



Model Development Phase Template

Date	30 June 2024
Team ID	-
Project Title	Golden Harvesting: A Predictive Model For Apple Quality Assurance
Maximum Marks	4 Marks

Initial Model Training Code, Model Validation and Evaluation Report

Initial Model Training Code:

```
# Decision tree model
from sklearn.tree import DecisionTreeClassifier
model1 = DecisionTreeClassifier()
model1.fit(x_train, y_train)
dt_pred = model1.predict(x_test)
acc_score = accuracy_score(y_test, dt_pred)
print("Accuracy of Decision tree is %.2f"%accuracy_score(y_test, dt_pred))
print("Confusion matrix of Decision tree is as follows:")
confusion_matrix(y_test, dt_pred)
```

```
# Random forest model
from sklearn.ensemble import RandomForestClassifier
model2 = RandomForestClassifier(n_estimators=100)
model2.fit(x_train, y_train)
rf_pred = model2.predict(x_test)
acc_score = accuracy_score(y_test, rf_pred)
print("Accuracy of Random forest is %.2f"%accuracy_score(y_test, rf_pred))
print("Confusion matrix of Random forest is as follows:")
confusion_matrix(y_test, rf_pred)
```





```
# XGBoost model
import xgboost as xgb
model3 = xgb.XGBClassifier()
model3.fit(x_train, y_train)
xgb_pred = model3.predict(x_test)
acc_score = accuracy_score(y_test, xgb_pred)
print("Accuracy of XGBoost is %.2f"%accuracy_score(y_test,xgb_pred))
print("Confusion matrix of XGBoost is as follows:")
confusion_matrix(y_test, xgb_pred)
```

```
# Logistic Regression Model
from sklearn.linear_model import LogisticRegression
model4 = LogisticRegression()
model4.fit(x_train, y_train)
lr_pred = model4.predict(x_test)
acc_score = accuracy_score(y_test, lr_pred)
print("Accuracy of Logistic Regression is %.2f"%accuracy_score(y_test, lr_pred))
print("Confusion matrix of Logistic Regression is as follows:")
confusion_matrix(y_test, lr_pred)
```

```
# Support Vector Machine model.
from sklearn.svm import SVC
model5 = SVC()
model5.fit(x_train,y_train)
svm_pred = model5.predict(x_test)
acc_score = accuracy_score(y_test, svm_pred)
print("Accuracy of Support Vector Machine is %.2f"%accuracy_score(y_test, svm_pred))
print("Confusion matrix of Support Vector Machine is as follows:")
confusion_matrix(y_test, svm_pred)
```

```
# K Nearest Neighbor model
from sklearn.neighbors import KNeighborsClassifier
model6 = KNeighborsClassifier()
model6.fit(x_train,y_train)
knn_pred = model6.predict(x_test)
acc_score = accuracy_score(y_test, knn_pred)
print("Accuracy of K Nearest Neighbor is %.2f"%accuracy_score(y_test, knn_pred))
print("Confusion matrix of K Nearest Neighbor is as follows:")
confusion_matrix(y_test, knn_pred)
```





```
# Naive Bayes model
from sklearn.naive_bayes import GaussianNB
model7 = GaussianNB()
model7.fit(x_train,y_train)
nb_pred = model7.predict(x_test)
acc_score = accuracy_score(y_test, nb_pred)
print("Accuracy of Naive Baye is %.2f"%accuracy_score(y_test, nb_pred))
print("Confusion matrix of Naive Baye is as follows:")
confusion_matrix(y_test, nb_pred)
```

Model Validation and Evaluation Report:

Model	Classification Report					Accuracy	Confusion Matrix
	print(classi	fication_rep	ort(dt_pr	ed, y_test))		
		precision	recall	f1-score	support		
Decision tree model	9	0.79	0.80	0.80	204		Confusion matrix of Decision tree
	1	0.79	0.79	0.80	394 406	81%	array([[317, 84],
	+:	0.01	0.73	0.00	400		[77, 322]], dtype=int64)
	accuracy			0.80	800		[//, 322]], dtype=int64)
	macro avg	0.80	0.80	0.80	899		
	weighted avg	0.80	0.80	0.80	800		
	print(classi	fication_rep	ort(rf_pr	ed,y_test))		
		precision	recall	f1-score	support		
Random	0	0.89	0.92	0.91	386		Confusion matrix of Random forest
forest model	1	0.93	0.89	0.91	414	91%	array([[357, 44],
Totest model	77.0	10000	(50,000)	0100	1070000		[29, 370]], dtype=int64)
	accuracy			0.91	800		[25, 570]], despe-2.1004/
	macro avg	0.91	0.91	0.91	899		
	weighted avg	0.91	0.91	0.91	800		
	print(classi	fication_rep	ort(xgb_p	red, y_test	:))		
		precision	recall	f1-score	support		
XGBoost			0.01	0.00	****		Confusion matrix of XGBoost
model	9	0.91 0.91	0.91 0.91	0.91 0.91	400 400	91%	array([[363, 38], [37, 362]], dtype=int64
model	L	0.74	0.51	0.91	400		
	accuracy			0.91	800		[37, 302]], dcype=Into4
	macro avg	0.91	0.91	0.91	800	1	





