

Surviving the Titanic

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Teaching the model with the passenger dataset

Dropping the non-essential components of the dataset

Determining the survival of passengers and evaluating the model simpl:

Agenda

- What is Supervised Learning?
- What is Classification? What are some of its solutions?
- What is Logistic Regression?
- Comparing Linear and Logistic Regression
- Logistic Regression applications
- Use Case Predicting the number in an image







What is Supervised Learning?

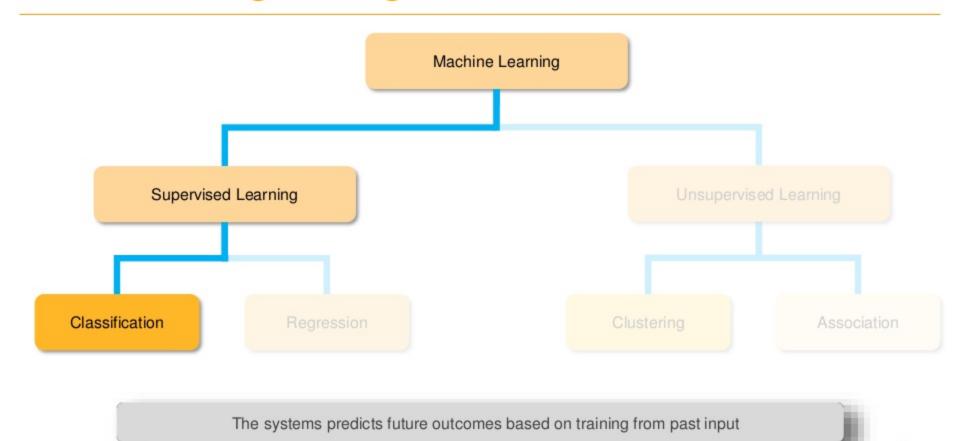


Teacher teaches child

Child recognizes an apple when she sees it again

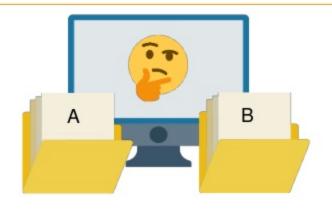
A model is able to make predictions based on past data

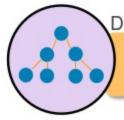
Where does Logistic Regression fit it?





A few Classification Solutions



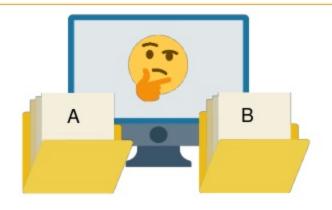


Decision Trees

We take decisions using a tree structure. Each branch node represents a choice, and leaf node represents a decision



A few Classification Solutions





Decision Trees

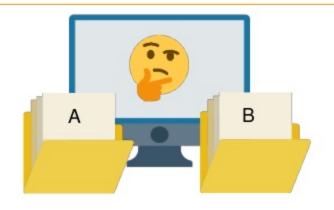
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K-Nearest Neighbor

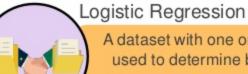
It helps determine what the given object is, based on its similarity to the objects it is compared to



A few Classification Solutions







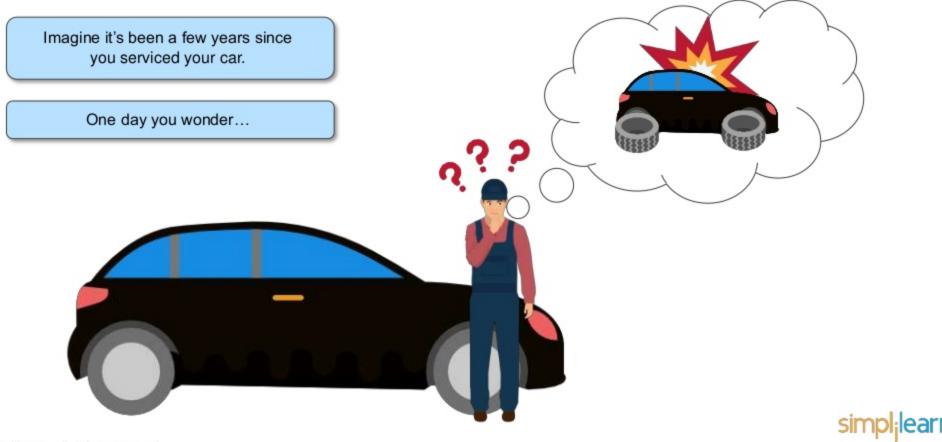
A dataset with one or more independent variables is used to determine binary output of the dependent variable

It helps determine what the given object is, based on its similarity to the objects it is compared to



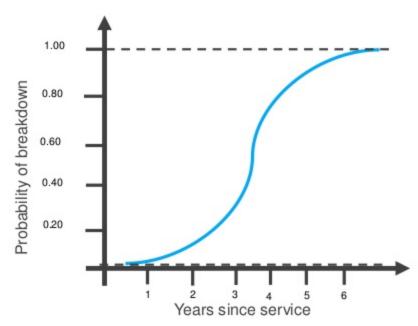


What is Logistic Regression?



What is Logistic Regression?



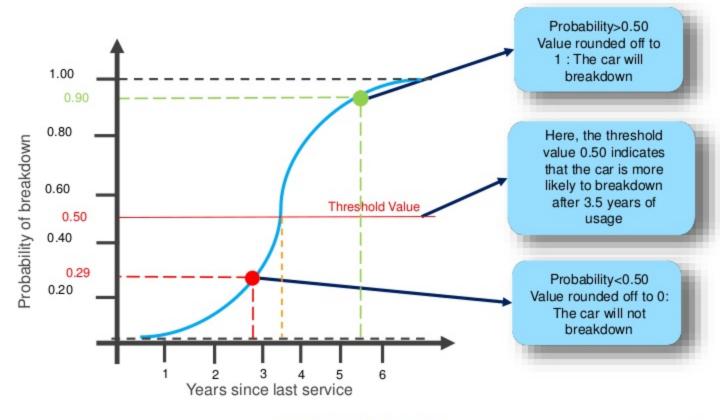


Regression model created based on other users' experience

It is a classification algorithm, used to predict binary outcomes for a given set of independent variables. The dependent variable's outcome is discrete.



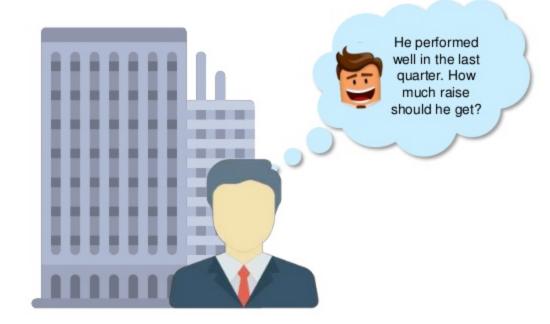
What is Logistic Regression?







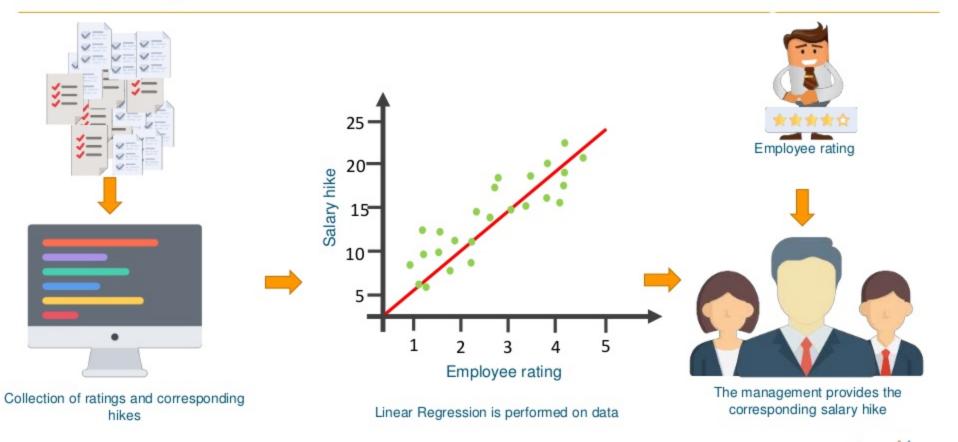
Linear Regression



It is a statistical method that helps find the relationship between an independent and dependent variable, both of which are continuous

