



## **Model Optimization and Tuning Phase Template**

Date	04 July 2024
Team ID	SWTID1720183095
Project Title	ECommerce Shipping Prediction Using Machine Learning
Maximum Marks	10 Marks

#### **Model Optimization and Tuning Phase**

Machine learning models are optimized and tuned during this phase to achieve optimal performance. To improve predicted accuracy and efficiency, it involves fine-tuning hyperparameters, comparing performance metrics, and defending the choice of the final model.

#### **Hyperparameter Tuning Documentation (6 Marks):**

Model	Tuned Hyperparameters	Optimal Values
SVM	c, kernel, gamma	1.0, rbf, 0.01
random forest	n_Estimators, Criterion, Max_Depth, Max_features	none,1e-9
KNN	n_neighbors, weights, algorithm, p	25, uniform, auto, 2
XGBoost	booster	gbtree





		Input layer: 16, 'random_uniform', 'relu'
ANN Units, kernel in	Units, kernel_initialiser, activation	First Hidden Layer:  16, 'random_uniform', 'relu'
	omes, kerner_minanser, activation	Second Hidden Layer: 8, 'random_uniform', 'relu'
		Output layer: 1, 'random_uniform', 'relu'

# **Performance Metrics Comparison Report (2 Marks):**

Model	:	(	Optim	ized I	Metric						
	Accuracy,F1 Score:					Accuracy,F1 Score:					
CVIV		precision	recall	f1-score	support	ţ	orecision	recall	f1-score	support	
SVM	0	0.56	0.89	0.68	895	0	0.56	0.88	0.69	895	
	1	0.87	0.51	0.65	1305	1	0.86	0.53	0.66	1305	
	accuracy			0.67	2200	accuracy			0.67	2200	
	macro avg	0.71	0.70	0.67	2200	macro avg	0.71	0.71	0.67	2200	
	weighted avg	0.74	0.67	0.66	2200	weighted avg	0.74	0.67	0.67	2200	





	Accurac	y,F1 S	core:			Accurac	y,F1 Sc	ore:		
random forest	Baseline Matr	ix: precision	recall	f1-score	support		precision	recall	f1-score	support
		ļ				0	0.58	0.69	0.63	895
	0	0.58			895	1	0.75	0.66	0.70	1305
	1	0.75	0.66	0.70	1305					
	accuracy			0.67	2200	accuracy	0.67	0.67	0.67	2200
	macro avg	0.66	0.67		2200	macro avg	0.67	0.67	0.67	2200
	weighted avg	0.68	0.67	0.67	2200	weighted avg	0.68	0.67	0.67	2200
	Accurac	y,F1 S	core:			Accurac	y,F1 Sc	ore:		
KNN										
IXIII	K-Nearest Neig	hbors (KNN)								
		precision	recall	f1-score	support	Optimized K-Nea				
	0	0.56	0.63	0.60	895	1	orecision	recatt 1	1-score s	upport
	1	0.73	0.66	0.69	1305	0	0.57	0.66	0.61	895
						1	0.74	0.66	0.70	1305
	accuracy	0.05	0.65	0.65	2200	accuracy			0.66	2200
	macro avg weighted avg	0.65 0.66	0.65 0.65	0.65 0.65	2200 2200	macro avg	0.66	0.66	0.66	2200
	weighted avg	0.00	0.03	0.03	2200	weighted avg	0.67	0.66	0.66	2200
	Accurac	y,F1 S	core:			Accurac	y,F1 Sc	ore:		
XGBoost	XGBoost Classifi	er:					precision	recall	f1-score	support
	pr	ecision r	ecall f1-	score supp	oort		,			
	۵	0.57	0.91	a 7a	205	0	0.56	0.88		895
	0 1	0.90	0.53		895 1305	1	0.86	0.53	0.66	1305
	accuracy	glas	-232		2200	accuracy			0.67	
	macro avg	0.73	0.72		2200	macro avg	0.71			
	weighted avg	0.76	0.69	0.68	2200	weighted avg	0.74	0.67	0.67	2200





	Accuracy,l	F1 Sco	re:			Accuracy,	F1 Sco	re:		
ANN	Classification Rep		ecall	f1-score	support	Classification p	Report:	recall	f1-score	support
	0	0.52 0.73	0.67 0.59	0.59 0.66	1312 1988	0 1	0.52 0.73	0.67 0.59	0.59 0.66	1312 1988
	accuracy macro avg weighted avg	0.63 0.65	0.63 0.63	0.63 0.62 0.63	3300 3300 3300	accuracy macro avg weighted avg	0.63 0.65	0.63 0.63	0.63 0.62 0.63	3300 3300 3300





### **Final Model Selection Justification (2 Marks):**

Final Model	Reasoning
XGBoost Classifier	The XGBoost Classifier model's better performance metrics led to its selection as the ultimate optimized model. It proved to be effective in creating accurate predictions, as evidenced by its highest accuracy of 69%.
	It also demonstrated a high precision score of 91.00%, demonstrating its consistency in accurately detecting true positives. The ensemble method used by XGBoost classifier reduces overfitting and enhances generalization to fresh data. Given these features, XGBoost is the best option for improving delivery time forecasts, which is in line with the project's goals.