

Importing the Dependencies

```
import numpy as np
import pandas as pd
import sklearn.datasets
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
from sklearn.model_selection import cross_val_score
import warnings
warnings.filterwarnings('ignore')
```

Data Collection & Processing

```
# loading the data from sklearn
breast_cancer_dataset = sklearn.datasets.load_breast_cancer()
```

```
print(breast_cancer_dataset)
```

[illegible]

```
# loading the data to a data frame
data_frame = pd.DataFrame(breast_cancer_dataset.data, columns = breast_cancer_dataset.feature_names)
```

```
# print the first 5 rows of the dataframe
data_frame.head()
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	...	worst radius	worst texture	worst perimeter
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419	0.07871	...	25.38	17.33	184.60
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812	0.05667	...	24.99	23.41	157.50
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069	0.05999	...	23.57	25.53	157.60
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597	0.09744	...	14.91	26.50	98.47
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809	0.05883	...	22.54	16.67	157.60

```
# adding the 'target' column to the data frame
data_frame['label'] = breast_cancer_dataset.target
```

```
# print last 5 rows of the dataframe
data_frame.tail()
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	...	worst radius	worst texture	worst perimeter	label
564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.1726	0.05623	...	26.40	166.10	166.10	2
565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.1752	0.05533	...	38.25	155.00	155.00	1
566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.1590	0.05648	...	34.12	126.70	126.70	1
567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.2397	0.07016	...	39.42	184.60	184.60	1
568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.1587	0.05884	...	30.37	59.16	59.16	0

5 rows × 31 columns

```
# number of rows and columns in the dataset
data_frame.shape
```

(569, 31)

```
# getting some information about the data
data_frame.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  label                                569 non-null    int64
dtypes: float64(30), int64(1)
memory usage: 137.9 KB
```

```
# checking for missing values
data_frame.isnull().sum()
```

```
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
label            0
dtype: int64
```

```
# statistical measures about the data
data_frame.describe()
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	...	
count	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	...	569
mean	14.127292	19.289649	91.969033	654.889104	0.096360	0.104341	0.088799	0.048919	0.181162	0.062798	...	0.062798
std	3.524049	4.301036	24.298981	351.914129	0.014064	0.052813	0.079720	0.038803	0.027414	0.007060	...	0.007060
min	6.981000	9.710000	43.790000	143.500000	0.052630	0.019380	0.000000	0.000000	0.106000	0.049960	...	0.049960
25%	11.700000	16.170000	75.170000	420.300000	0.086370	0.064920	0.029560	0.020310	0.161900	0.057700	...	0.057700
50%	13.370000	18.840000	86.240000	551.100000	0.095870	0.092630	0.061540	0.033500	0.179200	0.061540	...	0.061540
75%	15.780000	21.800000	104.100000	782.700000	0.105300	0.130400	0.130700	0.074000	0.195700	0.066120	...	0.066120
max	28.110000	39.280000	188.500000	2501.000000	0.163400	0.345400	0.426800	0.201200	0.304000	0.097440	...	0.097440

8 rows × 31 columns

```
# checking the distribution of Target Varibale
data_frame['label'].value_counts()
```

```
1    357
0    212
Name: label, dtype: int64
```

1 --> Benign
0 --> Malignant

```
data_frame.groupby('label').mean()
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	...	worst radius
label												
0	17.100000	04.001000	115.005000	070.070115	0.100000	0.115100	0.100735	0.007000	0.100000	0.000000		04.101011

Separating the features and target

```
X = data_frame.drop(columns='label', axis=1)
Y = data_frame['label']
```

```
print(X)
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	\
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	

	mean compactness	mean concavity	mean concave points	mean symmetry	\
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	

	mean fractal dimension	...	worst radius	worst texture	\
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	

	worst perimeter	worst area	worst smoothness	worst compactness	\
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	

	worst concavity	worst concave points	worst symmetry	\
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	

```
print(Y)
```

```
0    0
1    0
2    0
3    0
4    0
..
564  0
565  0
566  0
567  0
568  1
Name: label, Length: 569, dtype: int64
```

Splitting the data into training data & Testing data

```
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2, random_state=2)
```

```
print(X.shape, X_train.shape, X_test.shape)
```

```
(569, 30) (455, 30) (114, 30)
```

Model Training

Logistic Regression

```
model = LogisticRegression()
```

```
# training the Logistic Regression model using Training data
```

```
model.fit(X_train, Y_train)
```

```
▼ LogisticRegression  
LogisticRegression()
```

Model Evaluation

Accuracy Score

```
# accuracy on training data  
X_train_prediction = model.predict(X_train)  
training_data_accuracy = accuracy_score(Y_train, X_train_prediction)
```

```
print('Accuracy on training data = ', training_data_accuracy)
```

```
Accuracy on training data = 0.9472527472527472
```

```
# accuracy on test data  
X_test_prediction = model.predict(X_test)  
test_data_accuracy = accuracy_score(Y_test, X_test_prediction)
```

```
print('Accuracy on test data = ', test_data_accuracy)
```

```
Accuracy on test data = 0.9298245614035088
```

Building a Predictive System

```
input_data = (13.54,14.36,87.46,566.3,0.09779,0.08129,0.06664,0.04781,0.1885,0.05766,0.2699,0.7886,2.058,23.56,0.008462,0.0146,0.02387,0.
```

```
# change the input data to a numpy array  
input_data_as_numpy_array = np.asarray(input_data)
```

```
# reshape the numpy array as we are predicting for one datapoint  
input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)
```

```
prediction = model.predict(input_data_reshaped)  
print(prediction)
```

```
if (prediction[0] == 0):  
    print('The Breast cancer is Malignant')
```

```
else:  
    print('The Breast Cancer is Benign')
```

```
[1]  
The Breast Cancer is Benign  
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but LogisticRegression  
warnings.warn(
```



```
from sklearn.metrics import precision_score
from sklearn.metrics import recall_score
from sklearn.metrics import f1_score
```

```
# precision for training data predictions
precision_train = precision_score(Y_train, X_train_prediction)
print('Training data Precision =', precision_train)
```

Training data Precision = 0.952054794520548

```
# precision for test data predictions
precision_test = precision_score(Y_test, X_test_prediction)
print('Test data Precision =', precision_test)
```

Test data Precision = 0.9420289855072463

```
def precision_recall_f1_score(true_labels, pred_labels):

    precision_value = precision_score(true_labels, pred_labels)
    recall_value = recall_score(true_labels, pred_labels)
    f1_score_value = f1_score(true_labels, pred_labels)

    print('Precision =',precision_value)
    print('Recall =',recall_value)
    print('F1 Score =',f1_score_value)
```

```
# classification metrics for training data
precision_recall_f1_score(Y_train, X_train_prediction)
```

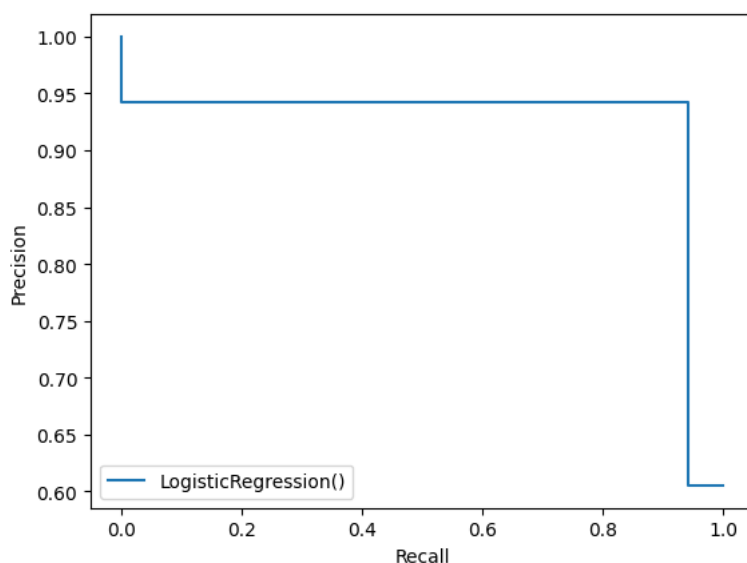
Precision = 0.952054794520548
Recall = 0.9652777777777778
F1 Score = 0.9586206896551724

```
# classification metrics for test data
precision_recall_f1_score(Y_test, X_test_prediction)
```

Precision = 0.9420289855072463
Recall = 0.9420289855072463
F1 Score = 0.9420289855072463

```
from sklearn.metrics import precision_recall_curve
from sklearn.metrics import PrecisionRecallDisplay

y_pred = model.predict(X_test)
prec, recall, _ = precision_recall_curve(Y_test, y_pred, pos_label=model.classes_[1])
pr_display = PrecisionRecallDisplay(precision=prec, recall=recall).plot(label=model)
```



```
from sklearn.metrics import confusion_matrix
cf_matrix = confusion_matrix(Y_test, X_test_prediction)

print(cf_matrix)
```

```
[[41  4]
 [ 4 65]]
```

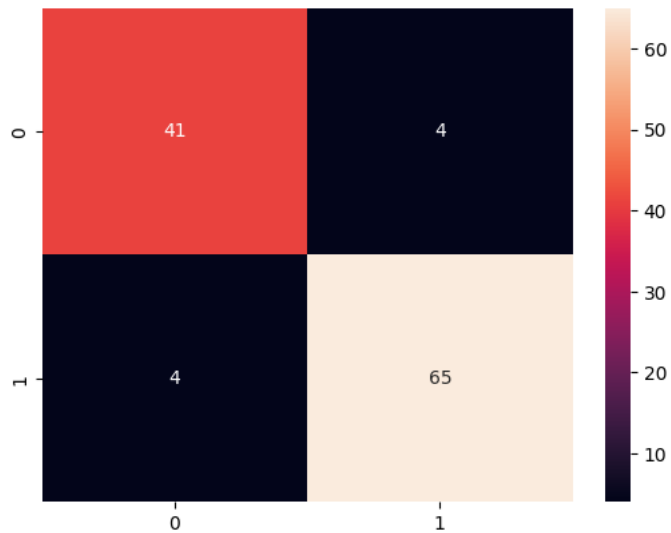
```
tn, fp, fn, tp = cf_matrix.ravel()
```

```
print(tn, fp, fn, tp)
```

```
41 4 4 65
```

```
import seaborn as sns
sns.heatmap(cf_matrix, annot=True)
```

<Axes: >



```
cv_score = cross_val_score(model, X_train, Y_train, cv=3)
print(cv_score)
```

```
[0.92763158 0.95394737 0.95364238]
```