```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib as plt
from sklearn.preprocessing import StandardScaler
from sklearn import svm
from sklearn.svm import SVC
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import classification_report
from sklearn.metrics import accuracy_score, confusion_matrix
from sklearn.model_selection import train_test_split, cross_val_score
```

df=pd.read_csv('forestfires.csv')
df

	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	•••	monthfeb	monthjar
0	mar	fri	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0		0	С
1	oct	tue	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0		0	С
2	oct	sat	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0		0	C
3	mar	fri	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2		0	C
4	mar	sun	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0		0	C
512	aug	sun	81.6	56.7	665.6	1.9	27.8	32	2.7	0.0		0	C
513	aug	sun	81.6	56.7	665.6	1.9	21.9	71	5.8	0.0		0	C
514	aug	sun	81.6	56.7	665.6	1.9	21.2	70	6.7	0.0		0	C
515	aug	sat	94.4	146.0	614.7	11.3	25.6	42	4.0	0.0		0	C
516	nov	tue	79.5	3.0	106.7	1.1	11.8	31	4.5	0.0		0	C
517 rows × 31 columns													

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 517 entries, 0 to 516
Data columns (total 31 columns):

#	Column	Non-Null Count	Dtype
0	month	517 non-null	object
1	day	517 non-null	object
2	FFMC	517 non-null	float64
3	DMC	517 non-null	float64
4	DC	517 non-null	float64
5	ISI	517 non-null	float64
6	temp	517 non-null	float64
7	RH	517 non-null	int64

wind	517	non-null	float64			
rain	517	non-null	float64			
area	517	non-null	float64			
dayfri	517	non-null	int64			
daymon	517	non-null	int64			
daysat	517	non-null	int64			
daysun	517	non-null	int64			
daythu	517	non-null	int64			
daytue	517	non-null	int64			
daywed	517	non-null	int64			
monthapr	517	non-null	int64			
monthaug	517	non-null	int64			
monthdec	517	non-null	int64			
monthfeb	517	non-null	int64			
monthjan	517	non-null	int64			
monthjul	517	non-null	int64			
monthjun	517	non-null	int64			
monthmar	517	non-null	int64			
monthmay	517	non-null	int64			
monthnov	517	non-null	int64			
monthoct	517	non-null	int64			
monthsep	517	non-null	int64			
size_category	517	non-null	object			
es: float64(8),	int64(20), object(3)					
	rain area dayfri daymon daysat daysun daythu daytue daywed monthapr monthaug monthfeb monthjan monthjul monthjul monthmar monthmay monthnov monthoct monthsep size_category	rain 517 area 517 dayfri 517 daymon 517 daysat 517 daysun 517 daythu 517 daytue 517 monthapr 517 monthapr 517 monthfeb 517 monthjan 517 monthjul 517 monthmar 517 monthmar 517 monthmay 517 monthnov 517 monthsep 517 size_category 517	rain 517 non-null area 517 non-null dayfri 517 non-null daysat 517 non-null daysat 517 non-null daysun 517 non-null daythu 517 non-null daythu 517 non-null daythu 517 non-null monthapr 517 non-null monthdec 517 non-null monthfeb 517 non-null monthjan 517 non-null monthjul 517 non-null monthmar 517 non-null monthmar 517 non-null monthmar 517 non-null monthmar 517 non-null monthmay 517 non-null monthnov 517 non-null monthoct 517 non-null monthsep 517 non-null size_category 517 non-null			

memory usage: 125.3+ KB

df.shape

(517, 31)

df.describe()

	FFMC	DMC	DC	ISI	temp	RH	win	
count	517.000000	517.000000	517.000000	517.000000	517.000000	517.000000	517.00000	
mean	90.644681	110.872340	547.940039	9.021663	18.889168	44.288201	4.01760	
std	5.520111	64.046482	248.066192	4.559477	5.806625	16.317469	1.79165	
min	18.700000	1.100000	7.900000	0.000000	2.200000	15.000000	0.40000	
25%	90.200000	68.600000	437.700000	6.500000	15.500000	33.000000	2.70000	
50%	91.600000	108.300000	664.200000	8.400000	19.300000	42.000000	4.00000	
75%	92.900000	142.400000	713.900000	10.800000	22.800000	53.000000	4.90000	
max	96.200000	291.300000	860.600000	56.100000	33.300000	100.000000	9.40000	
8 rows × 28 columns								

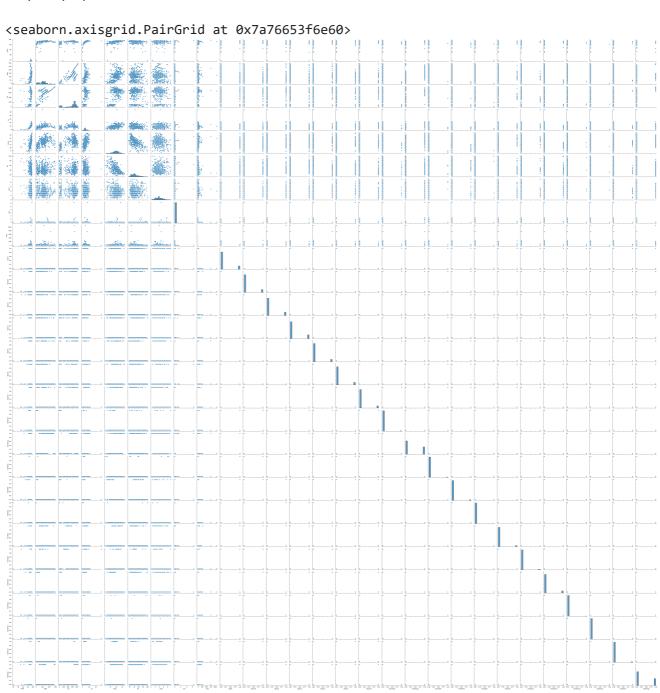
df.duplicated()

- 0 False
- 1 False
- False

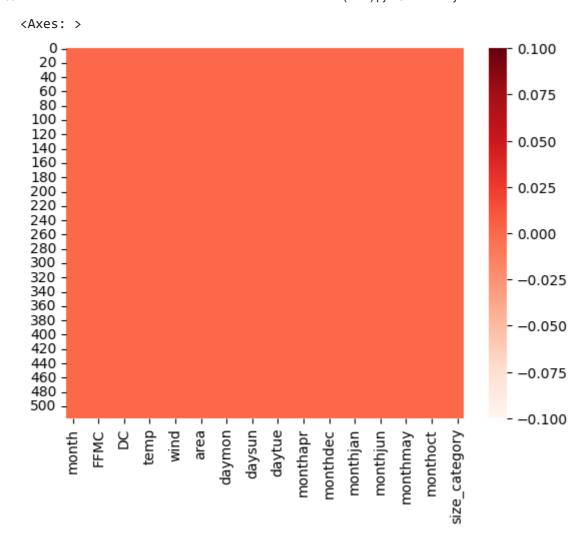
3 False
4 False
...
512 False
513 False
514 False
515 False
516 False

Length: 517, dtype: bool

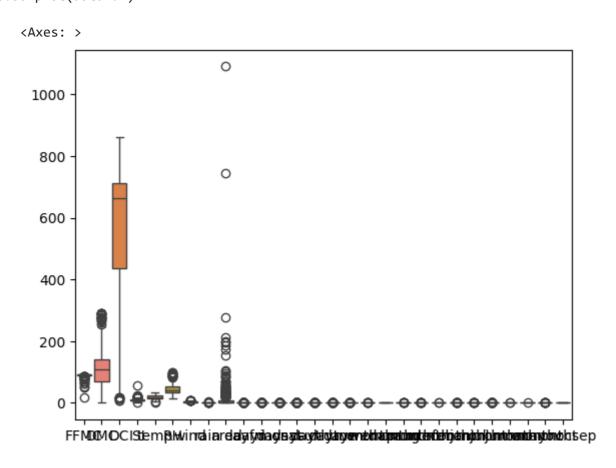
sns.pairplot(df)



sns.heatmap(df.isnull(),cmap='Reds')



sns.boxplot(data=df)



