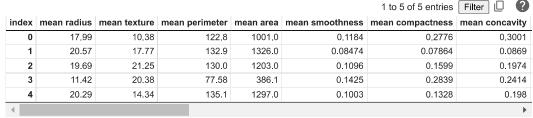
Importing the Dependencies

```
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
```

Data Collection & Processing

```
# loading the data from sklearn
breast_cancer_dataset = sklearn.datasets.load_breast_cancer()
print(breast cancer dataset)
    {'data': array([[1.799e+01, 1.038e+01, 1.228e+02, ..., 2.654e-01, 4.601e-01,
           [2.057e+01, 1.777e+01, 1.329e+02, ..., 1.860e-01, 2.750e-01,
            8.902e-02],
           [1.969e+01, 2.125e+01, 1.300e+02, ..., 2.430e-01, 3.613e-01,
            8.758e-02],
           [1.660e+01, 2.808e+01, 1.083e+02, ..., 1.418e-01, 2.218e-01,
            7.820e-02],
           [2.060e+01, 2.933e+01, 1.401e+02, ..., 2.650e-01, 4.087e-01,
            1.240e-01],
           [7.760e+00, 2.454e+01, 4.792e+01, ..., 0.000e+00, 2.871e-01,
            0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0,
           1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0,
           1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1,
           1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0,
           0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1,
           1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
           1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0,
           0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0,
           1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1,
           1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
           0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1,
           1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1,
           1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0,
           0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0,
           0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0,
           1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 1,
           1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0,
           1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1,
           1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0,
           1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1,
           1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1,
           1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
           1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1]), 'frame': None, 'target_names': array(['malignant', 'benign'], dtype='<
           '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'], dtype='<U23'), 'filename': 'breast cancer.csv', 'data module': 'sklearn.datasets.da
    4
```

```
# 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()
```

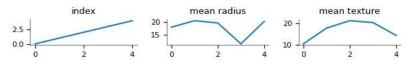


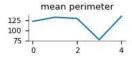
Show 25 ✔ per page



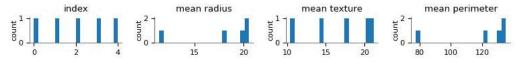
Like what you see? Visit the <u>data table notebook</u> to learn more about interactive tables. Warning: Total number of columns (30) exceeds max_columns (20) limiting to first (20) columns.

Values





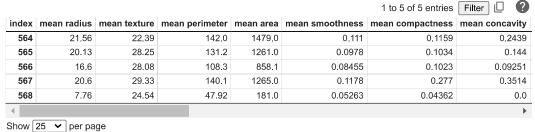
Distributions



2-d distributions



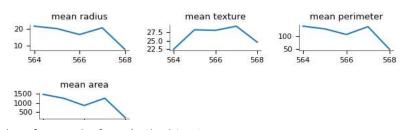
adding the 'target' column to the data frame
data_frame['label'] = breast_cancer_dataset.target
print last 5 rows of the dataframe
data_frame.tail()





Like what you see? Visit the data table notebook to learn more about interactive tables.

Values



number of rows and columns in the dataset
data_frame.shape
getting some information about the data

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 mean fractal dimension 0 radius error texture error a perimeter error 0 area error smoothness error 0 ${\tt compactness\ error}$ concavity error concave points error symmetry error fractal dimension error worst radius worst texture worst perimeter worst area worst smoothness worst compactness worst concavity 0 worst concave points worst symmetry 0 worst fractal dimension 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	
count	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	569.000000	Ę
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





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()

Separating the features and target

```
X = data_frame.drop(columns='label', axis=1)
Y = data_frame['label']
print(X)
print(Y)
    564
                        0.05623 ...
                                            25.450
                                                            26.40
                        0.05533 ...
                                                            38.25
     565
                                            23.690
                        0.05648 ...
     566
                                             18.980
                                                             34.12
    567
                        0.07016 ...
                                            25.740
                                                            39.42
                        0.05884 ...
                                             9.456
                                                            30.37
    568
         worst perimeter worst area worst smoothness worst compactness \
    0
                              2019.0
                                               0.16220
                                                                  0.66560
                  184.60
                                                                  0.18660
    1
                  158.80
                              1956.0
                                               0.12380
                  152.50
                              1709.0
                                               0.14440
                                                                  0.42450
    3
                   98.87
                               567.7
                                               0.20980
                                                                  0.86630
                  152.20
                                                                  0.20500
    4
                              1575.0
                                               0.13740
                  166.10
                              2027.0
                                               0.14100
                                                                  0.21130
     564
    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
                                        0.1860
                                                        0.2750
    1
                  0.2416
    2
                  0.4504
                                        0.2430
                                                        0.3613
                  0.6869
                                        0.2575
    3
                                                        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
                  0.0000
                                        0.0000
                                                        0.2871
         worst fractal dimension
    0
                         0.11890
    1
                         0.08902
                         0.08758
    2
    3
                         0.17300
    4
                         0.07678
                         0.07115
    564
                          0.06637
     566
                         0.07820
    567
                         0.12400
    568
                          0.07039
    [569 rows x 30 columns]
    0
           0
    1
           0
    2
           0
    3
           a
    4
           0
    564
           a
    565
           0
     566
     567
           0
    568
    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 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 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 training data = 0.9472527472527472
Accuracy on test data = 0.9298245614035088
```

Building a Predictive System

```
input_data = (13.54,14.36,87.46,566.3,0.09779,0.08129,0.06664,0.04781,0.1885,0.05766,0.2699,0.7886,2.058,23.56,0.008462,0.0146,0.02387,0.0131
# 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 wawarnings.warn(
```

✓ 0s completed at 12:22 PM