Predicting Credit Card Approvals 4/1/22, 7:02 PM

Predicting Credit Card Approvals

Exploratory Data Analysis (EDA)

```
In [10]:
           # Import pandas
           import pandas as pd
          # Load dataset
          df = pd.read_csv("creditcard.csv", header=None)
In [12]:
           # Inspect data
          df.head()
Out[12]:
                         2 3 4 5 6
                                         7 8 9 10 11 12
                                                               13
                                                                   14
                                                                      15
               30.83 0.000 u g w
                                      1.25
                                                           00202
                                                                    0
               58.67 4.460 u
                                 q h
                                      3.04
                                                           00043 560
            a 24.50 0.500 u
                                                           00280 824
                                   h
                                      1.50
               27.83
                     1.540
                                                            00100
                                       3.75
                                                                    3
                                                            00120
                20.17 5.625
                                       1.71
                                                                    0
In [13]:
           # Summary of Dataframe
          df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 690 entries, 0 to 689
Data columns (total 16 columns):

#	Colum	nn Non-	-Null	Count	Dtype
0	0	690	non-n	ull	object
1	1	690	non-n	ull	object
2	2	690	non-n	ull	float64
3	3	690	non-n	ull	object
4	4	690	non-n	ull	object
5	5	690	non-n	ull	object
6	6	690	non-n	ull	object
7	7	690	non-n	ull	float64
8	8	690	non-n	ull	object
9	9	690	non-n	ull	object
10	10	690	non-n	ull	int64
11	11	690	non-n	ull	object
12	12	690	non-n	ull	object
13	13	690	non-n	ull	object
14	14	690	non-n	ull	int64
15	15	690	non-n	ull	object
مــــــــــــــــــــــــــــــــــ	.a. £1	00+6112	٠,	+64(2)	ob + oo+ (11

dtypes: float64(2), int64(2), object(12)

memory usage: 86.4+ KB

In [14]:

Statistical details of dataframe
df.describe()

Out[14]:		2	7	10	14
	count	690.000000	690.000000	690.00000	690.000000
	mean	4.758725	2.223406	2.40000	1017.385507
	std	4.978163	3.346513	4.86294	5210.102598
	min	0.000000	0.000000	0.00000	0.000000
	25%	1.000000	0.165000	0.00000	0.000000
	50%	2.750000	1.000000	0.00000	5.000000
	75%	7.207500	2.625000	3.00000	395.500000
	max	28.000000	28.500000	67.00000	100000.000000

```
In [19]:
```

checking if there exists any missing values
df.isnull().any()

```
False
Out[19]:
                 False
          2
                 False
          3
                 False
                 False
          5
                 False
          6
                 False
                 False
          8
                 False
                 False
          10
                False
          11
                 False
          12
                False
          13
                False
                False
          14
          15
                False
          dtype: bool
In [21]:
           # Importing required libraries
           import matplotlib.pyplot as plt
           import seaborn as sns
           import numpy as np
           %matplotlib inline
In [23]:
           # Histogram for available data of columns 2,7,10,14
           df.hist()
          array([[<AxesSubplot:title={'center':'2'}>,
Out[23]:
                   <AxesSubplot:title={'center':'7'}>],
                  [<AxesSubplot:title={'center':'10'}>,
                   <AxesSubplot:title={'center':'14'}>]], dtype=object)
                        2
          300
                                     400
          200
                                     200
          100
                     10
                            20
                                               10
                                                      20
                                                  14
                       10
          600
                                     600
          400
                                     400
          200
                                     200
            0
                                       0
                                                 50000
                    20
                         40
                               60
                                                          100000
```

Data Preprocessing

```
In [24]:
          # Import train_test_split
          from sklearn.model_selection import train_test_split
          # Dropping the unnecessary features
          df1 = df \cdot drop([11,13], axis=1)
In [25]:
          # peforming train test split
          df1 train, df1 test = train test split(df1, test size=0.3, random state=1)
In [26]:
          # There are missing values in the dataframe in the form '?'.
          # This is handled in the following steps.
In [27]:
          # Replacing the '?'s with NaN
          df1_train = df1_train.replace('?',np.nan)
          df1_test = df1_test.replace('?',np.nan)
In [31]:
          # checking for null values/missing values in dataframe
          df1_train.isnull().sum()
                6
Out[31]:
                6
                0
          3
                5
          7
          8
          10
          12
          14
          15
                0
         dtype: int64
In [32]:
          # Filling the missing values in the dataframe with mean imputation
          df1 train.fillna(df1 train.mean(), inplace=True)
          df1_test.fillna(df1_train.mean(), inplace=True)
```

dtype: int64

ror. Select only valid columns before calling the reduction. dfl_train.fillna(dfl_train.mean(), inplace=True) /var/folders/q9/vmdwqts13cj9tk5 rqd7w6pr0000gn/T/ipykernel 57115/2568501163.py :3: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric only=None') is deprecated; in a future version this will raise TypeEr ror. Select only valid columns before calling the reduction. df1 test.fillna(df1 train.mean(), inplace=True) In [33]: # Number of missing values in train and test sets print(df1 train.isnull().sum()) print(df1 test.isnull().sum()) dtype: int64

/var/folders/q9/vmdwqts13cj9tk5_rqd7w6pr0000gn/T/ipykernel_57115/2568501163.py
:2: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric only=None') is deprecated; in a future version this will raise TypeEr

```
In [34]:
          # Iterate over each column of dataframe
          for col in df1_train.columns:
               # Checking if the column is of object type
               if dfl_train[col].dtypes == 'object':
                   # Impute with the most frequent value
                   df1_train = df1_train.fillna(df1_train[col].value_counts().index[0])
                   df1 test = df1 test.fillna(df1 train[col].value counts().index[0])
          # Counting the number of NaNs in the dataset after imputation
          print(df1 train.isnull().sum())
          print(df1 test.isnull().sum())
         0
                0
         1
                0
                0
         3
         5
         6
                0
         7
                0
                0
                0
         10
         12
                0
         14
                0
         15
         dtype: int64
                0
                0
         2
                0
         3
                0
         4
                0
         5
                0
         6
                0
                0
         9
         10
                0
         12
                0
         14
                0
         15
                0
         dtype: int64
In [35]:
          # Converting the categorical features
          dfl_train = pd.get_dummies(dfl_train)
          df1 test = pd.get dummies(df1 test)
          # Reindexing the columns of the test set
          df1_test = df1_test.reindex(columns=df1_train.columns, fill_value=0)
```

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```
In [36]: # Import MinMaxScaler
from sklearn.preprocessing import MinMaxScaler

# Segregating features and labels into separate variables
X_train, y_train = df1_train.iloc[:, :-1].values, df1_train.iloc[:, [-1]].values,
X_test, y_test = df1_test.iloc[:, :-1].values, df1_test.iloc[:, [-1]].values

# MinMaxScaler and using it to rescale X_train and X_test
scaler = MinMaxScaler(feature_range=(0, 1))
rescaledX_train = scaler.fit_transform(X_train)
rescaledX_test = scaler.transform(X_test)
```

Model Fitting

/Users/bdvvgangarajuabbireddy/opt/anaconda3/lib/python3.9/site-packages/sklear n/utils/validation.py:993: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), f or example using ravel().

```
y = column_or_1d(y, warn=True)
Out[37]: 0.7971014492753623
```

After scaling using Minmaxscaler, and fitting logistic regression resulted an acurracy of 79.71

Evaluating Performance of Model

```
In [38]: # Import confusion_matrix
    from sklearn.metrics import confusion_matrix

# Using logreg to predict instances from the test set and store it
y_pred = logreg.predict(rescaledX_test)

# Get the accuracy score of logreg model and print it
print("Accuracy of logistic regression classifier: ", logreg.score(rescaledX_
# Print the confusion matrix of the logreg model
confusion_matrix(y_test,y_pred)
Accuracy of logistic regression classifier: 1.0
```

Performing GridSearchCV for better performance and accuracy!!!

```
In [43]: # Import GridSearchCV
from sklearn.model_selection import GridSearchCV

# grid of values for tol and max_iter
tol = [0.01, 0.001, 0.0001]
max_iter = [100]

param_grid = dict(tol=tol, max_iter=max_iter)
```

Best Performing Model

```
In [46]: # GridSearchCV
model1 = GridSearchCV(estimator=logreg, param_grid=param_grid, cv=5)

# Fitting grid_model to the data
model1_result = grid_model.fit(rescaledX_train, y_train)

# Summarizing results
best_score, best_params = model1_result.best_score_, model1_result.best_param

# Performing best model on the test set
best_model = model1_result.best_estimator_
print("Accuracy of logistic regression classifier: ", best_model.score(rescaled)
```

Accuracy of logistic regression classifier: 1.0

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The model performed extremely well with an accuracy score of 1.0 making our model the best fit for predicting credit card approvals

In []:			