

Vashu Agarwal

E21CSEU0054 EB06 lab7 q1

```
In [1]: import numpy as np
import matplotlib.pyplot as mtp
import pandas as pd
```

```
In [2]: dataset = pd.read_csv("/Users/vashuagarwal/Downloads/BENNETT thing
```

```
In [3]: print(dataset.head())
```

	id	diagnosis	radius_mean	texture_mean	perimeter_mean
area_mean \					
0	842302	M	17.99	10.38	122.80
1001.0					
1	842517	M	20.57	17.77	132.90
1326.0					
2	84300903	M	19.69	21.25	130.00
1203.0					
3	84348301	M	11.42	20.38	77.58
386.1					
4	84358402	M	20.29	14.34	135.10
1297.0					
smoothness_mean					
ts_mean \					
0	0.11840		0.27760	0.3001	
0.14710					
1	0.08474		0.07864	0.0869	
0.07017					
2	0.10960		0.15990	0.1974	
0.12790					
3	0.14250		0.28390	0.2414	
0.10520					
4	0.10030		0.13280	0.1980	
0.10430					
...	texture_worst	perimeter_worst	area_worst	smoothness_worst	
st \					
0	...	17.33	184.60	2019.0	0.16
22					
1	...	23.41	158.80	1956.0	0.12
38					
2	...	25.53	152.50	1709.0	0.14
44					

```

3 ...          26.50          98.87          567.7          0.20
98
4 ...          16.67          152.20          1575.0          0.13
74

```

```

compactness_worst concavity_worst concave points_worst symme
try_worst \
0 0.6656 0.7119 0.2654
0.4601
1 0.1866 0.2416 0.1860
0.2750
2 0.4245 0.4504 0.2430
0.3613
3 0.8663 0.6869 0.2575
0.6638
4 0.2050 0.4000 0.1625
0.2364

```

```

fractal_dimension_worst Unnamed: 32
0 0.11890 NaN
1 0.08902 NaN
2 0.08758 NaN
3 0.17300 NaN
4 0.07678 NaN

```

[5 rows x 33 columns]

```

In [4]: f = set(["diagnosis"])
dataset["diagnosis"] = dataset["diagnosis"].map({'M':0, 'B':1}).astype
print(dataset.head)

```

```

<bound method NDFrame.head of
n texture_mean perimeter_mean \ id diagnosis radius_mea
0 842302 0 17.99 10.38 122.8
0
1 842517 0 20.57 17.77 132.9
0
2 84300903 0 19.69 21.25 130.0
0
3 84348301 0 11.42 20.38 77.5
8
4 84358402 0 20.29 14.34 135.1
0
.. ... .. ... ..
.
564 926424 0 21.56 22.39 142.0
0
565 926682 0 20.13 28.25 131.2
0
566 926954 0 16.60 28.08 108.3
0
567 927241 0 20.60 29.33 140.1
0

```

```

568      92751      1      7.76      24.54      47.9
2

```

```

      area_mean  smoothness_mean  compactness_mean  concavity_mean
\
0      1001.0      0.11840      0.27760      0.30010
1      1326.0      0.08474      0.07864      0.08690
2      1203.0      0.10960      0.15990      0.19740
3       386.1      0.14250      0.28390      0.24140
4      1297.0      0.10030      0.13280      0.19800
..      ...      ...      ...      ...
564     1479.0      0.11100      0.11590      0.24390
565     1261.0      0.09780      0.10340      0.14400
566       858.1      0.08455      0.10230      0.09251
567     1265.0      0.11780      0.27700      0.35140
568       181.0      0.05263      0.04362      0.00000

```

```

      concave points_mean  ...  texture_worst  perimeter_worst  are
a_worst \
0      0.14710  ...      17.33      184.60
2019.0
1      0.07017  ...      23.41      158.80
1956.0
2      0.12790  ...      25.53      152.50
1709.0
3      0.10520  ...      26.50      98.87
567.7
4      0.10430  ...      16.67      152.20
1575.0
..      ...  ...      ...      ...
...
564      0.13890  ...      26.40      166.10
2027.0
565      0.09791  ...      38.25      155.00
1731.0
566      0.05302  ...      34.12      126.70
1124.0
567      0.15200  ...      39.42      184.60
1821.0
568      0.00000  ...      30.37      59.16
268.6

```

```

      smoothness_worst  compactness_worst  concavity_worst  \
0      0.16220      0.66560      0.7119
1      0.12380      0.18660      0.2416
2      0.14440      0.42450      0.4504
3      0.20980      0.86630      0.6869
4      0.13740      0.20500      0.4000
..      ...      ...      ...
564      0.14100      0.21130      0.4107
565      0.11660      0.19220      0.3215
566      0.11390      0.30940      0.3403
567      0.16500      0.86810      0.9387

```

	0.08996	0.06444	0.0000	
	concave	points_worst	symmetry_worst	fractal_dimension_worst
\				
0		0.2654	0.4601	0.11890
1		0.1860	0.2750	0.08902
2		0.2430	0.3613	0.08758
3		0.2575	0.6638	0.17300
4		0.1625	0.2364	0.07678
..	
564		0.2216	0.2060	0.07115
565		0.1628	0.2572	0.06637
566		0.1418	0.2218	0.07820
567		0.2650	0.4087	0.12400
568		0.0000	0.2871	0.07039

```

    Unnamed: 32
0           NaN
1           NaN
2           NaN
3           NaN
4           NaN
..          ...
564         NaN
565         NaN
566         NaN
567         NaN
568         NaN

```

```
[569 rows x 33 columns]>
```

```
In [5]: x = dataset.iloc[:,2:-1].values
print(x)
```

```

[[1.799e+01 1.038e+01 1.228e+02 ... 2.654e-01 4.601e-01 1.189e-01]
 [2.057e+01 1.777e+01 1.329e+02 ... 1.860e-01 2.750e-01 8.902e-02]
 [1.969e+01 2.125e+01 1.300e+02 ... 2.430e-01 3.613e-01 8.758e-02]
 ...
 [1.660e+01 2.808e+01 1.083e+02 ... 1.418e-01 2.218e-01 7.820e-02]
 [2.060e+01 2.933e+01 1.401e+02 ... 2.650e-01 4.087e-01 1.240e-01]
 [7.760e+00 2.454e+01 4.792e+01 ... 0.000e+00 2.871e-01 7.039e-02]
]
```

```
In [6]: y = dataset.iloc[:,1].values
print(y)
```

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

```
In [ ]:
```

```
In [7]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.25)
```

```
In [ ]:
```

```
In [8]: from sklearn.preprocessing import StandardScaler
st_x= StandardScaler()
x_train = st_x.fit_transform(x_train)
x_test = st_x.transform(x_test)
```

In []:

```
In [14]: from sklearn.neighbors import KNeighborsClassifier
classifier = KNeighborsClassifier(n_neighbors=5, metric = 'minkowski')
classifier.fit(x_train, y_train)
```

```
Out[14]: KNeighborsClassifier()
```

```
In [15]: y_pred = classifier.predict(x_test)
```

```
In [16]: print(y_pred)
```

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

```
In [17]: from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
```

```
In [18]: print(cm)
```

```
[[47  6]
 [ 1 89]]
```

```
In [19]: from sklearn.metrics import accuracy_score
print("Accuracy of model {0}%".format(accuracy_score(y_test, y_pred)))

Accuracy of model 95.1048951048951%
```

In []: