

```
In [ ]: a= int(input("enter 1st no"))
b= int(input("enter 2nd no"))
c= int(input("enter 3rd no. "))
if a>b and a>c :
    print(a,"is greatest")
elif b>a and b>c :
    print(b,"is greatest")
else:
    print(c," is greatest")
```

```
In [5]: for i in range(1,21):
        if i%2==0:
            print(i,"is even")
        else:
            print(i, "is odd")
```

```
1 is odd
2 is even
3 is odd
4 is even
5 is odd
6 is even
7 is odd
8 is even
9 is odd
10 is even
11 is odd
12 is even
13 is odd
14 is even
15 is odd
16 is even
17 is odd
18 is even
19 is odd
20 is even
```

```
In [26]: num = input("Enter any no = ")
sum=0
for i in (num):
    sum += int(i)
print("Sum of digits is ",sum)
```

```
Enter any no = 12345
Sum of digits is  15
```

```
In [30]: num= input("enter no.")
last=len(num)-1
start=0
while(start<=last):
    if num[start]!=num[last]:
        print("no palindrom")
        last=0
        break
    start+=i
    last+-i
if last!=0:
    print(num, "a palindrome no.")
```

enter no.121

```
-----
TypeError                                Traceback (most recent call last)
C:\Users\AMRESH~1\AppData\Local\Temp\ipykernel_5468\223682299.py in <module>
      7         last=0
      8         break
----> 9     start+=i
     10     last+-i
     11 if last!=0:
```

TypeError: unsupported operand type(s) for +=: 'int' and 'str'

```
In [32]: Num = int(input("Enter a value:"))
Temp = num
Rev = 0
while(num>0):
    dig = num % 10
    revrev = rev * 10 + dig
    numnum = num // 10
if(temp == rev):
    print("This value is a palindrome number!")
else:
    print("This value is not a palindrome number!")
```

Enter a value:111

```
-----
TypeError                                Traceback (most recent call last)
C:\Users\AMRESH~1\AppData\Local\Temp\ipykernel_5468\2920106691.py in <module>
      2 Temp = num
      3 Rev = 0
----> 4 while(num>0):
      5     dig = num % 10
      6     revrev = rev * 10 + dig
```

TypeError: '>' not supported between instances of 'str' and 'int'

```
In [33]: import numpy as np
import pandas as pd
```

```
In [34]: df=pd.read_csv('spambase.csv')
```

```
In [35]: df
```

Out[35]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.40	0.41	0.42	0.778	0.4
0	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	0.00	0.94	...	0.000	0.132	0.0	0.372	0.18
1	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	0.64	0.25	...	0.010	0.143	0.0	0.276	0.18
2	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.000	0.137	0.0	0.137	0.00
3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.000	0.135	0.0	0.135	0.00
4	0.00	0.00	0.00	0.0	1.85	0.00	0.00	1.85	0.00	0.00	...	0.000	0.223	0.0	0.000	0.00
...
4595	0.31	0.00	0.62	0.0	0.00	0.31	0.00	0.00	0.00	0.00	...	0.000	0.232	0.0	0.000	0.00
4596	0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.000	0.000	0.0	0.353	0.00
4597	0.30	0.00	0.30	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.102	0.718	0.0	0.000	0.00
4598	0.96	0.00	0.00	0.0	0.32	0.00	0.00	0.00	0.00	0.00	...	0.000	0.057	0.0	0.000	0.00
4599	0.00	0.00	0.65	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.000	0.000	0.0	0.125	0.00

4600 rows × 58 columns



```
In [36]: print("Top 5 data in given ")
df.head(5)
```

Top 5 data in given

Out[36]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.40	0.41	0.42	0.778	0.43
0	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	0.00	0.94	...	0.00	0.132	0.0	0.372	0.180 0
1	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	0.64	0.25	...	0.01	0.143	0.0	0.276	0.184 0
2	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.00	0.137	0.0	0.137	0.000 0
3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.00	0.135	0.0	0.135	0.000 0
4	0.00	0.00	0.00	0.0	1.85	0.00	0.00	1.85	0.00	0.00	...	0.00	0.223	0.0	0.000	0.000 0

5 rows × 58 columns



In [37]: `df.tail(5)`

Out[37]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.40	0.41	0.42	0.778	0.43	0.4
4595	0.31	0.0	0.62	0.0	0.00	0.31	0.0	0.0	0.0	0.0	...	0.000	0.232	0.0	0.000	0.0	0.
4596	0.00	0.0	0.00	0.0	0.00	0.00	0.0	0.0	0.0	0.0	...	0.000	0.000	0.0	0.353	0.0	0.
4597	0.30	0.0	0.30	0.0	0.00	0.00	0.0	0.0	0.0	0.0	...	0.102	0.718	0.0	0.000	0.0	0.
4598	0.96	0.0	0.00	0.0	0.32	0.00	0.0	0.0	0.0	0.0	...	0.000	0.057	0.0	0.000	0.0	0.
4599	0.00	0.0	0.65	0.0	0.00	0.00	0.0	0.0	0.0	0.0	...	0.000	0.000	0.0	0.125	0.0	0.

5 rows × 58 columns



In [38]: `df.shape`

Out[38]: (4600, 58)

In [48]: `[row,col]=df.shape`

In [49]: `Data = df.iloc[0 : row, 0: (col-1)]`
`Label = df.iloc[0: row, (col-1)]`

In [50]: `Data.shape`

Out[50]: (4600, 57)

In [51]: `Label.value_counts()`

Out[51]: 0 2788
 1 1812
 Name: 1, dtype: int64

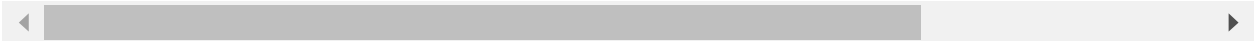
In [52]:

Data

Out[52]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.39	0.40	0.41	0.42	0.778
0	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	0.00	0.94	...	0.0	0.000	0.132	0.0	0.372
1	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	0.64	0.25	...	0.0	0.010	0.143	0.0	0.276
2	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.0	0.000	0.137	0.0	0.137
3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.0	0.000	0.135	0.0	0.135
4	0.00	0.00	0.00	0.0	1.85	0.00	0.00	1.85	0.00	0.00	...	0.0	0.000	0.223	0.0	0.000
...
4595	0.31	0.00	0.62	0.0	0.00	0.31	0.00	0.00	0.00	0.00	...	0.0	0.000	0.232	0.0	0.000
4596	0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.0	0.000	0.000	0.0	0.353
4597	0.30	0.00	0.30	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.0	0.102	0.718	0.0	0.000
4598	0.96	0.00	0.00	0.0	0.32	0.00	0.00	0.00	0.00	0.00	...	0.0	0.000	0.057	0.0	0.000
4599	0.00	0.00	0.65	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.0	0.000	0.000	0.0	0.125

4600 rows × 57 columns



In [53]:

Data.describe()

Out[53]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3
count	4600.000000	4600.000000	4600.000000	4600.000000	4600.000000	4600.000000	4600.000000
mean	0.104576	0.212922	0.280578	0.065439	0.312222	0.095922	0.114233
std	0.305387	1.290700	0.504170	1.395303	0.672586	0.273850	0.391480
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
25%	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
50%	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
75%	0.000000	0.000000	0.420000	0.000000	0.382500	0.000000	0.000000
max	4.540000	14.280000	5.100000	42.810000	10.000000	5.880000	7.270000

