Logistic_Regression

October 14, 2021

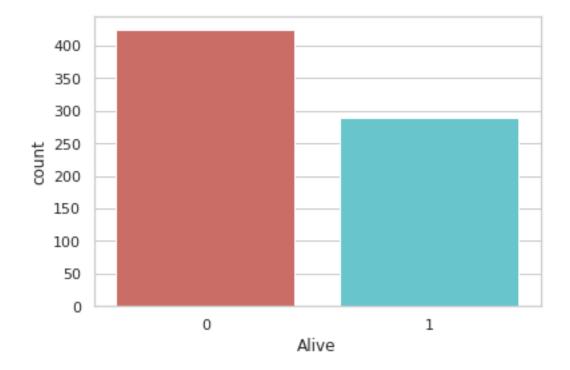
```
[314]: import pandas as pd
       import numpy as np
       import matplotlib.pyplot as plt
       from sklearn.linear_model import LogisticRegression
       from sklearn.model_selection import train_test_split
       import seaborn as sns
       sns.set(style="white")
       sns.set(style="whitegrid", color_codes=True)
       from sklearn import metrics
       from sklearn.metrics import classification_report, confusion_matrix
[315]: # load the SANK dataset:
       sank_data=pd.read_csv("Sank.csv",header=0)
[316]: print(sank_data.shape)
       print(list(sank_data.columns))
      (891, 5)
      ['Alive', 'Class', 'Sex', 'Age', 'Fare']
[317]: sank_data
[317]:
            Alive Class
                             Sex
                                   Age
                                           Fare
                0
                            male
                                 22.0
                                         7.2500
       1
                1
                       1 female
                                 38.0
                                        71.2833
       2
                1
                       3 female 26.0
                                        7.9250
                1
                       1 female 35.0
                                        53.1000
                0
                       3
                            male 35.0
                                         8.0500
       886
                0
                       2
                            male 27.0
                                       13.0000
       887
                       1 female 19.0
                                        30.0000
                1
                0
                       3 female
       888
                                  {\tt NaN}
                                        23.4500
       889
                1
                       1
                            male 26.0
                                        30.0000
       890
                            male 32.0
                                        7.7500
       [891 rows x 5 columns]
[318]: sank_data = sank_data.dropna()
```

[319]: sank_data

```
[319]:
             Alive
                    Class
                               Sex
                                      Age
                                               Fare
       0
                 0
                         3
                              male
                                     22.0
                                            7.2500
       1
                 1
                           female
                                     38.0
                                           71.2833
       2
                 1
                            female
                                     26.0
                                            7.9250
       3
                 1
                         1
                            female
                                     35.0
                                           53.1000
                 0
                         3
                                            8.0500
       4
                              male
                                     35.0
                 0
                                     39.0
       885
                         3
                           female
                                           29.1250
                 0
                         2
                                     27.0
                                           13.0000
       886
                              male
       887
                 1
                            female
                                           30.0000
                                     19.0
       889
                              male
                                     26.0
                                           30.0000
       890
                         3
                              male
                                     32.0
                                            7.7500
```

[714 rows x 5 columns]

```
[320]: sns.countplot(x="Alive",data=sank_data,palette='hls') plt.show()
```



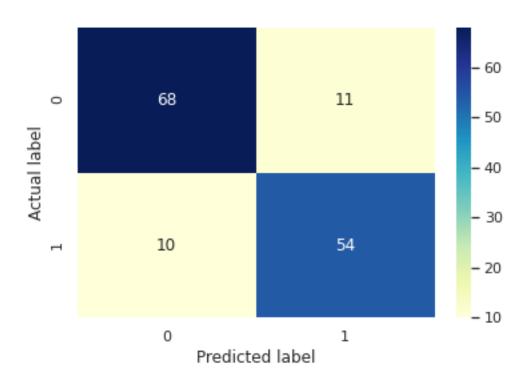
```
[321]: X = sank_data.loc[:, sank_data.columns != 'Alive']
y = sank_data.loc[:, sank_data.columns == 'Alive']
X
```

```
[321]:
            Class
                      Sex
                            Age
                                    Fare
                          22.0
                                  7.2500
       0
                3
                     male
       1
                1
                  female 38.0 71.2833
       2
                3
                   female 26.0
                                  7.9250
       3
                1
                   female 35.0
                                 53.1000
       4
                3
                     male
                          35.0
                                  8.0500
       885
                3
                   female
                          39.0
                                 29.1250
       886
                2
                     male 27.0 13.0000
       887
                1
                  female
                          19.0 30.0000
       889
                1
                           26.0 30.0000
                     male
       890
                3
                     male 32.0
                                  7.7500
       [714 rows x 4 columns]
[322]: d = {'male': 1, 'female': 0}
       X = X.replace({"Sex": d})
[323]: X
            Class
[323]:
                   Sex
                         Age
                                 Fare
       0
                3
                        22.0
                               7.2500
       1
                1
                     0 38.0 71.2833
       2
                3
                     0 26.0
                               7.9250
                     0 35.0
       3
                1
                              53.1000
                        35.0
                3
       4
                     1
                               8.0500
                     0 39.0
                              29.1250
       885
                3
                2
                     1 27.0 13.0000
       886
       887
                     0 19.0
                              30.0000
       889
                1
                        26.0
                              30.0000
       890
                3
                     1 32.0
                               7.7500
       [714 rows x 4 columns]
[324]: X_train, X_test, y_train, y_test=train_test_split(X,y,test_size=0.2,random_state=0)
[325]: logreg = LogisticRegression()
       # fit the model with data
       logreg = logreg.fit(X_train,y_train)
       y_pred=logreg.predict(X_test)
```

/opt/conda/lib/python3.9/site-packages/sklearn/utils/validation.py:63: DataConversionWarning: 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().
        return f(*args, **kwargs)
[326]: print("Model Score:",logreg.score(X_test,y_test))
      Model Score: 0.8531468531468531
[327]: cnfusion_matrix = metrics.confusion_matrix(y_test, y_pred)
       cnfusion_matrix
[327]: array([[68, 11],
              [10, 54]
[328]: # Heat Map Visualization
       # fig, ax = plt.subplots()
       tick_marks = np.arange(len(sank_data.Alive))
       plt.xticks(tick_marks, sank_data.Alive)
       plt.yticks(tick_marks, sank_data.Alive)
       sns.heatmap(pd.DataFrame(cnfusion_matrix), annot=True, cmap="YlGnBu",fmt='g')
       plt.title('Confusion matrix', y=1.1)
       plt.ylabel('Actual label')
       plt.xlabel('Predicted label')
[328]: Text(0.5, 12.5, 'Predicted label')
```

Confusion matrix



```
[329]: print(classification_report(y_test, logreg.predict(X_test)))
```

	precision	recall	f1-score	support
0	0.87	0.86	0.87	79
1	0.83	0.84	0.84	64
accuracy			0.85	143
macro avg	0.85	0.85	0.85	143
weighted avg	0.85	0.85	0.85	143

```
[330]: # Male = 1, Female = 0; As per our defined Dictionary d = {'male': 1, 'female': → 0}

# Class = 1, Male, Age: 28, Fare: 20.5
logreg.predict((np.array([1, 1, 28, 20.5]).reshape(1, -1)))
```

[330]: array([1])

```
[331]: # Class = 2, Male, Age: 70, Fare: 7.5
    logreg.predict((np.array([2, 1, 70, 7.5]).reshape(1, -1)))

[331]: array([0])

[332]: # Class = 3, Female, Age: 25, Fare: 6.76
    logreg.predict((np.array([3, 0, 25, 6.76]).reshape(1, -1)))

[332]: array([1])

[333]: # Class = 2, Female, Age: 43, Fare: 12.88
    logreg.predict((np.array([2, 0, 43, 12.88]).reshape(1, -1)))

[333]: array([1])

[334]: print("Accuracy:", metrics.accuracy_score(y_test, y_pred))
    print("Precision:", metrics.precision_score(y_test, y_pred))
```

Accuracy: 0.8531468531468531 Precision: 0.8307692307692308