

## breast-cancer-prediction

October 29, 2023

### 0.0.1 Breast Cancer Prediction :

```
[1]: # Import necessary libraries

import numpy as np
import pandas as pd
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report
```

```
[2]: # Load the breast cancer dataset

data = pd.read_csv("E:\\Dataset\\breast-cancer.csv")
```

```
[3]: data
```

[3]:	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	
..	...	...	...	...	...	...	
564	926424	M	21.56	22.39	142.00	1479.0	
565	926682	M	20.13	28.25	131.20	1261.0	
566	926954	M	16.60	28.08	108.30	858.1	
567	927241	M	20.60	29.33	140.10	1265.0	
568	92751	B	7.76	24.54	47.92	181.0	
	smoothness_mean	compactness_mean	concavity_mean	concave	points_mean	\	
0	0.11840	0.27760	0.30010		0.14710		
1	0.08474	0.07864	0.08690		0.07017		
2	0.10960	0.15990	0.19740		0.12790		
3	0.14250	0.28390	0.24140		0.10520		
4	0.10030	0.13280	0.19800		0.10430		
..	...	...	...		...		
564	0.11100	0.11590	0.24390		0.13890		

565	0.09780	0.10340	0.14400	0.09791
566	0.08455	0.10230	0.09251	0.05302
567	0.11780	0.27700	0.35140	0.15200
568	0.05263	0.04362	0.00000	0.00000

	...	radius_worst	texture_worst	perimeter_worst	area_worst	\
0	...	25.380	17.33	184.60	2019.0	
1	...	24.990	23.41	158.80	1956.0	
2	...	23.570	25.53	152.50	1709.0	
3	...	14.910	26.50	98.87	567.7	
4	...	22.540	16.67	152.20	1575.0	
..	...	...	...	...	...	
564	...	25.450	26.40	166.10	2027.0	
565	...	23.690	38.25	155.00	1731.0	
566	...	18.980	34.12	126.70	1124.0	
567	...	25.740	39.42	184.60	1821.0	
568	...	9.456	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	
568		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

[569 rows x 32 columns]

```
[4]: data.shape
```

```
[4]: (569, 32)
```

```
[5]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 569 entries, 0 to 568
Data columns (total 32 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   id                                     569 non-null    int64
1   diagnosis                             569 non-null    object
2   radius_mean                           569 non-null    float64
3   texture_mean                           569 non-null    float64
4   perimeter_mean                         569 non-null    float64
5   area_mean                             569 non-null    float64
6   smoothness_mean                        569 non-null    float64
7   compactness_mean                       569 non-null    float64
8   concavity_mean                         569 non-null    float64
9   concave points_mean                    569 non-null    float64
10  symmetry_mean                          569 non-null    float64
11  fractal_dimension_mean                 569 non-null    float64
12  radius_se                              569 non-null    float64
13  texture_se                             569 non-null    float64
14  perimeter_se                           569 non-null    float64
15  area_se                                569 non-null    float64
16  smoothness_se                          569 non-null    float64
17  compactness_se                         569 non-null    float64
18  concavity_se                           569 non-null    float64
19  concave points_se                      569 non-null    float64
20  symmetry_se                            569 non-null    float64
21  fractal_dimension_se                   569 non-null    float64
22  radius_worst                           569 non-null    float64
23  texture_worst                           569 non-null    float64
24  perimeter_worst                         569 non-null    float64
25  area_worst                             569 non-null    float64
26  smoothness_worst                       569 non-null    float64
27  compactness_worst                      569 non-null    float64
28  concavity_worst                        569 non-null    float64
29  concave points_worst                   569 non-null    float64
30  symmetry_worst                         569 non-null    float64
31  fractal_dimension_worst                 569 non-null    float64
dtypes: float64(30), int64(1), object(1)
memory usage: 142.4+ KB
```

```
[6]: data.isna().sum()
```

```
[6]: id 0
      diagnosis 0
      radius_mean 0
      texture_mean 0
      perimeter_mean 0
      area_mean 0
      smoothness_mean 0
      compactness_mean 0
      concavity_mean 0
      concave points_mean 0
      symmetry_mean 0
      fractal_dimension_mean 0
      radius_se 0
      texture_se 0
      perimeter_se 0
      area_se 0
      smoothness_se 0
      compactness_se 0
      concavity_se 0
      concave points_se 0
      symmetry_se 0
      fractal_dimension_se 0
      radius_worst 0
      texture_worst 0
      perimeter_worst 0
      area_worst 0
      smoothness_worst 0
      compactness_worst 0
      concavity_worst 0
      concave points_worst 0
      symmetry_worst 0
      fractal_dimension_worst 0
      dtype: int64
```

```
[7]: # Assuming 'diagnosis' is your target variable, and the rest are features
```

```
X = data.drop(columns=['diagnosis', 'id'])
```

```
[8]: X
```

```
[8]:      radius_mean  texture_mean  perimeter_mean  area_mean  smoothness_mean  \
0          17.99         10.38         122.80       1001.0         0.11840
1          20.57         17.77         132.90       1326.0         0.08474
2          19.69         21.25         130.00       1203.0         0.10960
3          11.42         20.38          77.58        386.1         0.14250
4          20.29         14.34         135.10       1297.0         0.10030
```

..	...	...	...	...	...
564	21.56	22.39	142.00	1479.0	0.11100
565	20.13	28.25	131.20	1261.0	0.09780
566	16.60	28.08	108.30	858.1	0.08455
567	20.60	29.33	140.10	1265.0	0.11780
568	7.76	24.54	47.92	181.0	0.05263

  

	compactness_mean	concavity_mean	concave	points_mean	symmetry_mean \
0	0.27760	0.30010		0.14710	0.2419
1	0.07864	0.08690		0.07017	0.1812
2	0.15990	0.19740		0.12790	0.2069
3	0.28390	0.24140		0.10520	0.2597
4	0.13280	0.19800		0.10430	0.1809
..	...	...		...	
564	0.11590	0.24390		0.13890	0.1726
565	0.10340	0.14400		0.09791	0.1752
566	0.10230	0.09251		0.05302	0.1590
567	0.27700	0.35140		0.15200	0.2397
568	0.04362	0.00000		0.00000	0.1587

  

	fractal_dimension_mean	...	radius_worst	texture_worst \
0	0.07871	...	25.380	17.33
1	0.05667	...	24.990	23.41
2	0.05999	...	23.570	25.53
3	0.09744	...	14.910	26.50
4	0.05883	...	22.540	16.67
..	...	...	...	...
564	0.05623	...	25.450	26.40
565	0.05533	...	23.690	38.25
566	0.05648	...	18.980	34.12
567	0.07016	...	25.740	39.42
568	0.05884	...	9.456	30.37

  

	perimeter_worst	area_worst	smoothness_worst	compactness_worst \
0	184.60	2019.0	0.16220	0.66560
1	158.80	1956.0	0.12380	0.18660
2	152.50	1709.0	0.14440	0.42450
3	98.87	567.7	0.20980	0.86630
4	152.20	1575.0	0.13740	0.20500
..	...	...	...	...
564	166.10	2027.0	0.14100	0.21130
565	155.00	1731.0	0.11660	0.19220
566	126.70	1124.0	0.11390	0.30940
567	184.60	1821.0	0.16500	0.86810
568	59.16	268.6	0.08996	0.06444

  

	concavity_worst	concave	points_worst	symmetry_worst \
--	-----------------	---------	--------------	------------------

0	0.7119	0.2654	0.4601
1	0.2416	0.1860	0.2750
2	0.4504	0.2430	0.3613
3	0.6869	0.2575	0.6638
4	0.4000	0.1625	0.2364
..	...	...	...
564	0.4107	0.2216	0.2060
565	0.3215	0.1628	0.2572
566	0.3403	0.1418	0.2218
567	0.9387	0.2650	0.4087
568	0.0000	0.0000	0.2871

	fractal_dimension_worst
0	0.11890
1	0.08902
2	0.08758
3	0.17300
4	0.07678
..	...
564	0.07115
565	0.06637
566	0.07820
567	0.12400
568	0.07039

[569 rows x 30 columns]

```
[9]: data.shape
```

```
[9]: (569, 32)
```

```
[10]: # Target variable
```

```
y = data['diagnosis']
```

```
[11]: y
```

```
[11]: 0      M
      1      M
      2      M
      3      M
      4      M
      ..
      564    M
      565    M
      566    M
      567    M
```

568 B  
Name: diagnosis, Length: 569, dtype: object

```
[12]: # Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
↪random_state=42)
```

```
[13]: X_train.shape
```

```
[13]: (455, 30)
```

```
[14]: X_test.shape
```

```
[14]: (114, 30)
```

```
[15]: y_train.shape
```

```
[15]: (455,)
```

```
[16]: y_test.shape
```

```
[16]: (114,)
```

```
[17]: # Create and train a Random Forest Classifier

clf = RandomForestClassifier(n_estimators=100, random_state=42)
clf.fit(X_train, y_train)
```

```
[17]: RandomForestClassifier(random_state=42)
```

```
[18]: # Make predictions on the test data

y_pred = clf.predict(X_test)
```

```
[ ]:
```

```
[19]: # Calculate accuracy
accuracy = accuracy_score(y_test, y_pred)
print(f"Accuracy: {accuracy}")
```

```
Accuracy: 0.9649122807017544
```

```
[20]: # Calculate accuracy and print a classification report
```

```
[21]: # Print a classification report with manually specified target names

target_names = ['benign', 'malignant']
```

```
report = classification_report(y_test, y_pred, target_names=target_names)
print("Classification Report:\n", report)
```

Classification Report:

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

[ ]: