Logistic Regression

Logistic regression is a statistical model that in its basic form uses a logistic function to model a binary dependent variable, although many more complex extensions exist. In regression analysis, logistic regression (or logit regression) is estimating the parameters of a logistic mod el (a form of binary regression).

1. Reading the Data

```
In [1]: import numpy as np import pandas as pd
```

Out[2]:

| | accepted | creditscore | amount | age | marital | health_ins | creditgrade |
|---|----------|-------------|----------|-----|---------|------------|-------------|
| 0 | 0 | 387 | 42580093 | 87 | 0 | 0 | А |
| 1 | 0 | 400 | 80816186 | 49 | 0 | 1 | Α |
| 2 | 0 | 360 | 37264552 | 46 | 1 | 0 | Α |
| 3 | 0 | 378 | 7209235 | 38 | 1 | 0 | Α |
| 4 | 0 | 387 | 71637479 | 51 | 1 | 0 | Α |

```
In [3]: df.shape
```

Out[3]: (2500, 7)

```
In [4]: df.describe()
```

Out[4]:

| | accepted | creditscore | amount | age | marital | health_ins |
|-------|-------------|-------------|--------------|-------------|-------------|-------------|
| count | 2500.000000 | 2500.000000 | 2.500000e+03 | 2500.000000 | 2500.000000 | 2500.000000 |
| mean | 0.518800 | 286.824400 | 3.364925e+07 | 58.156400 | 0.494800 | 0.440000 |
| std | 0.499746 | 74.884278 | 2.897636e+07 | 21.082678 | 0.500073 | 0.496486 |
| min | 0.000000 | 100.000000 | 5.113590e+05 | 18.000000 | 0.000000 | 0.000000 |
| 25% | 0.000000 | 246.000000 | 8.954224e+06 | 40.000000 | 0.000000 | 0.000000 |
| 50% | 1.000000 | 298.000000 | 2.360884e+07 | 61.000000 | 0.000000 | 0.000000 |
| 75% | 1.000000 | 343.250000 | 5.592593e+07 | 76.000000 | 1.000000 | 1.000000 |
| max | 1.000000 | 400.000000 | 9.995992e+07 | 90.000000 | 1.000000 | 1.000000 |

2. Loading the library

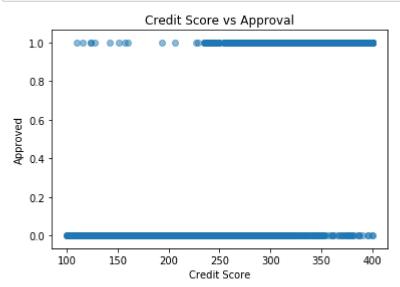
```
In [5]: from sklearn.preprocessing import MinMaxScaler
    from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LogisticRegression
    from sklearn.metrics import accuracy_score,roc_auc_score,confusion_matrix
    import matplotlib.pyplot as plt

In [6]: x=df['creditscore']
    y=df['accepted']

In [7]: print(y.value_counts())

    1 1297
    0 1203
    Name: accepted, dtype: int64
```

```
In [8]: plt.scatter(x,y,alpha=0.5)
    plt.title("Credit Score vs Approval")
    plt.xlabel("Credit Score")
    plt.ylabel("Approved")
    plt.show()
```



It shows that credit score is not only the reason behind accepting and rejecting the loan. There may be many other reason behind this.

3. Model Building [Using only 1 Column]

```
In [9]: xtrain,xtest,ytrain,ytest=train_test_split(x,y,test_size=0.7,random_state=0)
    xtrain=pd.DataFrame(xtrain)
    xtest=pd.DataFrame(xtest)
    ytrain=pd.DataFrame(ytrain)
    ytest=pd.DataFrame(ytest)
```

```
In [10]: #Transformation
         scale=MinMaxScaler()
         xtrain=scale.fit_transform(xtrain)
         xtest=scale.transform(xtest)
         C:\Users\Gaurav\Anaconda3\lib\site-packages\sklearn\preprocessing\data.py:334:
         DataConversionWarning: Data with input dtype int64 were all converted to float6
         4 by MinMaxScaler.
           return self.partial_fit(X, y)
         #Initializing the instance of the class
In [11]:
         model=LogisticRegression(random state=0, solver='lbfgs')
         #Fitting the model
         model.fit(xtrain,ytrain)
         C:\Users\Gaurav\Anaconda3\lib\site-packages\sklearn\utils\validation.py:761: Da
         taConversionWarning: A column-vector y was passed when a 1d array was expected.
         Please change the shape of y to (n_samples, ), for example using ravel().
           y = column or 1d(y, warn=True)
Out[11]: LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                   intercept_scaling=1, max_iter=100, multi_class='warn',
                   n_jobs=None, penalty='12', random_state=0, solver='lbfgs',
                   tol=0.0001, verbose=0, warm start=False)
```

4. Predicting the Value

Our model gives correct output for 80 input for every input in probability.

Now we will take all the columns as input and try to check whether our accuracy improves or not.

a. Model Building [Using all Columns]

```
In [15]: x=df[["creditscore", "amount", "age", "marital", "health_ins"]]
         y=df["accepted"]
In [16]: | xtrain,xtest,ytrain,ytest=train_test_split(x,y,test_size=0.7,random_state=0)
         xtrain=pd.DataFrame(xtrain)
         xtest=pd.DataFrame(xtest)
         ytrain=pd.DataFrame(ytrain)
         ytest=pd.DataFrame(ytest)
In [17]: #Transformation
         scale=MinMaxScaler()
         xtrain=scale.fit_transform(xtrain)
         xtest=scale.transform(xtest)
         C:\Users\Gaurav\Anaconda3\lib\site-packages\sklearn\preprocessing\data.py:334:
         DataConversionWarning: Data with input dtype int64 were all converted to float6
         4 by MinMaxScaler.
           return self.partial_fit(X, y)
In [18]: #Initializing the instance of the class
         model=LogisticRegression(random_state=0, solver='lbfgs')
         #Fitting the model
         model.fit(xtrain,ytrain)
         C:\Users\Gaurav\Anaconda3\lib\site-packages\sklearn\utils\validation.py:761: Da
         taConversionWarning: A column-vector y was passed when a 1d array was expected.
         Please change the shape of y to (n samples, ), for example using ravel().
           y = column_or_1d(y, warn=True)
Out[18]: LogisticRegression(C=1.0, class weight=None, dual=False, fit intercept=True,
                   intercept_scaling=1, max_iter=100, multi_class='warn',
                   n jobs=None, penalty='12', random state=0, solver='lbfgs',
                   tol=0.0001, verbose=0, warm start=False)
```

b. Predicting from Modal

We see that our accuracy have incresed from 80 to 95%.

```
In [21]: model.coef_
Out[21]: array([[ 5.35715507, -3.23045671, -0.12705459,  1.41530128,  3.79372072]])
```

We see that age have very small value for beta3 i.e it is not contributing for the data value. This gives us a variable importance. Now remove variable age and see the accuracy level. You will find that there is no significant in accuracy level.