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Connect

Importing the Dependencies

```
[ ] 1 import numpy as np
    2 import pandas as pd
    3 import sklearn.datasets
    4 from sklearn.model_selection import train_test_split
    5 from sklearn.linear_model import LogisticRegression
    6 from sklearn.metrics import accuracy_score
```

Data Collection & Processing

```
[ ] 1 # loading the data from sklearn
    2 breast_cancer_dataset = sklearn.datasets.load_breast_cancer()
```

```
[ ] 1 print(breast_cancer_dataset)
```

```
{'data': array([[1.799e+01, 1.038e+01, 1.228e+02, ..., 2.654e-01, 4.601e-01,
                  1.189e-01],
                 [2.057e+01, 1.777e+01, 1.329e+02, ..., 1.860e-01, 2.750e-01,
                  8.902e-02],
```

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

```
[ ] 1 # print the first 5 rows of the dataframe
    2 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	radius error	texture error	perimeter error	area error	smoothn er
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419	0.07871	1.0950	0.9053	8.589	153.40	0.006
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812	0.05667	0.5435	0.7339	3.398	74.08	0.005
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069	0.05999	0.7456	0.7869	4.585	94.03	0.006
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597	0.09744	0.4956	1.1560	3.445	27.23	0.009
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809	0.05883	0.7572	0.7813	5.438	94.44	0.011

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

```
[ ] 1 # print last 5 rows of the dataframe
    2 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	radius error	texture error	perimeter error	area error	smoothn er
564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.1726	0.05623	1.1760	1.256	7.673	158.70	0.0
565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.1752	0.05533	0.7655	2.463	5.203	99.04	0.0
566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.1590	0.05648	0.4564	1.075	3.425	48.55	0.0
567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.2397	0.07016	0.7260	1.595	5.772	86.22	0.0
568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.1587	0.05884	0.3857	1.428	2.548	19.15	0.0

```
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[ ] 1 # number of rows and columns in the dataset
    2 data_frame.shape

(569, 31)

[ ] 1 # getting some information about the data
    2 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
```

```
[ ] 1 # checking for missing values
    2 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
```

```
[ ] 1 # statistical measures about the data
    2 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  radius error  texture error
count  569.000000  569.000000  569.000000    569.000000    569.000000    569.000000    569.000000    569.000000    569.000000    569.000000    569.000000    569.000000
mean    14.127292   19.289649   91.969033    654.889104     0.096360     0.104341     0.088799     0.048919     0.181162     0.062798     0.405172     1.216853
std     3.524049    4.301036   24.298981    351.914129     0.014064     0.052813     0.079720     0.038803     0.027414     0.007060     0.277313     0.551648
min      6.981000    9.710000   43.790000   143.500000     0.052630     0.019380     0.000000     0.000000     0.106000     0.049960     0.111500     0.360200
25%     11.700000   16.170000   75.170000   420.300000     0.086370     0.064920     0.029560     0.020310     0.161900     0.057700     0.232400     0.833900
50%     13.370000   18.840000   86.240000   551.100000     0.095870     0.092630     0.061540     0.033500     0.179200     0.061540     0.324200     1.108000
75%     15.780000   21.800000   104.100000   782.700000     0.105300     0.130400     0.130700     0.074000     0.195700     0.066120     0.478900     1.474000
max     28.110000   39.280000   188.500000  2501.000000     0.163400     0.345400     0.426800     0.201200     0.304000     0.097440     2.873000     4.885000
```

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

```
{x}
1    357
0    212
Name: label, dtype: int64
```

1 -> Benign

0 -> Malignant

```
[ ] 1 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	radius error	texture error	perimeter error
label													
0	17.462830	21.604906	115.365377	978.376415	0.102898	0.145188	0.160775	0.087990	0.192909	0.062680	0.609083	1.210915	4.323929
1	12.146524	17.914762	78.075406	462.790196	0.092478	0.080085	0.046058	0.025717	0.174186	0.062867	0.284082	1.220380	2.000321

Separating the features and target

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

```
[ ] 1 print(X)
```

```

      mean radius  mean texture  ...  worst symmetry  worst fractal dimension
0             17.99         10.38  ...             0.4601                0.11890
1             20.57         17.77  ...             0.2750                0.08902
2             19.69         21.25  ...             0.3613                0.08758
3             11.42         20.38  ...             0.6638                0.17300
4             20.29         14.34  ...             0.2364                0.07678
..            ...          ...   ...             ...                  ...
564           21.56         22.39  ...             0.2060                0.07115
565           20.13         28.25  ...             0.2572                0.06637
566           16.60         28.08  ...             0.2218                0.07820
567           20.60         29.33  ...             0.4087                0.12400
568            7.76         24.54  ...             0.2871                0.07039
```

[569 rows x 30 columns]

Splitting the data into training data & Testing data

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

```
[ ] 1 print(X.shape, X_train.shape, X_test.shape)
```

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

Model Training

Logistic Regression

```
[ ] 1 model = LogisticRegression()
```

```
[ ] 1 # training the Logistic Regression model using Training data
2
3 model.fit(X_train, Y_train)
```

/usr/local/lib/python3.7/dist-packages/sklearn/linear_model/_logistic.py:818: ConvergenceWarning: lbfgs failed to converge (status=1):

Q

{x}

Model Evaluation

Accuracy Score

[]

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

[]

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

Accuracy on training data = 0.9494505494505494

[]

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

[]

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

Accuracy on test data = 0.9298245614035088

Q

{x}

Building a Predictive System

[]

1 input_data = (13.54,14.36,87.46,566.3,0.09779,0.08129,0.06664,0.04781,0.1885,0.05766,0.2699,0.7886,2.058,23.56,0.008462,0.0146,0.02387,0.0131!
2
3 # change the input data to a numpy array
4 input_data_as_numpy_array = np.asarray(input_data)
5
6 # reshape the numpy array as we are predicting for one datapoint
7 input_data_resaped = input_data_as_numpy_array.reshape(1,-1)
8
9 prediction = model.predict(input_data_resaped)
10 print(prediction)
11
12 if (prediction[0] == 0):
13 print('The Breast cancer is Malignant')
14
15 else:
16 print('The Breast Cancer is Benign')
17
18

[1]
The Breast Cancer is Benign