Creado por:

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Naives Bayes

```
In [1]: import warnings
        warnings.filterwarnings("ignore")
 In [2]: from sklearn import datasets
        wine = datasets.load wine()
 In [3]: print("Features: ", wine.feature_names)
        print("Labels: ", wine.target_names)
        Features: ['alcohol', 'malic_acid', 'ash', 'alcalinity_of_ash', 'magnesium', 'total_phenols', 'flavanoids', 'n
        onflavanoid_phenols', 'proanthocyanins', 'color_intensity', 'hue', 'od280/od315_of_diluted_wines', 'proline']
        Labels: ['class_0' 'class_1' 'class_2']
 In [4]: wine.data.shape
 Out[4]: (178, 13)
 In [6]: import pandas as pd
        df = pd.DataFrame(wine.data, columns=wine.feature_names)
            alcohol malic_acid ash alcalinity_of_ash magnesium total_phenols flavanoids nonflavanoid_phenols proanthocyanins color_intens
 Out[6]:
          0
             14.23
                      1.71 2.43
                                       15.6
                                               127.0
                                                                  3.06
                                                                                  0.28
                                                                                              2.29
                                                                                                         5
                                                          2.80
          1
             13.20
                      1.78 2.14
                                       11.2
                                               100.0
                                                          2.65
                                                                  2.76
                                                                                  0.26
                                                                                              1.28
                                                                                                         4
          2
             13.16
                      2.36 2.67
                                       18.6
                                               101.0
                                                          2.80
                                                                  3.24
                                                                                  0.30
                                                                                              2.81
                                                                                                         5
                                                                                                         7
          3
             14.37
                      1.95 2.50
                                       16.8
                                               113.0
                                                          3.85
                                                                  3.49
                                                                                  0.24
                                                                                              2.18
             13.24
                      2.59 2.87
                                       21.0
                                               118.0
                                                          2.80
                                                                  2.69
                                                                                  0.39
                                                                                              1.82
                                                                                                         4
        173
             13.71
                      5.65 2.45
                                       20.5
                                                95.0
                                                          1.68
                                                                  0.61
                                                                                  0.52
                                                                                              1.06
                                                                                                         7
        174
             13.40
                      3.91 2.48
                                       23.0
                                               102.0
                                                          1.80
                                                                  0.75
                                                                                  0.43
                                                                                              1.41
                                                                                                         7
             13.27
                      4.28 2.26
                                       20.0
                                               120.0
                                                          1.59
                                                                  0.69
                                                                                  0.43
                                                                                              1.35
                                                                                                         10
        175
                                       20.0
                                                                                                         9
        176
             13.17
                      2.59 2.37
                                               120.0
                                                          1.65
                                                                  0.68
                                                                                  0.53
                                                                                              1.46
                                                                                                         9
        177
             14.13
                      4.10 2.74
                                       24.5
                                                96.0
                                                          2.05
                                                                  0.76
                                                                                  0.56
                                                                                              1.35
       178 rows × 13 columns
 In [7]: wine.target
 2, 2])
 In [8]: | from sklearn.model_selection import train_test_split
        X_train, X_test, y_train, y_test = train_test_split(wine.data, wine.target, test_size=0.3, random_state=109)
In [12]: from sklearn.naive_bayes import GaussianNB
        gnb = GaussianNB()
        gnb.fit(X_train, y_train)
Out[12]:
       ▼ GaussianNB
        GaussianNB()
In [13]: y_pred = gnb.predict(X_test)
        y_pred
\texttt{Out[13]: array([0, \, 0, \, 1, \, 2, \, 0, \, 1, \, 0, \, 0, \, 1, \, 0, \, 2, \, 2, \, 2, \, 2, \, 0, \, 1, \, 1, \, 0, \, 0, \, 1, \, 2, \, 1,}
              0, 2, 0, 0, 1, 2, 0, 1, 2, 1, 1, 0, 1, 1, 0, 2, 2, 0, 2, 1, 0, 0,
              0, 2, 2, 0, 1, 1, 2, 0, 0, 2])
In [14]: y_test
Out[14]: array([0, 0, 1, 2, 0, 1, 0, 1, 1, 0, 1, 1, 2, 2, 0, 1, 1, 0, 0, 1, 2, 1,
              0, 2, 0, 0, 1, 2, 0, 1, 2, 1, 1, 0, 1, 1, 0, 2, 2, 0, 2, 0, 0, 0,
              0, 2, 2, 0, 1, 1, 2, 1, 0, 2])
In [15]: from sklearn import metrics
        print("Accuracy: ", metrics.accuracy_score(y_test, y_pred))
        Accuracy: 0.9074074074074
```