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TE IT

Batch: D

2018140059

ML - Exp 3 - Wine Quality Classification Daataset

```
#importing necessary libraries
import numpy as np
import pandas as pd
import warnings
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn import metrics
from sklearn.neural_network import MLPClassifier
import matplotlib.pyplot as plt
```

from google.colab import drive drive.mount("/content/gdrive")

Drive already mounted at /content/gdrive; to attempt to forcibly remount, call drive.mount("/content/gdrive", force_remount=True).

wine = pd.read_csv('/content/gdrive/My Drive/datasets/wine.csv',encoding= 'unicode_escape')

wine.head()

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol	quality
0	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4	bad
1	7.8	0.88	0.00	2.6	0.098	25.0	67.0	0.9968	3.20	0.68	9.8	bad
2	7.8	0.76	0.04	2.3	0.092	15.0	54.0	0.9970	3.26	0.65	9.8	bad
3	11.2	0.28	0.56	1.9	0.075	17.0	60.0	0.9980	3.16	0.58	9.8	good
4	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	0.56	9.4	bad

wine.isnull().sum()

fixed acidity volatile acidity citric acid residual sugar chlorides free sulfur dioxide total sulfur dioxide density sulphates alcohol quality

plt.subplots_adjust(left=0, bottom=0.5, right=0.9, top=0.9, wspace=0.5, hspace=0.8)

dtype: int64 plt.figure(figsize=(40,25)) plt.subplot(141) plt.title('Percentage of players from each country',fontsize = 20) wine['quality'].value_counts().plot.pie(autopct="%1.1f%%")

wine.head()

```
fixed acidity volatile acidity citric acid residual sugar chlorides free sulfur dioxide total sulfur dioxide density pH sulphates alcohol quality
         7.4
                       0.70
                                   0.00
                                                                             11.0
                                                                                                 34.0 0.9978 3.51
                                                                                                                       0.56
                                                   1.9
                                                          0.076
                                                                                                                                9.4
         7.8
                        0.88
                                   0.00
                                                          0.098
                                                                              25.0
                                                                                                 67.0 0.9968 3.20
                                                                                                                       0.68
                                                                                                                                9.8
                                                                                                 54.0 0.9970 3.26
        7.8
                        0.76
                                   0.04
                                                  2.3
                                                          0.092
                                                                              15.0
                                                                                                                                9.8
                                   0.56
                                                                             17.0
        11.2
                                                          0.075
                                                                                                 60.0 0.9980 3.16
                                                                             11.0
                                   0.00
                                                          0.076
                                                                                                 34.0 0.9978 3.51
                                                                                                                                9.4
                                                  1.9
```

Y = wine['quality']

```
9
0 0
1 0
2 0
3 1
4 0
...
1594 0
1595 1
1596 1
1597 0
1598 1
Name: quality, Length: 1599, dtype: int64
```

name: quartey, tengen: 1999, acype:

X = wine.drop(['quality'],axis = 1)

X

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	sulphates	alcohol
0	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	9.4
1	7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20	0.68	9.8
2	7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26	0.65	9.8
3	11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16	0.58	9.8
4	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	0.56	9.4
1594	6.2	0.600	0.08	2.0	0.090	32.0	44.0	0.99490	3.45	0.58	10.5
1595	5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52	0.76	11.2
1596	6.3	0.510	0.13	2.3	0.076	29.0	40.0	0.99574	3.42	0.75	11.0
1597	5.9	0.645	0.12	2.0	0.075	32.0	44.0	0.99547	3.57	0.71	10.2
1598	6.0	0.310	0.47	3.6	0.067	18.0	42.0	0.99549	3.39	0.66	11.0

1599 rows × 11 columns

from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
training_set_scaled = sc.fit_transform(X)

num_val = 0.2

X_train, X_test, y_train, y_test = train_test_split(training_set_scaled, Y, test_size=num_val, random_state=23)

mlp = MLPClassifier(hidden_layer_sizes=(15,15,15), activation='relu', solver='adam', max_iter=1000)
mlp.fit(X_train,y_train)

/usr/local/lib/python3.6/dist-packages/sklearn/neural_network/_multilayer_perceptron.py:571: ConvergenceWarning: Stochastic Optimizer: Maximum iterations (1000) reached and the optimization hasn't converged yet. % self.max_iter, ConvergenceWarning)

y_pred = mlp.predict(X_test)

print("Accuracy:",metrics.accuracy_score(y_test, y_pred))

predict_train = mlp.predict(X_train)
predict_test = mlp.predict(X_test)

from sklearn.metrics import classification_report,confusion_matrix

cf_matrix = confusion_matrix(y_train,predict_train)

print(confusion_matrix(y_train,predict_train)) print(classification_report(y_train,predict_train))

[[550 48] [37 644]]	precision	recall	f1-score	support
	brecision	recarr	11-30016	support
0	0.94	0.92	0.93	598
1	0.93	0.95	0.94	681
accuracy			0.93	1279
macro avg	0.93	0.93	0.93	1279
weighted avg	0.93	0.93	0.93	1279

mlp.predict([[7.4, 0.700, 0.00, 1.9, 0.076, 11.0, 34.0, 0.99780, 3.51, 0.56, 9.4]])
array([0])

sns.heatmap(cf_matrix, annot=True)

<matplotlib.axes._subplots.AxesSubplot at 0x7f3429696908>

