

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
from sklearn.svm import SVC

import warnings
warnings.filterwarnings('ignore')
```

```
In [2]: data = pd.read_csv('forestfires.csv')
data
```

Out[2]:

	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	oct	tue	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	...	0	0	
2	oct	sat	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	...	0	0	
3	mar	fri	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	...	0	0	
4	mar	sun	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	...	0	0	
...	...	...	...	...	...	...	...	...	...	...	...	...	...	
512	aug	sun	81.6	56.7	665.6	1.9	27.8	32	2.7	0.0	...	0	0	
513	aug	sun	81.6	56.7	665.6	1.9	21.9	71	5.8	0.0	...	0	0	
514	aug	sun	81.6	56.7	665.6	1.9	21.2	70	6.7	0.0	...	0	0	
515	aug	sat	94.4	146.0	614.7	11.3	25.6	42	4.0	0.0	...	0	0	
516	nov	tue	79.5	3.0	106.7	1.1	11.8	31	4.5	0.0	...	0	0	

517 rows × 31 columns

```
In [3]: data.shape
```

Out[3]: (517, 31)

```
In [4]: data.isna().sum()
```

```
Out[4]: month                0
        day                  0
        FPMC                 0
        DMC                  0
        DC                   0
        ISI                  0
        temp                 0
        RH                   0
        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 [5]: data.dtypes

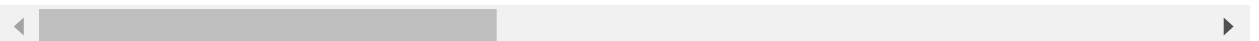
```
Out[5]: month          object
day          object
FFMC         float64
DMC          float64
DC           float64
ISI          float64
temp         float64
RH           int64
wind         float64
rain         float64
area         float64
dayfri       int64
daymon       int64
daysat      int64
daysun      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
dtype: object
```

In [7]: data.describe()

Out[7]:

	FFMC	DMC	DC	ISI	temp	RH	wind	
<b>count</b>	517.000000	517.000000	517.000000	517.000000	517.000000	517.000000	517.000000	517.00
<b>mean</b>	90.644681	110.872340	547.940039	9.021663	18.889168	44.288201	4.017602	0.02
<b>std</b>	5.520111	64.046482	248.066192	4.559477	5.806625	16.317469	1.791653	0.29
<b>min</b>	18.700000	1.100000	7.900000	0.000000	2.200000	15.000000	0.400000	0.00
<b>25%</b>	90.200000	68.600000	437.700000	6.500000	15.500000	33.000000	2.700000	0.00
<b>50%</b>	91.600000	108.300000	664.200000	8.400000	19.300000	42.000000	4.000000	0.00
<b>75%</b>	92.900000	142.400000	713.900000	10.800000	22.800000	53.000000	4.900000	0.00
<b>max</b>	96.200000	291.300000	860.600000	56.100000	33.300000	100.000000	9.400000	6.40

8 rows × 28 columns



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

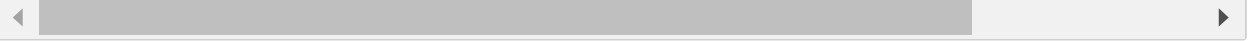
Out[8]:

	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	...	monthfeb	monthjan	monthjul
0	7	0	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	...	0	0	0
1	10	5	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	...	0	0	0
2	10	2	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	...	0	0	0
3	7	0	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	...	0	0	0
4	7	3	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	...	0	0	0

5 rows × 31 columns



```
In [13]: data = data.drop(['dayfri', 'daymon', 'daysat', 'daysun', 'daythu', 'daytue', 'day',
                        'monthaug', 'monthdec', 'monthfeb', 'monthjan', 'monthjul', 'mon',
                        'monthmay', 'monthnov', 'monthoct', 'monthsep'], axis = 1)
```



```
In [14]: data.head()
```

Out[14]:

	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	area	size_category
0	7	0	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	0.0	1
1	10	5	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	0.0	1
2	10	2	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	0.0	1
3	7	0	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	0.0	1
4	7	3	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	0.0	1

```
In [16]: std = StandardScaler()
X = data.drop('size_category',axis = 1)
std = std.fit_transform(X)
X_scaled = pd.DataFrame(data = std,columns=X.columns)
X_scaled
```

Out[16]:

	month	day	FFMC	DMC	DC	ISI	temp	RH	wir
0	0.284222	-1.423121	-0.805959	-1.323326	-1.830477	-0.860946	-1.842640	0.411724	1.4986
1	0.970871	1.176715	-0.008102	-1.179541	0.488891	-0.509688	-0.153278	-0.692456	-1.7417
2	0.970871	-0.383187	-0.008102	-1.049822	0.560715	-0.509688	-0.739383	-0.692456	-1.5182
3	0.284222	-1.423121	0.191362	-1.212361	-1.898266	-0.004756	-1.825402	3.233519	-0.0098
4	0.284222	0.136781	-0.243833	-0.931043	-1.798600	0.126966	-1.291012	3.356206	-1.2389
...	...	...	...	...	...	...	...	...	...
512	-1.089076	0.136781	-1.640083	-0.846648	0.474768	-1.563460	1.536084	-0.753800	-0.7361
513	-1.089076	0.136781	-1.640083	-0.846648	0.474768	-1.563460	0.519019	1.638592	0.9957
514	-1.089076	0.136781	-1.640083	-0.846648	0.474768	-1.563460	0.398350	1.577248	1.4986
515	-1.089076	-0.383187	0.680957	0.549003	0.269382	0.500176	1.156839	-0.140366	-0.0098
516	0.741988	1.176715	-2.020879	-1.685913	-1.780442	-1.739089	-1.222058	-0.815143	0.2695

517 rows × 11 columns



```
In [17]: X = X_scaled
y = data[['size_category']]
```

```
In [19]: X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.20,random_state=12,stratify=y)
```



```
In [20]: X_train.shape,y_train.shape
```

Out[20]: ((413, 11), (413, 1))

```
In [21]: X_test.shape,y_test.shape
```

Out[21]: ((104, 11), (104, 1))

```
In [22]: model_linear = SVC(kernel = 'linear',C=30)
model_linear.fit(X_train,y_train)
```

Out[22]: SVC(C=30, kernel='linear')

```
In [23]: y_pred_train_lr = model_linear.predict(X_train)
y_pred_test_lr = model_linear.predict(X_test)
```

```
In [24]: t('Accuracy Score :',accuracy_score(y_train,y_pred_train_lr))
t('\n Confusion Matrix : \n ',confusion_matrix(y_train,y_pred_train_lr))
t('\n Classification Report :\n ',classification_report(y_train,y_pred_train_lr))
```

Accuracy Score : 0.9927360774818402

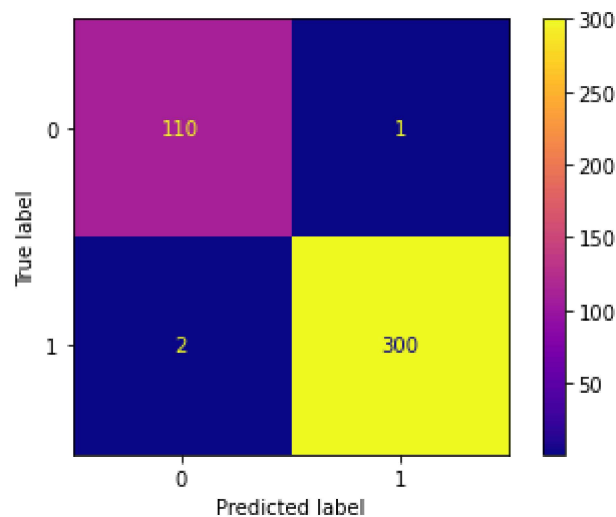
Confusion Matrix :

```
[[110  1]
 [ 2 300]]
```

Classification Report :

	precision	recall	f1-score	support
0	0.98	0.99	0.99	111
1	1.00	0.99	1.00	302
accuracy			0.99	413
macro avg	0.99	0.99	0.99	413
weighted avg	0.99	0.99	0.99	413

```
In [25]: plot_confusion_matrix(model_linear, X_train, y_train, cmap='plasma')
plt.show()
```



```
In [26]: print('Accuracy Score :',accuracy_score(y_test,y_pred_test_lr))
print('\n Confusion Matrix : \n ',confusion_matrix(y_test,y_pred_test_lr))
print('\n Classification Report :\n ',classification_report(y_test,y_pred_test_lr))
```

Accuracy Score : 0.9615384615384616

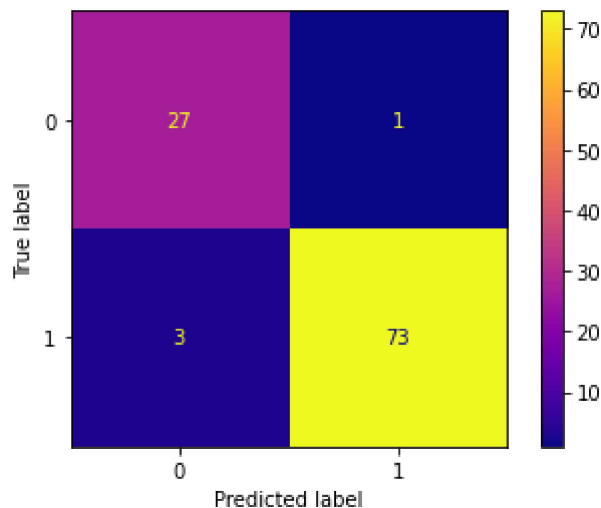
Confusion Matrix :

```
[[27  1]
 [ 3 73]]
```

Classification Report :

	precision	recall	f1-score	support
0	0.90	0.96	0.93	28
1	0.99	0.96	0.97	76
accuracy			0.96	104
macro avg	0.94	0.96	0.95	104
weighted avg	0.96	0.96	0.96	104

```
In [27]: plot_confusion_matrix(model_linear, X_test, y_test, cmap='plasma')
plt.show()
```



```
In [28]: model_rbf = SVC(kernel='rbf',C=30,gamma=0.1)
model_rbf.fit(X_train,y_train)
```

Out[28]: SVC(C=30, gamma=0.1)

```
In [31]: y_pred_train_rbf = model_rbf.predict(X_train)
y_pred_test_rbf = model_rbf.predict(X_test)
```

```
In [33]: print('Accuracy score :',accuracy_score(y_train,y_pred_train_rbf))
print('Confision matrix :\n',confusion_matrix(y_train,y_pred_train_rbf))
print('Classification report :\n',classification_report(y_train,y_pred_train_rbf))
```

Accuracy score : 0.9757869249394673

Confision matrix :

```
[[102  9]
 [ 1 301]]
```

Classification report :

	precision	recall	f1-score	support
0	0.99	0.92	0.95	111
1	0.97	1.00	0.98	302
accuracy			0.98	413
macro avg	0.98	0.96	0.97	413
weighted avg	0.98	0.98	0.98	413

```
In [35]: print('Accuracy score :',accuracy_score(y_test,y_pred_test_rbf))
print('Confision matrix :\n',confusion_matrix(y_test,y_pred_test_rbf))
print('Classification report :\n',classification_report(y_test,y_pred_test_rbf))
```

Accuracy score : 0.9519230769230769

Confision matrix :