8 rows × 57 columns



```
In [56]: from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
import numpy as np

x= Data
y= Label

x_train, x_test, y_train, y_test = train_test_split(x,y, test_size = 0.2)

knn = KNeighborsClassifier(n_neighbors=4)
knn.fit(x_train, y_train)
print(knn.score(x_test, y_test))

0.7956521739130434
```

```
In [57]: y_test.value_counts()
```

```
Out[57]: 0    563
         1    357
         Name: 1, dtype: int64
```

```
In [58]: y_train.value_counts()
```

```
Out[58]: 0    2225
         1    1455
         Name: 1, dtype: int64
```

```
In [59]: y_train.value_counts() + y_test.value_counts()
```

```
Out[59]: 0    2788
         1    1812
         Name: 1, dtype: int64
```

```
In [60]: Label.count()
```

```
Out[60]: 4600
```

```

In [61]: from sklearn import neighbors, datasets, preprocessing
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
from sklearn.metrics import classification_report
from sklearn.metrics import confusion_matrix

iris = datasets.load_iris()
X, y = iris.data[:, :], iris.target
Xtrain, Xtest, y_train, y_test = train_test_split(X, y, stratify = y, random_state=42)

scaler = preprocessing.StandardScaler().fit(Xtrain)
Xtrain = scaler.transform(Xtrain)
Xtest = scaler.transform(Xtest)

knn = neighbors.KNeighborsClassifier(n_neighbors=3)
knn.fit(Xtrain, y_train)
y_pred = knn.predict(Xtest)

print(accuracy_score(y_test, y_pred))
print(classification_report(y_test, y_pred))
print(confusion_matrix(y_test, y_pred))

```

0.9777777777777777

	precision	recall	f1-score	support
0	1.00	1.00	1.00	15
1	0.94	1.00	0.97	15
2	1.00	0.93	0.97	15
accuracy			0.98	45
macro avg	0.98	0.98	0.98	45
weighted avg	0.98	0.98	0.98	45

```

[[15  0  0]
 [ 0 15  0]
 [ 0  1 14]]

```

```

In [68]: def factorial(n):
        if n == 0:
            return 1
        else:
            return n * factorial(n - 1)

```

```

In [72]: n = int(input("enter no. "))

```

enter no.5

```

In [73]: print(factorial(n))

```

120

```
In [75]: def UniqueList(UL):
        s = []
        for a in UL:
            if a not in s:
                s.append(a)
        return s
```

```
In [76]: print(UniqueList([6,3,5,3,5,3,4,6,2,9,6,9,1,7,7,10,8]))

[6]
```

```
In [77]: import numpy as np
A = np.array([5,10,15,20,25,30,35,40,45,50])
print("Entered Array List : \n", A)
ReversedArray = np.flip(A)
print("\nReversed Array List : \n", ReversedArray)
```

```
Entered Array List :
[ 5 10 15 20 25 30 35 40 45 50]
```

```
Reversed Array List :
[50 45 40 35 30 25 20 15 10  5]
```

```
In [78]: import pandas as pd
dig = {'Dairymilk': [40, 50, 90, 100, 250],
'DarkChocolate': [300, 200, 100, 250, 500],
'Perk': [50, 60, 45, 80, 90],
'Fuse': [100, 150, 300, 200, 250],
'KitKat': [110, 200, 150, 60, 70]}
```

```
In [79]: df = pd.DataFrame(data=dig)
```

```
In [80]: df
```

Out[80]:

	Dairymilk	DarkChocolate	Perk	Fuse	KitKat
0	40	300	50	100	110
1	50	200	60	150	200
2	90	100	45	300	150
3	100	250	80	200	60
4	250	500	90	250	70

```
In [81]: import pandas as pd
df = pd.read_csv("USA_Housing.csv")
```


In [82]: df

Out[82]:

	Avg. Area Income	Avg. Area House Age	Avg. Area Number of Rooms	Avg. Area Number of Bedrooms	Area Population	Price	Address
0	79545.45857	5.682861	7.009188	4.09	23086.80050	1.059034e+06	208 Michael Ferry Ap 674\nLaurabury, N 3701.
1	79248.64245	6.002900	6.730821	3.09	40173.07217	1.505891e+06	188 Johnson View Suite 079\nLak Kathleen, CA.
2	61287.06718	5.865890	8.512727	5.13	36882.15940	1.058988e+06	9127 Elizabet Stravenue\nDanielstowr WI 06482.
3	63345.24005	7.188236	5.586729	3.26	34310.24283	1.260617e+06	USS Barnett\nFPO A 4482
4	59982.19723	5.040555	7.839388	4.23	26354.10947	6.309435e+05	USNS Raymond\nFPO AE 0938
...
4995	60567.94414	7.830362	6.137356	3.46	22837.36103	1.060194e+06	USNS Williams\nFPO AP 30153-765
4996	78491.27543	6.999135	6.576763	4.02	25616.11549	1.482618e+06	PSC 9258, Bo 8489\nAPO AA 4299 335
4997	63390.68689	7.250591	4.805081	2.13	33266.14549	1.030730e+06	4215 Tracy Garde Suite 076\nJoshualanc VA 01.
4998	68001.33124	5.534388	7.130144	5.44	42625.62016	1.198657e+06	USS Wallace\nFPO A 7331
4999	65510.58180	5.992305	6.792336	4.07	46501.28380	1.298950e+06	37778 George Ridge Apt. 509\nEast Holl NV 2.

5000 rows × 7 columns



In [83]: dfFirstFifty = df.head(50)

In [84]: dfFirstFifty

	Income	House Age	Number of Rooms	Number of Bedrooms	Population	Price	Address
43	70421.47049	6.907083	7.440505	6.32	43183.92549	1.744932e+06	54111 Davidson Rd 346
44	62614.42062	5.499310	7.440505	6.32	26888.57956	1.153871e+06	FieldnWest Debc SD 45
45	66394.87159	7.069512	7.204640	3.18	39741.07751	1.499989e+06	71956 Jen FallnBrooketown, 67485-C
46	73946.85107	4.863154	7.537182	6.35	35261.12702	1.109588e+06	8034 Pierce Prairie S 727nDevonfurt, NE
47	69144.74571	7.296224	5.928223	3.22	19030.61549	9.801773e+05	Unit 8108 5159nDPO AP 04
48	77278.69703	6.238891	6.919204	2.13	21725.95429	1.323952e+06	08639 Ga PortnAnthonybury 17
49	86754.19663	6.604440	6.252455	4.02	43017.44076	1.662495e+06	91863 Curtis Pointnl Richard, AK 99996-7

In [85]: df_column = df[['Avg. Area House Age', 'Avg. Area Number of Bedrooms', 'Price']]

In [86]: df_column

Out[86]:

	Avg. Area House Age	Avg. Area Number of Bedrooms	Price
0	5.682861	4.09	1.059034e+06
1	6.002900	3.09	1.505891e+06
2	5.865890	5.13	1.058988e+06
3	7.188236	3.26	1.260617e+06
4	5.040555	4.23	6.309435e+05
...
4995	7.830362	3.46	1.060194e+06
4996	6.999135	4.02	1.482618e+06
4997	7.250591	2.13	1.030730e+06
4998	5.534388	5.44	1.198657e+06
4999	5.992305	4.07	1.298950e+06

5000 rows × 3 columns

In [87]: import pandas as pd
df = pd.read_csv("spambase.csv")

