



Model Development Phase Template

Date	15 July 2024
Team ID	739699
Project Title	Telecom Customer Churn Prediction
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]], dtype=int64)
   [54]: #logistic Regression
         from sklearn.linear_model import LogisticRegression
         model=LogisticRegression()
         model.fit(x_train,y_train)
         accuracy_score(model.predict(x_test),y_test)
   [54]: 0.807
56]: #Decision Tree classifier]
     from sklearn.tree import DecisionTreeClassifier
      classifier= DecisionTreeClassifier(criterion='entropy', random_state=42)
      classifier.fit(x_train, y_train)
      pred=classifier.predict(x_test)
      dtc_acc=accuracy_score(pred,y_test)
      dtc_acc
56]: 0.7835
8]:
    #random forest classifier
     from sklearn.ensemble import RandomForestClassifier
     rc=RandomForestClassifier(random_state=42)
     rc.fit(x train,y train)
     pred=rc.predict(x_test)
     rfc_acc=accuracy_score(y_test,pred)
     rfc_acc
```

3]: 0.864





```
#kNeighborsClassifier
[67]:
      from sklearn.neighbors import KNeighborsClassifier
      knn=KNeighborsClassifier()
      knn.fit(x_train,y_train)
[68]:
[68]: TWeighborsClassifier
      KNeighborsClassifier()
      knn_acc=accuracy_score(knn.predict(x_test),y_test)
[69]:
      knn_acc
[69]: 0.8345
[71]: #naive bayes classifier
      from sklearn.naive_bayes import GaussianNB
      gnb = GaussianNB()
      gnb.fit(x_train, y_train)
      nb_acc=accuracy_score(gnb.predict(x_test),y_test)
      nb_acc
[71]: 0.8275
```

Model Validation and Evaluation Report:

Model	Classific	ation Report	Accuracy	Confusion Matrix
svm	82]: print(classification_rep precision 0 1.00 1 0.00 accuracy macro avg 0.50 weighted avg 1.00	ort(svm_pred,y_test)) recall f1-score support 0.80 0.89 2000 0.00 0.00 0 0.80 2000 0.40 0.44 2000 0.80 0.89 2000	79	[83]: errey([1395, 405], [63]: errey([1395, 405], [6], 6]], dtypa-irtid)





Logistic regression	[84]: print(classification_report(model.predict(x_test),y_test)) precision recall f1-score support 0 0.96 0.88 0.92 1733 1 0.49 0.75 0.60 267 accuracy 0.86 2000 macro avg 0.73 0.82 0.76 2000 weighted avg 0.90 0.86 0.88 2000	80	[85]: confusion_matrix(model.predict(x_test),y_test) [85]: array([[1528, 205],
Decision Tree	[86]: print(classification_report(pred,y_test)) precision recell f1-score support 0 0.96 0.88 0.92 1733 1 0.49 0.79 0.80 767 accuracy 0.86 2000 macro avg 0.73 0.82 0.76 2000 weighted avg 0.90 0.86 0.88 2000	78	[87]: confusion_matrix(pred,y_test) [87]: array([[1528, 205],
Random Forest	[88]: print(classification_report(pred,y_test)) precision recall f1-score support 0 0.96 0.88 0.92 1733 1 0.49 0.75 0.60 267 accuracy macro avg 0.73 0.82 0.76 2000 weighted avg 0.90 0.86 0.88 2000	86	[61]: rfc_con=confusion_matrix(pred,y_test) rfc_con [61]: array([[1528, 205],
knn	print(classification_report(knn.predict(x_test),y_test))	83	[70]: knn_consconfusion_matrix(knn.predict(x_test),y_test) knn_con [70]: array([[1496, 232],
Naïve bayes		82	[72]: rb.conccefusion_matrix(gab.predict(x_test),y_test) rbcos [72]: arrey([[1548, 256],