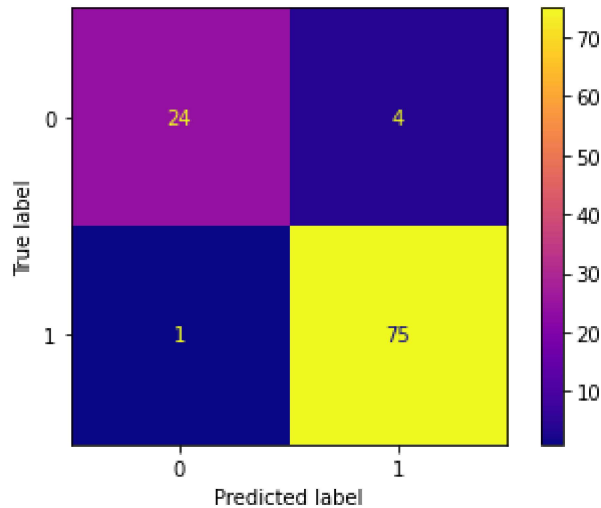
```
[[24  4]
 [ 1 75]]
```

Classification report :

	precision	recall	f1-score	support
0	0.96	0.86	0.91	28
1	0.95	0.99	0.97	76
accuracy			0.95	104
macro avg	0.95	0.92	0.94	104
weighted avg	0.95	0.95	0.95	104



```
In [36]: plot_confusion_matrix(model_rbf, X_test, y_test, cmap='plasma')
plt.show()
```



```
In [44]: model_poly = SVC(kernel='poly')
model_poly.fit(X_train,y_train)
```

```
Out[44]: SVC(kernel='poly')
```

```
In [45]: y_pred_poly_train = model_poly.predict(X_train)
y_pred_poly_test = model_poly.predict(X_test)
```

```
In [46]: print('Accuracy Score :',accuracy_score(y_train,y_pred_poly_train))
print('\n Confusion Matrix : \n ',confusion_matrix(y_train,y_pred_poly_train))
print('\n Classification Report :\n ',classification_report(y_train,y_pred_poly_train))
```

Accuracy Score : 0.8038740920096852

Confusion Matrix :

```
[[ 31  80]
 [  1 301]]
```

Classification Report :

	precision	recall	f1-score	support
0	0.97	0.28	0.43	111
1	0.79	1.00	0.88	302
accuracy			0.80	413
macro avg	0.88	0.64	0.66	413
weighted avg	0.84	0.80	0.76	413

```
In [47]: t('Accuracy Score :',accuracy_score(y_test,y_pred_poly_test))
t('\n Confusion Matrix : \n ',confusion_matrix(y_test,y_pred_poly_test))
t('\n Classification Report :\n ',classification_report(y_test,y_pred_poly_test))
```

Accuracy Score : 0.75

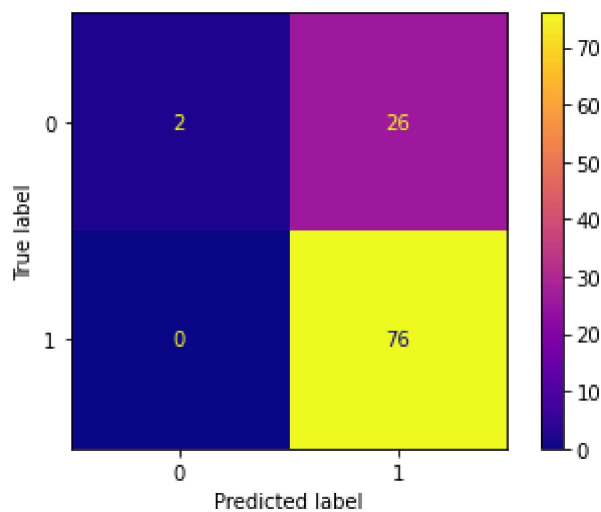
Confusion Matrix :

```
[[ 2 26]
 [ 0 76]]
```

Classification Report :

	precision	recall	f1-score	support
0	1.00	0.07	0.13	28
1	0.75	1.00	0.85	76
accuracy			0.75	104
macro avg	0.87	0.54	0.49	104
weighted avg	0.81	0.75	0.66	104

```
In [48]: plot_confusion_matrix(model_poly, X_test, y_test, cmap='plasma')
plt.show()
```



In [ ]: