

```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
```

```
df=pd.read_csv('downloads/heart_disease_uci.csv')
```

```
df
```

| | id | age | sex | dataset | cp | trestbps | chol | |
|-------|-----|-----|-----|---------|---------------|-----------------|-------|-------|
| fbs \ | 0 | 1 | 63 | Male | Cleveland | typical angina | 145.0 | 233.0 |
| True | 1 | 2 | 67 | Male | Cleveland | asymptomatic | 160.0 | 286.0 |
| False | 2 | 3 | 67 | Male | Cleveland | asymptomatic | 120.0 | 229.0 |
| False | 3 | 4 | 37 | Male | Cleveland | non-anginal | 130.0 | 250.0 |
| False | 4 | 5 | 41 | Female | Cleveland | atypical angina | 130.0 | 204.0 |
| False | .. | ... | ... | ... | ... | ... | ... | |
| ... | 915 | 916 | 54 | Female | VA Long Beach | asymptomatic | 127.0 | 333.0 |
| True | 916 | 917 | 62 | Male | VA Long Beach | typical angina | NaN | 139.0 |
| False | 917 | 918 | 55 | Male | VA Long Beach | asymptomatic | 122.0 | 223.0 |
| True | 918 | 919 | 58 | Male | VA Long Beach | asymptomatic | NaN | 385.0 |
| True | 919 | 920 | 62 | Male | VA Long Beach | atypical angina | 120.0 | 254.0 |
| False | | | | | | | | |

| | restecg | thalch | exang | oldpeak | slope | ca \ |
|-----|------------------|--------|-------|---------|-------------|------|
| 0 | lv hypertrophy | 150.0 | False | 2.3 | downsloping | 0.0 |
| 1 | lv hypertrophy | 108.0 | True | 1.5 | flat | 3.0 |
| 2 | lv hypertrophy | 129.0 | True | 2.6 | flat | 2.0 |
| 3 | normal | 187.0 | False | 3.5 | downsloping | 0.0 |
| 4 | lv hypertrophy | 172.0 | False | 1.4 | upsloping | 0.0 |
| .. | ... | ... | ... | ... | ... | ... |
| 915 | st-t abnormality | 154.0 | False | 0.0 | NaN | NaN |
| 916 | st-t abnormality | NaN | NaN | NaN | NaN | NaN |
| 917 | st-t abnormality | 100.0 | False | 0.0 | NaN | NaN |
| 918 | lv hypertrophy | NaN | NaN | NaN | NaN | NaN |
| 919 | lv hypertrophy | 93.0 | True | 0.0 | NaN | NaN |

| | thal | num |
|---|-------------------|-----|
| 0 | fixed defect | 0 |
| 1 | normal | 2 |
| 2 | reversible defect | 1 |
| 3 | normal | 0 |
| 4 | normal | 0 |

```

..          ...  ...
915          NaN  1
916          NaN  0
917    fixed defect  2
918          NaN  0
919          NaN  1

```

[920 rows x 16 columns]

```
df.head()
```

```

   id  age  sex  dataset  cp  trestbps  chol  fbs
\
0   1   63  Male  Cleveland  typical angina  145.0  233.0  True
1   2   67  Male  Cleveland  asymptomatic  160.0  286.0  False
2   3   67  Male  Cleveland  asymptomatic  120.0  229.0  False
3   4   37  Male  Cleveland  non-anginal  130.0  250.0  False
4   5   41  Female  Cleveland  atypical angina  130.0  204.0  False

```

```

   restecg  thalch  exang  oldpeak  slope  ca  \
0  lv hypertrophy  150.0  False  2.3  downsloping  0.0
1  lv hypertrophy  108.0  True  1.5  flat  3.0
2  lv hypertrophy  129.0  True  2.6  flat  2.0
3  normal  187.0  False  3.5  downsloping  0.0
4  lv hypertrophy  172.0  False  1.4  upsloping  0.0

```

```

   thal  num
0  fixed defect  0
1  normal  2
2  reversable defect  1
3  normal  0
4  normal  0

```

```
df.shape
```

```
(920, 16)
```

```
df.drop(['id', 'dataset'], axis=1, inplace=True)
```

```
df.isnull().sum()
```

```

age          0
sex          0
cp           0
trestbps     59
chol         30

```

```
fbs          90
restecg      2
thalch       55
exang        55
oldpeak      62
slope        309
ca           611
thal         486
num          0
dtype: int64
```

```
df=df.dropna()
```

```
df.isnull().sum()
```

```
age          0
sex          0
cp           0
trestbps     0
chol         0
fbs          0
restecg      0
thalch       0
exang        0
oldpeak      0
slope        0
ca           0
thal         0
num          0
dtype: int64
```

```
x=df.drop("num",axis=1)
```

```
y=df.num
```

```
x.head()
```

```
   age  sex          cp  trestbps  chol  fbs  lv
0  63  Male  typical angina    145.0  233.0  True  lv
hypertrophy
1  67  Male  asymptomatic    160.0  286.0  False  lv
hypertrophy
2  67  Male  asymptomatic    120.0  229.0  False  lv
hypertrophy
3  37  Male  non-anginal    130.0  250.0  False
normal
4  41  Female  atypical angina    130.0  204.0  False  lv
hypertrophy

   thalch  exang  oldpeak  slope  ca  thal
0  150.0  False    2.3  downsloping  0.0  fixed defect
```

```
1  108.0  True      1.5      flat  3.0      normal
2  129.0  True      2.6      flat  2.0  reversable  defect
3  187.0  False     3.5  downsloping  0.0      normal
4  172.0  False     1.4  upsloping  0.0      normal
```

```
from sklearn.preprocessing import LabelEncoder
```

```
le=LabelEncoder()
x.sex=le.fit_transform(x.sex)
x.cp=le.fit_transform(x.cp)
x.fbs=le.fit_transform(x.fbs)
x.restecg=le.fit_transform(x.restecg)
x.exang=le.fit_transform(x.exang)
x.slope=le.fit_transform(x.slope)
x.thal=le.fit_transform(x.thal)
```

```
import seaborn as sns
import matplotlib.pyplot as plt
```

```
plt.figure(figsize=(10,8))
corr=x.corr()
plot=sns.heatmap(corr.round(2),annot=True)
plot.set_title("Correlation map")
Text(0.5, 1.0, 'Correlation map')
```



```

from sklearn.naive_bayes import MultinomialNB
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis

```

```

MNB=MultinomialNB()
KNN=KNeighborsClassifier(10)
SVM=SVC(C=3,kernel='poly')
LDS=LinearDiscriminantAnalysis()
DT=DecisionTreeClassifier()

```

```

from sklearn.model_selection import train_test_split
xTrain,xTest,yTrain,yTest=train_test_split(x,y,test_size=0.2)

```

```

import time
start=time.time()
KNN.fit(xTrain,yTrain)
KNNScore=KNN.score(xTest,yTest)
end=time.time()
KNNTime=end-start
print(f"KNN score:{KNNScore*100}%\nKNN time:{KNNTime} seconds")

```

```
KNN score:50.0%
KNN time:0.018689870834350586 seconds

start=time.time()
MNB.fit(xTrain,yTrain)
MNBScore=MNB.score(xTest,yTest)
end=time.time()
MNBTime=end-start
print(f"MNB score:{MNBScore*100}%\nMNB time:{MNBTime} seconds")
```

```
MNB score:58.33333333333336%
MNB time:0.017225027084350586 seconds
```

```
start=time.time()
SVM.fit(xTrain,yTrain)
SVMScore=SVM.score(xTest,yTest)
end=time.time()
SVMTime=end-start
print(f"SVM score:{SVMScore*100}%\nSVM time:{SVMTime} seconds")
```

```
SVM score:53.33333333333336%
SVM time:0.030105113983154297 seconds
```

```
start=time.time()
LDS.fit(xTrain,yTrain)
LDSScore=LDS.score(xTest,yTest)
end=time.time()
LDSTime=end-start
print(f"LDS score:{LDSScore*100}%\nLDS time:{LDSTime} seconds")
```

```
LDS score:60.0%
LDS time:0.02342510223388672 seconds
```

```
start=time.time()
DT.fit(xTrain,yTrain)
DTScore=DT.score(xTest,yTest)
end=time.time()
DTTime=end-start
print(f"DT score:{DTScore*100}%\nDT time:{DTTime} seconds")
```

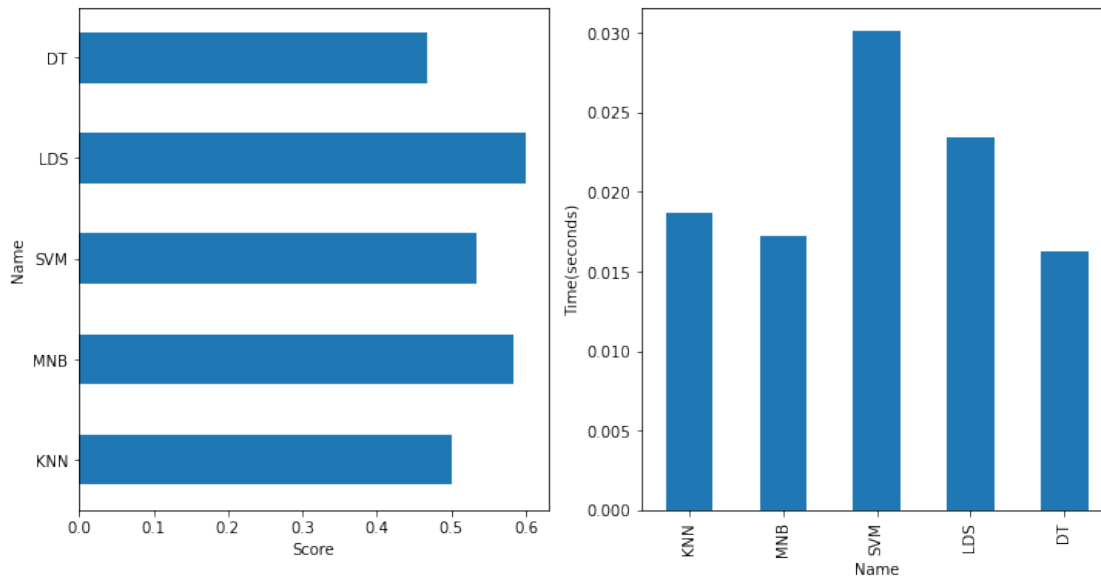
```
DT score:46.666666666666664%
DT time:0.016269207000732422 seconds
```

```
li=[[ "KNN",KNNScore,KNNTime], ['MNB',MNBScore,MNBTime],
 ['SVM',SVMScore,SVMTime], ["LDS",LDSScore,LDSTime],
 ['DT',DTScore,DTTime]]
newdf=pd.DataFrame(li,columns=["Name","Score","Time"])
newdf
```

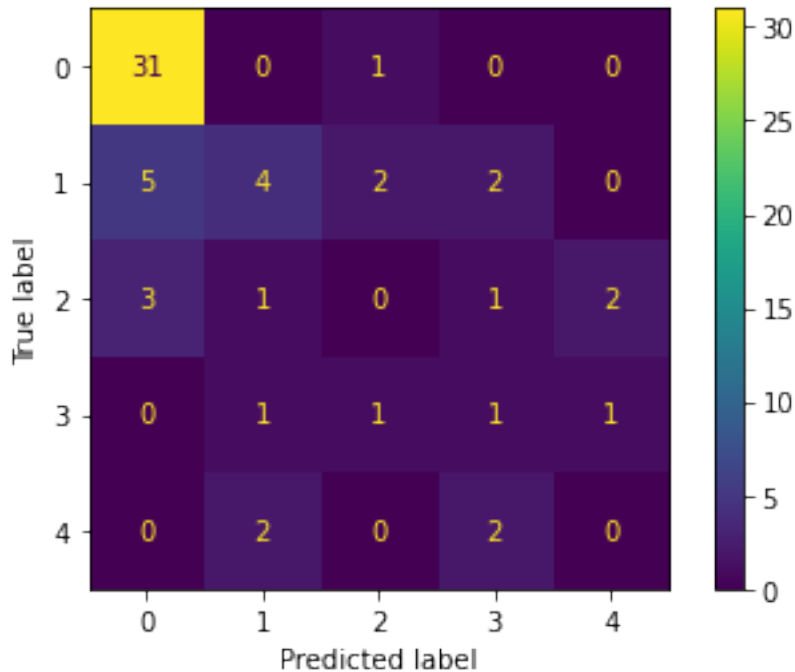
| | Name | Score | Time |
|---|------|----------|----------|
| 0 | KNN | 0.500000 | 0.018690 |
| 1 | MNB | 0.583333 | 0.017225 |

```
2 SVM 0.533333 0.030105
3 LDS 0.600000 0.023425
4 DT 0.466667 0.016269
```

```
fig, ax =plt.subplots(1,2)
newdf.plot.barh(x='Name',y='Score',ax=ax[0],legend=False)
ax[0].set_xlabel("Score")
newdf.plot.bar(x="Name",y="Time",ax=ax[1],legend=False)
ax[1].set_ylabel("Time(seconds)")
fig.set_size_inches(12, 6)
```



```
from sklearn.metrics import ConfusionMatrixDisplay
ConfusionMatrixDisplay.from_predictions(yTest,LDS.predict(xTest))
<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at
0x7fed692ef8e0>
```



```
from sklearn.metrics import f1_score
```

```
KNNf1=f1_score(yTest,KNN.predict(xTest),average='micro')
SVMf1=f1_score(yTest,SVM.predict(xTest),average='micro')
MNBf1=f1_score(yTest,MNB.predict(xTest),average='micro')
LDSf1=f1_score(yTest,LDS.predict(xTest),average='micro')
DTf1=f1_score(yTest,DT.predict(xTest),average='micro')
```

```
li=[["KNN",KNNf1],['MNB',MNBf1],['SVM',SVMf1],["LDS",LDSf1],
['DT',DTf1]]
```

```
newdf=pd.DataFrame(li,columns=["Name","F1 Score"])
newdf
```

| | Name | F1 Score |
|---|------|----------|
| 0 | KNN | 0.500000 |
| 1 | MNB | 0.583333 |
| 2 | SVM | 0.533333 |
| 3 | LDS | 0.600000 |
| 4 | DT | 0.466667 |

```
newdf.plot.bar(x="Name",y="F1 Score",legend=False)
```

```
<AxesSubplot:xlabel='Name'>
```