

```
In [1]: import pandas as pd  
import numpy as np
```

```
In [3]: from sklearn.datasets import load_breast_cancer
```

## Loading of breast cancer data

```
In [6]: data = load_breast_cancer()  
data  
print(data.feature_names)
```

```
['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']
```

```
In [8]: print(data.target_names)
```

```
['malignant' 'benign']
```

```
In [10]: # Convert to DataFrame  
df=pd.DataFrame(data.data,columns=data.feature_names)  
df
```

Out[10]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	r
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.30010	0.14710	0.2419	0.0	
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.08690	0.07017	0.1812	0.0	
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.19740	0.12790	0.2069	0.0	
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.24140	0.10520	0.2597	0.0	
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.19800	0.10430	0.1809	0.0	
...	...	...	...	...	...	...	...	...	...	...	
564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.1726	0.0	
565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.1752	0.0	
566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.1590	0.0	
567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.2397	0.0	
568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.1587	0.0	

569 rows × 30 columns



```
In [12]: ## Adding the target to the DataFrame(df)  
df['target']=data.target
```

df

Out[12]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	r fr dimer
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.30010	0.14710	0.2419	0.0
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.08690	0.07017	0.1812	0.0
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.19740	0.12790	0.2069	0.0
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.24140	0.10520	0.2597	0.0
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.19800	0.10430	0.1809	0.0
...	...	...	...	...	...	...	...	...	...	...
564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.1726	0.0
565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.1752	0.0
566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.1590	0.0
567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.2397	0.0
568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.1587	0.0

569 rows × 31 columns



In [14]: df.head(5)

Out[14]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	me frac dimensi
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419	0.078
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812	0.056
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069	0.059
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597	0.097
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809	0.058

5 rows × 31 columns



## preprocessing of data

In [17]: 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  target             569 non-null    int32
dtypes: float64(30), int32(1)
memory usage: 135.7 KB
```

In [19]: `df.isnull().sum()`

```
Out[19]: 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  
target                  0  
dtype: int64
```

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

```
Out[21]: 0
```

```
In [23]: ## There are no missing values and duplicate values in the dataset.
```

```
In [25]: df.describe()
```

Out[25]:

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave point
<b>count</b>	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000
<b>mean</b>	14.127292	19.289649	91.969033	654.889104	0.096360	0.104341	0.088799	0.048911
<b>std</b>	3.524049	4.301036	24.298981	351.914129	0.014064	0.052813	0.079720	0.038801
<b>min</b>	6.981000	9.710000	43.790000	143.500000	0.052630	0.019380	0.000000	0.000000
<b>25%</b>	11.700000	16.170000	75.170000	420.300000	0.086370	0.064920	0.029560	0.020311
<b>50%</b>	13.370000	18.840000	86.240000	551.100000	0.095870	0.092630	0.061540	0.033501
<b>75%</b>	15.780000	21.800000	104.100000	782.700000	0.105300	0.130400	0.130700	0.074001
<b>max</b>	28.110000	39.280000	188.500000	2501.000000	0.163400	0.345400	0.426800	0.201201

8 rows × 31 columns



In [27]:

```
df=df.select_dtypes(include='number')
df
correlation=df.corr()
correlation
print("\nCorrelation:")
print(correlation)
```



## Correlation:

	mean radius	mean texture	mean perimeter	mean area	\
mean radius	1.000000	0.323782	0.997855	0.987357	
mean texture	0.323782	1.000000	0.329533	0.321086	
mean perimeter	0.997855	0.329533	1.000000	0.986507	
mean area	0.987357	0.321086	0.986507	1.000000	
mean smoothness	0.170581	-0.023389	0.207278	0.177028	
mean compactness	0.506124	0.236702	0.556936	0.498502	
mean concavity	0.676764	0.302418	0.716136	0.685983	
mean concave points	0.822529	0.293464	0.850977	0.823269	
mean symmetry	0.147741	0.071401	0.183027	0.151293	
mean fractal dimension	-0.311631	-0.076437	-0.261477	-0.283110	
radius error	0.679090	0.275869	0.691765	0.732562	
texture error	-0.097317	0.386358	-0.086761	-0.066280	
perimeter error	0.674172	0.281673	0.693135	0.726628	
area error	0.735864	0.259845	0.744983	0.800086	
smoothness error	-0.222600	0.006614	-0.202694	-0.166777	
compactness error	0.206000	0.191975	0.250744	0.212583	
concavity error	0.194204	0.143293	0.228082	0.207660	
concave points error	0.376169	0.163851	0.407217	0.372320	
symmetry error	-0.104321	0.009127	-0.081629	-0.072497	
fractal dimension error	-0.042641	0.054458	-0.005523	-0.019887	
worst radius	0.969539	0.352573	0.969476	0.962746	
worst texture	0.297008	0.912045	0.303038	0.287489	
worst perimeter	0.965137	0.358040	0.970387	0.959120	
worst area	0.941082	0.343546	0.941550	0.959213	
worst smoothness	0.119616	0.077503	0.150549	0.123523	
worst compactness	0.413463	0.277830	0.455774	0.390410	
worst concavity	0.526911	0.301025	0.563879	0.512606	
worst concave points	0.744214	0.295316	0.771241	0.722017	
worst symmetry	0.163953	0.105008	0.189115	0.143570	
worst fractal dimension	0.007066	0.119205	0.051019	0.003738	
target	-0.730029	-0.415185	-0.742636	-0.708984	

	mean smoothness	mean compactness	mean concavity	\
mean radius	0.170581	0.506124	0.676764	
mean texture	-0.023389	0.236702	0.302418	
mean perimeter	0.207278	0.556936	0.716136	
mean area	0.177028	0.498502	0.685983	
mean smoothness	1.000000	0.659123	0.521984	
mean compactness	0.659123	1.000000	0.883121	
mean concavity	0.521984	0.883121	1.000000	
mean concave points	0.553695	0.831135	0.921391	
mean symmetry	0.557775	0.602641	0.500667	
mean fractal dimension	0.584792	0.565369	0.336783	
radius error	0.301467	0.497473	0.631925	
texture error	0.068406	0.046205	0.076218	
perimeter error	0.296092	0.548905	0.660391	
area error	0.246552	0.455653	0.617427	
smoothness error	0.332375	0.135299	0.098564	
compactness error	0.318943	0.738722	0.670279	
concavity error	0.248396	0.570517	0.691270	
concave points error	0.380676	0.642262	0.683260	
symmetry error	0.200774	0.229977	0.178009	
fractal dimension error	0.283607	0.507318	0.449301	
worst radius	0.213120	0.535315	0.688236	
worst texture	0.036072	0.248133	0.299879	
worst perimeter	0.238853	0.590210	0.729565	
worst area	0.206718	0.509604	0.675987	
worst smoothness	0.805324	0.565541	0.448822	
worst compactness	0.472468	0.865809	0.754968	
worst concavity	0.434926	0.816275	0.884103	

worst concave points	0.503053	0.815573	0.861323
worst symmetry	0.394309	0.510223	0.409464
worst fractal dimension	0.499316	0.687382	0.514930
target	-0.358560	-0.596534	-0.696360

mean radius	0.822529	0.147741	\
mean texture	0.293464	0.071401	
mean perimeter	0.850977	0.183027	
mean area	0.823269	0.151293	
mean smoothness	0.553695	0.557775	
mean compactness	0.831135	0.602641	
mean concavity	0.921391	0.500667	
mean concave points	1.000000	0.462497	
mean symmetry	0.462497	1.000000	
mean fractal dimension	0.166917	0.479921	
radius error	0.698050	0.303379	
texture error	0.021480	0.128053	
perimeter error	0.710650	0.313893	
area error	0.690299	0.223970	
smoothness error	0.027653	0.187321	
compactness error	0.490424	0.421659	
concavity error	0.439167	0.342627	
concave points error	0.615634	0.393298	
symmetry error	0.095351	0.449137	
fractal dimension error	0.257584	0.331786	
worst radius	0.830318	0.185728	
worst texture	0.292752	0.090651	
worst perimeter	0.855923	0.219169	
worst area	0.809630	0.177193	
worst smoothness	0.452753	0.426675	
worst compactness	0.667454	0.473200	
worst concavity	0.752399	0.433721	
worst concave points	0.910155	0.430297	
worst symmetry	0.375744	0.699826	
worst fractal dimension	0.368661	0.438413	
target	-0.776614	-0.330499	

mean fractal dimension	...	worst texture	\
mean radius	-0.311631	...	0.297008
mean texture	-0.076437	...	0.912045
mean perimeter	-0.261477	...	0.303038
mean area	-0.283110	...	0.287489
mean smoothness	0.584792	...	0.036072
mean compactness	0.565369	...	0.248133
mean concavity	0.336783	...	0.299879
mean concave points	0.166917	...	0.292752
mean symmetry	0.479921	...	0.090651
mean fractal dimension	1.000000	...	-0.051269
radius error	0.000111	...	0.194799
texture error	0.164174	...	0.409003
perimeter error	0.039830	...	0.200371
area error	-0.090170	...	0.196497
smoothness error	0.401964	...	-0.074743
compactness error	0.559837	...	0.143003
concavity error	0.446630	...	0.100241
concave points error	0.341198	...	0.086741
symmetry error	0.345007	...	-0.077473
fractal dimension error	0.688132	...	-0.003195
worst radius	-0.253691	...	0.359921
worst texture	-0.051269	...	1.000000
worst perimeter	-0.205151	...	0.365098

worst area	-0.231854	...	0.345842
worst smoothness	0.504942	...	0.225429
worst compactness	0.458798	...	0.360832
worst concavity	0.346234	...	0.368366
worst concave points	0.175325	...	0.359755
worst symmetry	0.334019	...	0.233027
worst fractal dimension	0.767297	...	0.219122
target	0.012838	...	-0.456903

	worst perimeter	worst area	worst smoothness \
mean radius	0.965137	0.941082	0.119616
mean texture	0.358040	0.343546	0.077503
mean perimeter	0.970387	0.941550	0.150549
mean area	0.959120	0.959213	0.123523
mean smoothness	0.238853	0.206718	0.805324
mean compactness	0.590210	0.509604	0.565541
mean concavity	0.729565	0.675987	0.448822
mean concave points	0.855923	0.809630	0.452753
mean symmetry	0.219169	0.177193	0.426675
mean fractal dimension	-0.205151	-0.231854	0.504942
radius error	0.719684	0.751548	0.141919
texture error	-0.102242	-0.083195	-0.073658
perimeter error	0.721031	0.730713	0.130054
area error	0.761213	0.811408	0.125389
smoothness error	-0.217304	-0.182195	0.314457
compactness error	0.260516	0.199371	0.227394
concavity error	0.226680	0.188353	0.168481
concave points error	0.394999	0.342271	0.215351
symmetry error	-0.103753	-0.110343	-0.012662
fractal dimension error	-0.001000	-0.022736	0.170568
worst radius	0.993708	0.984015	0.216574
worst texture	0.365098	0.345842	0.225429
worst perimeter	1.000000	0.977578	0.236775
worst area	0.977578	1.000000	0.209145
worst smoothness	0.236775	0.209145	1.000000
worst compactness	0.529408	0.438296	0.568187
worst concavity	0.618344	0.543331	0.518523
worst concave points	0.816322	0.747419	0.547691
worst symmetry	0.269493	0.209146	0.493838
worst fractal dimension	0.138957	0.079647	0.617624
target	-0.782914	-0.733825	-0.421465

	worst compactness	worst concavity \
mean radius	0.413463	0.526911
mean texture	0.277830	0.301025
mean perimeter	0.455774	0.563879
mean area	0.390410	0.512606
mean smoothness	0.472468	0.434926
mean compactness	0.865809	0.816275
mean concavity	0.754968	0.884103
mean concave points	0.667454	0.752399
mean symmetry	0.473200	0.433721
mean fractal dimension	0.458798	0.346234
radius error	0.287103	0.380585
texture error	-0.092439	-0.068956
perimeter error	0.341919	0.418899
area error	0.283257	0.385100
smoothness error	-0.055558	-0.058298
compactness error	0.678780	0.639147
concavity error	0.484858	0.662564
concave points error	0.452888	0.549592
symmetry error	0.060255	0.037119

fractal dimension error	0.390159	0.379975
worst radius	0.475820	0.573975
worst texture	0.360832	0.368366
worst perimeter	0.529408	0.618344
worst area	0.438296	0.543331
worst smoothness	0.568187	0.518523
worst compactness	1.000000	0.892261
worst concavity	0.892261	1.000000
worst concave points	0.801080	0.855434
worst symmetry	0.614441	0.532520
worst fractal dimension	0.810455	0.686511
target	-0.590998	-0.659610

	worst concave points	worst symmetry	\
mean radius	0.744214	0.163953	
mean texture	0.295316	0.105008	
mean perimeter	0.771241	0.189115	
mean area	0.722017	0.143570	
mean smoothness	0.503053	0.394309	
mean compactness	0.815573	0.510223	
mean concavity	0.861323	0.409464	
mean concave points	0.910155	0.375744	
mean symmetry	0.430297	0.699826	
mean fractal dimension	0.175325	0.334019	
radius error	0.531062	0.094543	
texture error	-0.119638	-0.128215	
perimeter error	0.554897	0.109930	
area error	0.538166	0.074126	
smoothness error	-0.102007	-0.107342	
compactness error	0.483208	0.277878	
concavity error	0.440472	0.197788	
concave points error	0.602450	0.143116	
symmetry error	-0.030413	0.389402	
fractal dimension error	0.215204	0.111094	
worst radius	0.787424	0.243529	
worst texture	0.359755	0.233027	
worst perimeter	0.816322	0.269493	
worst area	0.747419	0.209146	
worst smoothness	0.547691	0.493838	
worst compactness	0.801080	0.614441	
worst concavity	0.855434	0.532520	
worst concave points	1.000000	0.502528	
worst symmetry	0.502528	1.000000	
worst fractal dimension	0.511114	0.537848	
target	-0.793566	-0.416294	

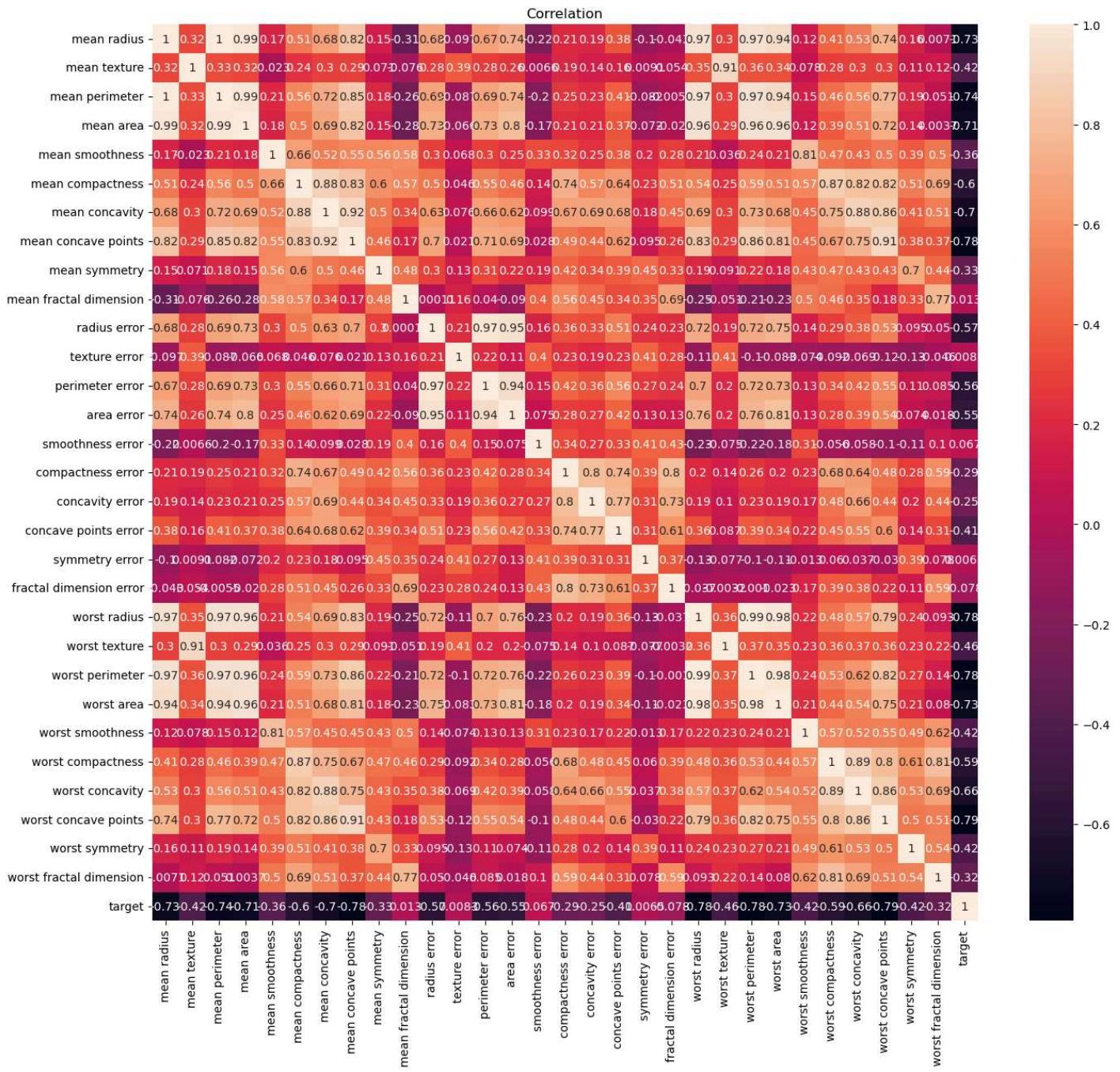
	worst fractal dimension	target
mean radius	0.007066	-0.730029
mean texture	0.119205	-0.415185
mean perimeter	0.051019	-0.742636
mean area	0.003738	-0.708984
mean smoothness	0.499316	-0.358560
mean compactness	0.687382	-0.596534
mean concavity	0.514930	-0.696360
mean concave points	0.368661	-0.776614
mean symmetry	0.438413	-0.330499
mean fractal dimension	0.767297	0.012838
radius error	0.049559	-0.567134
texture error	-0.045655	0.008303
perimeter error	0.085433	-0.556141
area error	0.017539	-0.548236
smoothness error	0.101480	0.067016

```
compactness error          0.590973 -0.292999
concavity error            0.439329 -0.253730
concave points error       0.310655 -0.408042
symmetry error             0.078079  0.006522
fractal dimension error    0.591328 -0.077972
worst radius                0.093492 -0.776454
worst texture               0.219122 -0.456903
worst perimeter              0.138957 -0.782914
worst area                  0.079647 -0.733825
worst smoothness             0.617624 -0.421465
worst compactness            0.810455 -0.590998
worst concavity              0.686511 -0.659610
worst concave points         0.511114 -0.793566
worst symmetry                0.537848 -0.416294
worst fractal dimension      1.000000 -0.323872
target                      -0.323872  1.000000
```

[31 rows x 31 columns]

```
In [29]: import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [31]: # Visualize the correlation matrix
plt.figure(figsize=(16,14))
sns.heatmap(correlation, annot=True)
plt.title("Correlation")
plt.show()
```



```
In [33]: ## Correlation is analyzed to understand relationships between features.
```

```
##High correlation between features might indicate redundancy and affect some models (like Logistic Regression)
```

```
In [35]: from sklearn.model_selection import train_test_split
```

```
In [37]: ## features and target
```

```
X = df.drop("target", axis=1)
y = df["target"]
```

```
In [39]: from sklearn.preprocessing import StandardScaler
```

```
In [41]: scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
X_scaled
```

```
Out[41]: array([[ 1.09706398, -2.07333501,  1.26993369, ... ,  2.29607613,
   2.75062224,  1.93701461],
   [ 1.82982061, -0.35363241,  1.68595471, ... ,  1.0870843 ,
  -0.24388967,  0.28118999],
   [ 1.57988811,  0.45618695,  1.56650313, ... ,  1.95500035,
  1.152255 ,  0.20139121],
   ... ,
   [ 0.70228425,  2.0455738 ,  0.67267578, ... ,  0.41406869,
  -1.10454895, -0.31840916],
   [ 1.83834103,  2.33645719,  1.98252415, ... ,  2.28998549,
  1.91908301,  2.21963528],
   [-1.80840125,  1.22179204, -1.81438851, ... , -1.74506282,
  -0.04813821, -0.75120669]])
```

```
In [43]: # Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.2, random_state=42)
X_train, X_test, y_train, y_test
```

```
Out[43]: (array([[-1.44798723, -0.45602336, -1.36665103, ..., 0.91959172,
   2.14719008, 1.85943247],
   [ 1.97750799, 1.69418666, 2.08961867, ..., 2.6752184 ,
   1.9368786 , 2.46346488],
   [-1.40708919, -1.26351565, -1.34976305, ..., -0.97048581,
   0.61676962, 0.05287682],
   ...,
   [ 0.04621146, -0.57470379, -0.06874782, ..., -1.23756033,
   -0.71628161, -1.26047806],
   [-0.04183295, 0.07687501, -0.03497186, ..., 1.03683652,
   0.45013821, 1.19444266],
   [-0.5530585 , 0.28631105, -0.60751564, ..., -0.61357437,
   -0.33448538, -0.84042616]]),
array([[ -0.47069438, -0.16048584, -0.44810956, ..., -0.19956318,
   0.18320441, 0.19695794],
   [ 1.36687747, 0.47014935, 1.30288585, ..., 0.97897545,
   -0.56582801, -1.00057787],
   [ 0.37850807, 0.04429607, 0.40082046, ..., 0.56024403,
   -0.10314275, -0.20813168],
   ...,
   [-0.74050787, -1.01451947, -0.74550281, ..., -0.28209134,
   -0.3830188 , -0.3245049 ],
   [ 0.02633046, 1.99205126, 0.02393013, ..., -0.49358878,
   -1.63518099, -0.33170895],
   [ 1.87526288, 2.75300221, 1.80128727, ..., 0.78102969,
   -0.05299156, -0.0978542 ]]),
68 1
181 0
63 1
248 1
60 1
..
71 1
106 1
270 1
435 0
102 1
Name: target, Length: 455, dtype: int32,
204 1
70 0
131 0
431 1
540 1
..
486 1
75 0
249 1
238 1
265 0
Name: target, Length: 114, dtype: int32)
```

In [104...]

```
# Function to train and evaluate models
def train_and_evaluate_model(model, model_name):
    model.fit(X_train, y_train)
    y_pred = model.predict(X_test)
    accuracy = accuracy_score(y_test, y_pred)
    print(f"{model_name} Accuracy: {accuracy:.2f}")
    print(f"{model_name} Classification Report:\n{classification_report(y_test, y_pred)}\n")
    print(f"{model_name} Confusion Matrix:\n{confusion_matrix(y_test, y_pred)}\n")
    return y_pred
```

```
In [106...]: # Initialize results dictionary
results = {
    "Model": [],
    "Accuracy": [],
    "Precision": [],
    "Recall": [],
    "F1_Score": []
}
```

## LogisticRegression

```
In [108...]: from sklearn.linear_model import LogisticRegression
```

```
In [110...]: from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score, classification_report
```

```
In [112...]: # Logistic Regression
log_reg = LogisticRegression()
log_reg
y_pred_log_reg = train_and_evaluate_model(log_reg, "Logistic Regression")
y_pred_log_reg
```

Logistic Regression Accuracy: 0.97

Logistic Regression Classification Report:

	precision	recall	f1-score	support
0	0.98	0.95	0.96	43
1	0.97	0.99	0.98	71
accuracy			0.97	114
macro avg	0.97	0.97	0.97	114
weighted avg	0.97	0.97	0.97	114

Logistic Regression Confusion Matrix:

```
[[41  2]
 [ 1 70]]
```

```
Out[112...]: array([1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1,
 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 1,
 1, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
 0, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1,
 0, 1, 0, 0])
```

```
In [142...]: ## Logistic Regression is classification model. It works well for binary & multiclass classification.
```

## DecisionTreeClassifier

```
In [115...]: from sklearn.tree import DecisionTreeClassifier
```

```
In [117...]: # Decision Tree Classifier
dt = DecisionTreeClassifier()
dt
y_pred_dt = train_and_evaluate_model(dt, "Decision Tree")
y_pred_dt
```

```
Decision Tree Accuracy: 0.94
```

```
Decision Tree Classification Report:
```

	precision	recall	f1-score	support
0	0.91	0.93	0.92	43
1	0.96	0.94	0.95	71
accuracy			0.94	114
macro avg	0.93	0.94	0.93	114
weighted avg	0.94	0.94	0.94	114

```
Decision Tree Confusion Matrix:
```

```
[[40  3]
 [ 4 67]]
```

```
Out[117... array([1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1,
 0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1,
 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0,
 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0,
 0, 1, 1, 0])
```

```
In [144... #Decision Tree splits data based on feature thresholds to make decisions. It handles non-linear relationships well.
```

## RandomForestClassifier

```
In [120... from sklearn.ensemble import RandomForestClassifier
```

```
In [122... rf = RandomForestClassifier()
rf
y_pred_rf= train_and_evaluate_model(rf, "Random Forest")
y_pred_rf
```

```
Random Forest Accuracy: 0.96
```

```
Random Forest Classification Report:
```

	precision	recall	f1-score	support
0	0.98	0.93	0.95	43
1	0.96	0.99	0.97	71
accuracy			0.96	114
macro avg	0.97	0.96	0.96	114
weighted avg	0.97	0.96	0.96	114

```
Random Forest Confusion Matrix:
```

```
[[40  3]
 [ 1 70]]
```

```
Out[122... array([1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1,
 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1,
 1, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 1,
 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0,
 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0,
 0, 1, 1, 0])
```

## # Support Vector Machine (SVM)

```
In [125...]: from sklearn.svm import SVC
```

```
In [127...]: svm = SVC()
svm
y_pred_svm = train_and_evaluate_model(svm, "SVM")
y_pred_svm
```

SVM Accuracy: 0.97

SVM Classification Report:

	precision	recall	f1-score	support
0	0.98	0.95	0.96	43
1	0.97	0.99	0.98	71
accuracy			0.97	114
macro avg	0.97	0.97	0.97	114
weighted avg	0.97	0.97	0.97	114

SVM Confusion Matrix:

```
[[41  2]
 [ 1 70]]
```

```
Out[127...]: array([1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1,
 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1,
 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0,
 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1,
 0, 1, 1, 0])
```

## KNeighborsClassifier

```
In [130...]: from sklearn.neighbors import KNeighborsClassifier
```

```
In [132...]: knn = KNeighborsClassifier()
knn
y_pred_knn = train_and_evaluate_model(knn, "k-NN")
y_pred_knn
```

k-NN Accuracy: 0.95

k-NN Classification Report:

	precision	recall	f1-score	support
0	0.93	0.93	0.93	43
1	0.96	0.96	0.96	71
accuracy			0.95	114
macro avg	0.94	0.94	0.94	114
weighted avg	0.95	0.95	0.95	114

k-NN Confusion Matrix:

```
[[40  3]
 [ 3 68]]
```

```
In [134]: # Step 3: Model Evaluation and Comparison
predictions = {
    "Logistic Regression": y_pred_log_reg,
    "Decision Tree": y_pred_dt,
    "Random Forest": y_pred_rf,
    "SVM": y_pred_svm,
    "k-NN": y_pred_knn
}
```

```
In [136]: for model_name, y_pred in predictions.items():
    results["Model"].append(model_name)
    results["Accuracy"].append(accuracy_score(y_test, y_pred))
    results["Precision"].append(precision_score(y_test, y_pred))
    results["Recall"].append(recall_score(y_test, y_pred))
    results["F1_Score"].append(f1_score(y_test, y_pred))
```

```
In [138...]: # Convert results to a DataFrame for better visualization  
results_df = pd.DataFrame(results)  
print("\nModel Evaluation Results:")  
print(results_df)
```

Model Evaluation Results:					
	Model	Accuracy	Precision	Recall	F1_Score
0	Logistic Regression	0.973684	0.972222	0.985915	0.979021
1	Decision Tree	0.938596	0.957143	0.943662	0.950355
2	Random Forest	0.964912	0.958904	0.985915	0.972222
3	SVM	0.973684	0.972222	0.985915	0.979021
4	k-NN	0.947368	0.957746	0.957746	0.957746

```
In [140...]: ## algorithm performed the best and which one performed the worst?  
#Best Performing Models Logistic Regression & SVM both models performed similarly across all  
#Best Performing Model is Logistic Regression  
#Worst Performing Model is Decision Tree
```

In [ ]: