Creado por:

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14.13

In [7]: | sc = StandardScaler()

4.10 2.74

24.5

In [8]: **from** sklearn.discriminant_analysis **import** LinearDiscriminantAnalysis **as** LDA

-2.629893 -3.660171

0.831072

3.078393

Linear Discriminant Analysis (LDA)

```
In [2]: import numpy as np
        import matplotlib.pyplot as plt
        import pandas as pd
        from sklearn.model selection import train test split
        from sklearn.preprocessing import StandardScaler
In [3]: dataset = pd.read csv("Wine.csv")
```

```
dataset
                 Alcohol Malic_Acid Ash Ash_Alcanity Magnesium Total_Phenols Flavanoids Nonflavanoid_Phenols Proanthocyanins
Out[3]:
             0
                   14.23
                                 1.71 2.43
                                                       15.6
                                                                     127
                                                                                     2.80
                                                                                                 3.06
                                                                                                                          0.28
                                                                                                                                             2.29
                                                                                                                                                              5.6
             1
                   13.20
                                 1.78 2.14
                                                       11.2
                                                                     100
                                                                                     2.65
                                                                                                 2.76
                                                                                                                          0.26
                                                                                                                                             1.28
                                                                                                                                                              4.3
             2
                   13.16
                                 2.36 2.67
                                                       18.6
                                                                     101
                                                                                     2.80
                                                                                                 3.24
                                                                                                                          0.30
                                                                                                                                             2.81
             3
                   14.37
                                                       16.8
                                                                                     3.85
                                                                                                                                             2.18
                                                                                                                                                              7.8
                                 1.95 2.50
                                                                     113
                                                                                                 3.49
                                                                                                                          0.24
                                                       21.0
                                                                                     2.80
                                                                                                 2.69
              4
                   13.24
                                 2.59 2.87
                                                                     118
                                                                                                                          0.39
                                                                                                                                             1.82
                                                                                                                                                              4.3
                                                       20.5
                                                                                     1.68
                                                                                                 0.61
                                                                                                                          0.52
                                                                                                                                                              7.7
           173
                   13.71
                                 5.65 2.45
                                                                      95
                                                                                                                                             1.06
           174
                   13.40
                                 3.91 2.48
                                                       23.0
                                                                     102
                                                                                     1.80
                                                                                                 0.75
                                                                                                                          0.43
                                                                                                                                             1.41
                                                                                                                                                              7.3
           175
                   13.27
                                 4.28 2.26
                                                       20.0
                                                                     120
                                                                                     1.59
                                                                                                 0.69
                                                                                                                          0.43
                                                                                                                                             1.35
                                                                                                                                                             10.2
                                                       20.0
                                                                                                                                                              9.3
                   13.17
                                 2.59 2.37
                                                                     120
                                                                                     1.65
                                                                                                 0.68
                                                                                                                          0.53
                                                                                                                                             1.46
           176
```

2.05

0.76

0.56

1.35

	178 rows × 14 columns
In [4]:	dataset.shape
Out[4]:	(178, 14)
In [5]:	<pre>X = dataset.iloc[:, 0:13].values y = dataset.iloc[:, 13].values</pre>
In [6]:	<pre>X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)</pre>

X_train = sc.fit_transform(X_train) X_test = sc.transform(X_test)

96

```
lda = LDA(n_components=2)
        X_train = lda.fit_transform(X_train, y_train)
        X_test = lda.transform(X_test)
In [9]:
        principal_Df = pd.DataFrame(data = X_train
                       columns = ['principal component 1', 'principal component 2'])
```

principal Df principal component 1 principal component 2 Out[9]:

```
0
                   3.573156
                                           1.940189
  1
                   0.854759
                                           -2.081830
                   0.621737
                                            3.062345
  3
                   4.807864
                                            2.006387
  4
                  -3.857976
                                            0.149873
137
                   1.686471
                                           -3.834276
```

In [10]: **from** sklearn.linear model **import** LogisticRegression classifier = LogisticRegression(random_state=0)

In [11]: y_pred = classifier.predict(X_test)

[0, 16, 0],

plt.legend() plt.show()

142 rows × 2 columns

-0.902058

-0.191056

-4.206327

4.529108

138

139

140

141

```
classifier.fit(X_train, y_train)
Out[10]: ▼
                  LogisticRegression
         LogisticRegression(random_state=0)
```

y_pred Out[11]: array([1, 3, 2, 1, 2, 2, 1, 3, 2, 2, 3, 3, 1, 2, 3, 2, 1, 1, 2, 1, 2, 1, 1, 2, 2, 2, 2, 2, 3, 1, 1, 2, 1, 1, 1])

```
In [12]: from sklearn.metrics import confusion_matrix
         cm = confusion_matrix(y_test, y_pred)
```

```
cm
Out[12]: array([[14, 0, 0],
```

```
[0, 0, 6]]
In [13]: from matplotlib.colors import ListedColormap
         X_set, y_set = X_train, y_train
```

plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).reshape(X1.shape),

alpha = 0.75, cmap = ListedColormap(('red', 'green', 'blue')))

