

## Model Development Phase Template

Date	July 2024
Team ID	Team-739662
Project Title	Ecommerce Shipping Prediction using Machine Learning
Maximum Marks	10 Marks

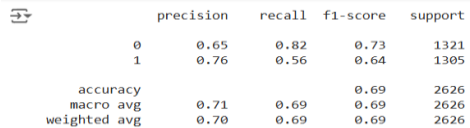
### Initial Model Training Code, Model Validation and Evaluation Report

The initial model training code will be showcased in the future through a screenshot. The model validation and evaluation report will include a summary and training and validation performance metrics for multiple models, presented through respective screenshots.

#### Initial Model Training Code (5 marks):

Paste the screenshot of the model training code

#### Model Validation and Evaluation Report (5 marks):

Model	Summary	Training and Validation Performance Metrics
Model 1	Logistic regression model typically include accuracy, precision, recall, F1 score to evaluate its predictive performance and generalization capability.	<p>LOGISTIC REGRESSION</p> <pre>[ ] #LOGISTIC REGRESSION # importing the library from sklearn.linear_model import LogisticRegression #initializing the model lr=LogisticRegression() #fit the model lr.fit(x_train,y_train) #predict the model predic=lr.predict(x_test) #finding accuracy,classification report from sklearn.metrics import classification_report print(classification_report(y_test,predic))</pre>  <pre>               precision    recall  f1-score   support      0       0.65       0.82       0.73       1321     1       0.76       0.56       0.64       1305   accuracy          0.69       2626  macro avg         0.71       0.69       0.69       2626  weighted avg      0.70       0.69       0.69       2626 </pre>


Model 2	Decision tree classifier model commonly include accuracy, precision, recall, F1 score which help assess the model's prediction accuracy and generalizability	<pre>#decision tree classifier from sklearn.tree import DecisionTreeClassifier dec=DecisionTreeClassifier() dec.fit(x_train,y_train) predi=dec.predict(x_test) from sklearn.metrics import classification_report print(classification_report(y_test,predi))</pre> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>0</td><td>0.71</td><td>0.69</td><td>0.70</td><td>1321</td></tr><tr><td>1</td><td>0.69</td><td>0.71</td><td>0.70</td><td>1305</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.70</td><td>2626</td></tr><tr><td>macro avg</td><td>0.70</td><td>0.70</td><td>0.70</td><td>2626</td></tr><tr><td>weighted avg</td><td>0.70</td><td>0.70</td><td>0.70</td><td>2626</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.71	0.69	0.70	1321	1	0.69	0.71	0.70	1305	accuracy			0.70	2626	macro avg	0.70	0.70	0.70	2626	weighted avg	0.70	0.70	0.70	2626
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Model 3	Random forest classifier model often encompass accuracy, precision, recall, F1 score to measure its prediction quality and robustness.	<p>RANDOM FOREST CLASSIFIER</p> <pre>[45] #random forest classifier from sklearn.ensemble import RandomForestClassifier rfc=RandomForestClassifier() rfc.fit(x_train,y_train) pred=rfc.predict(x_test) from sklearn.metrics import classification_report print(classification_report(y_test,pred))</pre> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>0</td><td>0.69</td><td>0.87</td><td>0.77</td><td>1321</td></tr><tr><td>1</td><td>0.82</td><td>0.61</td><td>0.70</td><td>1305</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.74</td><td>2626</td></tr><tr><td>macro avg</td><td>0.76</td><td>0.74</td><td>0.73</td><td>2626</td></tr><tr><td>weighted avg</td><td>0.76</td><td>0.74</td><td>0.73</td><td>2626</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.69	0.87	0.77	1321	1	0.82	0.61	0.70	1305	accuracy			0.74	2626	macro avg	0.76	0.74	0.73	2626	weighted avg	0.76	0.74	0.73	2626
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Model 4	K-nearest neighbors classifier model typically include accuracy, precision, recall, F1 score to evaluate its prediction performance and generalization ability	<p>K-NEAREST NEIGHBORS</p> <pre>[48] #knn from sklearn.neighbors import KNeighborsClassifier knn=KNeighborsClassifier() knn.fit(x_train,y_train) p=knn.predict(x_test) from sklearn.metrics import classification_report print(classification_report(y_test,p))</pre> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>0</td><td>0.69</td><td>0.79</td><td>0.74</td><td>1321</td></tr><tr><td>1</td><td>0.75</td><td>0.65</td><td>0.69</td><td>1305</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.72</td><td>2626</td></tr><tr><td>macro avg</td><td>0.72</td><td>0.72</td><td>0.72</td><td>2626</td></tr><tr><td>weighted avg</td><td>0.72</td><td>0.72</td><td>0.72</td><td>2626</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.69	0.79	0.74	1321	1	0.75	0.65	0.69	1305	accuracy			0.72	2626	macro avg	0.72	0.72	0.72	2626	weighted avg	0.72	0.72	0.72	2626
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Model 5	XGBoost classifier model typically include accuracy, precision, recall, F1 score used to evaluate the model's predictive performance and ability to generalize	<p>XGBOOST CLASSIFIER</p> <pre>[49] #XGBOOST CLASSIFICATION from xgboost import XGBClassifier xg=XGBClassifier() xg.fit(x_train,y_train) p=xg.predict(x_test) from sklearn.metrics import classification_report print(classification_report(y_test,p))</pre> <table><thead><tr><th></th><th>precision</th><th>recall</th><th>f1-score</th><th>support</th></tr></thead><tbody><tr><td>0</td><td>0.71</td><td>0.80</td><td>0.75</td><td>1321</td></tr><tr><td>1</td><td>0.77</td><td>0.66</td><td>0.71</td><td>1305</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.73</td><td>2626</td></tr><tr><td>macro avg</td><td>0.74</td><td>0.73</td><td>0.73</td><td>2626</td></tr><tr><td>weighted avg</td><td>0.74</td><td>0.73</td><td>0.73</td><td>2626</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.71	0.80	0.75	1321	1	0.77	0.66	0.71	1305	accuracy			0.73	2626	macro avg	0.74	0.73	0.73	2626	weighted avg	0.74	0.73	0.73	2626
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Model 6

Ridge classifier model typically include accuracy, precision, recall, F1 score, and mean squared error to evaluate its prediction performance and generalization.

RIDGE CLASSIFIER

```
[50] #RIDGE CLASSIFIER
from sklearn.linear_model import RidgeClassifier
rg=RidgeClassifier()
rg.fit(x_train,y_train)
p=rg.predict(x_test)
from sklearn.metrics import classification_report
print(classification_report(y_test,p))
```



	precision	recall	f1-score	support
0	0.65	0.80	0.72	1321
1	0.74	0.56	0.64	1305
accuracy			0.68	2626
macro avg	0.69	0.68	0.68	2626
weighted avg	0.69	0.68	0.68	2626