



Model Optimization and Tuning Phase Template

Date	03-10-2024
Team ID	LTVIP2024TMID24892
Project Title	Liver Patient Identification – prediction of liver patient
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
Random Forest	<pre>rf_classifier = RandomForestClassifier() param_grid = { 'n_estimators': [50, 100, 200], 'criterion': ['gini', 'entropy'], 'max_depth': [None, 10, 20, 30], 'min_samples_split': [2, 5, 10], 'min_samples_leaf': [1, 2, 4], }</pre>	accuracy = accuracy_score(y_test, y_pres) print(f(optimal hyperparameters: (best_parame)') print(f'Accuracy on Test Set: {accuracy}') Optimal Hyperparameters: {'criterion': 'entropy', 'max_depth': None, 'min_samples_leaf': 1, Accuracy on Test Set: 8.8343849844585388
SVM	<pre>svm_classifier = svm.SVC() # Define the hyperparameters and their possible values param_grid = { 'kernel': ['linear', 'rbf', 'poly'], 'C': [0.1, 1, 10], 'gamma': ['scale', 'auto'] }</pre>	accuracy = accuracy_score(y_test, y_pred) print(f'Accuracy with Best Parameters: {accuracy}') Best Parameters: {'C': 10, 'gamma': 'scale', 'kernel': 'rbf'} Accuracy with Best Parameters: 0.77070063369426752





```
knn_classifier = KNeighborsClassifier()

# Define the hyperparameters and their possible values

param_grid = {

    'n_neighbors': [3, 5, 7, 9],

    'weights': ['uniform', 'distance'],

    'p': [1, 2]
}

knn_classifier = KNeighborsClassifier()

# Couracy = accuracy_score(y_test, y_pred)

print(f'Accuracy on Test Set: (accuracy)')

# Optimal Hyperparameters: ('n_neighbors': 3, 'p': 1, 'weights': 'distance')

Accuracy on Test Set: 0.777078635942676
```

Performance Metrics Comparison Report (2 Marks):

Model		Baseline	Metric, o	ptimal m	etrics	
	print(clas	ssification_	report(y_t	est,y_pred)))	
	Accuracy: 0.8	40764331 <i>2</i> 10	1911			
		precision		f1-score	support	
Random Forest	0	0.93	0.72	0.81	76	
Kandom i orest	2	0.79	0.95	0.86	81	
	accuracy			0.84		
	macro avg					
	weighted avg	0.86	0.84	0.84	157	
	nnint/al		nonont/v	toot v ppo	4))	
	print(cia	assification	_report(y_	test,y_pre	a))	
	Accuracy: 0.	73248407643	3121			
		precision	recall	f1-score	support	
SVM	ø	0.81	0.58	0.68	76	
S V IVI	2	0.69	0.88	0.77	81	
	accuracy			0.73	157	
	macro avg	0.75	0.73	0.72	157	
	weighted avg	0.75	0.73	0.73	157	





	<pre>print('Accuracy:', accuracy_score(y_test,y_pred)) print(classification_report(y_test,y_pred))</pre>				
	Accuracy: 0.7643	312101910	829		
	pr	ecision	recall	f1-score	support
KNN	0	0.87	0.61	0.71	76
	2	0.87 0.71			76 81
	_	0.,_	0.51	0.00	01
	accuracy			0.76	157
	macro avg	0.79	0.76	0.76	157
	weighted avg	0.79	0.76	0.76	157

Final Model Selection Justification (2 Marks):

print(classification_report(y_test,y_pred)) Accuracy: 0.8407643312101911	Final Model	Reasoning
confusion_matrix(y_test,y_pred) array([[55, 21],	Random Forest	Accuracy: 0.8407643312101911

NOTE: I have done other models like Gradient Boosting Classifier, AdaBoost Classifier these model will be available in the lliver.ipynb file.



