80	0.74	0.80 0.79 0.79	94 169 169		
	Loan will not be Approved accuracy macro avg weighted avg confusion_matrix(y_test,y) array([[64, 11],	0.86 0.80 0.80	0.74	0.80 0.79 0.79	94 169 169		

Final Model Selection Justification (2 Marks):

Final Model	Reasoning





Gradient Boosting	The Gradient Boosting 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.
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