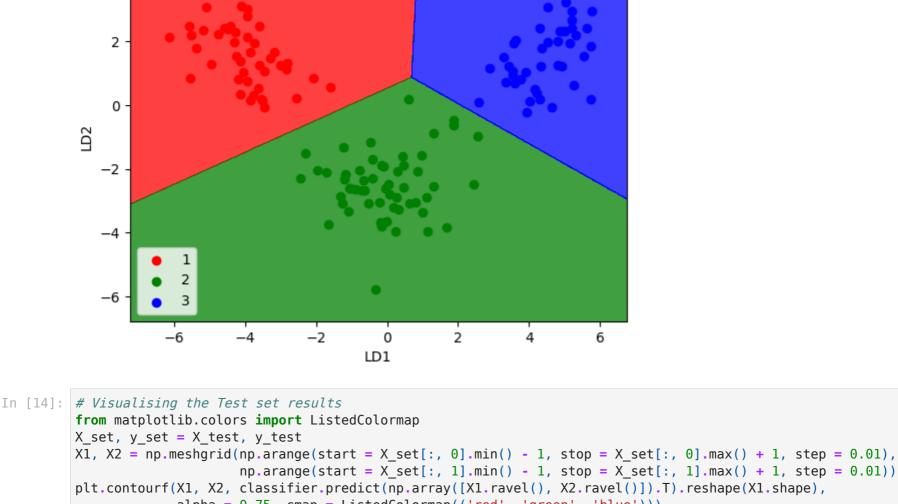
X1, $X2 = np.meshgrid(np.arange(start = X_set[:, 0].min() - 1, stop = X_set[:, 0].max() + 1, step = 0.01),$

 $np.arange(start = X_set[:, 1].min() - 1, stop = X_set[:, 1].max() + 1, step = 0.01))$

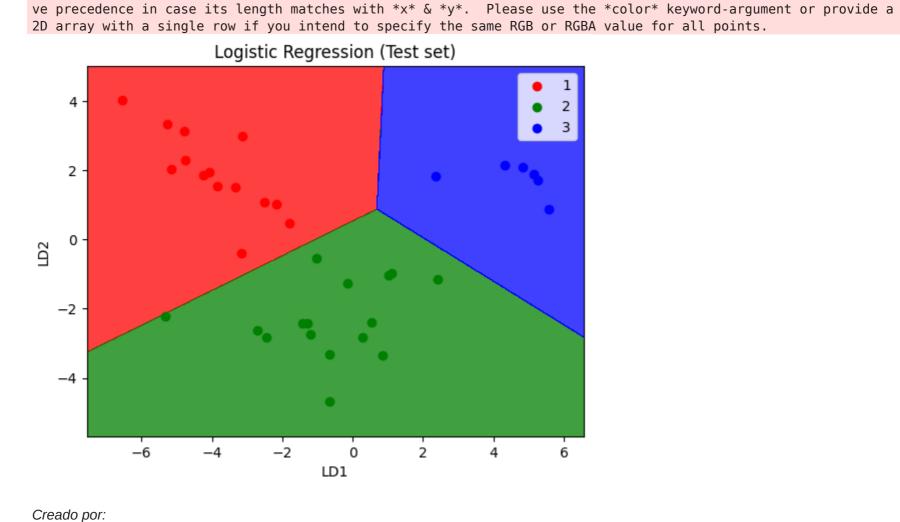
```
plt.xlim(X1.min(), X1.max())
plt.ylim(X2.min(), X2.max())
for i, j in enumerate(np.unique(y_set)):
    plt.scatter(X_set[y_set == j, 0], X_set[y_set == j, 1],
                c = ListedColormap(('red', 'green', 'blue'))(i), label = j)
plt.title('Logistic Regression (Training set)')
plt.xlabel('LD1')
plt.ylabel('LD2')
```

2D array with a single row if you intend to specify the same RGB or RGBA value for all points. *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will ha ve precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA value for all points. *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will ha ve precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a 2D array with a single row if you intend to specify the same RGB or RGBA value for all points. Logistic Regression (Training set)

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```
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                c = ListedColormap(('red', 'green', 'blue'))(i), label = j)
plt.title('Logistic Regression (Test set)')
plt.xlabel('LD1')
plt.ylabel('LD2')
plt.legend()
plt.show()
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ve precedence in case its length matches with *x* & *y*. Please use the *color* keyword-argument or provide a
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