```
In [1]: import os
   import numpy as np
   import pandas as pd
   %matplotlib inline
   import matplotlib.pyplot as plt
   import seaborn as sns
```

In [3]: | df=pd.read_csv('C:/Users/User/Downloads/heart.csv')

In [4]: df.head()

Out[4]:

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	са	thal	target
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1	1
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	1
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	1
3	56	1	1	120	236	0	1	178	0	8.0	2	0	2	1
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	1

In [6]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 303 entries, 0 to 302
Data columns (total 14 columns):

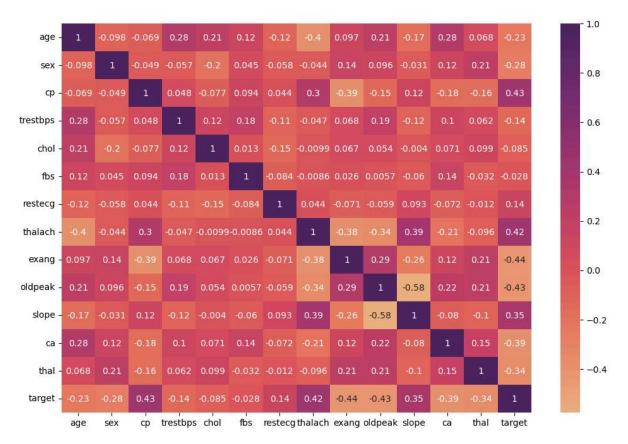
	(, •
#	Column	Non-	-Null Count	Dtype
0	age	303	non-null	int64
1	sex	303	non-null	int64
2	ср	303	non-null	int64
3	trestbps	303	non-null	int64
4	chol	303	non-null	int64
5	fbs	303	non-null	int64
6	restecg	303	non-null	int64
7	thalach	303	non-null	int64
8	exang	303	non-null	int64
9	oldpeak	303	non-null	float64
10	slope	303	non-null	int64
11	ca	303	non-null	int64
12	thal	303	non-null	int64
13	target	303	non-null	int64
44	Cl+c	4/4\	·-+ C4 (43)	

dtypes: float64(1), int64(13)

memory usage: 33.3 KB

```
In [5]: plt.figure(figsize=(12,8))
sns.heatmap(df.corr(),square=False,cmap='flare',annot=True)
```

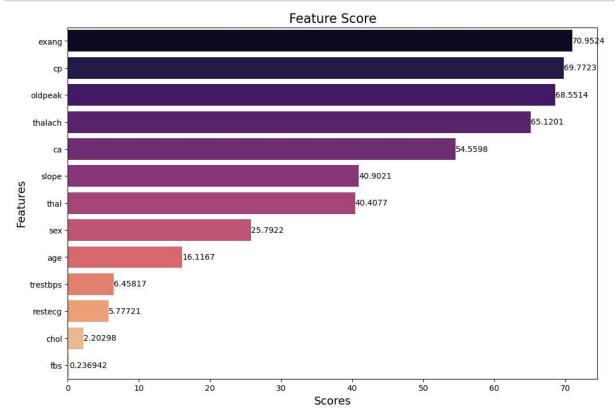
Out[5]: <Axes: >



```
In [7]: #creating features (x) and targets(y) variables
    x=df.iloc[:,:-1]
    y=df.iloc[:,-1]
```

- In [15]: #Finding the best features for the model
 from sklearn.feature_selection import SelectKBest, f_classif
 selector=SelectKBest(f_classif,k=13)
 x_selected=selector.fit_transform(x,y)
- In [18]: #list of all column ans score inx variable
 selected_features=x.columns[selector.get_support()]
 feature_scores=selector.scores_[selector.get_support()]
- In [19]: #Creating dataframe for feature and scores
 feature_score_df=pd.DataFrame({'Features': selected_features, 'Scores': feature
- In [21]: #sort the dataframe in descending order
 feature_score_df=feature_score_df.sort_values(by='Scores', ascending=False)

```
In [25]: #plotting a barplot for better understanding
    plt.figure(figsize=(12,8))
    ax=sns.barplot(x=feature_score_df['Scores'],y=feature_score_df['Features'], pal
    plt.title('Feature Score', fontsize=16)
    plt.xlabel('Scores',fontsize=14)
    plt.ylabel('Features',fontsize=14)
    for lab in ax.containers:
        ax.bar_label(lab)
```



```
In [26]: #dropping the low scored values
x=x.drop(['fbs','chol','restecg','trestbps'],axis=1)
```

In [31]: from sklearn.preprocessing import StandardScaler
 scaler=StandardScaler()
 x=scaler.fit_transform(x)

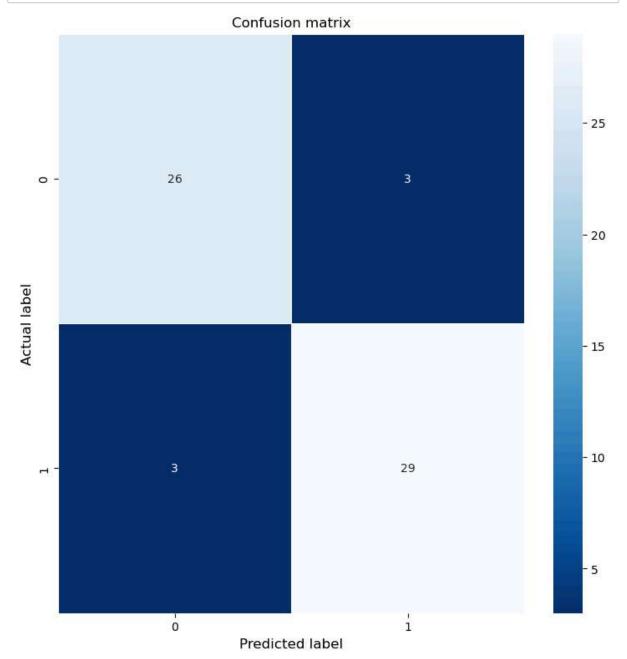
```
In [29]: |x.head
Out[29]: <bound method NDFrame.head of
                                                                               thalach
                                                    age
                                                              sex
                                                                          ср
         exang
                  oldpeak
                              slope \
         0
               0.952197    0.681005    1.973123    0.015443    -0.696631    1.087338    -2.274579
         1
              -1.915313 \quad 0.681005 \quad 1.002577 \quad 1.633471 \quad -0.696631 \quad 2.122573 \quad -2.274579
         2
              -1.474158 -1.468418 0.032031 0.977514 -0.696631 0.310912 0.976352
         3
               0.180175  0.681005  0.032031  1.239897  -0.696631  -0.206705
                                                                              0.976352
         4
               0.290464 -1.468418 -0.938515 0.583939 1.435481 -0.379244
                                                                              0.976352
                               . . .
                                         . . .
                                                    . . .
                                                              . . .
         298 0.290464 -1.468418 -0.938515 -1.165281 1.435481 -0.724323 -0.649113
         299 -1.033002 0.681005 1.973123 -0.771706 -0.696631 0.138373 -0.649113
         300 1.503641 0.681005 -0.938515 -0.378132 -0.696631 2.036303 -0.649113
          301 0.290464 0.681005 -0.938515 -1.515125 1.435481 0.138373 -0.649113
          302 0.290464 -1.468418 0.032031 1.064975 -0.696631 -0.896862 -0.649113
                             thal
                     ca
              -0.714429 -2.148873
         0
         1
              -0.714429 -0.512922
         2
              -0.714429 -0.512922
         3
              -0.714429 -0.512922
              -0.714429 -0.512922
          . .
                    . . .
         298 -0.714429 1.123029
         299 -0.714429 1.123029
          300 1.244593 1.123029
         301 0.265082 1.123029
          302 0.265082 -0.512922
          [303 \text{ rows x 9 columns}]
In [32]:
         from sklearn.model selection import train test split
         x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=4
         from sklearn.neighbors import KNeighborsClassifier
In [37]:
         knn=KNeighborsClassifier(n_neighbors=11)
         knn.fit(x_train,y_train)
Out[37]: KNeighborsClassifier(n_neighbors=11)
         In a Jupyter environment, please rerun this cell to show the HTML representation or trust
         the notebook.
         On GitHub, the HTML representation is unable to render, please try loading this page with
         nbviewer.org.
In [38]: y_predict=knn.predict(x_test)
In [39]: from sklearn.metrics import accuracy score
         print(accuracy_score(y_test,y_predict))
         0.9016393442622951
```

In [40]: from sklearn.metrics import classification_report
 print(classification_report(y_test,y_predict))

	precision	recall	f1-score	support
0	0.90	0.90	0.90	29
1	0.91	0.91	0.91	32
accuracy			0.90	61
macro avg	0.90	0.90	0.90	61
weighted avg	0.90	0.90	0.90	61

> [[26 3] [3 29]]

```
In [47]: plt.figure(figsize=(9,9))
    plt.title('Confusion matrix')
    sns.heatmap(cm,annot=True,fmt='g',linewidth=0.5,cmap='Blues_r',)
    plt.xlabel('Predicted label',fontsize=12)
    plt.ylabel('Actual label',fontsize=12)
    plt.show()
```



```
In [ ]:
```