| | <pre>import pandas as pd Bank=pd.read_excel("E:\Data science training\R AND PYTHON KPMG\stat and ml\ASSIGNMENT\Bank Marketing.xlsx") Bank.head()</pre> |
|------------------|---|
|]: . | Age Job Marital Status Account Balance Own House Personal Loan No of campaigns Subscription 0 59 unemployed married 0 0 0 0 1 36 Others married 3057 0 0 0 1 2 47 blue-collar divorced 126 1 0 0 1 |
|]:[| 3 43 management divorced 388 1 0 0 1 4 34 self-employed single 462 0 0 0 1 Bank.info() |
| | <pre>class 'pandas.core.frame.DataFrame'> RangeIndex: 4521 entries, 0 to 4520 Data columns (total 8 columns): # Column Non-Null Count Dtype</pre> |
| | 0 Age 4521 non-null int64 1 Job 4521 non-null object 2 Marital Status 4521 non-null object 3 Account Balance 4521 non-null int64 |
| | 4 Own House 4521 non-null int64 5 Personal Loan 4521 non-null int64 6 No of campaigns 4521 non-null int64 7 Subscription 4521 non-null int64 dtypes: int64(6), object(2) memory usage: 282.7+ KB |
|]: | <pre>Bank.isnull().sum()</pre> Age 0 |
| | Job 0 Marital Status 0 Account Balance 0 Own House 0 Personal Loan 0 No of campaigns 0 |
| | Subscription 0 dtype: int64 cat_col = ['Job', 'Marital Status'] |
|]: | <pre>Bank_dummy = pd.get_dummies(Bank,columns=cat_col,drop_first=True)</pre> |
|]:]:]: [| Bank_dummy.shape (4521, 12) |
|]: | Bank_dummy . head() Age Account Balance House Personal Loan Subscription Job_blue-collar Job_management Job_self-employed Job_unemployed Status_married Status_ 0 59 0 0 0 0 0 0 0 0 0 1 1 1 |
| | 1 36 3057 0 0 0 1 0 0 0 0 1 2 47 126 1 0 0 1 1 0 0 0 0 0 3 43 388 1 0 0 1 0 1 0 0 0 0 |
|]: | 4 34 462 0 0 0 1 0 0 1 0 0 1 0 0 #Identifying the input and output variables Y = Bank_dummy[['Subscription']] |
|]:[| <pre>X = Bank_dummy.drop(columns=['Subscription']) X.head()</pre> |
|]: . | Age Account Balance Own House Personal Loan No of campaigns Job_blue-collar Job_management Job_self-employed Job_unemployed Marital Status_married Marital Status_single 1 36 3057 0 0 0 0 0 0 0 0 0 2 47 126 1 0 0 0 0 0 0 0 0 0 |
| | 3 43 388 1 0 0 0 1 0 0 0 0 4 34 462 0 0 0 0 0 1 0 0 0 1 |
|]: | <pre>##Spillting the data into train and test from sklearn.model_selection import train_test_split x_train, x_test, y_train, y_test = train_test_split(X, Y, test_size=0.2, random_state=42) len(x_train), len(x_test), len(y_train), len(y_test)</pre> |
| | (3616, 905, 3616, 905) Building the Random Forest model: |
|]: | <pre>from sklearn.ensemble import RandomForestClassifier model = RandomForestClassifier(n_estimators=1000, random_state=42) rf = model.fit(x_train, y_train) print('The model has been built successfully!! yeah')</pre> |
| | <pre>C:\Users\Pratik\AppData\Local\Temp/ipykernel_5368/3858015865.py:3: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Pl ange the shape of y to (n_samples,), for example using ravel(). rf = model.fit(x_train,y_train) The model has been built successfully!! yeah</pre> |
|]: | # Predicting the model on test data y_test Subscription |
| | 2398 0 800 0 2288 0 2344 0 |
| | 3615 0 3589 0 |
| | 3128 0 3259 0 4239 0 2233 0 |
| | <pre>905 rows × 1 columns y_test['Prediction'] = model.predict(x_test)</pre> |
|]: | y_test Subscription Prediction 2398 0 0 |
| | 800 0 0 2288 0 0 2344 0 0 800 0 0 8015 0 0 |
| | 3615 0 0 3589 0 0 3128 0 0 |
| | 3259 0 0 4239 0 0 2233 0 0 |
|]: | 905 rows × 2 columns from sklearn.metrics import confusion_matrix, accuracy_score print(confusion_matrix(y_test['Subscription'], y_test['Prediction'])) |
|]: [| <pre>[[777 26] [95 7]] print(accuracy_score(y_test['Subscription'], y_test['Prediction']))</pre> |
|]: | 0.8662983425414365 ## My Model Accuracy Score is 0.8662 or 86.62% correctly predicted |
|]: | 2) What is the accuracy of Support Vector Machine? How does changing the model affect the accuracy #Identifying the input and output variables Y1 = Bank_dummy[['Subscription']] X1 = Bank_dummy.drop(columns=['Subscription']) |
|]: []: [| X1.head() Age Account Balance Own House Personal Loan No of campaigns Job_blue-collar Job_management Job_self-employed Job_unemployed Marital Status_married Marital Status_single |
| | 0 59 0 0 0 0 0 0 1 1 0 1 36 3057 0 0 0 0 0 0 0 1 0 2 47 126 1 0 0 1 0 0 0 0 0 |
|]: [| 3 43 388 1 0 0 0 1 0 0 0 0 4 34 462 0 0 0 0 1 0 0 0 1 #Spillting the data into train and test |
| | <pre>from sklearn.model_selection import train_test_split x1_train,x1_test,y1_train,y1_test = train_test_split(X1,Y1,test_size=0.2,random_state=42) len(x1_train),len(x1_test),len(y1_train),len(y1_test)</pre> |
|]: | (3616, 905, 3616, 905) from sklearn.svm import SVC model1 = SVC(random_state=42) |
|]: | <pre># fitting the Model svc_model = model1.fit(x1_train,y1_train) print('The model has been built successfully!! yeah')</pre> |
| | <pre>C:\Users\Pratik\anaconda3\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was passed when a 1d array was ex Please change the shape of y to (n_samples,), for example using ravel(). y = column_or_1d(y, warn=True) The model has been built successfully!! yeah</pre> |
|]: | #PRediction of test data y1_test Subscription |
| | 2398 0 800 0 2288 0 2344 0 |
| | 2344 0 3615 0 3589 0 |
| | 3128 0 3259 0 4239 0 2233 0 |
| ;]:[| 905 rows × 1 columns |
|]: [| <pre>y1_test['Prediction'] = svc_model.predict(x1_test) y1_test</pre> Subscription Prediction |
|]: | Subscription Prediction 2398 0 800 0 2288 0 |
| | 2344 0 0 0 3615 0 0 |
| | 3589 0 0 3128 0 0 3259 0 0 4239 0 0 |
| | 2233 0 0 0 905 rows × 2 columns |
|]: | <pre>from sklearn.metrics import confusion_matrix,accuracy_score print(confusion_matrix(y1_test['Subscription'],y1_test['Prediction'])) [[803 0] [102 0]]</pre> |
|]: | <pre>[102 0]] print(accuracy_score(y1_test['Subscription'], y1_test['Prediction'])) 0.887292817679558</pre> |
|]: | ##Here we can see that using SVM acurracy of my model increase it's predict More no of Data correctly but # it's only Predict one kind of Data Correctly which is those doesn't have subscription("0"). model feature importances |
| 1. | model.feature_importances_ array([0.2979414 , 0.47291481, 0.02827675, 0.01718792, 0.09330933, |
| | <pre>forest_importance = pd.Series(model.feature_importances_,index=X.columns) forest_importance.sort_values(ascending=False).head(4)</pre> |
|]: | Account Balance 0.472915 Age 0.297941 |
|]: | Account Balance 0.472915 Age 0.297941 No of campaigns 0.093309 Own House 0.028277 dtype: float64 ## 3) above 4 columns are most importance factor in Random Forest Model.Here Customer Account Balance and Age are |
|]: | Age 0.297941 No of campaigns 0.093309 Own House 0.028277 dtype: float64 ## 3) above 4 columns are most importance factor in Random Forest Model.Here Customer Account Balance and Age are # most important Factor For getting Subscription. # 4)SVM Model only Predict one kind of Data Correctly which is those doesn't have subscription("0"). # so though it's have a better accuracy than Random Forest but Random Forest overall have a better Model bcz it's predict both the Data correctly |
| .]: | Age 0.297941 No of campaigns 0.093309 Own House 0.028277 dtype: float64 ## 3) above 4 columns are most importance factor in Random Forest Model.Here Customer Account Balance and Age are # most important Factor For getting Subscription. # 4)SVM Model only Predict one kind of Data Correctly which is those doesn't have subscription("0"). |