Υ

```
df1=df.iloc[:,2:]
df1
```

	FFMC	DMC	DC	ISI	temp	RH	wind	rain	area	dayfri	• • •	monthfeb	month
0	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	0.00	1		0	
1	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	0.00	0		0	
2	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	0.00	0		0	
3	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	0.00	1		0	
4	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	0.00	0		0	
512	81.6	56.7	665.6	1.9	27.8	32	2.7	0.0	6.44	0		0	
513	81.6	56.7	665.6	1.9	21.9	71	5.8	0.0	54.29	0		0	
514	81.6	56.7	665.6	1.9	21.2	70	6.7	0.0	11.16	0		0	
515	94.4	146.0	614.7	11.3	25.6	42	4.0	0.0	0.00	0		0	
516	79.5	3.0	106.7	1.1	11.8	31	4.5	0.0	0.00	0		0	
	_												

517 rows × 29 columns

 $https://colab.research.google.com/drive/14SiwnXEG_P_FCnxSClOkeVyqVblAZLZL\#scrollTo=rKlgbxaQwiiF\&printMode=true$

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                                       'large', 'large', 'small', 'small', 'large', 'small', 'small', 'small', 'small', 'small', 'small', 'small', 'large', 'small', 'large',
                                       'large', 'small', 'small', 'small', 'small', 'small',
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                                        'small', 'small', 'large', 'small', 'small', 'small'
                                       'small', 'large', 'small', 'small', 'large', 'small', 'large',
                                       'small', 'small', 'small', 'large', 'large', 'small', 'large',
                                       'small', 'small', 'large', 'large', 'small', 'large', 'large', 'large', 'large', 'small', 'large', 'small', 'large', 'small', 'sm
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                                        'large', 'small', 'large', 'small', 'small', 'small'
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                                       'small', 'small', 'small', 'large', 'small', 'small',
                                        'small', 'small', 'small', 'large', 'small', 'small', 'small'
                                       'small', 'small', 'large', 'large', 'small', 'small',
                                       'small', 'small', 'large', 'small', 'small', 'small',
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                                        'small', 'small', 'small', 'large', 'small', 'small', 'large'
                                       'large', 'small', 'large', 'small', 'large', 'large', 'small',
                                       'small', 'large', 'small', 'large', 'small', 'small', 'small',
                                        'small', 'large', 'small', 'small', 'large', 'large',
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                                       'small', 'small', 'small', 'large', 'large', 'small', 'small',
                                       'large', 'large', 'small', 'small', 'small', 'small',
                                       'large', 'small', 'small', 'small', 'small', 'small',
                                       'small', 'large', 'large', 'small', 'small'], dtype=object)
X_train, X_test, y_train, y_test = train_test_split(X,Y, test_size = 0.3)
X_train.shape, y_train.shape, X_test.shape, y_test.shape
                ((361, 28), (361,), (156, 28), (156,))
```

clf = SVC()
param_grid = [{'kernel':['rbf'], 'gamma':[50,5,10,0.5], 'C':[15,14,13,12,11,10,0.1,0.
gsv = GridSearchCV(clf,param_grid,cv=10)
gsv.fit(X_train,y_train)

```
gsv.best_params_ , gsv.best_score_
     ({'C': 15, 'gamma': 0.5, 'kernel': 'rbf'}, 0.7451951951951952)
clf = SVC(C = 15, gamma = 50)
clf.fit(X_train , y_train)
y_pred = clf.predict(X_test)
acc = accuracy_score(y_test, y_pred) * 100
print("Accuracy =", acc)
confusion_matrix(y_test, y_pred)
    Accuracy = 71.7948717948718
     array([[ 1, 44],
            [ 0, 111]])
clf = SVC(kernel= "linear")
clf.fit(X_train , y_train)
y_pred = clf.predict(X_test)
acc = accuracy_score(y_test, y_pred) * 100
print("Accuracy =", acc)
confusion_matrix(y_test, y_pred)
     Accuracy = 98.71794871794873
     array([[ 44, 1],
            [ 1, 110]])
clf = SVC(kernel= "poly")
clf.fit(X train , y train)
y_pred = clf.predict(X_test)
acc = accuracy_score(y_test, y_pred) * 100
print("Accuracy =", acc)
confusion_matrix(y_test, y_pred)
     Accuracy = 76.92307692307693
     array([[ 9, 36],
            [ 0, 111]])
```