4)Implement any 2 Classification techniques using any data analytics tool.

```
In [8]:
          import pandas as pd # to load dataset
          import matplotlib.pyplot as plt
          import seaborn as sns
          import pylab as pl
          from sklearn.model selection import train test split # for splitting dataset
          from sklearn.preprocessing import MinMaxScaler # for scaling
          from sklearn.linear model import LogisticRegression # machine Learning lib/model, #
          from sklearn.tree import DecisionTreeClassifier # get accuracy by Decision Tree class
          from sklearn.neighbors import KNeighborsClassifier # get accuracy by KNN classifier
          from sklearn.naive bayes import GaussianNB # get accuracy by GNB classifier
          df=pd.read_csv('fruit_data.csv')
 In [9]:
In [10]:
          df.shape
          (59, 7)
Out[10]:
In [11]:
          df.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 59 entries, 0 to 58
          Data columns (total 7 columns):
           #
               Column
                              Non-Null Count
                                               Dtype
               -----
                              -----
                                               ____
           0
              fruit label
                                               int64
                              59 non-null
               fruit_name
                              59 non-null
           1
                                               object
           2
               fruit_subtype 59 non-null
                                               object
           3
              mass
                              59 non-null
                                               int64
           4
                                               float64
              width
                              59 non-null
           5
                                               float64
              height
                              59 non-null
               color_score
                              59 non-null
                                               float64
          dtypes: float64(3), int64(2), object(2)
          memory usage: 3.4+ KB
In [12]:
          df.describe()
Out[12]:
                fruit_label
                               mass
                                        width
                                                 height color_score
                 59.000000
                           59.000000
                                     59.000000
                                               59.000000
                                                          59.000000
          count
                 2.542373 163.118644
                                      7.105085
                                               7.693220
                                                          0.762881
          mean
            std
                 1.208048
                           55.018832
                                      0.816938
                                               1.361017
                                                          0.076857
                  1.000000
                           76.000000
                                      5.800000
                                               4.000000
                                                          0.550000
           min
           25%
                  1.000000 140.000000
                                      6.600000
                                               7.200000
                                                          0.720000
           50%
                  3.000000 158.000000
                                      7.200000
                                               7.600000
                                                          0.750000
           75%
                  4.000000 177.000000
                                      7.500000
                                               8.200000
                                                          0.810000
```

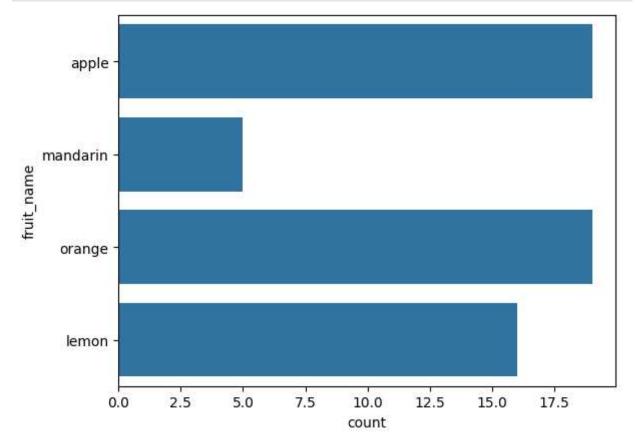
file:///D:/DS lab/DS4.html

0.930000

9.600000 10.500000

4.000000 362.000000

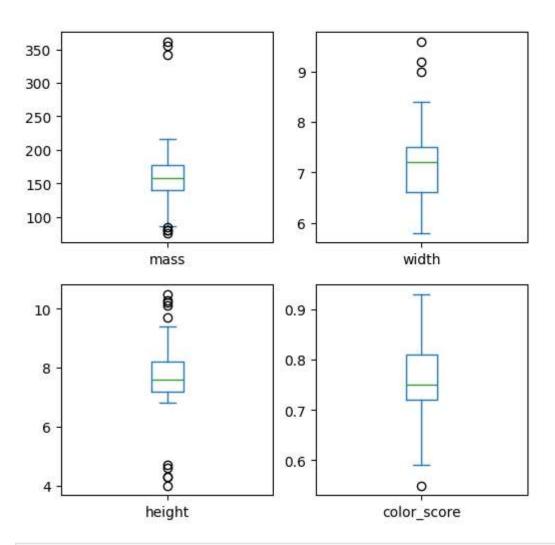
max



In [16]: df.drop('fruit_label',axis=1).plot(kind='box', subplots=True, layout=(2,2), sharex=Fal
 plt.savefig('fruits_box')
 plt.show()

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Box Plot for each input variable

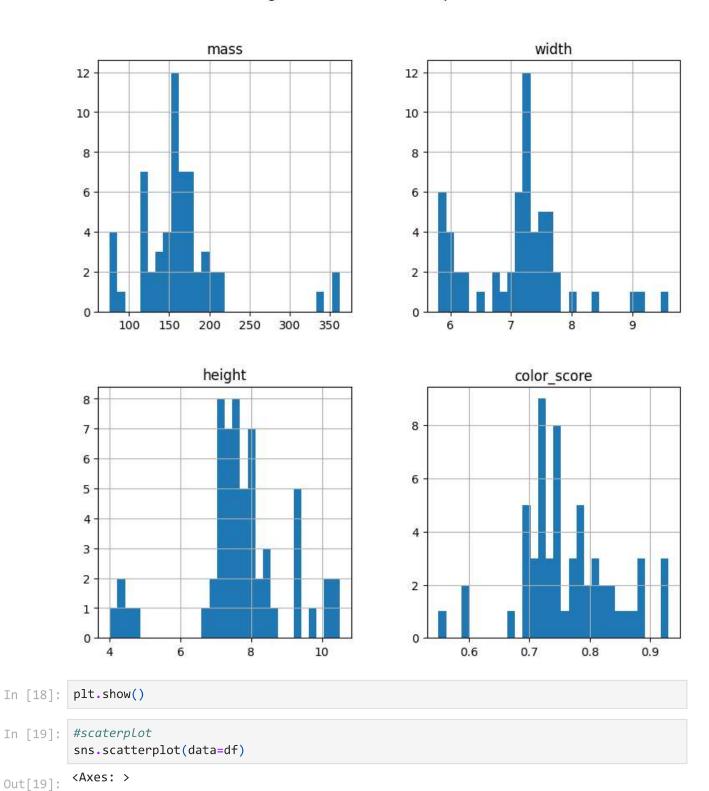


```
In [17]: import pylab as pl

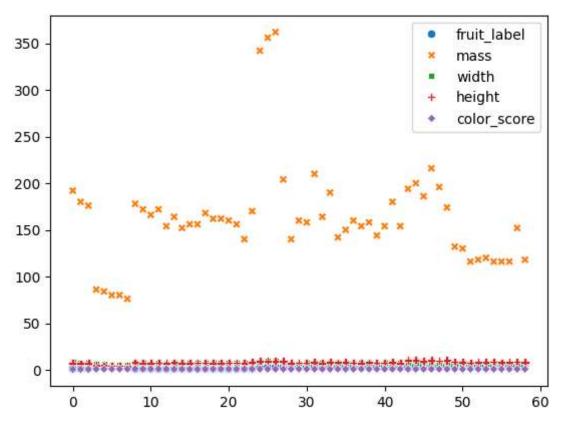
df.drop('fruit_label', axis=1).hist(bins=30, figsize=(9,9))
pl.suptitle("Histogram for each numeric input variable")
plt.savefig('fruits_hist')
```

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Histogram for each numeric input variable



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```
#preparing data with scaling
In [20]:
         import pandas as pd
         from sklearn.model_selection import train_test_split
         from sklearn.preprocessing import MinMaxScaler
         feature_names = ['mass', 'width', 'height', 'color_score']
         x=df[feature names]
         y=df['fruit_label']
         x_train, x_test, y_train, y_test = train_test_split(x,y, random_state=0)
         print(x_train[:3]) # to check output
         scaler = MinMaxScaler()
         x_train=scaler.fit_transform(x_train)
         x_test= scaler.transform(x_test)
         print("\nAfter scaling\n")
         print(x_train[:3]) # to check output
             mass width height color_score
         42
              154
                     7.2
                             7.2
                                         0.82
                                         0.72
         48
              174
                     7.3
                            10.1
         7
               76
                     5.8
                             4.0
                                         0.81
         After scaling
         [[0.27857143 0.41176471 0.49230769 0.72972973]
          [0.35
                      0.44117647 0.93846154 0.45945946]
          [0.
                                 0.
                                            0.7027027 ]]
In [21]: from sklearn.neighbors import KNeighborsClassifier
```

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KNN method

```
DS4
         knn = KNeighborsClassifier()
         knn.fit(x_train, y_train)
         #print score of train data
         print('Accuracy of KNN classifier on training set:{:.2f}'
               .format(knn.score(x train, y train)))
         #print score of test data
         print('Accuracy of KNN Classifier on test set:{:.2f}'
               .format(knn.score(x_test, y_test)))
         Accuracy of KNN classifier on training set:0.95
         Accuracy of KNN Classifier on test set:1.00
In [22]: from sklearn.naive_bayes import GaussianNB
         # Gaussian Naive bayes
         gnb = GaussianNB()
         gnb.fit(x_train, y_train)
         #print score of train data
         print('Accuracy of GNB classifier on training set:{:.2f}'
```

Accuracy of GNB classifier on training set:0.86 Accuracy of GNB Classifier on test set:0.67

.format(gnb.score(x_test, y_test)))

.format(gnb.score(x_train, y_train)))

print('Accuracy of GNB Classifier on test set:{:.2f}'

#print score of test data

```
In [ ]:
```

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