



Model Development Phase Template

Date	05 July 2024
Team ID	739938
Project Title	Anticipating Business Bankruptcy
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:

```
#importing and building svc
from sklearn.svm import SVC
svm= SVC(kernel='rbf', random_state=0)
svm.fit(x_train,y_train)

y_pred_svc=svm.predict(x_test)

print('Training Set:', svm.score(x_train,y_train))
print('Testing Set:', svm.score(x_test,y_test))

accuracy_SVC=svm.score(x_test,y_test)
print('Accuracy_SVC=svm.score(x_test,y_test))

#importing and building decision classifier
```

```
#importing and building decision classifier
!pip install scikit-learn # Install scikit-learn if you haven't alrefrom sklearn.tree import DecisionTreeClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import r2_score,mean_squared_error,mean_absolute_error
dt = DecisionTreeClassifier() # Now you can create an instance
```





```
dt.fit(x_train,y_train)
 y_pred_dt=dt.predict(x_test)
 print('Training Set: ',dt.score(x_train,y_train))
 print('Test Set: ',dt.score(x_test,y_test))
from sklearn import metrics # Import the metrics module
dt = DecisionTreeClassifier()
print("Accuracy:",metrics.accuracy_score(y_test,y_pred_dt)*100) # Now you can use metrics.accuracy_score
accuracy dt=accuracy score(y test,y pred dt)
print('Accuracy_DT: ', accuracy_dt*100)
#importing and building random forest
from sklearn.ensemble import RandomForestClassifier
rand forest= RandomForestClassifier(random state=42)
rand_forest.fit(x_train,y_train)
                                                                                                                                                          Python
predictionRF=rand_forest .predict(x_test)
#checking the accuracy on the training set print('Training set :', rand_forest.score(x_train,y_train)) #checking the accuracy on the testing set
print('Testing set :', rand_forest.score(x_test,y_test))
                                                                                                                                                          Python
accuracy_RF=rand_forest.score(x_test, y_test)
print('Accuracy_RF: ', accuracy_RF*100)
```

Model Validation and Evaluation Report:

Model	Cla	ssifica	tion F	Repor	t		F1 Scor e	Confusion Matrix
Support vector classifier	# Assuming print(class	y_test is y sification_r precision 0.63 0.66 0.65 0.65	your tr (vereport(y_te	ariable) y	_test: Any	r predicted labels	64%	print(confusion_matrix(y_test, y_pred_dt)) [[1730 261] [192 1871]]





Random Forest	<pre>from sklearn.metrics import classification_report # Assuming y_test is your true labels and predictionRF is your predicted labels print(classification_report(y_test, predictionRF))</pre>	94%	<pre>print(confusion_matrix(y_test, predictionRF)) [[1860 131] [98 1965]]</pre>
	precision recall f1-score support 0 0.95 0.93 0.94 1991 1 0.94 0.95 0.94 2063 accuracy 0.94 4054 macro avg 0.94 0.94 0.94 4054 weighted avg 0.94 0.94 0.94 4054		
Decision Tree	<pre>from sklearn.metrics import classification_report # Assuming y_test is your true labels and predictionRF is your predicted labels print(classification_report(y_test,y_pred_dt)) precision recall f1-score support</pre>	88%	<pre>print(confusion_matrix(y_test, y_pred_dt)) [[1730 261] [192 1871]]</pre>
	0 0.90 0.87 0.88 1991 1 0.88 0.91 0.89 2063 accuracy 0.89 4054 macro avg 0.89 0.89 0.89 4054 weighted avg 0.89 0.89 0.89 4054		