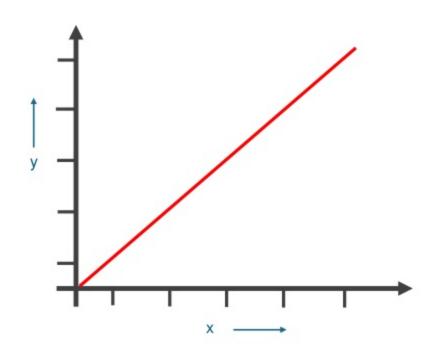


Linear Regression



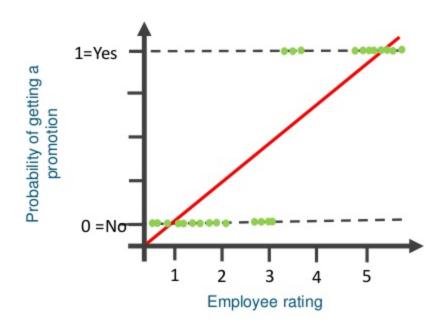


Here's the graph of how linear regression would be, for a given scenario





What if you wanted to know whether the employee would get a promotion or not based on their rating

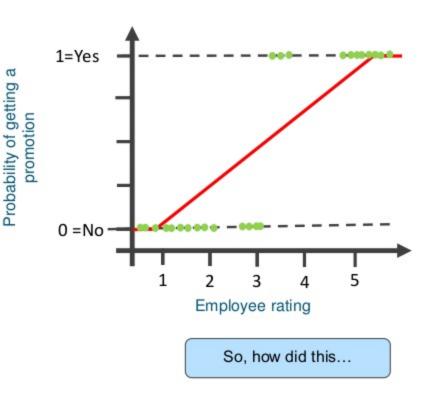


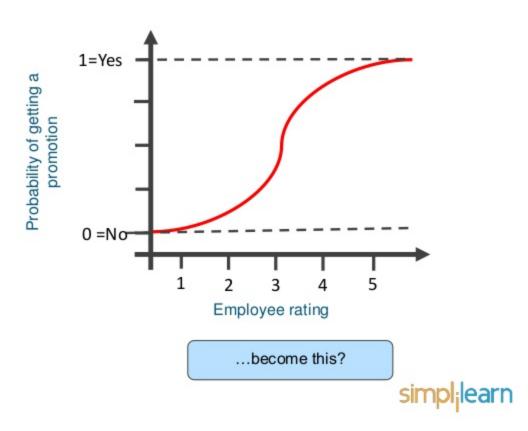


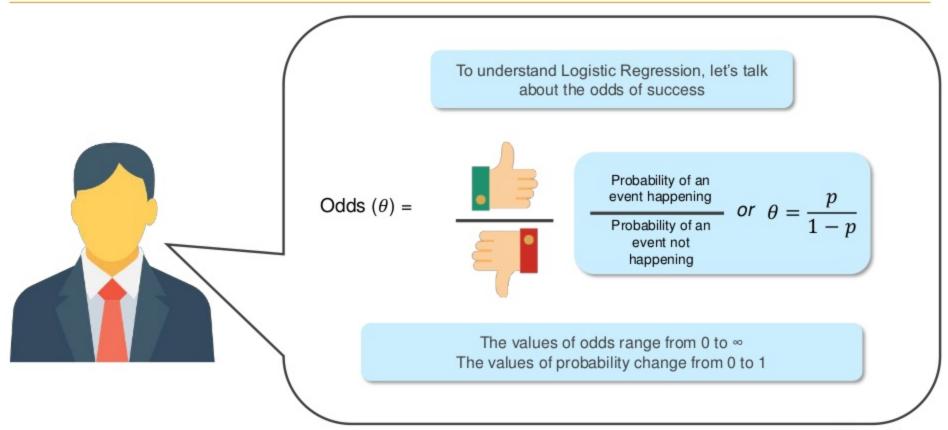
This graph would not be able to make such a prediction. So we clip the line at 0 and 1.

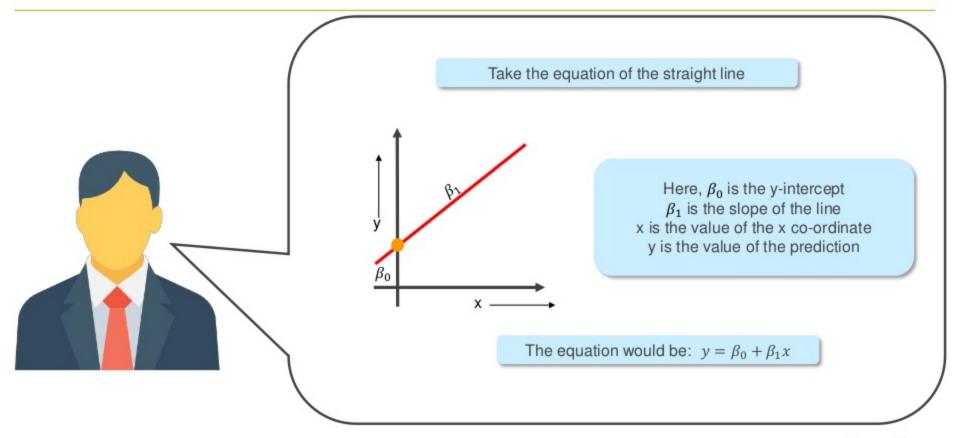


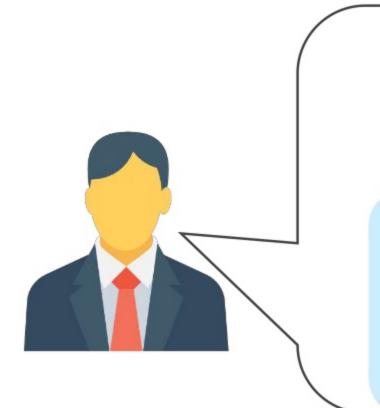












Now, we predict the odds of success

$$\log\left(\frac{p(x)}{1-P(x)}\right) = \beta_0 + \beta_1 x$$

Exponentiating both sides:
$$e^{ln}\left(\frac{p(x)}{1-p(x)}\right) = e^{\beta_0 + \beta_1 x}$$

$$\left(\frac{p(x)}{1 - p(x)}\right) = e^{\beta_0 + \beta_1 x}$$

Let Y =
$$e^{\beta_0 + \beta_1 x}$$

Then
$$\frac{p(x)}{1-p(x)} = Y$$

$$p(x) = Y(1 - p(x))$$

$$p(x) = Y - Y(p(x))$$

$$p(x) + Y(p(x)) = Y$$

$$p(x)(1+Y)=Y$$

$$p(x) = \frac{Y}{1 + Y}$$

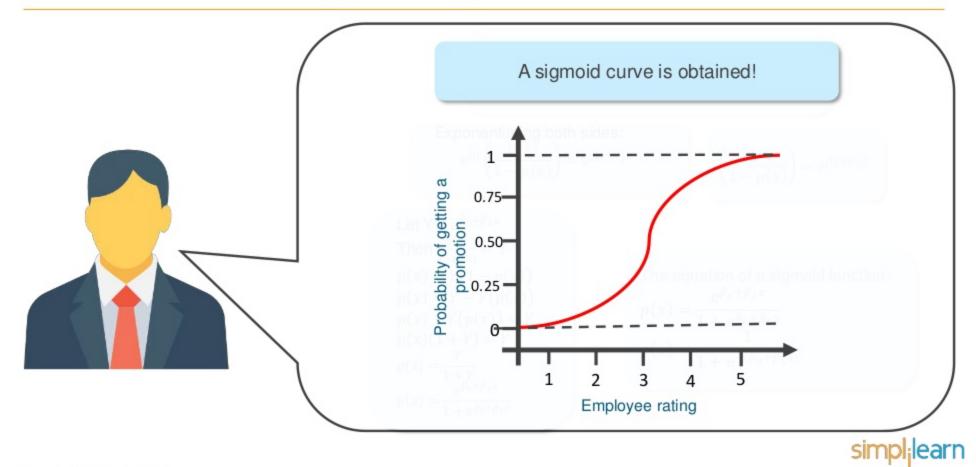
$$p(x) = \frac{e^{\beta_0 + \beta_1 x}}{1 + e^{\beta_0 + \beta_1 x}}$$

The equation of a sigmoid function:

$$p(x) = \frac{e^{\beta_0 + \beta_1 x}}{1 + e^{\beta_0 + \beta_1 x}}$$

$$p(x) = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x)}}$$







Linear Regression Used to solve Regression Problems

Linear Regression

 Used to solve Regression Problems

Logistic Regression

 Used to solve Classification Problems



Linear Regression

- Used to solve Regression Problems
- The response variables are continuous in nature

Logistic Regression

 Used to solve Classification Problems



Linear Regression

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- Used to solve Classification Problems
- The response variable is categorical in nature



Linear Regression

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- It helps estimate the dependent variable when there is a change in the independent variable.

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- It helps calculate the possibility of a particular event taking place.



Linear Regression

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- The response variables are continuous in nature
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- Is a straight line.

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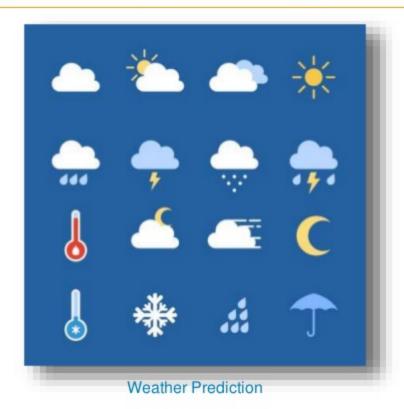
Logistic Regression

- Used to solve Classification Problems
- The response variable is categorical in nature
- It helps calculate the possibility of a particular event taking place.

An S-curve. (S = Sigmoid)



Logistic Regression Applications



Helps determine the kind of weather that can be expected



Logistic Regression Applications

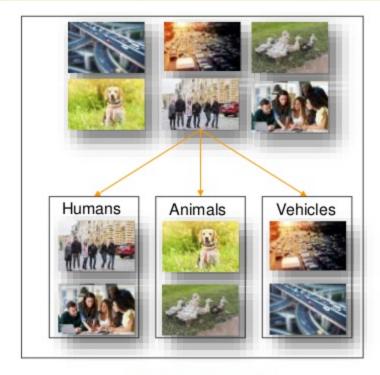
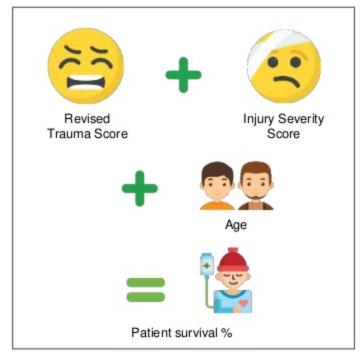


Image Categorization

Identifies the different components that are present in the image, and helps categorize them



Logistic Regression Applications



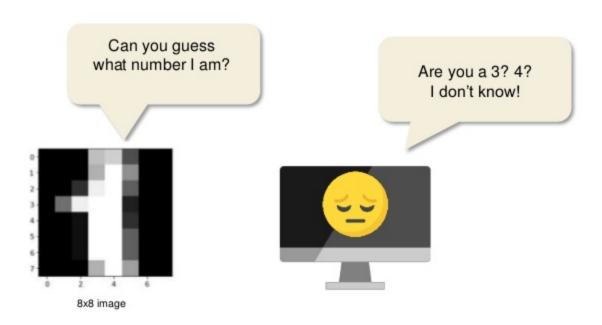
Healthcare (TRISS)

Determines the possibility of patient survival, taking age, ISS and RTS into consideration



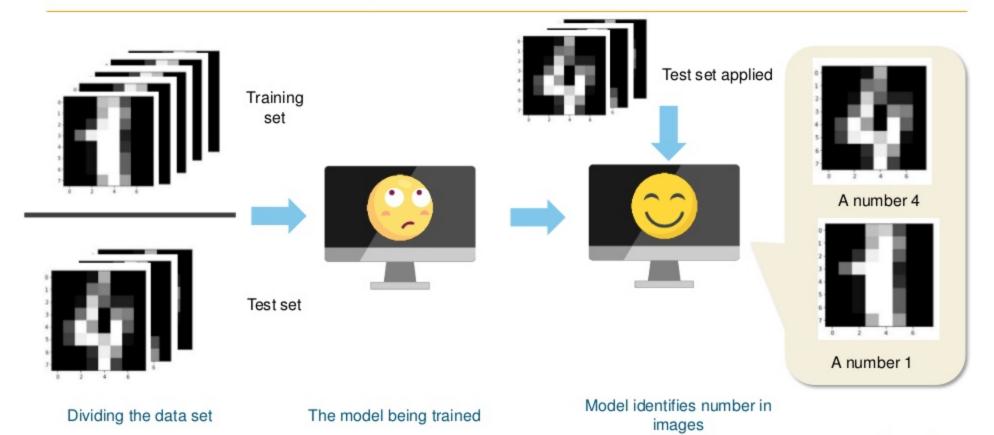


Use Case – Predicting numbers





Use Case – Predicting numbers





Importing libraries and their associated methods

```
from sklearn.datasets import load_digits
from sklearn.model_selection import train_test_split
import numpy as np

import matplotlib.pyplot as plt
import seaborn as sns
from sklearn import metrics
%matplotlib inline
digits = load_digits()
```

Determining the total number of images and labels

```
print("Image Data Shape", digits.data.shape)
print("Label Data Shape", digits.target.shape)

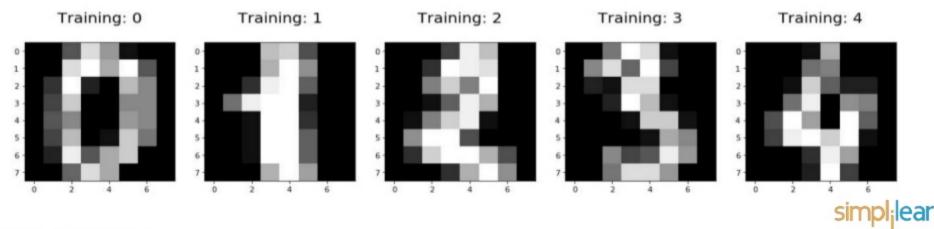
Image Data Shape (1797, 64)
Label Data Shape (1797,)
```



Displaying some of the images and labels

```
import numpy as np
import matplotlib.pyplot as plt

plt.figure(figsize=(20,4))
for index, (image, label) in enumerate(zip(digits.data[0:5], digits.target[0:5])):
   plt.subplot(1, 5, index + 1)
   plt.imshow(np.reshape(image, (8,8)), cmap=plt.cm.gray)
   plt.title('Training: %i\n' % label, fontsize = 20)
```



Dividing dataset into Training and Test set

```
from sklearn.model selection import train test split
x train, x test, y train, y test = train test split(digits.data, digits.target, test size=0.23, random state=2)
print(x_train.shape)
(1383, 64)
print(y train.shape)
(1383,)
print(x test.shape)
(414, 64)
print(y_test.shape)
(414,)
```



Import the Logistic Regression model

```
from sklearn.linear_model import LogisticRegression
```

Making an instance of the model and training it

Predicting the output of the first element of the test set

verbose=0, warm start=False)

```
print(logisticRegr.predict(x_test[0].reshape(1,-1)))
```

[4]

Predicting the output of the first 10 elements of the test set

```
logisticRegr.predict(x_test[0:10])
array([4, 0, 9, 1, 8, 7, 1, 5, 1, 6])
```



Predicting for the entire dataset

```
predictions = logisticRegr.predict(x_test)
```

Determining the accuracy of the model

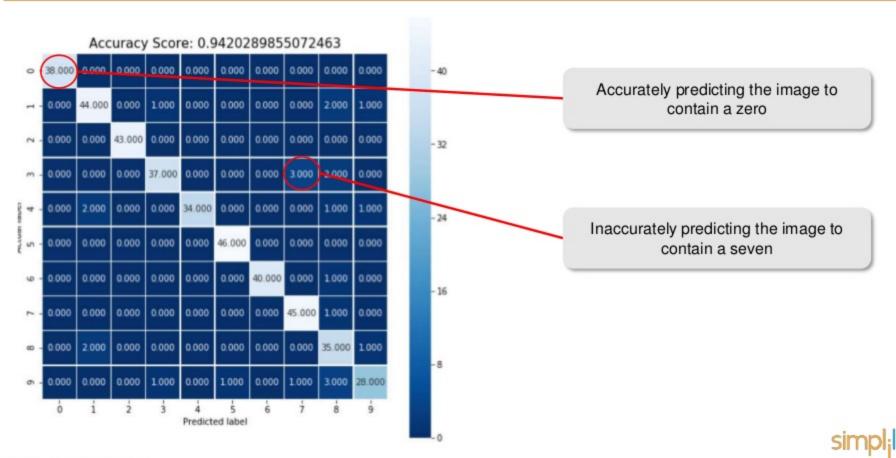
```
score = logisticRegr.score(x_test, y_test)
print(score)
```

0.9420289855072463

Representing the confusion matrix in a heat map

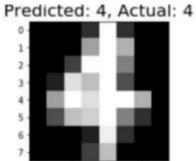
```
plt.figure(figsize=(9,9))
sns.heatmap(cm, annot=True, fmt=".3f", linewidths=.5, square = True, cmap = 'Blues_r');
plt.ylabel('Actual label');
plt.xlabel('Predicted label');
all_sample_title = 'Accuracy Score: {0}'.format(score)
plt.title(all_sample_title, size = 15);
```

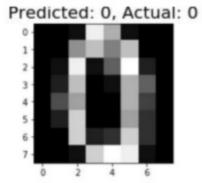


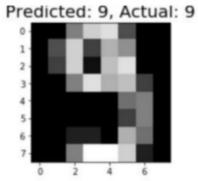


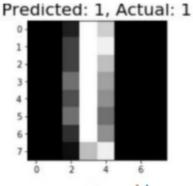
Presenting predictions and actual output

```
index = 0
misclassifiedIndex = []
for predict, actual in zip(predictions, y_test):
    if predict==actual:
        misclassifiedIndex.append(index)
    index +=1
plt.figure(figsize=(20,3))
for plotIndex, wrong in enumerate(misclassifiedIndex[0:4]):
    plt.subplot(1,4, plotIndex +1)
    plt.imshow(np.reshape(x_test[wrong], (8,8)),cmap=plt.cm.gray)
    plt.title("Predicted: {}, Actual: {}" .format(predictions[wrong], y_test[wrong]), fontsize=20)
```

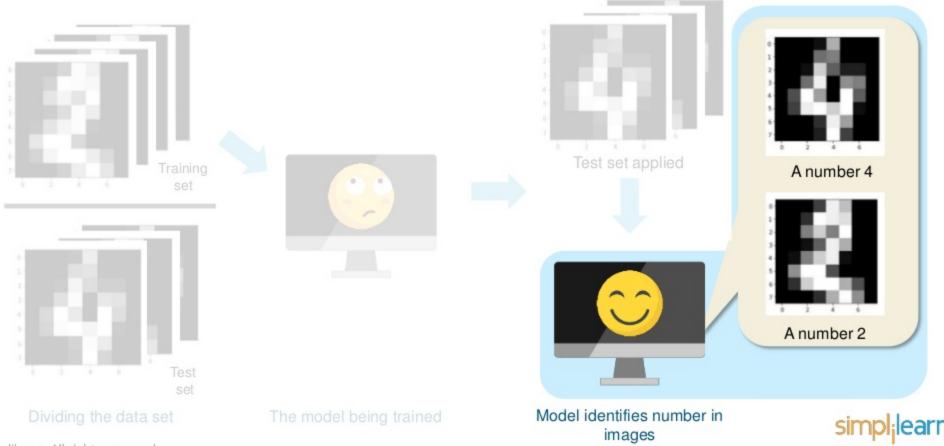








Use Case – Predicting numbers



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Here's What You've Learnt so Far

