```
In [1]:
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          from matplotlib.colors import ListedColormap
In [2]:
          zero_one_colourmap = ListedColormap(('red', 'green'))
In [3]: | forest_data = pd.read_csv('forestfires.csv')
          forest_data
Out[3]:
                month day FFMC
                                     DMC
                                             DC
                                                   ISI temp RH
                                                                   wind
                                                                         rain ... monthfeb monthjan mont
             0
                   mar
                          fri
                               86.2
                                      26.2
                                            94.3
                                                   5.1
                                                          8.2
                                                               51
                                                                     6.7
                                                                          0.0
                                                                                          0
                                                                                                     0
             1
                               90.6
                                      35.4
                                           669.1
                                                        18.0
                                                                                                     0
                   oct
                         tue
                                                   6.7
                                                               33
                                                                     0.9
                                                                          0.0
                                                                                          0
             2
                   oct
                         sat
                               90.6
                                      43.7
                                           686.9
                                                   6.7
                                                        14.6
                                                               33
                                                                     1.3
                                                                          0.0
             3
                         fri
                               91.7
                                      33.3
                                            77.5
                                                   9.0
                                                         8.3
                                                               97
                                                                     4.0
                                                                          0.2
                                                                                          0
                                                                                                     0
                   mar
                                           102.2
             4
                               89.3
                                     51.3
                                                   9.6
                                                         11.4
                                                               99
                                                                     1.8
                                                                          0.0
                                                                                          0
                                                                                                     0
                   mar
                        sun
             ...
                                 ...
                                                    ...
                                                           ...
                                                                           ...
           512
                                           665.6
                               81.6
                                      56.7
                                                   1.9
                                                        27.8
                                                               32
                                                                     2.7
                                                                          0.0
                                                                                          0
                                                                                                     0
                   aug
                        sun
           513
                               81.6
                                      56.7
                                           665.6
                                                   1.9
                                                        21.9
                                                               71
                                                                     5.8
                                                                          0.0
                   aug
                        sun
           514
                                                                          0.0 ...
                               81.6
                                           665.6
                                                   1.9
                                                        21.2
                                                               70
                                                                     6.7
                                                                                                     0
                                      56.7
                   aug
                        sun
           515
                                    146.0
                                          614.7
                                                        25.6
                                                               42
                                                                                          0
                                                                                                     0
                   aug
                         sat
                               94.4
                                                  11.3
                                                                     4.0
                                                                          0.0
           516
                               79.5
                                       3.0
                                          106.7
                                                   1.1
                                                         11.8
                                                               31
                                                                     4.5
                                                                          0.0 ...
                                                                                                     0
                   nov
                        tue
          517 rows × 31 columns
In [4]: # initial analysis
```

forest_data.shape

Out[4]: (517, 31)

In [5]: forest_data.isna().sum() Out[5]: month 0 day 0 FFMC 0 DMC 0 DC 0 ISI 0 temp 0 RH0 wind 0 rain 0 area 0 dayfri 0 daymon 0 daysat 0 daysun 0 daythu 0 daytue 0 daywed 0 monthapr 0 monthaug 0 monthdec 0 monthfeb 0 monthjan 0 monthjul 0 monthjun 0 monthmar 0 monthmay 0 monthnov 0 monthoct 0 monthsep 0 size_category 0 dtype: int64

```
In [6]: forest_data.dtypes
Out[6]: month
                            object
                           object
        day
        FFMC
                          float64
        DMC
                          float64
        DC
                          float64
        ISI
                          float64
                          float64
        temp
        RH
                            int64
        wind
                          float64
        rain
                          float64
                          float64
        area
        dayfri
                            int64
        daymon
                            int64
        daysat
                            int64
        daysun
                            int64
        daythu
                            int64
        daytue
                            int64
        daywed
                            int64
        monthapr
                            int64
        monthaug
                            int64
        monthdec
                            int64
        monthfeb
                            int64
        monthjan
                            int64
        monthjul
                            int64
        monthjun
                            int64
        monthmar
                            int64
        monthmay
                            int64
        monthnov
                            int64
        monthoct
                            int64
        monthsep
                            int64
        size_category
                            object
```

```
In [7]: # converting category column to numeric form.
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
```

dtype: object

```
In [8]: forest_data['month'] = le.fit_transform(forest_data['month'])
    forest_data['day'] = le.fit_transform(forest_data['day'])
    forest_data['size_category'] = le.fit_transform(forest_data['size_category'])
    forest_data
```

Out[8]:		month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	 monthfeb	monthjan	month
	0	7	0	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	 0	0	
	1	10	5	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	 0	0	
	2	10	2	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	 0	0	
	3	7	0	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	 0	0	
	4	7	3	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	 0	0	
	512	1	3	81.6	56.7	665.6	1.9	27.8	32	2.7	0.0	 0	0	
	513	1	3	81.6	56.7	665.6	1.9	21.9	71	5.8	0.0	 0	0	
	514	1	3	81.6	56.7	665.6	1.9	21.2	70	6.7	0.0	 0	0	
	515	1	2	94.4	146.0	614.7	11.3	25.6	42	4.0	0.0	 0	0	
	516	9	5	79.5	3.0	106.7	1.1	11.8	31	4.5	0.0	 0	0	
4														- N

```
In [9]: forest_data.dtypes
 Out[9]: month
                             int32
         day
                             int32
         FFMC
                           float64
         DMC
                           float64
         DC
                           float64
         ISI
                           float64
                           float64
         temp
         RH
                             int64
         wind
                           float64
         rain
                           float64
                           float64
         area
         dayfri
                             int64
         daymon
                             int64
         daysat
                             int64
         daysun
                             int64
         daythu
                             int64
         daytue
                             int64
         daywed
                             int64
         monthapr
                             int64
         monthaug
                             int64
         monthdec
                             int64
         monthfeb
                             int64
         monthjan
                             int64
         monthjul
                             int64
         monthjun
                             int64
         monthmar
                             int64
         monthmay
                             int64
         monthnov
                             int64
         monthoct
                             int64
         monthsep
                             int64
         size_category
                             int32
         dtype: object
In [10]: #splitting the data for training and testing.
         X = forest_data.drop(['size_category'], axis=1)
         y= forest_data['size_category']
```

In [11]: X

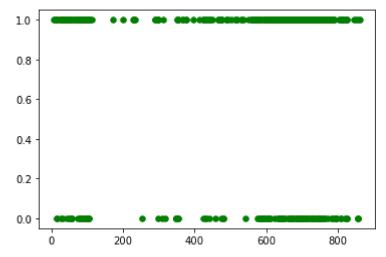
Out	[11]	
out	ובבן	

	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	 monthdec	monthfeb	mon
0	7	0	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	 0	0	
1	10	5	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	 0	0	
2	10	2	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	 0	0	
3	7	0	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	 0	0	
4	7	3	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	 0	0	
512	1	3	81.6	56.7	665.6	1.9	27.8	32	2.7	0.0	 0	0	
513	1	3	81.6	56.7	665.6	1.9	21.9	71	5.8	0.0	 0	0	
514	1	3	81.6	56.7	665.6	1.9	21.2	70	6.7	0.0	 0	0	
515	1	2	94.4	146.0	614.7	11.3	25.6	42	4.0	0.0	 0	0	
516	9	5	79.5	3.0	106.7	1.1	11.8	31	4.5	0.0	 0	0	

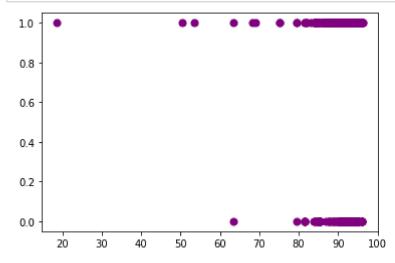
517 rows × 30 columns

```
In [12]: y
Out[12]: 0
                 1
         1
                 1
                 1
         3
                 1
         4
                 1
         512
                 0
         513
                 0
         514
                 0
          515
                 1
         516
                 1
         Name: size_category, Length: 517, dtype: int32
```

```
In [77]: plt.scatter(forest_data['DC'], y, s=30 , alpha = 1, c= 'green')
plt.show()
```



```
In [68]: plt.scatter(forest_data['FFMC'], y, s=50, c= 'purple')
plt.show()
```



Model Building

```
In [32]: from sklearn.model_selection import train_test_split
    from sklearn.metrics import confusion_matrix
    from sklearn.metrics import plot_confusion_matrix
    from sklearn.metrics import accuracy_score

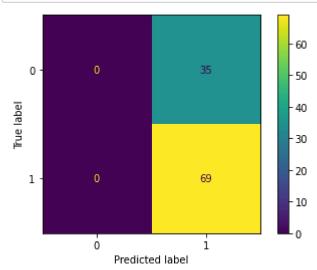
In [33]: X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.20, random_stat)
In [34]: X_train.shape, X_test.shape, y_train.shape, y_test.shape
Out[34]: ((413, 30), (104, 30), (413,), (104,))
In [35]: from sklearn.svm import SVC
```

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```
In [43]:
         classifier = SVC(kernel = 'linear', C = 0.01, gamma = 0.1)
          classifier.fit(X_train, y_train)
          y_test_pred = classifier.predict(X_test)
In [44]: | accuracy_score(y_test, y_test_pred)
Out[44]: 0.9903846153846154
In [45]: confusion_matrix(y_test, y_test_pred)
Out[45]: array([[35, 0],
                 [ 1, 68]], dtype=int64)
         plot_confusion_matrix(classifier, X_test, y_test, cmap = 'plasma')
In [46]:
          plt.show()
                                                  60
             0
                                                  50
                                                  40
          Fue label
                                                  30
                                                  20
             1 .
                                    68
                                                 10
                                     1
                        Predicted label
```

Non-Linear classification

```
In [82]: plot_confusion_matrix(classifier, X_test, y_test, cmap = 'viridis')
plt.show()
```



Scaling the data.

```
In [52]: from sklearn.preprocessing import StandardScaler
         scaler = StandardScaler()
         scaled X = scaler.fit transform(X)
In [53]: X train, X test, y train, y test= train test split(scaled X, y, test size=0.20, rand
In [54]: X train.shape, y train.shape, X test.shape, y test.shape
Out[54]: ((413, 30), (413,), (104, 30), (104,))
In [55]: X_train
Out[55]: array([[ 0.28422225, 1.69668174, -0.08063453, ..., -0.04402255,
                 -0.17285971, -0.70608125],
                [0.05533922, 0.65674759, 0.42709293, ..., -0.04402255,
                 -0.17285971, -0.70608125],
                [1.19975437, 0.13678051, -0.18943327, ..., -0.04402255,
                 -0.17285971, 1.41626761],
                [-1.31795895, -1.42312073, -1.38621943, ..., -0.04402255,
                 -0.17285971, -0.70608125],
                [-1.08907592, 0.65674759, 0.22762857, ..., -0.04402255,
                 -0.17285971, -0.70608125],
                [-0.63130986, -0.90315365, -1.07795633, ..., -0.04402255,
                 -0.17285971, -0.70608125]])
```

```
In [56]: X_test
Out[56]: array([[-0.17354381, 0.13678051,
                                             0.59029104, \ldots, -0.04402255,
                  -0.17285971, -0.70608125],
                 [-1.08907592, -0.38318657, 0.51775855, ..., -0.04402255,
                 -0.17285971, -0.70608125],
                 [1.19975437, 0.65674759, 0.08256358, ..., -0.04402255,
                  -0.17285971, 1.41626761],
                 [1.19975437, -1.42312073, 0.26389482, ..., -0.04402255,
                 -0.17285971, 1.41626761],
                 [1.19975437, 0.65674759, 0.31829419, ..., -0.04402255,
                 -0.17285971, 1.41626761],
                 [1.19975437, 1.69668174, 0.35456044, ..., -0.04402255,
                  -0.17285971,
                                1.41626761]])
 In [ ]:
In [57]: | classifier_1 = SVC(C=0.1, gamma = 0.1, kernel='rbf')
         classifier_1
         classifier 1.fit(X train, y train)
         y test pred = classifier.predict(X test)
In [58]: | accuracy_score(y_test,y_test_pred)
Out[58]: 0.6634615384615384
In [59]: |confusion_matrix(y_test,y_test_pred)
Out[59]: array([[ 0, 35],
                 [ 0, 69]], dtype=int64)
In [98]:
         plot_confusion_matrix(classifier, X_test, y_test, cmap = 'viridis')
         plt.show()
                                                60
            0
                                                50
          Frue label
                                                20
            1 .
                                   69
                                                10
                     0
                                   1
                        Predicted label
 In [ ]:
```