In [88]: df

Out[88]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.40	0.41	0.42	0.778	0.4
0	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	0.00	0.94	...	0.000	0.132	0.0	0.372	0.18
1	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	0.64	0.25	...	0.010	0.143	0.0	0.276	0.18
2	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.000	0.137	0.0	0.137	0.00
3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.000	0.135	0.0	0.135	0.00
4	0.00	0.00	0.00	0.0	1.85	0.00	0.00	1.85	0.00	0.00	...	0.000	0.223	0.0	0.000	0.00
...
4595	0.31	0.00	0.62	0.0	0.00	0.31	0.00	0.00	0.00	0.00	...	0.000	0.232	0.0	0.000	0.00
4596	0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.000	0.000	0.0	0.353	0.00
4597	0.30	0.00	0.30	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.102	0.718	0.0	0.000	0.00
4598	0.96	0.00	0.00	0.0	0.32	0.00	0.00	0.00	0.00	0.00	...	0.000	0.057	0.0	0.000	0.00
4599	0.00	0.00	0.65	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.000	0.000	0.0	0.125	0.00

4600 rows × 58 columns



```
In [89]: [row, col] = df.shape
Data = df.iloc[0 : row, 0 : (col - 1)]
Label = df.iloc[0 : row, (col - 1)]
```

In [90]: Data

Out[90]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.39	0.40	0.41	0.42	0.778
0	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	0.00	0.94	...	0.0	0.000	0.132	0.0	0.372
1	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	0.64	0.25	...	0.0	0.010	0.143	0.0	0.276
2	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.0	0.000	0.137	0.0	0.137
3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.0	0.000	0.135	0.0	0.135
4	0.00	0.00	0.00	0.0	1.85	0.00	0.00	1.85	0.00	0.00	...	0.0	0.000	0.223	0.0	0.000
...
4595	0.31	0.00	0.62	0.0	0.00	0.31	0.00	0.00	0.00	0.00	...	0.0	0.000	0.232	0.0	0.000
4596	0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.0	0.000	0.000	0.0	0.353
4597	0.30	0.00	0.30	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.0	0.102	0.718	0.0	0.000
4598	0.96	0.00	0.00	0.0	0.32	0.00	0.00	0.00	0.00	0.00	...	0.0	0.000	0.057	0.0	0.000
4599	0.00	0.00	0.65	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.0	0.000	0.000	0.0	0.125

4600 rows × 57 columns




```
In [97]: from sklearn.metrics import accuracy_score
accuracy_score(y_test,predictions)
```

```
Out[97]: 0.5956521739130435
```

```
In [98]: df = pd.read_csv("heart_failure_clinical_records_dataset.csv")
```

```
In [99]: [row,col]= df.shape
Data = df.iloc[0:row, 0:(col -1)]
Label = df.iloc[0:row,(col -1)]
```

```
In [101]: from sklearn.model_selection import train_test_split
from sklearn.cluster import KMeans
X=Data
y=Label
X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.20,random_state=42)
kmns=KMeans(n_clusters=2,random_state=0)
kmns.fit(X_train,y_train)
```

```
Out[101]: KMeans(n_clusters=2, random_state=0)
```

```
In [102]: predictions = kmns.predict(X_test)
print(predictions)
```

```
[0 1 1 1 0 0 0 1 0 0 0 0 0 0 1 0 1 1 0 0 0 0 0 0 1 0 0 0 0 0 0 1 0 1 1 1
 0 0 0 0 0 1 0 0 1 0 1 0 1 0 1 1 0 0 0 0 0 0]
```

```
In [103]: from sklearn.metrics import accuracy_score
accuracy_score(y_test, predictions)
```

```
Out[103]: 0.5833333333333334
```

```
In [104]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
In [105]: df = pd.read_csv('spambase.csv')
```

In [106]: df

Out[106]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.40	0.41	0.42	0.778	0.4
0	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	0.00	0.94	...	0.000	0.132	0.0	0.372	0.18
1	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	0.64	0.25	...	0.010	0.143	0.0	0.276	0.18
2	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.000	0.137	0.0	0.137	0.00
3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.000	0.135	0.0	0.135	0.00
4	0.00	0.00	0.00	0.0	1.85	0.00	0.00	1.85	0.00	0.00	...	0.000	0.223	0.0	0.000	0.00
...
4595	0.31	0.00	0.62	0.0	0.00	0.31	0.00	0.00	0.00	0.00	...	0.000	0.232	0.0	0.000	0.00
4596	0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.000	0.000	0.0	0.353	0.00
4597	0.30	0.00	0.30	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.102	0.718	0.0	0.000	0.00
4598	0.96	0.00	0.00	0.0	0.32	0.00	0.00	0.00	0.00	0.00	...	0.000	0.057	0.0	0.000	0.00
4599	0.00	0.00	0.65	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.000	0.000	0.0	0.125	0.00

4600 rows × 58 columns

```
In [107]: [row,col]=df.shape
Data = df.iloc[0:row, 0: (col-1)]
Label=df.iloc[0:row, (col-1)]
```

In [108]: # DT

```
In [110]: ort warnings
warnings.filterwarnings("ignore")
from sklearn.tree import DecisionTreeClassifier
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix
from sklearn.metrics import precision_score, recall_score, f1_score, accuracy_score
```

```
In [115]: X_train, X_test, y_train, y_test = train_test_split(Data, Label, test_size =0.1)
```

```
In [116]: dt = DecisionTreeClassifier(random_state = 0, max_depth = 2)
dt.fit(X_train, y_train )
y_pred = dt.predict(X_test)
```

```
In [117]: print('Accuracy: %.4f' %accuracy_score(y_test, y_pred))
```

Accuracy: 0.8370

```
In [118]: # knn
```

```
In [119]: from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
X = Data
y = Label
X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.2)
knn = KNeighborsClassifier(n_neighbors = 4)
knn.fit( X_train, y_train)
print(knn.score(X_test,y_test))
```

0.7978260869565217

```
In [120]: # k-Means
```

```
In [124]: learn.model_selection import train_test_split
learn.cluster import KMeans
a
el
,X_test,y_train,y_test = train_test_split(X,y,test_size = 0.20, random_state=42)
Means(n_clusters = 2, random_state = 0)
t( X_train, y_train)
```

```
Out[124]: KMeans(n_clusters=2, random_state=0)
```

```
In [125]: import numpy as np
import pandas as pd
df=pd.read_csv('spambase.csv')
df
```

Out[125]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.40	0.41	0.42	0.778	0.4
0	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	0.00	0.94	...	0.000	0.132	0.0	0.372	0.18
1	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	0.64	0.25	...	0.010	0.143	0.0	0.276	0.18
2	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.000	0.137	0.0	0.137	0.00
3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.000	0.135	0.0	0.135	0.00
4	0.00	0.00	0.00	0.0	1.85	0.00	0.00	1.85	0.00	0.00	...	0.000	0.223	0.0	0.000	0.00
...
4595	0.31	0.00	0.62	0.0	0.00	0.31	0.00	0.00	0.00	0.00	...	0.000	0.232	0.0	0.000	0.00
4596	0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.000	0.000	0.0	0.353	0.00
4597	0.30	0.00	0.30	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.102	0.718	0.0	0.000	0.00
4598	0.96	0.00	0.00	0.0	0.32	0.00	0.00	0.00	0.00	0.00	...	0.000	0.057	0.0	0.000	0.00
4599	0.00	0.00	0.65	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.000	0.000	0.0	0.125	0.00

4600 rows × 58 columns



```
In [126]: df.head()
```

Out[126]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.40	0.41	0.42	0.778	0.43
0	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	0.00	0.94	...	0.00	0.132	0.0	0.372	0.180
1	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	0.64	0.25	...	0.01	0.143	0.0	0.276	0.184
2	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.00	0.137	0.0	0.137	0.000
3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.00	0.135	0.0	0.135	0.000
4	0.00	0.00	0.00	0.0	1.85	0.00	0.00	1.85	0.00	0.00	...	0.00	0.223	0.0	0.000	0.000

5 rows × 58 columns



In [127]: `df.tail()`

Out[127]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.40	0.41	0.42	0.778	0.43	0.4
4595	0.31	0.0	0.62	0.0	0.00	0.31	0.0	0.0	0.0	0.0	...	0.000	0.232	0.0	0.000	0.0	0.
4596	0.00	0.0	0.00	0.0	0.00	0.00	0.0	0.0	0.0	0.0	...	0.000	0.000	0.0	0.353	0.0	0.
4597	0.30	0.0	0.30	0.0	0.00	0.00	0.0	0.0	0.0	0.0	...	0.102	0.718	0.0	0.000	0.0	0.
4598	0.96	0.0	0.00	0.0	0.32	0.00	0.0	0.0	0.0	0.0	...	0.000	0.057	0.0	0.000	0.0	0.
4599	0.00	0.0	0.65	0.0	0.00	0.00	0.0	0.0	0.0	0.0	...	0.000	0.000	0.0	0.125	0.0	0.

5 rows × 58 columns



In [128]: `df.shape`

Out[128]: (4600, 58)

In [129]: `[row, col] = df.shape`
`Data = df.iloc[0 : row, 0 : (col - 1)]`
`Label = df.iloc[0 : row, (col - 1)]`
`Data`

Out[129]:

	0	0.64	0.64.1	0.1	0.32	0.2	0.3	0.4	0.5	0.6	...	0.39	0.40	0.41	0.42	0.778
0	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	0.00	0.94	...	0.0	0.000	0.132	0.0	0.372
1	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	0.64	0.25	...	0.0	0.010	0.143	0.0	0.276
2	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.0	0.000	0.137	0.0	0.137
3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	0.31	0.63	...	0.0	0.000	0.135	0.0	0.135
4	0.00	0.00	0.00	0.0	1.85	0.00	0.00	1.85	0.00	0.00	...	0.0	0.000	0.223	0.0	0.000
...
4595	0.31	0.00	0.62	0.0	0.00	0.31	0.00	0.00	0.00	0.00	...	0.0	0.000	0.232	0.0	0.000
4596	0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.0	0.000	0.000	0.0	0.353
4597	0.30	0.00	0.30	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.0	0.102	0.718	0.0	0.000
4598	0.96	0.00	0.00	0.0	0.32	0.00	0.00	0.00	0.00	0.00	...	0.0	0.000	0.057	0.0	0.000
4599	0.00	0.00	0.65	0.0	0.00	0.00	0.00	0.00	0.00	0.00	...	0.0	0.000	0.000	0.0	0.125

4600 rows × 57 columns



In [130]: Label

```
Out[130]: 0      1
          1      1
          2      1
          3      1
          4      1
          ..
         4595    0
         4596    0
         4597    0
         4598    0
         4599    0
          Name: 1, Length: 4600, dtype: int64
```

In [131]: Label.value_counts()

```
Out[131]: 0      2788
          1      1812
          Name: 1, dtype: int64
```

```
In [132]: from sklearn.neighbors import KNeighborsClassifier
          from sklearn.model_selection import train_test_split
          import numpy as np
          X = Data
          y = Label
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2)
          knn = KNeighborsClassifier(n_neighbors = 4)
          knn.fit(X_train, y_train)
          print(knn.score(X_test, y_test))
```

0.8152173913043478

```
In [133]: knn = KNeighborsClassifier(n_neighbors = 2)
          knn.fit(X_train, y_train)
          print(knn.score(X_test, y_test))
```

0.825

```
In [134]: knn = KNeighborsClassifier(n_neighbors = 5)
          knn.fit(X_train, y_train)
          print(knn.score(X_test, y_test))
```

0.8152173913043478

```
In [135]: knn = KNeighborsClassifier(n_neighbors = 10)
          knn.fit(X_train, y_train)
          print(knn.score(X_test, y_test))
```

0.7934782608695652

```
In [136]: knn = KNeighborsClassifier(n_neighbors = 20)
knn.fit(X_train, y_train)
print(knn.score(X_test, y_test))
```

