Logistic Regression Project

In this application we will be working on dataset called exam scores. We will try to create a model that will predict whether or not the candiate admitted or not admitted based on two exam scores.

Import Libraries

Import a required libraries

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import classification_report
from sklearn.metrics import confusion_matrix
%matplotlib inline
```

Get the Data

Read in the examscores.csv file and set it to a data frame called examscores.

```
In [3]: df =pd.read_csv("examscores.csv")
```

Check the head of examscores

```
In [5]: df.head()
```

Out[5]:

	score1	score2	result
0	34.623660	78.024693	0
1	30.286711	43.894998	0
2	35.847409	72.902198	0
3	60.182599	86.308552	1
4	79.032736	75.344376	1

Use info and describe() on examscores

Logistic Regression

In []:

Now it's time to do a train test split, and train our model!

Split the data into training set and testing set using train_test_split

```
1. Define X with score1 and score2 and Y data with result
```

```
2. Use train_test_split() with X and Y
```

```
In [13]: X=df[["score1", "score2"]]
y=df["result"]
The [14]: X that a V that A that A that test online(Y v tost of sale)
```

```
In [14]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, random_state=42)
```

Train and fit a logistic regression model on the training set.

Predictions and Evaluations

Now predict values for the testing data.

```
In [16]: predictions = logmodel.predict(X_test)
```

Create a classification report and confusion matrix for the model.

```
In [17]: print(confusion_matrix(y_test, predictions))
    [[ 7    3]
       [ 0    23]]
```

In [18]: print(classification_report(y_test,predictions))

```
precision
                         recall f1-score
                                           support
         0
                 1.00
                           0.70
                                     0.82
                                                10
         1
                 0.88
                           1.00
                                     0.94
                                                23
avg / total
                 0.92
                           0.91
                                     0.90
                                                33
```