

# Deena 20104016

## Basic Analysis using Numpy and Pandas

### Import Libraries

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

### Importing Dataset

```
In [2]: df=pd.read_csv("8_BreastCancerPrediction.csv")  
df
```

Out[2]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_
0	842302	M	17.99	10.38	122.80	1001.0	0.
1	842517	M	20.57	17.77	132.90	1326.0	0.
2	84300903	M	19.69	21.25	130.00	1203.0	0.
3	84348301	M	11.42	20.38	77.58	386.1	0.
4	84358402	M	20.29	14.34	135.10	1297.0	0.
...	...	...	...	...	...	...	...
564	926424	M	21.56	22.39	142.00	1479.0	0.
565	926682	M	20.13	28.25	131.20	1261.0	0.
566	926954	M	16.60	28.08	108.30	858.1	0.
567	927241	M	20.60	29.33	140.10	1265.0	0.
568	92751	B	7.76	24.54	47.92	181.0	0.

569 rows × 33 columns



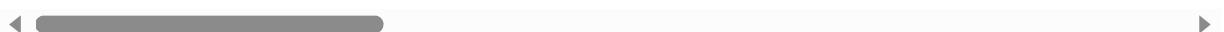
### To display first 10 rows

In [3]: df.head(10)

Out[3]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mean	concave points_mean	symmetry_mean	fractal_dimension_mean	radius_se	texture_se	perimeter_se	area_se	smoothness_se	compactness_se	concavity_se	concave points_se	symmetry_se	fractal_dimension_se	radius_worst	texture_worst	perimeter_worst	area_worst	smoothness_worst	compactness_worst	concavity_worst	concave points_worst	symmetry_worst	fractal_dimension_worst
0	842302	M	17.99	10.38	122.80	1001.0	0.118	0.285	0.037	0.000	0.132	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
1	842517	M	20.57	17.77	132.90	1326.0	0.084	0.244	0.045	0.000	0.125	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
2	84300903	M	19.69	21.25	130.00	1203.0	0.109	0.277	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
3	84348301	M	11.42	20.38	77.58	386.1	0.142	0.205	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
4	84358402	M	20.29	14.34	135.10	1297.0	0.100	0.244	0.045	0.000	0.125	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
5	843786	M	12.45	15.70	82.57	477.1	0.121	0.229	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
6	844359	M	18.25	19.98	119.60	1040.0	0.094	0.244	0.035	0.000	0.125	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
7	84458202	M	13.71	20.83	90.20	577.9	0.118	0.229	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
8	844981	M	13.00	21.82	87.50	519.8	0.121	0.229	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
9	84501001	M	12.46	24.04	83.97	475.9	0.118	0.229	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	

10 rows × 33 columns



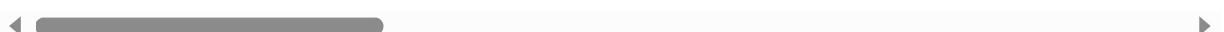
## To display last 5 rows

In [4]: df.tail(5)

Out[4]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	compactness_mean	concavity_mean	concave points_mean	symmetry_mean	fractal_dimension_mean	radius_se	texture_se	perimeter_se	area_se	smoothness_se	compactness_se	concavity_se	concave points_se	symmetry_se	fractal_dimension_se	radius_worst	texture_worst	perimeter_worst	area_worst	smoothness_worst	compactness_worst	concavity_worst	concave points_worst	symmetry_worst	fractal_dimension_worst
564	926424	M	21.56	22.39	142.00	1479.0	0.111	0.229	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
565	926682	M	20.13	28.25	131.20	1261.0	0.091	0.229	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
566	926954	M	16.60	28.08	108.30	858.1	0.084	0.229	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
567	927241	M	20.60	29.33	140.10	1265.0	0.111	0.229	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	
568	92751	B	7.76	24.54	47.92	181.0	0.052	0.229	0.035	0.000	0.120	0.006	1.386	0.424	3.088	7.569	0.053	0.019	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000

5 rows × 33 columns



## Satistical Summary

In [5]: df.describe()

Out[5]:

	id	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean
count	5.690000e+02	569.000000	569.000000	569.000000	569.000000	569.000000
mean	3.037183e+07	14.127292	19.289649	91.969033	654.889104	0.09636
std	1.250206e+08	3.524049	4.301036	24.298981	351.914129	0.01406
min	8.670000e+03	6.981000	9.710000	43.790000	143.500000	0.05263
25%	8.692180e+05	11.700000	16.170000	75.170000	420.300000	0.08637
50%	9.060240e+05	13.370000	18.840000	86.240000	551.100000	0.09587
75%	8.813129e+06	15.780000	21.800000	104.100000	782.700000	0.10530
max	9.113205e+08	28.110000	39.280000	188.500000	2501.000000	0.16340

8 rows × 32 columns



## To find shape and size

In [6]: df.shape

Out[6]: (569, 33)

In [7]: df.size

Out[7]: 18777

## To fill the null values

In [8]: df.isna()

Out[8]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	
0	False	False	False	False	False	False	False	Fals
1	False	False	False	False	False	False	False	Fals
2	False	False	False	False	False	False	False	Fals
3	False	False	False	False	False	False	False	Fals
4	False	False	False	False	False	False	False	Fals
...	...	...	...	...	...	...	...	.
564	False	False	False	False	False	False	False	Fals
565	False	False	False	False	False	False	False	Fals
566	False	False	False	False	False	False	False	Fals
567	False	False	False	False	False	False	False	Fals
568	False	False	False	False	False	False	False	Fals

569 rows × 33 columns



## To fill missing values

In [9]: df.dropna()

Out[9]:

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	con
0	rows	×	33	columns				



## columns

In [10]: df.columns

Out[10]: Index(['id', 'diagnosis', 'radius\_mean', 'texture\_mean', 'perimeter\_mean', 'area\_mean', 'smoothness\_mean', 'compactness\_mean', 'concavity\_mean', 'concave points\_mean', 'symmetry\_mean', 'fractal\_dimension\_mean', 'radius\_se', 'texture\_se', 'perimeter\_se', 'area\_se', 'smoothness\_se', 'compactness\_se', 'concavity\_se', 'concave points\_se', 'symmetry\_se', 'fractal\_dimension\_se', 'radius\_worst', 'texture\_worst', 'perimeter\_worst', 'area\_worst', 'smoothness\_worst', 'compactness\_worst', 'concavity\_worst', 'concave points\_worst', 'symmetry\_worst', 'fractal\_dimension\_worst', 'Unnamed: 32'], dtype='object')

## to print a particular column

In [11]: `data=df[["radius_mean", "texture_mean"]]`  
data

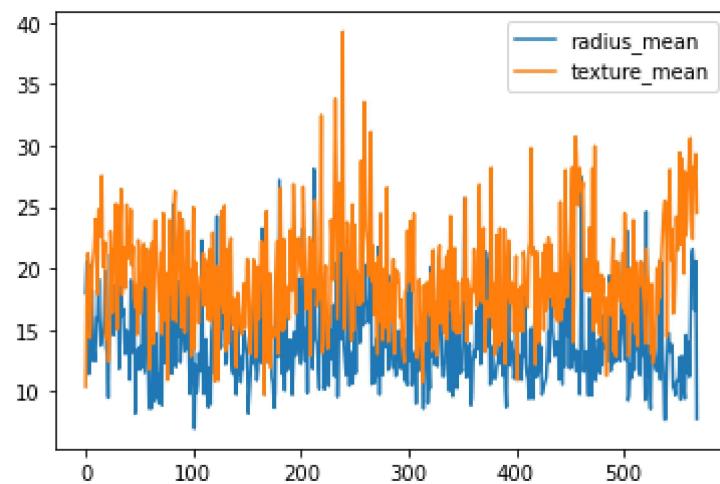
	radius_mean	texture_mean
0	17.99	10.38
1	20.57	17.77
2	19.69	21.25
3	11.42	20.38
4	20.29	14.34
...	...	...
564	21.56	22.39
565	20.13	28.25
566	16.60	28.08
567	20.60	29.33
568	7.76	24.54

569 rows × 2 columns

## line plot

In [12]: `data.plot.line()`

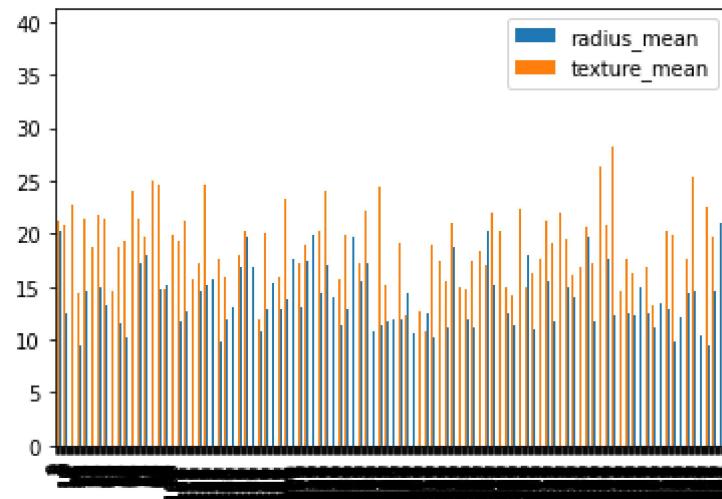
Out[12]: <AxesSubplot:>



## bar plot

```
In [13]: data.plot.bar()
```

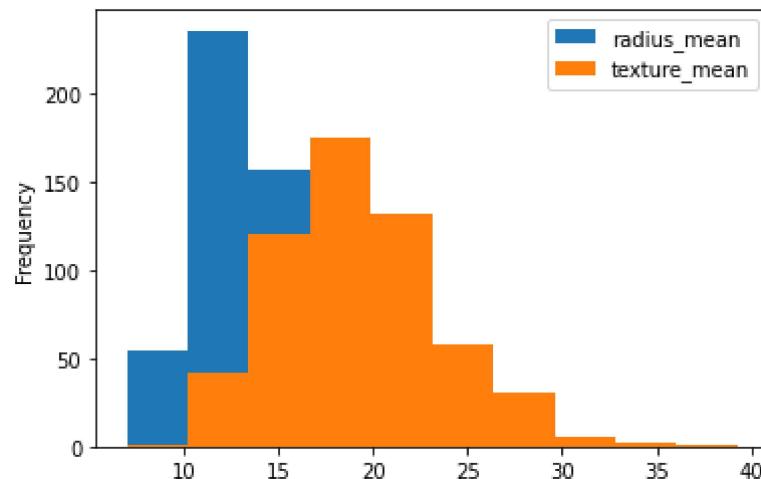
```
Out[13]: <AxesSubplot:>
```



## hist plot

```
In [14]: data.plot.hist()
```

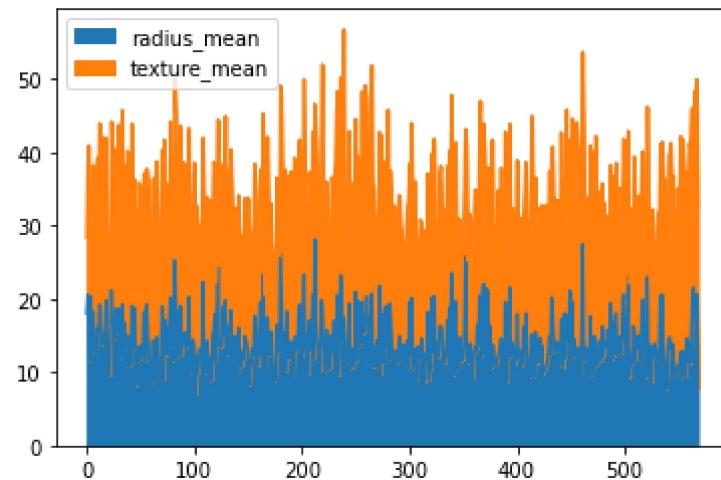
```
Out[14]: <AxesSubplot:ylabel='Frequency'>
```



## Area plot

```
In [15]: data.plot.area()
```

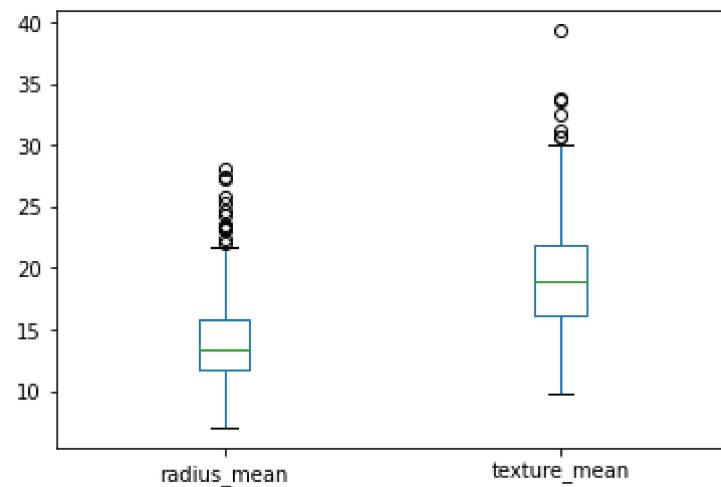
```
Out[15]: <AxesSubplot:>
```



## Box plot

```
In [16]: data.plot.box()
```

```
Out[16]: <AxesSubplot:>
```

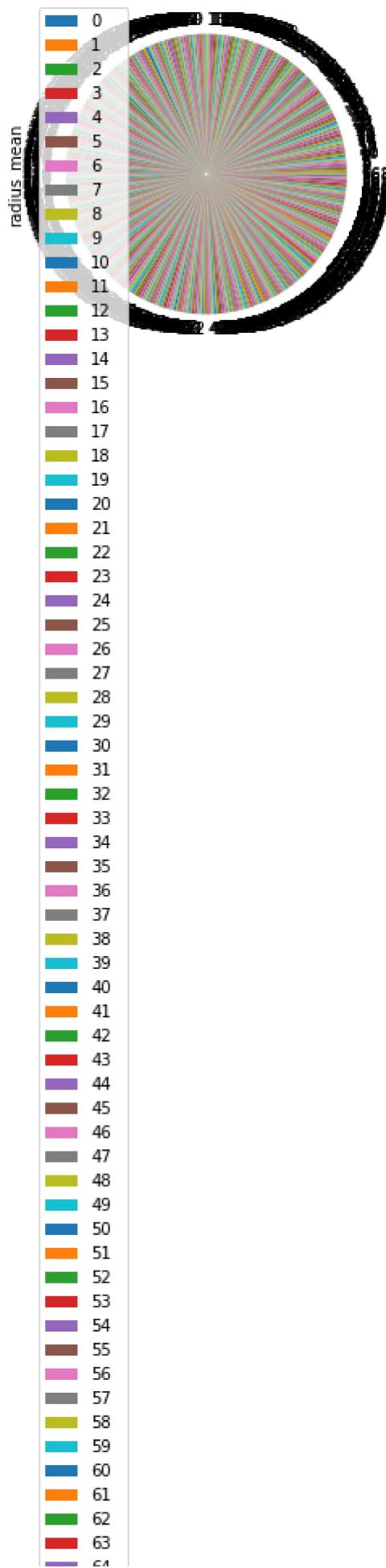


## pie plot

```
In [17]: data.plot.pie(y="radius_mean")
```

```
Out[17]: <AxesSubplot:ylabel='radius_mean'>
```





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```
In [18]: data.plot.scatter(x="radius_mean",y="texture_mean")
```

```
Out[18]: <AxesSubplot:xlabel='radius_mean', ylabel='texture_mean'>
```