	print(classif	fication_rep	ort(lr_pr	ed,y_test)))		
T		precision	recall	f1-score	support		
Logistic Regression	0	0.76	0.75	0.75	402	75%	Confusion matrix of Logistic Regression
Model	1	0.75	0.75	0.75	398	73%	array([[303, 98],
Model	accuracy			0.75	800		[99, 300]], dtype=int64)
	macro avg	0.75	0.75	0.75	800		
	weighted avg	0.75	0.75	0.75	800		
	print(classif	fication rep	ort(svm_p	red,y test))		
	070	precision	recall	f1-score	support		
Support				ACT CHARGE			AND COMES PROPERTY AND ADDRESS
Vector	9	0.91	0.90	0.90	402	90%	Confusion matrix of Support Vector Machine
Machine	1	0.90	0.90	0.90	398		array([[363, 38], [39, 360]], dtype=int64)
model	accuracy			0.90	888		
	macro avg	0.90	0.90	0.90	800		
	weighted avg	0.90	0.90	0.90	800		
	weighted avg		ort(knn_p				
K Nearest	print(classif	fication_rep precision	ort(knn_p recall	red,y_test) f1-score) support		
K Nearest Neighbor	print(classif	fication_rep precision 0.90	ort(knn_p recall 0.90	red,y_test) f1-score 0.90) support	90%	
K Nearest Neighbor model	print(classif	fication_rep precision	ort(knn_p recall	red,y_test) f1-score) support	90%	array([[360, 41],
Neighbor	print(classif	fication_rep precision 0.90	ort(knn_p recall 0.90	red,y_test) f1-score 0.90) support	90%	
Neighbor	print(classif	fication_rep precision 0.90	ort(knn_p recall 0.90	red,y_test) f1-score 0.90 0.90) support 398 402	90%	
Neighbor	print(classif	fication_rep precision 0.90 0.90	ort(knn_p recall 0.90 0.90	red,y_test) f1-score 0.90 0.90) support 398 402 800	90%	array([[360, 41],
Neighbor	print(classif	fication_rep precision 0.90 0.90 0.90 0.90	ort(knn_p recall 0.90 0.90 0.90	red,y_test) f1-score 0.90 0.90 0.90 0.90 0.90	support 398 402 800 800 800	90%	array([[360, 41],
Neighbor	print(classif	fication_rep precision 0.90 0.90 0.90 0.90	ort(knn_p recall 0.90 0.90 0.90 0.90	red,y_test) f1-score 0.90 0.90 0.90 0.90 0.90	support 398 402 800 800 800	90%	array([[360, 41],
Neighbor model	print(classif	fication_rep precision 0.90 0.90 0.90 fication_rep	ort(knn_p recall 0.90 0.90 0.90 0.90	f1-score 0.90 0.90 0.90 0.90 0.90 0.90 f1-score	support 398 402 800 800 800		array([[360, 41],
Neighbor	print(classif	fication_rep precision 0.90 0.90 0.90	ort(knn_p recall 0.90 0.90 0.90 0.90	f1-score 0.90 0.90 0.90 0.90 0.90 f1-score 0.77	support 398 402 800 800 800	90%	array([[360, 41], [38, 361]], dtype=int64) Confusion matrix of Naive Baye
Neighbor model	print(classif	fication_rep precision 0.90 0.90 0.90 fication_rep precision 0.78	0.76	f1-score 0.90 0.90 0.90 0.90 f1-score 0.77 0.76	support 398 402 800 800 800 support 413 387		array([[360, 41], [38, 361]], dtype=int64)
Neighbor model	print(classif	fication_rep precision 0.90 0.90 0.90 0.70 fication_rep precision 0.78 0.75	0.76 0.77	f1-score 0.90 0.90 0.90 0.90 f1-score 0.77 0.76	support 398 402 800 800 support 413 387 800		array([[360, 41],
Neighbor model	print(classif	fication_rep precision 0.90 0.90 0.90 fication_rep precision 0.78	0.76	f1-score 0.90 0.90 0.90 0.90 f1-score 0.77 0.76	support 398 402 800 800 800 support 413 387		array([[360, 41],