



Model Optimization and Tuning Phase Template

Date	18 July 2024
Team ID	SWTID1720190389
Project Title	E-Commerce Shipping Prediction
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>param_grids = { 'LogisticRegression': { 'solver': ['liblinear', 'lbfgs'], 'C': [0.01, 0.1, 1, 10, 100] },</pre>	ACC> 0.6803092314688495 R2CV> 0.6884973017849729 MEAN SQUARED ERROR> 0.5581242677173489 ROC> (array([0. ,0.30954995, 1.]), array([0.
GaussianNB	<pre>param_grids = { 'GaussianNB': { 'var_smoothing': [1e-09, 1e-08, 1e-07, 1e-06, 1e-05] },</pre>	ACC> 0.6821282401091405 R2CV> 0.6712224989622251 MEAN SQUARED ERROR> 0.5733912286020558 ROC> (array([0. , 0.13391877, 1.]), array([0
KNN	param_grids = {	ACC> 0.6630286493860846 R2CV> 0.6652988792029887 MEAN SQUARED ERROR> 0.5785335952189908 ROC> (array([0. , 0.37211855, 1.]), arr





	'KNeighborsClassifier': { 'n_neighbors': [3, 5, 7, 9, 11], 'weights': ['uniform', 'distance'] },	
Decision Tree	<pre>param_grids = { 'DecisionTreeClassifier': { 'max_depth': [3, 5, 7, 9, None], 'min_samples_split': [2, 5, 10], 'min_samples_leaf': [1, 2, 4] },</pre>	ACC> 0.6912232833105957 R2CV> 0.6898713158987132 MEAN SQUARED ERROR> 0.556891986027171 ROC> (array[[0. , 0.09110867, 1.]), array([{
Random Forest	<pre>param_grids = { 'RandomForestClassifier': { 'n_estimators': [50, 100, 200], 'max_depth': [3, 5, 7, 9, None], 'min_samples_split': [2, 5, 10], 'min_samples_leaf': [1, 2, 4] },</pre>	ACC> 0.6966803092314688 R2CV> 0.6816791199667911 MEAN SQUARED ERROR> 0.5641993265089996 ROC> (array([0. , 0.09110867, 1.]), array
Gradient Boosting	<pre>param_grids = { 'GradientBoostingClassifier': { 'n_estimators': [50, 100, 200], 'learning_rate': [0.01, 0.1, 0.05], 'max_depth': [3, 5, 7, 9] },</pre>	ACC> 0.694406548431105 R2CV> 0.6775778331257782 MEAN SQUARED ERROR> 0.5678223021986911 ROC> (array([0. , 0.09659715, 1.]), arra
XGBoost	<pre>param_grids = { 'XGBClassifier': { 'n_estimators': [50, 100, 200], 'learning_rate': [0.01, 0.1, 0.05], 'max_depth': [3, 5, 7, 9] },</pre>	ACC> 0.700318326512051 R2CV> 0.6934952262349523 MEAN SQUARED ERROR> 0.5536287327849302 ROC> (array([0. , 0.07244786, 1.]), array
CatBoost	param_grids = {	ACC> 0.6998635743519782 R2CV> 0.6885014528850146 MEAN SQUARED ERROR> 0.5581205489094497 ROC> (array([0. , 0.08122942, 1.]), array([





```
'CatBoostClassifier': {
    'iterations': [50, 100, 200],
    'learning_rate': [0.01, 0.1, 0.05],
    'depth': [3, 5, 7, 9]
}
```

Performance Metrics Comparison Report (2 Marks):

Model	Optimized Metric					
	LogisticRegre	n_report		f1-score	support	
	0 1	0.60 0.75		0.64 0.71		
Logistic Regression	accuracy macro avg weighted avg confusion_mat [[629 282] [421 867]]				2199	
	GaussianNB:					
GaussianNB	0 1 accuracy	0.58 0.85 0.72 0.74	0.87 0.55	0.69	911 1288 2199	





	VNoighbaraCla	ssifie				
	KNeighborsCla	ssitler:				
classification_report>						
		precision	recall	f1-score	support	
	0			0.61		
	1	0.72	0.69	0.71	1288	
KNN	accuracy			0.66	2199	
	_	0.66	0.66	0.66	2199	
	weighted avg					
	confusion_mat	rix>				
	[[572 339]					
	[402 886]]					
	DecisionTreeC	lassifier:				
	3					
	classification	n_report> precision		f1 scane	support	
		precision	recall	T1-score	Support	
	0	0.58	0.91	0.71	911	
	1	0.89	0.54	0.67	1288	
Decision Tree						
	accuracy	0.74	0.70	0.69	2199	
	_	0.74			2199 2199	
	weighted avg	0.76	60.09	6.03	2133	
	confusion_matrix> [[828 83]					
	[596 692]]					





	1					
	RandomForestC	lassifier:				
	classification	n nenont>				
	crassification			f1 ccons	suppost	
		precision	recall	11-score	support	
	0	0.59	0.91	0.71	911	
	1	0.89	0.55		1288	
Random Forest		2.33	3.33	3.50		
Kandom Potest	accuracy			0.70	2199	
		0.74	0.73	0.70		
	weighted avg					
	confusion_mat	rix>				
	[[828 83]					
	[584 704]]					
	Chadianta	inaCl:C'				
	GradientBoost	ingciassitie	r:			
	classificatio	n renort>				
	C1033111C0C10	precision		f1-score	support	
		p. 20232011	. ccull	5007 6	-apport	
	0	0.58	0.90	0.71	911	
	1	0.89	0.55		1288	
Gradient Boosting						
22	accuracy			0.69	2199	
	_	0.74			2199	
	weighted avg	0.76	0.69	0.69	2199	
	confusion_mat	rix>				
	[[823 88]					
1	[584 704]]					





	XGBClassifier					
		•				
	classification_report>					
		precision	recall	f1-score	support	
	0	0.59	0.93	0.72	911	
	1	0.91	0.54	0.68	1288	
XGBoost						
Addoost	accuracy			0.70	2199	
	macro avg		0.73	0.70	2199	
	weighted avg					
	weighted dvg	0.70	0.70	0.70	2133	
	confusion_mat	niv>				
	[[845 66]	.I-IX>				
	[593 695]]					
	CatBoostClass	sifier:				
	classification	on_report>				
		precision	recall	f1-score	support	
	0	0.59	0.92	0.72	911	
	1	0.90	0.55	0.68	1288	
CatBoost						
Cuiboost	accuracy			0.70	2199	
	macro avg	0.75	0.73	0.70	2199	
	weighted avg	0.77	0.70	0.70	2199	
confusion_matrix>						
	[[837 74]					
[586 702]]						





Model Evaluation after Hyper Parameter Tuning

	Name	Accuracy	f1_score	Recall	Precision
0	LogisticRegression	68.03	71.15	67.31	75.46
1	GaussianNB	68.21	67.04	55.20	85.35
2	KNeighbors Classifier	66.30	70.51	68.79	72.33
3	DecisionTreeClassifier	69.12	67.09	53.73	89.29
4	Random Forest Classifier	69.67	67.86	54.66	89.45
5	${\sf Gradient Boosting Classifier}$	69.44	67.69	54.66	88.89
6	XGBClassifier	70.03	67.84	53.96	91.33
7	CatBoostClassifier	69.99	68.02	54.50	90.46

Final Model Selection Justification (2 Marks):

Final Model	Reasoning
Model 1 (or other)	Explanation of why this model was chosen as the final optimized model