



Model Development Phase

Date	10 July 2024
Team ID	SWTID1720084775
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 is showcased in the following table through screenshots. The model validation and evaluation report include classification reports, accuracy, and confusion matrices for multiple models, presented through respective screenshots.

Initial Model Training Code:

```
from sklearn.linear_model import LogisticRegression
lr=LogisticRegression()

from sklearn.neighbors import KNeighborsClassifier
knn=KNeighborsClassifier(n_neighbors=7)

from sklearn.svm import SVC
model = SVC(kernel="linear")

from sklearn.naive_bayes import GaussianNB
nb = GaussianNB()

from sklearn.ensemble import RandomForestClassifier
rf = RandomForestClassifier(n_estimators=7, criterion='entropy',random_state=0)

import xgboost as xgb
xg = xgb.XGBClassifier()
```





```
#Fitting x_train and y_train
    lr.fit(x train, y train)
    knn.fit(x_train, y_train)
    model.fit(x train, y train)
    nb.fit(x_train, y_train)
    rf.fit(x_train, y_train)
    xg.fit(x_train, y_train)
₹
                                     XGBClassifier
    XGBClassifier(base score=None, booster=None, callbacks=None,
                   colsample_bylevel=None, colsample_bynode=None,
                   colsample_bytree=None, device=None, early_stopping_rounds=None,
                   enable categorical=False, eval metric=None, feature types=None,
                   gamma=None, grow_policy=None, importance_type=None,
                   interaction_constraints=None, learning_rate=None, max bin=None,
                   max_cat_threshold=None, max_cat_to_onehot=None,
                   max delta step=None, max depth=None, max leaves=None,
                   min child weight=None, missing=nan, monotone constraints=None,
                   multi strategy=None, n estimators=None, n jobs=None,
                   num parallel tree=None, random state=None, ...)
```

```
pred_lr = lr.predict(x_test)
    pred_knn = knn.predict(x_test)
    pred_model = model.predict(x_test)
    pred_nb = nb.predict(x_test)
    pred_rf = rf.predict(x_test)
    pred_xg = xg.predict(x_test)
```





```
[85] from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
[86] cr_lr = classification_report(y_test, pred_lr)
        cr knn = classification report(y test, pred knn)
        cr_svc = classification_report(y_test, pred_model)
        cr_nb = classification_report(y_test, pred_nb)
        cr_rf = classification_report(y_test, pred_rf)
        cr_xg = classification_report(y_test, pred_xg)
[87] as_lr = accuracy_score(y_test, pred_lr)
        as_knn = accuracy_score(y_test, pred_knn)
        as_svc = accuracy_score(y_test, pred_model)
        as_nb = accuracy_score(y_test, pred_nb)
        as_rf = accuracy_score(y_test, pred_rf)
        as_xg = accuracy_score(y_test, pred_xg)
[88] cm_lr = confusion_matrix(y_test, pred_lr)
        cm_knn = confusion_matrix(y_test, pred_knn)
        cm_svc = confusion_matrix(y_test, pred_model)
        cm nb = confusion matrix(y test, pred nb)
        cm_rf = confusion_matrix(y_test, pred_rf)
        cm_xg = confusion_matrix(y_test, pred_xg)
```

# Specify the Column Names while initializing the Table myTable = PrettyTable(["Metrics", "Logistic Regression", "KNN", "SVC", "Naive Bayes", "Random Forest", "XGBoost"]) # Add rows myTable.add_row(["classification Report", cr_lr, cr_knn, cr_svc, cr_nb, cr_rf, cr_xg]) myTable.add_row(["Accuracy Score", as_lr, as_knn, as_svc, as_nb, as_rf, as_xg]) myTable.add_row(["Confusion Matrix", cm_lr, cm_knn, cm_svc, cm_nb, cm_rf, cm_xg]) print(myTable)												
€	+ Metrics	+				+ KNN						
	Classification Report		precision	recall	f1-score	support	 	precision	recall	f1-score	support	
	i	0	0.63	0.67	0.65	872	j ø	0.62	0.72	0.67	872	
	İ		0.66	0.63	0.64	903] 1 	0.68	0.58	0.63	903	
	i	accuracy			0.65	1775	accuracy			0.65	1775	aco
		macro avg	0.65	0.65	0.65	1775	macro avg	0.65	0.65	0.65	1775	macı
		weighted avg	0.65	0.65	0.65	1775	weighted avg 	0.65	0.65	0.65	1775	weight
	Accuracy Score	0.6467605633802817				0.6473239436619719						
	Confusion Matrix [[583 289]				[[627 245] [381 522]]							





Model Validation and Evaluation Report:

Model	Classification Report	Accuracy	Confusion Matrix
Logistic Regressio	precision recall f1-score support 0 0.63 0.67 0.65 872 1 0.66 0.63 0.64 903 accuracy 0.65 1775 macro avg 0.65 0.65 0.65 1775 weighted avg 0.65 0.65 0.65 1775	0.64676056	[[583 289] [338 565]]
K-Nearest Neighbours (KNN)	precision recall f1-score support 0 0.62 0.72 0.67 872 1 0.68 0.58 0.63 903 accuracy 0.65 1775 macro avg 0.65 0.65 0.65 1775 weighted avg 0.65 0.65 0.65 1775	0.64732394	[[627 245] [381 522]]
Support Vector Machine (SVM)	precision recall f1-score support 0 0.64 0.75 0.69 872 1 0.71 0.59 0.64 903 accuracy 0.67 1775 macro avg 0.67 0.67 0.67 1775 weighted avg 0.67 0.67 0.67 1775	0.66760563	[[654 218] [372 531]]
Gaussian Naive Bayes	precision recall f1-score support 0 0.65 0.73 0.68 872 1 0.70 0.61 0.65 903 accuracy 0.67 1775 macro avg 0.67 0.67 0.67 1775 weighted avg 0.67 0.67 0.67 1775	0.66985915	[[636 236] [350 553]]
Random Forest Generator	precision recall f1-score support 0 0.63 0.66 0.65 872 1 0.66 0.63 0.64 903 accuracy 0.64 1775 macro avg 0.64 0.64 0.64 1775 weighted avg 0.64 0.64 0.64 1775	0.64394366	[[575 297] [335 568]]
XG Boost	precision recall f1-score support 0 0.64 0.79 0.71 872 1 0.74 0.58 0.65 903 accuracy 0.68 0.68 1775 macro avg 0.69 0.68 0.68 1775 weighted avg 0.69 0.68 0.68 1775	0.67943662	[[686 186] [383 520]]