

## Model Development Phase Template

Date	04 June 2024
Team ID	SWTID1720183095
Project Title	Ecommerce Shipping Prediction Using Machine Learning
Maximum Marks	4 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 classification reports, accuracy, and confusion matrices for multiple models, presented through respective screenshots.

### Initial Model Training Code:

```
0]: #supportvectormachine
svm_model = svm.SVC(gamma='auto',C=5,kernel='rbf')
svm_model.fit(X_train,y_train)
y_pred = svm_model.predict(X_test)
```

```
32]: #randomforestclassifier
params = {'n_estimators':[100,150], 'criterion':['gini', 'entropy']}
#Hyper parameter tuning
rf_model =GridSearchCV(estimator=RandomForestClassifier(),param_grid=params,scoring='accuracy', cv=5)
rf_model = rf_model.fit(X_train,y_train)
y_pred=rf_model.predict(X_test)
```

```
34]: #artificialneuralnetwork
ann = Sequential()
ann.add(Dense(14,input_dim=8,activation='relu'))
ann.add(Dense(8,activation='relu'))
ann.add(Dense(8,activation='relu'))
ann.add(Dense(1,activation='sigmoid'))
ann.compile(loss="binary_crossentropy", optimizer='SGD',metrics=['accuracy'])
```

```
1]: # Logistic Regression
logreg_model = LogisticRegression()
logreg_model.fit(X_train, y_train)
y_pred = logreg_model.predict(X_test)
print("Logistic Regression:")
```

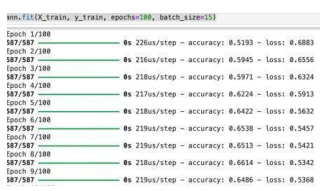
```
1]: # XGBoost Classifier
params = {
    'objective': 'binary:logistic',
    'max_depth': 3,
    'learning_rate': 0.1,
    'n_estimators': 100
}
xgb_model = xgb.XGBClassifier(**params)
xgb_model.fit(X_train, y_train)
y_pred = xgb_model.predict(X_test)
print("XGBoost Classifier:")
```

```
# K-Nearest Neighbors (KNN) Classifier
from sklearn.neighbors import KNeighborsClassifier

knn_model = KNeighborsClassifier()
knn_model.fit(X_train, y_train)
y_pred = knn_model.predict(X_test)
print("K-Nearest Neighbors (KNN) Classifier:")
```

### Model Validation and Evaluation Report:

Model	Classification Report	Accuracy	Confusion Matrix

Support Vector Machine	<pre>y_pred = svm_model.predict(X_test) print(classification_report(y_test,y_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>0.55</td><td>0.85</td><td>0.67</td><td>895</td></tr><tr><td>1</td><td>0.83</td><td>0.53</td><td>0.65</td><td>1385</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.66</td><td>2200</td></tr><tr><td>macro avg</td><td>0.69</td><td>0.69</td><td>0.66</td><td>2200</td></tr><tr><td>weighted avg</td><td>0.72</td><td>0.66</td><td>0.66</td><td>2200</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.55	0.85	0.67	895	1	0.83	0.53	0.65	1385	accuracy			0.66	2200	macro avg	0.69	0.69	0.66	2200	weighted avg	0.72	0.66	0.66	2200	66%	Confusion Matrix: [[1139 173] [ 977 1011]]
	precision	recall	f1-score	support																													
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Random Forest Classifier	<pre>y_pred=rf_model.predict(X_test) print(classification_report(y_test,y_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>0.58</td><td>0.66</td><td>0.61</td><td>895</td></tr><tr><td>1</td><td>0.74</td><td>0.67</td><td>0.70</td><td>1385</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.67</td><td>2200</td></tr><tr><td>macro avg</td><td>0.66</td><td>0.66</td><td>0.66</td><td>2200</td></tr><tr><td>weighted avg</td><td>0.67</td><td>0.67</td><td>0.67</td><td>2200</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.58	0.66	0.61	895	1	0.74	0.67	0.70	1385	accuracy			0.67	2200	macro avg	0.66	0.66	0.66	2200	weighted avg	0.67	0.67	0.67	2200	68%	Confusion Matrix: [[1009 303] [ 774 1214]]
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Artificial Neural Network	<pre>history=nnf.train(x_train,y_train, epochs=100, batch_size=32)</pre> 	67%	Confusion Matrix: [[ 884 428] [ 806 1182]]																														
Logistic Classifier	<pre>print Logistic Regression report print(classification_report(y_test, y_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>0.56</td><td>0.58</td><td>0.57</td><td>895</td></tr><tr><td>1</td><td>0.70</td><td>0.69</td><td>0.70</td><td>1385</td></tr><tr><td>accuracy</td><td></td><td></td><td>0.64</td><td>2200</td></tr><tr><td>macro avg</td><td>0.63</td><td>0.63</td><td>0.63</td><td>2200</td></tr><tr><td>weighted avg</td><td>0.65</td><td>0.64</td><td>0.65</td><td>2200</td></tr></tbody></table>		precision	recall	f1-score	support	0	0.56	0.58	0.57	895	1	0.70	0.69	0.70	1385	accuracy			0.64	2200	macro avg	0.63	0.63	0.63	2200	weighted avg	0.65	0.64	0.65	2200	68%	Confusion Matrix: [[ 870 442] [ 781 1207]]
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