



Model Optimization and Tuning Phase Template

Date	7 July 2024
Team ID	team-739896
Project Title	Identifying Airline Passenger Satisfaction 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
Decision Tree	<pre>from sklearn.model_selection import GridSearchCV param_grid = { 'criterion': ['gini', 'entropy'], 'max_depth': [None, 5, 10, 15], 'min_samples_split': [2, 5, 10], 'min_samples_leaf': [1, 2, 4] } tree = DecisionTreeClassifier() grid_search = GridSearchCV(estimator=tree, param_grid=param_grid,</pre>	<pre>grid_search= GridSearchCV(estimator= tree,param_gr grid_search=grid_search.fit(X_train,Y_train) print("Best accuracy=",grid_search.best_score_) print("Best parameters=",grid_search.best_params_) warnings.warn(Best accuracy= 0.9244474806826352 Best parameters= {'criterion': 'entropy',</pre>





Performance Metrics Comparison Report (2 Marks):

Model	Optimized Metric						
Decision Tree	Decision Tree						





		domForest cl			
	Model accura Accuracy in	Percentage	94.5%	f1-score	
		0.93 0.96			
Random Forest	accuracy macro avg weighted avg	0.95			4209
	cm=confusion_	_matrix(Y_te	st,Y_pred)		
	array([[1119,	48], 861]])			

Final Model Selection Justification (2 Marks):

Final Model	Reasoning
	The Random Forest 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
Random Forest	its selection as the final model.