



Model Optimization and Tuning Phase Template

Date	13 july 2024
Team ID	739952
Project Title	Prediction and Analysis of Liver Patient Data 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
Logistic		
Regression	<pre>from sklearn.linear_model import LogisticRegression lr = LogisticRegression(random_state=42) lr.fit(x_train, y_train)</pre>	
	LogisticRegression LogisticRegression(random_state=42)	<pre>lr_acc = accuracy_score(y_pred_ir, y_test) lr_acc. a.7606837636837686</pre>





K neighbors		
Classifier	<pre>free sklearn.neighbors impert KNeighborsClassifier kon=KNeighborsClassifier(n_neighbors=6, weights='unifurm', algorithm='bd_true', lasf_siz=28)</pre>	
	kon.fit(x_train,y_train)	accuracy score(y test,y pred)
	- KNeighborsClassifier	
	KNeighborsClassifier(algorithm='kd_tree', leaf_size=20, n_neighbors=6)	0.7692307692307693
RandomForest		
Classifier	rfwEardowForestClassifier(n_estimator=v500, criterion="entrupy",rendom_stater(E) rf.fit(n_train,y_train)	
	s fundosforestflassifiar	accuracy_score(y_test,y_pred)
	RandomForestClassifier(criterions'entropy', n_estimators=300, random_state=18)	0.7606837606837606
SVC		
	<pre>model = SVC(kerrol**rbf*,rockom_state=100,gamma*_auts*,verbose=1,decisiom_functiom_shape*_mos*)</pre>	
	model_fit(x_tnein,y_tnein)	
	[TIDOM]	accuracy_score(pred,y_test)
	SMC(Secision_function_shape='ovo', gamma='auto', random_statu=180, varbose=2)	0.7808219178082192

Performance Metrics Comparison Report (2 Marks):

Model	Baseline Metric	Optimized Metric





Logistic										
Regression						print(classific	stion reco	ntív test	.v pred ir	W
	print(classific	ation_repo	ert(y_test,	y_prud))			recision	- 07.6	f1-score	support
	p	rectator	recall	fi-score	support		Trebuil	1-1-011-1	Variotis	ang-port
	1 2	0.75 8.45	0.91 0.19	0.83 0.27	128 47	1 2	0.56	0.92 0.30	0.85 0.39	87 30
	accuracy macro avg weighted avg	8.60 0.67	0.55 0.72	8.72 8.55 8.68	175 175 175	accuracy macro avg weighted avg	0.68 0.73	0.61 0.76	0.76 0.62 0.73	117 117 117
	confusion_matri	x(y_test,y	_pred_lr)							
	[[117 11] [38 9]]					array([[88, 7] [21, 9]	,], dtype=i	nt64)		
K neighbors										
Classifier							print(classification_report(y_test,y_pred))			
	print(classific	ation_repo	rt(y_test,	ypned_kon)	F	p	recision	recall	f1-score	support
		recision	recall	f3-score	support	1	0.77	0.99	0.86	86
	i	0.81 0.42	8.88 8.43	0.88 8.43	109 37	2	0.83	0.16	0,27	31
	accuracy			8.71	146	accuracy macro avg	88.6	0.57	0.77	117
	macro avg weighted avg	0.61 0.71	0.71	8.61 8.71	146 148	weighted avg	0.78	0.77	0.71	117
	confusion_matri	н(y_test,)	gred_knn)			confusion_matri	c(y_test,y	_pred)		
	acray([[87, 22] [21, 26]], dtype-i	at64)			array([[85, 1] [26, 5]	, dtype=i	int64)		
RandomForest										
Classifier						print(classific	ation repo	et(y test	y pred))	
	print(classifi	cation_re	port(y_tes	t,ypred_m	fc))		recision		fl-score	support
	- 1	precision	recall	F1-score	e support					
	1 2	0.88 0.46	9.85 8.37			1 2	0.82 0.54	0.87 0.43	0.84 0.48	87 30
	accuracy			0.7	117	accuracy	100	11 (2015)28	0.76	117
		0.63		0.6	1 117	macro avg	0.68 8.75	0.65 0.76	8,75	117 117
	matro avg weighted avg	0.71	0.73	8.7	6. 347					
	matro avg	0.71			***	confusion_matri	x(y_test,y	_pred)		





SVC						
						classification_report(pred,y_text)
	print(classificat	tion_repor	t(y_test	y_pred_sw	033	144441 1444111 1444 1 (b. 144) Tearly
	pro	ecision	recall	f1-score	support.	(100)
	1 2	0.74 0.88	1.00 0.00	8.85 8.86	117 38	precision recall fi-score support\n\n 1
	accuracy macro aug weighted aug	8.37 8.55	0.58 0.74	0.74 0.43 0.63	117 117 117	0.78 106\n macro avg 8.50 0.20 0.64 160\maclghtmd avg 1.00 0.78 8.68 146\n'
	confusion_matrix	(y_test,y_	pred_sviii)		confusium_matrix(pred,y_test)
	array([[87, 8], [30, 8]]		rtf43			

Final Model Selection Justification (2 Marks):

Final Model	Reasoning					
	SVC is selected as for its Effective in High-Dimensional Spaces, Robust to					
	Overfitting handle both linear and non-linear classification problems by employing kernel functions, making it a versatile and powerful tool for a wide range of applications					
SVC						