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Week 7 Assignment: Presenting and Interpreting Results of Predictive Models
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           #loading the required packages
           import pandas as pd
           import numpy as np
           import matplotlib.pyplot as plt
           import random
           import statsmodels.tools.tools as stattools
           import sklearn
           from sklearn.linear model import LinearRegression
           from sklearn.model selection import train test split
           from sklearn.preprocessing import StandardScaler
           from sklearn.linear model import LogisticRegression
           from sklearn.metrics import confusion matrix
           from sklearn.tree import DecisionTreeClassifier, export graphviz
           from sklearn.ensemble import RandomForestClassifier
           from sklearn.metrics import accuracy score
           #Importing the required data sets
           loans_training = pd.read_csv('C:/School/DSC-530/DataSets/Loans Training')
           loans_test = pd.read_csv('C:/School/DSC-530/DataSets/Loans_Test')
In [40]:
           #Convert T and F to 1 and 0
           loans training["Approval"]=loans training["Approval"].apply(lambda x: 1 if x in("T") else 0)
           features = loans training.columns
           print(features)
          Index(['Approval', 'Debt-to-Income Ratio', 'FICO Score', 'Request Amount',
                 'Interest'],
                dtype='object')
In [43]:
           # Extractng the Predictors
          X = loans training[['Debt-to-Income Ratio','FICO Score','Request Amount']]
           print(X.columns)
           # Extractng the vector
           y = loans training[['Approval']]
           print(y.columns)
          Index(['Debt-to-Income Ratio', 'FICO Score', 'Request Amount'], dtype='object')
          Index(['Approval'], dtype='object')
         Using the training data set, create a C5.0 model (Model 1) to predict a loan applicant's Approval using Debt-to-Income Ratio, FICO Score,
         and Request Amount. Obtain the predicted responses.
           #Creating the C5.0 Model (Model1)
           c50 model1 = DecisionTreeClassifier(criterion='entropy', max leaf nodes=5).fit(X,y)
           #Obtaining the classification
           c50 model1.predict(X)
Out[33]: array([0, 0, 0, ..., 0, 0], dtype=int64)
         Evaluate Model 1 using the test data set. Construct a contingency table to compare the actual and predicted values of Approval.
In [45]:
           #Convert T and F to 1 and 0 on the test data
           loans\_test["Approval"] = loans\_test["Approval"] . apply(lambda x: 1 if x in("T") else 0)
           features = loans test.columns
           print(features)
          Index(['Approval',
                               Debt-to-Income Ratio', 'FICO Score', 'Request Amount',
                 'Interest'],
                dtype='object')
In [47]:
           # Extractng the Predictors for the test data
           X test = loans test[['Debt-to-Income Ratio','FICO Score','Request Amount']]
           print(X.columns)
           # Extractng the vector
           y_test = loans_training[['Approval']]
           print(y.columns)
          Index(['Debt-to-Income Ratio', 'FICO Score', 'Request Amount'], dtype='object')
          Index(['Approval'], dtype='object')
In [54]:
           #Evaluating model 1
           y predicted = c50 model1.predict(X test)
           #Creating a contingency table to compare the actual and predicted values of Approval
           ypred = pd.crosstab(loans_test['Approval'], y_predicted, rownames= ['Actual'], colnames= ['Predicted'])
           ypred
Out[73]: Predicted
                       0
                             1
            Actual
                0 17665
                          7269
                     751 24013
           ypred['Total'] = ypred.sum(axis=1); ypred.loc['Total'] = ypred.sum(); ypred
Out[69]: Predicted
                             1 Total
            Actual
                0 17665
                          7269 24934
                     751 24013 24764
             Total 18416 31282 49698
           #Calculation for the below following can be found on the attached excel spreadsheet.
           # 1 . Accuracy
           # 2 . Error rate
           # 3 . Sensitivity
           # 4 . Specificity
           # 5 . Precision
           # 6 . F1
           # 7 . F2
           # 8 . F0.5
         Calculate the mean Request Amount per loan applicant from the training data set. Set this value to be the cost of a false positive.
           loans training['Request Amount'].mean()
Out[83]: 13427.080145307447
           1. Compute the mean of the Interest per loan applicant from the training data set. Set the negative of that value to be the cost of a true
             positive.
           loans training['Interest'].mean()
          6042.186065388351
         Obtain the simplified data-driven cost matrix.
In [84]:
           y pred = classifier.predict(xtest)
           cm = confusion_matrix(ytest, y_pred)
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print ("Confusion Matrix : \n", cm)

Confusion Matrix : [[14436 4199] [3159 15782]]

DSC-530: Predictive Modeling