0.7771739130434783

```
In [137]: knn = KNeighborsClassifier(n_neighbors = 2)
knn.fit(X_train, y_train)
print(knn.score(X_test, y_test))
```

0.825

```
In [138]: #precision in knn
from sklearn.metrics import precision_score
y_pred=k_means.predict(X_test)
precision_score(y_test,y_pred,average=None,zero_division=1)[0]
```

```
-----
NameError                                Traceback (most recent call last)
C:\Users\AMRESH~1\AppData\Local\Temp\ipykernel_5468\2936095963.py in <module>
      1 #precision in knn
      2 from sklearn.metrics import precision_score
----> 3 y_pred=k_means.predict(X_test)
      4 precision_score(y_test,y_pred,average=None,zero_division=1)[0]

NameError: name 'k_means' is not defined
```

In [139]: `#Recall_score in knn`

```
from sklearn.metrics import recall_score
recall_score(y_test,y_pred,average='macro',zero_division=1)
```

```
-----
ValueError                                Traceback (most recent call last)
C:\Users\AMRESH~1\AppData\Local\Temp\ipykernel_5468\2000363971.py in <module>
      2
      3 from sklearn.metrics import recall_score
----> 4 recall_score(y_test,y_pred,average='macro',zero_division=1)

~\anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, *
kwargs)
      61         extra_args = len(args) - len(all_args)
      62         if extra_args <= 0:
--> 63             return f(*args, **kwargs)
      64
      65         # extra_args > 0

~\anaconda3\lib\site-packages\sklearn\metrics\_classification.py in recall_sc
ore(y_true, y_pred, labels, pos_label, average, sample_weight, zero_division)
    1772     array([0.5, 1. , 1. ])
    1773     """
-> 1774     _, r, _, _ = precision_recall_fscore_support(y_true, y_pred,
    1775                                                    labels=labels,
    1776                                                    pos_label=pos_label,

~\anaconda3\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, *
kwargs)
      61         extra_args = len(args) - len(all_args)
      62         if extra_args <= 0:
--> 63             return f(*args, **kwargs)
      64
      65         # extra_args > 0

~\anaconda3\lib\site-packages\sklearn\metrics\_classification.py in precision
_recall_fscore_support(y_true, y_pred, beta, labels, pos_label, average, warn
_for, sample_weight, zero_division)
    1462     if beta < 0:
    1463         raise ValueError("beta should be >=0 in the F-beta score")
-> 1464     labels = _check_set_wise_labels(y_true, y_pred, average, labels,
    1465                                     pos_label)
    1466

~\anaconda3\lib\site-packages\sklearn\metrics\_classification.py in _check_se
t_wise_labels(y_true, y_pred, average, labels, pos_label)
    1275         str(average_options))
    1276
-> 1277     y_type, y_true, y_pred = _check_targets(y_true, y_pred)
    1278     # Convert to Python primitive type to avoid NumPy type / Python s
    1279     # comparison. See https://github.com/numpy/numpy/issues/6784 (htt
    ps://github.com/numpy/numpy/issues/6784)

~\anaconda3\lib\site-packages\sklearn\metrics\_classification.py in _check_ta
```

```
rgets(y_true, y_pred)
    81     y_pred : array or indicator matrix
    82     """
---> 83     check_consistent_length(y_true, y_pred)
    84     type_true = type_of_target(y_true)
    85     type_pred = type_of_target(y_pred)

~\anaconda3\lib\site-packages\sklearn\utils\validation.py in check_consistent_length(*arrays)
    317     uniques = np.unique(lengths)
    318     if len(uniques) > 1:
--> 319         raise ValueError("Found input variables with inconsistent num
bers of"
    320                             " samples: %r" % [int(l) for l in lengths])
    321

ValueError: Found input variables with inconsistent numbers of samples: [920,
460]
```

In []: