qbn3llhwk

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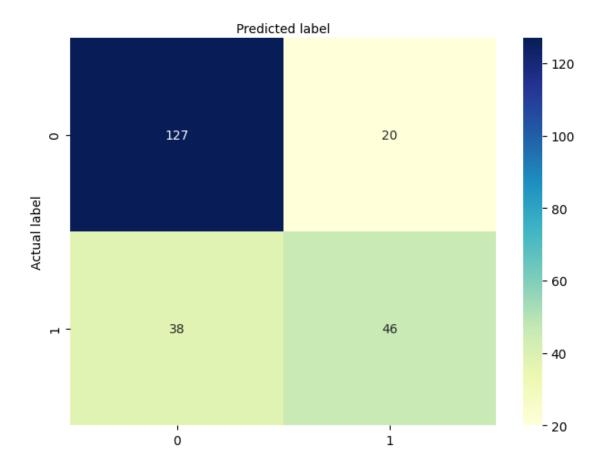
[]: import pandas as pd

```
import numpy as np
     from sklearn.linear_model import LogisticRegression
     from sklearn.model_selection import train_test_split
     import seaborn as sns
     import matplotlib.pyplot as plt
     dataset= pd.read_csv("diabetes_dataset.csv")
[]: dataset.describe()
[]:
            Pregnancies
                             Glucose
                                                      SkinThickness
                                      BloodPressure
                                                                         Insulin
             768.000000
                          768.000000
                                          768.000000
                                                         768.000000
                                                                      768.000000
     mean
               3.845052
                          120.894531
                                          69.105469
                                                          20.536458
                                                                       79.799479
     std
               3.369578
                           31.972618
                                          19.355807
                                                          15.952218
                                                                      115.244002
    min
               0.000000
                            0.000000
                                            0.000000
                                                           0.000000
                                                                        0.000000
     25%
               1.000000
                           99.000000
                                          62.000000
                                                           0.000000
                                                                        0.000000
     50%
               3.000000
                          117.000000
                                                          23.000000
                                                                       30.500000
                                          72.000000
     75%
               6.000000
                          140.250000
                                          80.000000
                                                          32.000000
                                                                      127.250000
              17.000000
                          199.000000
                                                                      846.000000
     max
                                          122.000000
                                                          99.000000
                         DiabetesPedigreeFunction
                                                           Age
                                                                   Outcome
            768.000000
                                       768.000000
                                                    768.000000
                                                                768.000000
     count
     mean
             31.992578
                                         0.471876
                                                     33.240885
                                                                   0.348958
     std
              7.884160
                                                     11.760232
                                                                   0.476951
                                         0.331329
                                                     21.000000
     min
              0.000000
                                         0.078000
                                                                   0.000000
     25%
             27.300000
                                         0.243750
                                                     24.000000
                                                                   0.000000
     50%
             32.000000
                                         0.372500
                                                     29.000000
                                                                   0.000000
     75%
             36.600000
                                         0.626250
                                                     41.000000
                                                                   1.000000
     max
             67.100000
                                         2.420000
                                                     81.000000
                                                                   1.000000
[]: from sklearn.model_selection import train_test_split
     X= dataset.drop("Outcome", axis=1)
     y= dataset[["Outcome"]]
     X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.30,__
      →random state=7)
```

```
[]: model = LogisticRegression()
     model.fit(X_train, y_train)
     y_predict = model.predict(X_test)
     model_score= model.score (X_test, y_test)
    model score
    /usr/local/lib/python3.10/dist-packages/sklearn/utils/validation.py:1143:
    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().
      y = column_or_1d(y, warn=True)
    /usr/local/lib/python3.10/dist-packages/sklearn/linear_model/_logistic.py:458:
    ConvergenceWarning: lbfgs failed to converge (status=1):
    STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
    Increase the number of iterations (max_iter) or scale the data as shown in:
        https://scikit-learn.org/stable/modules/preprocessing.html
    Please also refer to the documentation for alternative solver options:
        https://scikit-learn.org/stable/modules/linear_model.html#logistic-
    regression
      n_iter_i = _check_optimize_result(
[]: 0.7489177489177489
[]: from sklearn import metrics
     cnf_matrix = metrics.confusion_matrix(y_test, y_predict)
     print(cnf_matrix)
    [[127 20]
     [ 38 46]]
[]: class_names=[0,1]
     fig, ax = plt.subplots()
     tick_marks = np.arange(len(class_names))
     plt.xticks(tick marks, class names)
     plt.yticks(tick_marks, class_names)
     sns.heatmap(pd.DataFrame(cnf_matrix), annot=True, cmap="YlGnBu" ,fmt='g')
     ax.xaxis.set_label_position("top")
     plt.tight_layout()
     plt.title('Confusion matrix', y=1.1)
     plt.ylabel('Actual label')
     plt.xlabel('Predicted label')
```

[]: Text(0.5, 427.9555555555555, 'Predicted label')

Confusion matrix



```
[]: true_neg, false_pos, false_neg, true_pos = cnf_matrix.ravel()
true_neg, false_pos, false_neg, true_pos
total = true_neg + false_pos + false_neg + true_pos
```

```
[]: accuracy = (true_pos + true_neg)/total print(accuracy)
```

0.7489177489177489

```
[]: precision = true_pos/(true_pos + false_pos)
print(precision)
```

0.6969696969697

```
[]: recall = true_pos/(true_pos + false_neg)
print(recall)
```

0.5476190476190477

```
[]: f1_score = (2*precision*recall)/(precision+recall)
print(f1_score)
```

0.61333333333333333