



Model Optimization and Tuning Phase Report

Date	15 July 2024
Team ID	739884
Project Title SmartLender -	Automotive Kickstart
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
Logistic Regression		<pre>pred=lr.predict(x_test) pred array([1, 1, 1,, 3, 1, 1]) lg_ac=accuracy_score(y_test,pred) lg_f1 = f1_score(y_test, pred, average='weighted') # Options: 'micro', 'macro', 'weighted' lg_r2r2_score(y_test,pred) print(lg_ac) print(lg_f1) print(lg_f1) print(lg_r2) 0.8842576462899887 0.7560252728477291 0.5764852951733073</pre>
RandomForest		kpred=knn.predict(x_test) kpred array([1, 3, 1,, 3, 3, 1]) knn_ac=accuracy_score(y_test,kpred) knm_f1 = f1_score(y_test, kpred, average='weighted') # Options: 'micro', 'mocro', 'weighted' knm_r2=r2_score(y_test,kpred) print(knn_r2) print(knn_f1) print(knn_f1) print(knn_r2) 0.82737326999383 0.7963496160215812 0.679318237664017

KNN	 rpred:rf.predict(x_test) rpred array([1, 1, 1,, 3, 3, 1]) rf.ac:accuracy_score(y_test,rpred) rf.fl = fl_score(y_test, rpred, average='weighted') # Options: 'micro', 'macro', 'weighted' rf.fr2r2, Score(y_test,rpred) print(rf.ac) print(rf.ac) print(rf.fr) print(rf.fr) 0.8595590294914034 0.8252275461008338 0.7824805763876369
SVM	 <pre>spred=rf.predict(x_test) spred array([1, 1, 1,, 3, 3, 1]) svm_ac=accuracy_score(y_test,spred) svm_f1 = f1_score(y_test, spred, average='weighted') # Options: 'micro', 'mocro', 'weighted' svm_r2=r2_score(y_test, rpred) print(svm_ac) print(svm_ac) print(svm_r2) 0.8595590249414934 0.82553275461898338 0.7824896763876369</pre>





Performance Metrics Comparison Report (2 Marks):

Model	Optimised Metric					
Logistic regression	<pre>print('\n\n', classification_report(y_test,pred))</pre>					
		precision	recall	f1-score	support	
	0	0.00	0.00	0.00	7766	
	1	0.75	0.95	0.84	39471	
	2	1.00	0.01	0.02	552	
	3	0.91	0.84	0.87	26831	
	4	0.00	0.00	0.00	351	
	accuracy			0.80	74971	
	macro avg	0.53	0.36	0.35	74971	
	weighted avg	0.73	0.80	0.76	74971	

Random Forest	<pre>print('\n\n', classification_report(y_test,kpred))</pre>			kpred))	
		precision	recall	f1-score	support
	0	0.22	0.07	0.11	7766
	1	0.81	0.91	0.86	39471
	2	0.12	0.01	0.01	552
	3	0.92	0.95	0.93	26831
	4	0.00	0.00	0.00	351
	accuracy			0.83	74971
	macro avg	0.41	0.39	0.38	74971
	weighted avg	0.78	0.83	0.80	74971
KNN	print('\n\n\n	', classifica	ation repo	ort(v test.	rpred))
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		precision	recall	f1-score	support
	0	0.28	0.07	0.12	7766
	1	0.83	0.94	0.88	39471
	2	0.26	0.03	0.06	552
	3	0.95	0.99	0.97	26831
	4	0.10	0.01	0.03	351
	accuracy			0.86	74971
	macro avg	0.49	0.41	0.41	74971
	weighted avg	0.81	0.86	0.83	74971
SVM	<pre>print('\n\n\n',</pre>	classificatio	on_report(y	_test,spred))
	ggle output scrolling	recision r	ecall f1-	score sup	port
		0.38	0.07	0.40 7	766
	0				766 471
	2				552
	3				831
	4	0.10	0.01	0.03	351
	accuracy			0.86 74	971
	macro avg	0.49			971
	weighted avg				971
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Final Model Selection Justification (2 Marks):

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