```
import pandas as pd
         import numpy as np
         from sklearn.datasets import load_wine
In [8]:
         winedata = load_wine()
In [9]:
         winedata.keys()
        dict_keys(['data', 'target', 'frame', 'target_names', 'DESCR', 'feature_names'])
Out[9]
In [10]:
         dfwine = pd.DataFrame(winedata["data"], columns =winedata ["feature_names"])
In [11]:
         dfwine
            alcohol malic_acid ash alcalinity_of_ash magnesium total_phenols flavanoids nonflavanoid_phenols proanthocyanins color_intensity hue od280/od315_of_diluted_wines proline
Out[11]:
                                                          2.80
             14.23
                      1.71 2.43
                                       15.6
                                               127.0
                                                                  3.06
                                                                                  0.28
                                                                                               2.29
                                                                                                          5.64 1.04
                                                                                                                                   3.92 1065.0
             13.20
                      1.78 2.14
                                       11.2
                                               100.0
                                                          2.65
                                                                  2.76
                                                                                  0.26
                                                                                               1.28
                                                                                                          4.38 1.05
                                                                                                                                   3.40 1050.0
                                               101.0
                                                          2.80
             13.16
                      2.36 2.67
                                                                                  0.30
                                                                                               2.81
                                                                                                          5.68 1.03
                                                                                                                                   3.17 1185.0
                                       18.6
                                                                  3.24
             14.37
                      1.95 2.50
                                       16.8
                                               113.0
                                                          3.85
                                                                  3.49
                                                                                  0.24
                                                                                               2.18
                                                                                                          7.80 0.86
                                                                                                                                   3.45 1480.0
             13.24
                                               118.0
                                                                                  0.39
                                                                                               1.82
                      2.59 2.87
                                       21.0
                                                          2.80
                                                                  2.69
                                                                                                          4.32 1.04
                                                                                                                                   2.93
                                                                                                                                       735.0
        173
             13.71
                      5.65 2.45
                                       20.5
                                                95.0
                                                          1.68
                                                                  0.61
                                                                                  0.52
                                                                                               1.06
                                                                                                          7.70 0.64
                                                                                                                                   1.74
                                                                                                                                        740.0
                      3.91 2.48
                                                                                                          7.30 0.70
        174
             13.40
                                       23.0
                                               102.0
                                                          1.80
                                                                  0.75
                                                                                  0.43
                                                                                               1.41
                                                                                                                                   1.56
                                                                                                                                         750.0
             13.27
                                               120.0
                                                          1.59
                                                                                  0.43
        175
                      4.28 2.26
                                       20.0
                                                                  0.69
                                                                                               1.35
                                                                                                         10.20 0.59
                                                                                                                                   1.56
                                                                                                                                         835.0
        176
             13.17
                      2.59 2.37
                                       20.0
                                               120.0
                                                          1.65
                                                                  0.68
                                                                                  0.53
                                                                                               1.46
                                                                                                          9.30 0.60
                                                                                                                                   1.62
                                                                                                                                         840.0
                                                                                  0.56
        177
             14.13
                      4.10 2.74
                                       24.5
                                                96.0
                                                          2.05
                                                                  0.76
                                                                                               1.35
                                                                                                          9.20 0.61
                                                                                                                                   1.60
                                                                                                                                         560.0
       178 rows × 13 columns
In [12]:
         X = dfwine.copy()
In [13]:
        X.isnull().sum()
        alcohol
                                    0
Out[13]:
        malic_acid
        alcalinity_of_ash
        magnesium
        total_phenols
        flavanoids
        nonflavanoid_phenols
        proanthocyanins
        color_intensity
        hue
                                    0
        od280/od315_of_diluted_wines
                                    0
        proline
        dtype: int64
         X["alcohol"].skew()
        -0.051482331077134784
       y = winedata["target"]
In [16]:
        2, 2])
         from sklearn.model_selection import train_test_split
In [18]:
        Xtrain, Xtest, ytrain, ytest = train_test_split(X,y, test_size=0.2)
In [19]:
         Xtrain.shape, Xtest.shape, X.shape
        ((142, 13), (36, 13), (178, 13))
Out[19]
In [22]:
         from sklearn.tree import DecisionTreeClassifier
         dtent = DecisionTreeClassifier(criterion="entropy")
In [26]:
         dtgini = DecisionTreeClassifier()
         dtent.fit(Xtrain,ytrain)
        DecisionTreeClassifier(criterion='entropy')
In [28]:
         dtgini.fit(Xtrain, ytrain)
        DecisionTreeClassifier()
Out[28]
In [29]:
         dtent.score(Xtest, ytest)
        0.916666666666666
Out[29]:
In [30]:
         dtgini.score(Xtest,ytest)
        \tt 0.94444444444444444
Out[30]
In [31]:
         from sklearn.model_selection import KFold, cross_val_score
         kf = KFold(n_splits=10)
In [33]:
         giniscore = cross_val_score(DecisionTreeClassifier(), X,y, cv= kf)
         giniscore
        array([1.
                       , 0.83333333, 0.88888889, 0.61111111, 0.77777778,
              0.9444444, 0.94444444, 0.72222222, 0.88235294, 0.94117647])
In [35]:
         giniscore.mean()
        0.8545751633986928
Out[35]:
In [36]:
         entroscore = cross_val_score(DecisionTreeClassifier(criterion="entropy"), X, y, cv = kf)
In [37]:
         entroscore.mean()
        {\tt 0.8872549019607842}
Out[37]:
```

In [6]:

In []