

Project : Restaurant Revenue Prediction



In this project, you are going to work on the "Restaurant revenue" dataset which contains 100 observations of restaurants with 8 attributes.

Problem Statement:

Create a machine learning model to predict revenue of the restaurant based on the features present in the dataset.

Dataset Attributes:

- 1. ID Restaurant ID
- Name Name of the Restaurant
- Franchise Restaurant has franchise or not
- 4. Category specific type of category provided by restaurant
- No_of_item Different types of items provided by restaurant
- 6. Order_Placed Order placed by customer to restaurant (in lacs)
- 7. Revenue Total amount of income generated by the restaurant

Tasks To Be Performed:

In this project, you have to predict the restaurant revenue based on the independent features using a machine learning algorithm of your own choice.

Import the libraries 🗸

```
In [1]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
```

Load the dataset <

In [2]: df=pd.read_csv("revenue_prediction.csv")
 df.head()

Out[2]:

	ld	Name	Franchise	Category	City	No_Of_Item	Order_Placed	Revenue
0	101	HungryHowie'sPizza	Yes	Mexican	Bengaluru	55	5.5	5953753
1	102	CharleysPhillySteaks	No	Varied Menu	Gurugram	72	6.8	7223131
2	103	Chuy's	Yes	Chicken	Pune	25	1.9	2555379
3	104	O'Charley's	Yes	Italian/Pizza	Mumbai	18	2.5	2175511
4	105	PolloTropical	Yes	Pizza	Noida	48	4.2	4816715

Examining missing values in a dataset 🗸



Drop irrelevant columns 🗸

```
In [4]: df=df.drop(columns=["Id","Name","Franchise","Category","City","No_Of_Item"])
df.head()
```

Out[4]:

	Order_Placed	Revenue		
0	5.5	5953753		
1	6.8	7223131		
2	1.9	2555379		
3	2.5	2175511		
4	4.2	4816715		

```
In [6]: df.shape 

/
```

Out[6]: (100, 2)

Create independent feature and dependent variable,

```
In [7]: x=df.iloc[:,:-1].values
y=df.iloc[:,-1].values
```

Split the dataset into Training set and Test set 🗸

```
In [8]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
```

```
In [9]: print(x_train) /
```

[[6.8] [1.9] [2.1] [4.8] [3.3] [3.1] [6.7] [6.3] [5.3] [2.4]

> 5.] 7.7] 7.2] 4.6] 3.3] 4.9]

```
print(x_test)
In [10]:
          [[5.7]
           [5.4]
           [3.8]
           [3.6]
           [4.]
           [3.7]
           [3.8]
           [3.5]
           [3.6]
           [5.5]
           [5.2]
           [2.9]
           [3.1]
           [2.2]
           [3.3]
           [4.2]
           [4.3]
           [1.7]
           [1.6]
           [1.9]]
         print(y_train) /
In [11]:
          6941173
                     2025297
                               2967425
                                        4952255
                                                  3452382
                                                                     6836483
                                                            3918918
                                                                              6412623
            4544227
                     2525375
                               3861370
                                        7865428
                                                  7495092
                                                            7705945
                                                                     3410878
                                                                              4517319
                     1756069
                               3903884 19696939
                                                  8630682
                                                            4100886
            1882131
                                                                     6491607
                                                                              1270499
            6135276
                     1521934
                                                            5107746
                               4066618
                                        2778621
                                                  5286212
                                                                     7904084
                                                                               2175511
            7513524
                     5461700
                               5152497
                                        3871344
                                                  5966635
                                                            2551252
                                                                     3028267
                                                                               2364478
                                                           4052733
                                                                     4250758
            2390534
                     4567678
                               4651866
                                        2732645
                                                  2156098
                                                                               3836721
            4350573
                     6694797
                               6313221
                                        4554237
                                                  3004429
                                                           7201784
                                                                     2344689
                                                                               3258837
           16549064
                     3261924
                               3753720
                                        3600467
                                                  1619683
                                                            4250553
                                                                     2740687
                                                                               2447890
            7223131
                     1763231
                                849870
                                        2555379
                                                  1999068
                                                           5906596
                                                                     2083447
                                                                               5500818
            3351383
                     3807496
                               5025735
                                        3164972
                                                  4136425
                                                           1099097
                                                                     6782425
                                                                               3273041]
```

```
In [12]: print(y_test)
         [5966193 5595267 3982767 3752885 4264176 3818055 4956086 3727364 3945135
          5953753 4937526 2408199 3347767 2371202 3426169 4816715 4590423 1904842
          2234634 2358644]
```

Training the Simple Linear Regression model on the Training set 🧹



```
In [13]: from sklearn.linear model import LinearRegression
         regressor=LinearRegression()
         regressor.fit(x train,y train)
```

Out[13]: LinearRegression LinearRegression()

Predicting the Test set results

1347673.99433668, 1713778.57302608])

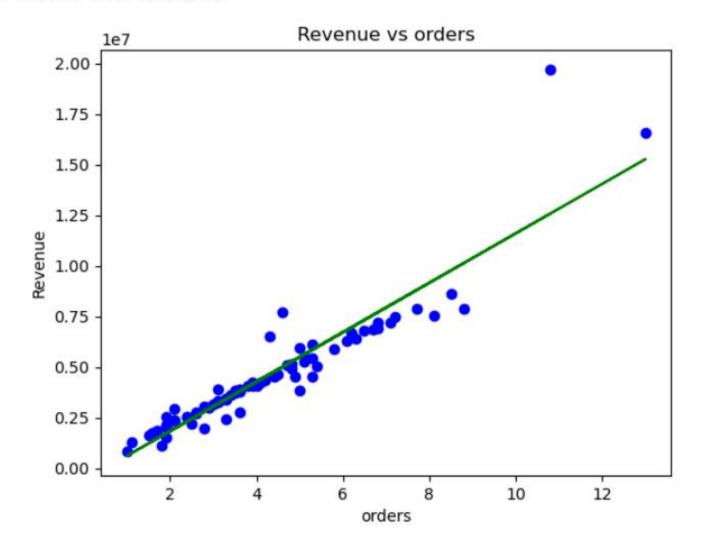
```
In [14]: y pred=regressor.predict(x test)
         y pred
Out[14]: array([6351103.23642516, 5984998.65773576, 4032440.90472562,
                3788371.18559936, 4276510.62385189, 3910406.04516249,
                4032440.90472562, 3666336.32603622, 3788371.18559936,
                6107033.5172989 , 5740928.93860949 , 2934127.16865742 ,
```

3178196.88778369, 2079883.15171548, 3422266.60690995, 4520580.34297816, 4642615.20254129, 1469708.85389981,

Visualising the Training set results

```
In [15]: plt.scatter(x_train,y_train,color="blue")
         plt.plot(x_train,regressor.predict(x_train),color="green")
         plt.title("Revenue vs orders")
         plt.xlabel("orders")
         plt.ylabel("Revenue")
```

Out[15]: Text(0, 0.5, 'Revenue')

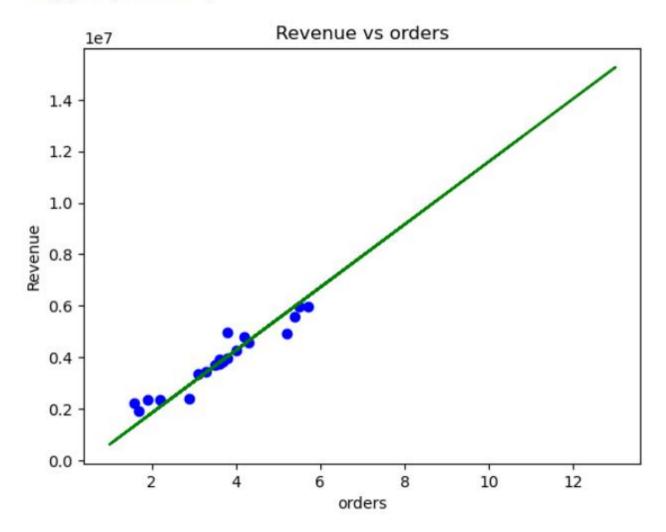


Visualising the Test set results



```
In [16]: plt.scatter(x_test,y_test,color="blue")
    plt.plot(x_train,regressor.predict(x_train),color="green")
    plt.title("Revenue vs orders")
    plt.xlabel("orders")
    plt.ylabel("Revenue")
```

Out[16]: Text(0, 0.5, 'Revenue')



Accuracy of the Model

In [17]: from sklearn.metrics import r2_score
 score=r2_score(y_pred,y_test)
 score

Out[17]: 0.9109612329066814