

```
In [13]: import pandas as pd
import numpy as np
from sklearn.preprocessing import MinMaxScaler

import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import skew
```

```
In [15]: df=pd.read_csv("breast_cancer.csv")
```

```
In [16]: df.head()
```

Out[16]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	..
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419	0.07871	..
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812	0.05667	..
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069	0.05999	..
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597	0.09744	..
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809	0.05883	..

5 rows × 31 columns

```
In [17]: df.tail()
```

Out[17]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	..
564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.1726	0.05623	..
565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.1752	0.05533	..
566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.1590	0.05648	..
567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.2397	0.07016	..
568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.1587	0.05884	..

5 rows × 31 columns

```
In [18]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 569 entries, 0 to 568
```

```
Data columns (total 31 columns):
```

#	Column	Non-Null Count	Dtype
0	mean radius	569 non-null	float64
1	mean texture	569 non-null	float64
2	mean perimeter	569 non-null	float64
3	mean area	569 non-null	float64
4	mean smoothness	569 non-null	float64
5	mean compactness	569 non-null	float64
6	mean concavity	569 non-null	float64
7	mean concave points	569 non-null	float64
8	mean symmetry	569 non-null	float64
9	mean fractal dimension	569 non-null	float64
10	radius error	569 non-null	float64
11	texture error	569 non-null	float64
12	perimeter error	569 non-null	float64
13	area error	569 non-null	float64
14	smoothness error	569 non-null	float64
15	compactness error	569 non-null	float64
16	concavity error	569 non-null	float64
17	concave points error	569 non-null	float64
18	symmetry error	569 non-null	float64
19	fractal dimension error	569 non-null	float64
20	worst radius	569 non-null	float64
21	worst texture	569 non-null	float64
22	worst perimeter	569 non-null	float64
23	worst area	569 non-null	float64
24	worst smoothness	569 non-null	float64
25	worst compactness	569 non-null	float64
26	worst concavity	569 non-null	float64
27	worst concave points	569 non-null	float64
28	worst symmetry	569 non-null	float64
29	worst fractal dimension	569 non-null	float64
30	outcome	569 non-null	int64

```
dtypes: float64(30), int64(1)
```

```
memory usage: 137.9 KB
```

```
In [19]: df.describe()
```

```
Out[19]:
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	
count	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	5
mean	14.127292	19.289649	91.969033	654.889104	0.096360	0.104341	0.088799	0.048919	
std	3.524049	4.301036	24.298981	351.914129	0.014064	0.052813	0.079720	0.038803	
min	6.981000	9.710000	43.790000	143.500000	0.052630	0.019380	0.000000	0.000000	
25%	11.700000	16.170000	75.170000	420.300000	0.086370	0.064920	0.029560	0.020310	
50%	13.370000	18.840000	86.240000	551.100000	0.095870	0.092630	0.061540	0.033500	
75%	15.780000	21.800000	104.100000	782.700000	0.105300	0.130400	0.130700	0.074000	
max	28.110000	39.280000	188.500000	2501.000000	0.163400	0.345400	0.426800	0.201200	

```
8 rows × 31 columns
```

```
In [20]: df.isnull()
```

Out[20]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension
0	False	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	False
...	...	...	...	...	...	...	...	...	...	...
564	False	False	False	False	False	False	False	False	False	False
565	False	False	False	False	False	False	False	False	False	False
566	False	False	False	False	False	False	False	False	False	False
567	False	False	False	False	False	False	False	False	False	False
568	False	False	False	False	False	False	False	False	False	False

569 rows × 31 columns

In [21]: `df.isnull().sum()`

```
Out[21]: mean radius          0
mean texture          0
mean perimeter        0
mean area             0
mean smoothness       0
mean compactness      0
mean concavity        0
mean concave points   0
mean symmetry         0
mean fractal dimension 0
radius error          0
texture error         0
perimeter error       0
area error            0
smoothness error      0
compactness error     0
concavity error       0
concave points error  0
symmetry error        0
fractal dimension error 0
worst radius          0
worst texture         0
worst perimeter       0
worst area            0
worst smoothness      0
worst compactness     0
worst concavity       0
worst concave points  0
worst symmetry        0
worst fractal dimension 0
outcome              0
dtype: int64
```

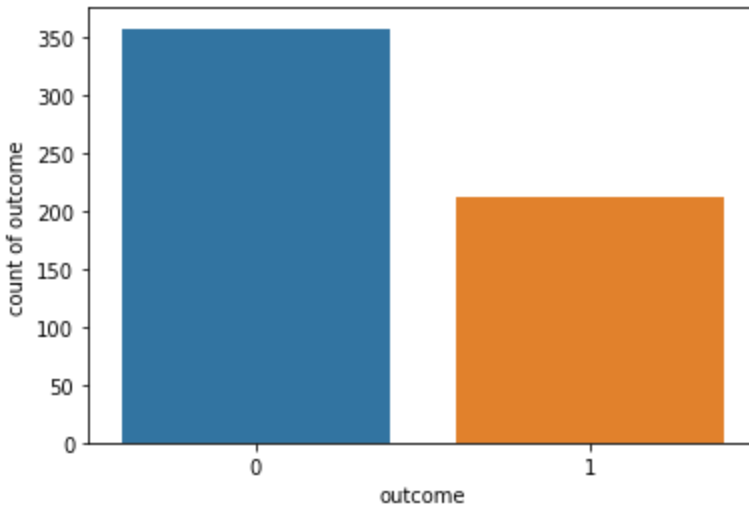
In [23]: `df["outcome"].value_counts()`

```
Out[23]: 0    357
1    212
Name: outcome, dtype: int64
```

```
In [24]: sns.countplot(df["outcome"])
plt.xlabel("outcome")
plt.ylabel("count of outcome")
plt.show()
```

C:\Users\kokila periyasamy\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

```
warnings.warn(
```



```
In [26]: x=df.iloc[:, :-1]
y=df.iloc[:, -1]
```

```
In [27]: X.shape
```

```
Out[27]: (569, 30)
```

```
In [28]: y.shape
```

```
Out[28]: (569,)
```

```
In [38]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=99)
```

```
In [39]: from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_absolute_error, r2_score
forest_model=RandomForestRegressor(n_estimators=750,max_depth=4,max_leaf_nodes=500,random_state=1)
```

```
In [40]: forest_model.fit(X_train, y_train)
```

```
Out[40]: RandomForestRegressor(max_depth=4, max_leaf_nodes=500, n_estimators=750,
                                random_state=1)
```

```
In [43]: forest_model.feature_importances_
```

```
Out[43]: array([0.00268249, 0.01581475, 0.00211393, 0.01592123, 0.00323533,
                0.00130309, 0.00458619, 0.08535125, 0.00101596, 0.00209099,
                0.0052539 , 0.00149769, 0.00382338, 0.02354548, 0.00100886,
                0.0017764 , 0.00406216, 0.00228771, 0.0011848 , 0.00530255,
                0.02044271, 0.01611963, 0.25185632, 0.07679379, 0.00688009,
                0.0023232 , 0.00808895, 0.42939713, 0.00254625, 0.00169379])
```

```
df.columns
```

```
In [44]: df.columns
```

```
Out[44]: Index(['mean radius', 'mean texture', 'mean perimeter', 'mean area',
        'mean smoothness', 'mean compactness', 'mean concavity',
        'mean concave points', 'mean symmetry', 'mean fractal dimension',
        'radius error', 'texture error', 'perimeter error', 'area error',
        'smoothness error', 'compactness error', 'concavity error',
        'concave points error', 'symmetry error', 'fractal dimension error',
        'worst radius', 'worst texture', 'worst perimeter', 'worst area',
        'worst smoothness', 'worst compactness', 'worst concavity',
        'worst concave points', 'worst symmetry', 'worst fractal dimension',
        'outcome'],
        dtype='object')
```

```
In [49]: power_preds=forest_model.predict(X_test)
```

```
In [51]: power_preds
```

```
Out[51]: array([0.01174317, 0.00507125, 0.0405616 , 0.00507125, 0.00515642,
        0.99894507, 0.75169602, 0.9989984 , 0.00537539, 0.01106544,
        0.9989984 , 0.06463776, 0.87146066, 0.00687874, 0.00507125,
        0.96111096, 0.9989984 , 0.29701253, 0.02709731, 0.0058881 ,
        0.28437355, 0.00507125, 0.00607125, 0.00507125, 0.9989984 ,
        0.00578553, 0.00507125, 0.00507125, 0.00507125, 0.00507125,
        0.9989984 , 0.00507125, 0.0088516 , 0.00607125, 0.00576466,
        0.02980223, 0.00507125, 0.01346395, 0.00507125, 0.88928831,
        0.02108799, 0.00517381, 0.06143323, 0.85816726, 0.00520633,
        0.04370317, 0.02119105, 0.00510377, 0.99633173, 0.01442264,
        0.35763056, 0.97624259, 0.56402574, 0.00507125, 0.99638228,
        0.00507125, 0.00592062, 0.00507125, 0.00507125, 0.00507125,
        0.9140906 , 0.9979984 , 0.9989984 , 0.00507125, 0.99533173,
        0.00507125, 0.00507125, 0.00684142, 0.0050654 , 0.9989984 ,
        0.07283112, 0.9989984 , 0.96468437, 0.07879092, 0.97826305,
        0.9989984 , 0.18051927, 0.9989984 , 0.96948153, 0.04340454,
        0.00507125, 0.00507125, 0.00507125, 0.08939125, 0.9989984 ,
        0.00510377, 0.00915486, 0.00510377, 0.33110267, 0.00507125,
        0.9989984 , 0.02788602, 0.00507125, 0.09251526, 0.67496218,
        0.9989984 , 0.9989984 , 0.32203922, 0.00507125, 0.91574758,
        0.00607125, 0.00507125, 0.1387028 , 0.41724875, 0.14574524,
        0.00507125, 0.60220517, 0.00507125, 0.00507125, 0.00507125,
        0.47925458, 0.94709742, 0.00520633, 0.00520633, 0.00640458,
        0.00510377, 0.92696388, 0.00520633, 0.00694294, 0.02653579,
        0.00507125, 0.0427797 , 0.99766507, 0.0547102 , 0.22950025,
        0.00507125, 0.82774075, 0.00607125, 0.00641641, 0.99633173,
        0.00507125, 0.90658357, 0.00507125, 0.00517381, 0.07333024,
        0.01211246, 0.0050654 , 0.99639234, 0.01590958, 0.01153497,
        0.00507125, 0.00517381, 0.00507125])
```

```
In [70]: from sklearn.metrics import mean_absolute_error, mean_squared_error, explained_variance
```

```
In [74]: forest_model.score(X_train,y_train)
```

```
Out[74]: 0.9543176382164082
```

```
In [76]: ytrain_preds=forest_model.predict(X_train)
```

```
In [77]: mean_absolute_error(y_train,ytrain_preds)
```

```
Out[77]: 0.041683460079334614
```

```
mean_squared_error(y_train,ytrain_preds)
```

```
In [78]: mean_squared_error(y_train,ytrain_preds)
```

Out[78]: 0.010910843798342484

In [ ]: