

## Logistic regression:

Logistic regression is the appropriate regression analysis to conduct when the dependent variable is dichotomous (binary).

Like all regression analyses, logistic regression is a predictive analysis.

Logistic regression is used to describe data and to explain the relationship between one dependent binary variable

and one or more nominal, ordinal, interval or ratio-level independent variables.

## How does Logistic Regression work?

- Logistic regression works in the following steps:

### ***1, Prepare the data:***

- The data should be in a format where each row represents a single observation
- and each column represents a different variable. The target variable (the variable you want to predict)
- should be binary (yes/no, true/false, 0/1).

### ***2, Train the model:***

- We teach the model by showing it the training data.
- This involves finding the values of the model parameters that minimize the error in the training data.

### ***3, Evaluate the model:***

- The model is evaluated on the held-out test data to assess its performance on unseen data.
- Use the model to make predictions: After the model has been trained and assessed,
- it can be used to forecast outcomes on new data.

In [1]:

```
1 import matplotlib.pyplot as plt
2 %matplotlib inline
3 from sklearn.datasets import load_digits
4
5 import warnings
6 warnings.filterwarnings('ignore')
```

```
In [2]: 1 digits = load_digits()
```

```
In [3]: 1 dir(digits)
```

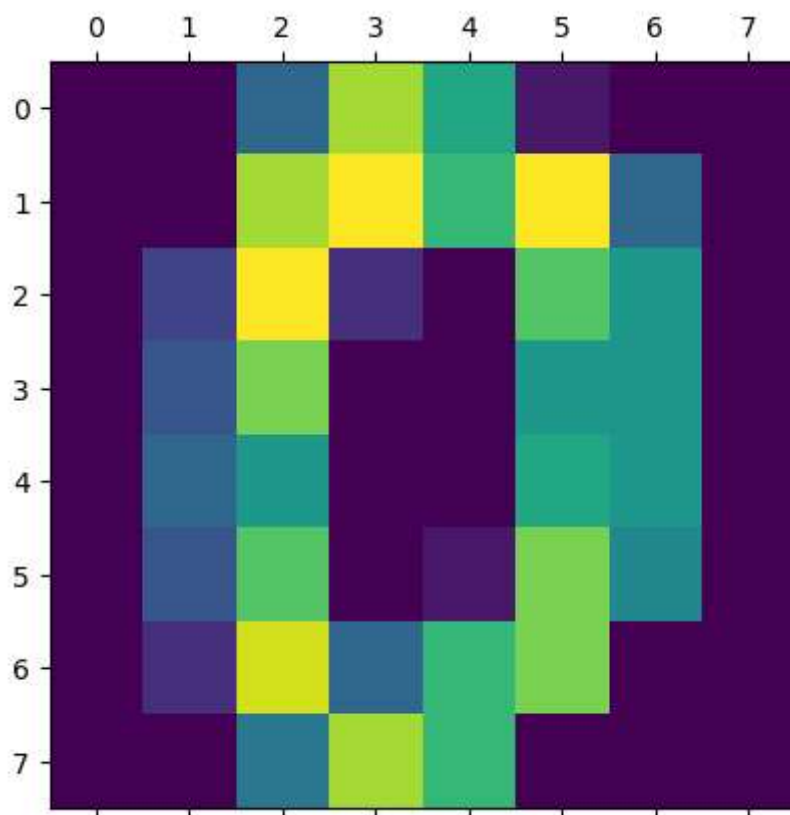
```
Out[3]: ['DESCR', 'data', 'feature_names', 'frame', 'images', 'target', 'target_names']
```

```
In [4]: 1 digits.data[0]
```

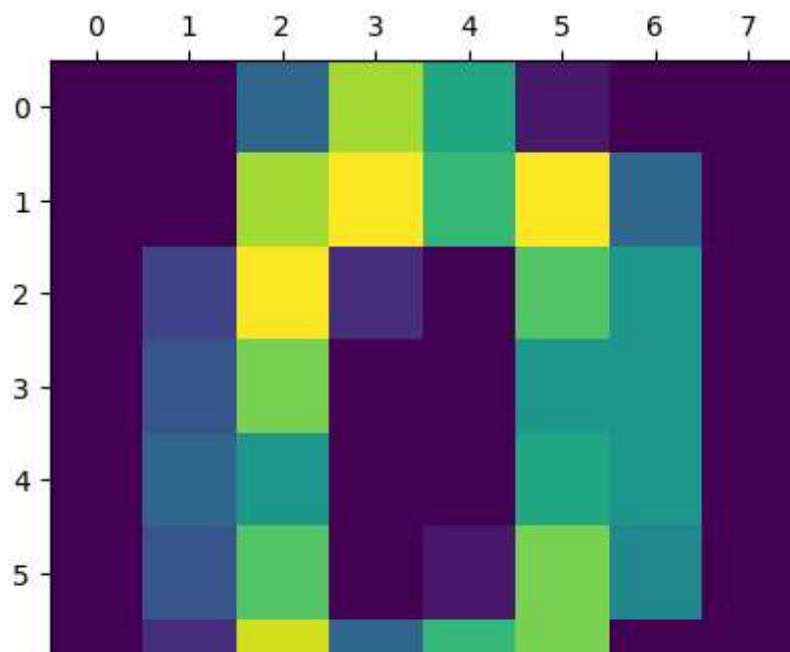
```
Out[4]: array([ 0.,  0.,  5., 13.,  9.,  1.,  0.,  0.,  0.,  0., 13., 15., 10.,
                15.,  5.,  0.,  0.,  3., 15.,  2.,  0., 11.,  8.,  0.,  0.,  4.,
                12.,  0.,  0.,  8.,  8.,  0.,  0.,  5.,  8.,  0.,  0.,  9.,  8.,
                 0.,  0.,  4., 11.,  0.,  1., 12.,  7.,  0.,  0.,  2., 14.,  5.,
                10., 12.,  0.,  0.,  0.,  0.,  6., 13., 10.,  0.,  0.,  0.])
```

```
In [5]: 1 plt.gray
        2 plt.matshow(digits.images[0])
```

```
Out[5]: <matplotlib.image.AxesImage at 0x1a9ad456c20>
```



```
In [6]: 1 plt.gray
        2 for i in range(6):
        3     plt.matshow(digits.images[i])
```



```
In [7]: 1 digits.target[0:5]
```

```
Out[7]: array([0, 1, 2, 3, 4])
```

```
In [8]: 1 # Split the data into a training set and a test set
        2 from sklearn.model_selection import train_test_split
        3 X_train, X_test, y_train, y_test = train_test_split(digits.data ,digits.ta
```

```
In [9]: 1 len(X_train)
```

```
Out[9]: 1437
```

```
In [10]: 1 len(X_test)
```

```
Out[10]: 360
```

```
In [11]: 1 from sklearn.linear_model import LogisticRegression
        2
        3 model = LogisticRegression()
        4
```

```
In [12]: 1 model.fit(X_train, y_train)
```

```
Out[12]: ▼ LogisticRegression
          LogisticRegression()
```

```
In [13]: 1 model.score(X_test, y_test)
```

```
Out[13]: 0.9694444444444444
```

```
In [14]: 1 model.predict(digits.data[0:5])
```

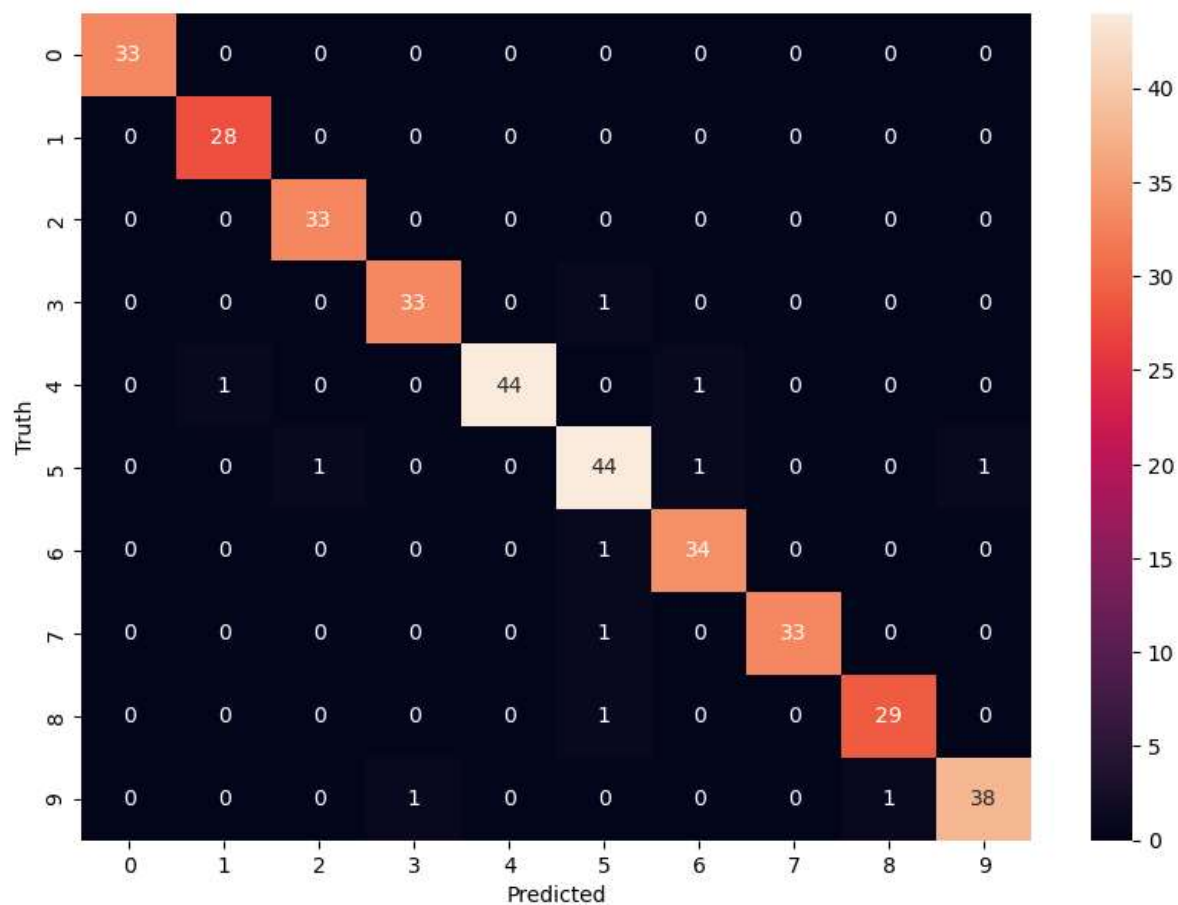
```
Out[14]: array([0, 1, 2, 3, 4])
```

```
In [15]: 1 y_predicted =model.predict(X_test)
2 from sklearn.metrics import confusion_matrix
3
4 cm = confusion_matrix(y_test, y_predicted)
5 cm
```

```
Out[15]: array([[33,  0,  0,  0,  0,  0,  0,  0,  0,  0],
 [ 0, 28,  0,  0,  0,  0,  0,  0,  0,  0],
 [ 0,  0, 33,  0,  0,  0,  0,  0,  0,  0],
 [ 0,  0,  0, 33,  0,  1,  0,  0,  0,  0],
 [ 0,  1,  0,  0, 44,  0,  1,  0,  0,  0],
 [ 0,  0,  1,  0,  0, 44,  1,  0,  0,  1],
 [ 0,  0,  0,  0,  0,  1, 34,  0,  0,  0],
 [ 0,  0,  0,  0,  0,  1,  0, 33,  0,  0],
 [ 0,  0,  0,  0,  0,  1,  0,  0, 29,  0],
 [ 0,  0,  0,  1,  0,  0,  0,  0,  1, 38]], dtype=int64)
```

```
In [16]: 1 import seaborn as sns
2
3 plt.figure(figsize = (10, 7))
4 sns.heatmap(cm , annot = True)
5 plt.xlabel("Predicted")
6 plt.ylabel("Truth")
```

Out[16]: Text(95.7222222222221, 0.5, 'Truth')



```
In [ ]: 1
```