

LINEAR REGRESSION

Simple Linear Regression

Simple linear regression is a statistical machine learning method that allows us to summarize and study relationship between two continuous (quantative) variables. Linear regression attempts to model the relationshi between two variables by fitting a linear equation to obsered data. One variable is considered to br an explanatory variable and the other is considered to br a dependent variable.

Simple Linear regression model

Simple Linear Equation

$E(y) = \beta_0 + \beta_1 x$... (Pure Mathematical Notations)

Estimate of the Simple Linear Equation

$$\hat{y} = b_0 + b_1 x$$

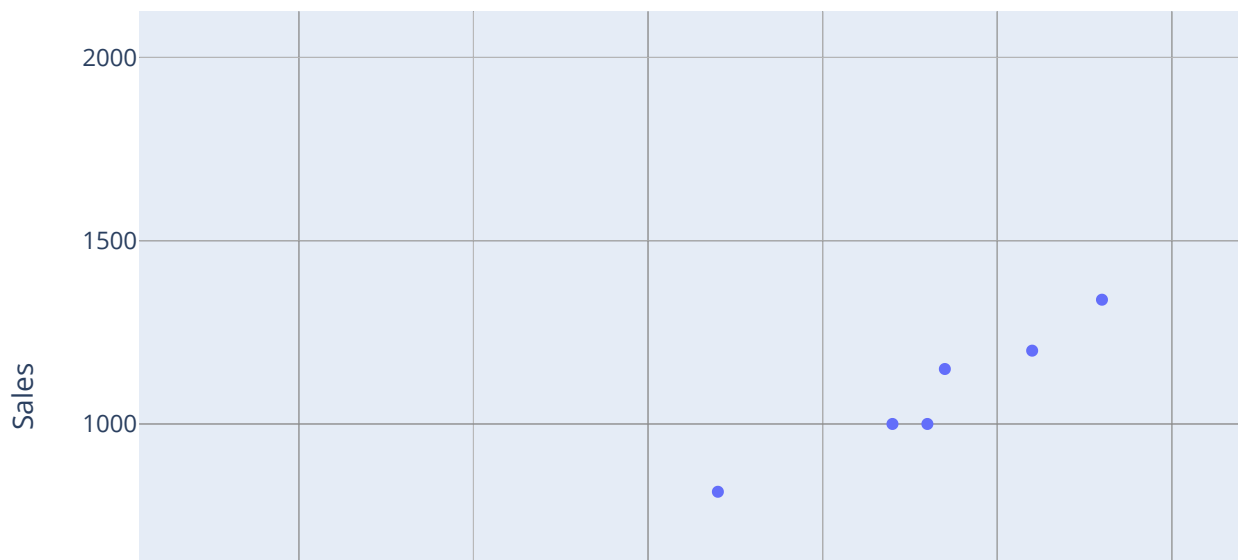
```
In [1]: import pandas as pd
import plotly.graph_objects as go

burger = pd.read_csv('E:\\Machine Learning\\Datasets\\Burger.csv')
burger.head()
```

Out[1]:

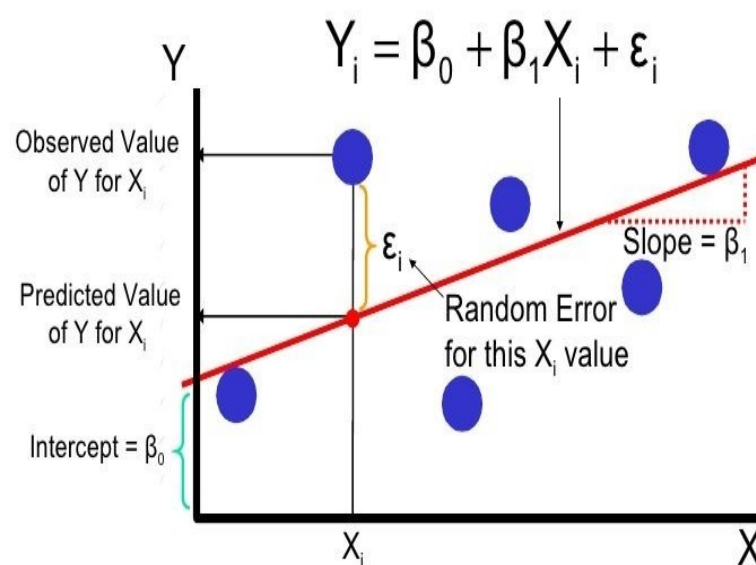
	Promote	Sales
0	23	554
1	56	1339
2	34	815
3	25	609
4	67	1600

```
In [2]: import plotly.express as px
fig = px.scatter(burger, x = 'Promote', y= 'Sales')
fig.show()
```



Simple Linear Regression Model

(continued)



STEP 1 : PREPROCESS THE DATA

- Import the libraries
- Import the dataset
- Read the Data
- Check for null values in columns
- Split the DataSet

```
In [3]: ## Import the Libraries
import pandas as pd
import numpy as np

from sklearn import linear_model
from sklearn import datasets
from sklearn.model_selection import train_test_split
```

```
In [4]: ## Import the dataset
data = datasets.load_boston()
df = pd.DataFrame(data.data, columns = data.feature_names)
df.head()
```

Out[4]:

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	B	LSTAT
0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1.0	296.0	15.3	396.90	4.98
1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2.0	242.0	17.8	396.90	9.14
2	0.02729	0.0	7.07	0.0	0.469	7.185	61.1	4.9671	2.0	242.0	17.8	392.83	4.03
3	0.03237	0.0	2.18	0.0	0.458	6.998	45.8	6.0622	3.0	222.0	18.7	394.63	2.94
4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3.0	222.0	18.7	396.90	5.33

```
In [5]: target = pd.DataFrame(data.target, columns=['MEDV'])
target
```

Out[5]:

	MEDV
0	24.0
1	21.6
2	34.7
3	33.4
4	36.2
...	...
501	22.4
502	20.6
503	23.9
504	22.0
505	11.9

506 rows × 1 columns

```
In [6]: # Check for null values in columns
print('NUM in RM : ',df['RM'].isnull().values.any())#chek if there are NULL in the
print('NUM in MEDV : ',target['MEDV'].isnull().values.any())
```

```
NUM in RM : False
NUM in MEDV : False
```

```
In [7]: X= pd.DataFrame(df['RM'])
y = pd.DataFrame(target['MEDV'])
```

```
In [8]: #Split the data...
#split the data in 80/20 proportion
X_train ,X_test, y_train, y_test = train_test_split(X,y,test_size =0.2, random_stat
```

```
In [9]: print('o X : ',X.shape)
print("X_train :",X_train.shape)
print("X_test : ",X_test.shape)
print('\no y : ',y.shape)
print("y_train :",y_train.shape)
print("y_test: ",y_test.shape)
```

```
o X : (506, 1)
X_train : (404, 1)
X_test : (102, 1)
```

```
o y : (506, 1)
y_train : (404, 1)
y_test: (102, 1)
```

STEP 2 : FITTING SIMPLE LINEAR REGRESSION MODEL TO THE TRAINING SET

To fit the dataset into the model we will use LinearRegression class from sklearn.linear_model library. Then we make object of lm from LinearRegression(). We will fit the regressor object into to our dataset using fit() method of LinearRegression.

```
In [10]: lm = linear_model.LinearRegression()
model = lm.fit(X_train,y_train)
print('coeficent',lm.coef_)
print('Intercept',lm.intercept_)
print('R square',lm.score(X_train,y_train))
```

```
coeficent [[9.37638431]]
Intercept [-36.47618963]
R square 0.4970800097843844
```

STEP 3 : PREDICTING THE RESULT

Now we will predict the observations from our type set. We will save the output in vector `y_pred`. To predict the result we predict method of `LinearRegression` class on the regressor we trained in the previous step.

```
In [11]: y_pred = lm.predict(X_test)
print('y_pred : ',y_pred[0:5]) ## Printing only first five y_pred elements
print(len(y_pred))
```

```
y_pred : [[22.90445223]
 [21.80741526]
 [23.2795076 ]
 [13.67809006]
 [21.95743741]]
102
```

STEP 4 : VISUALIZATION

The final Step is to visualize the results. We will use `plotly` library, to make scatter plot of our training set results and test set results to see how closer our model predicted the values..

```
In [12]: X_test
```

Out[12]:

	RM
329	6.333
371	6.216
219	6.373
403	5.349
78	6.232
...	...
56	6.383
455	6.525
60	5.741
213	6.375
108	6.474

102 rows × 1 columns

```
In [13]: y_predict = pd.DataFrame(data = y_pred,columns = ['y_pred']) #store the data of y_pred  
y_predict
```

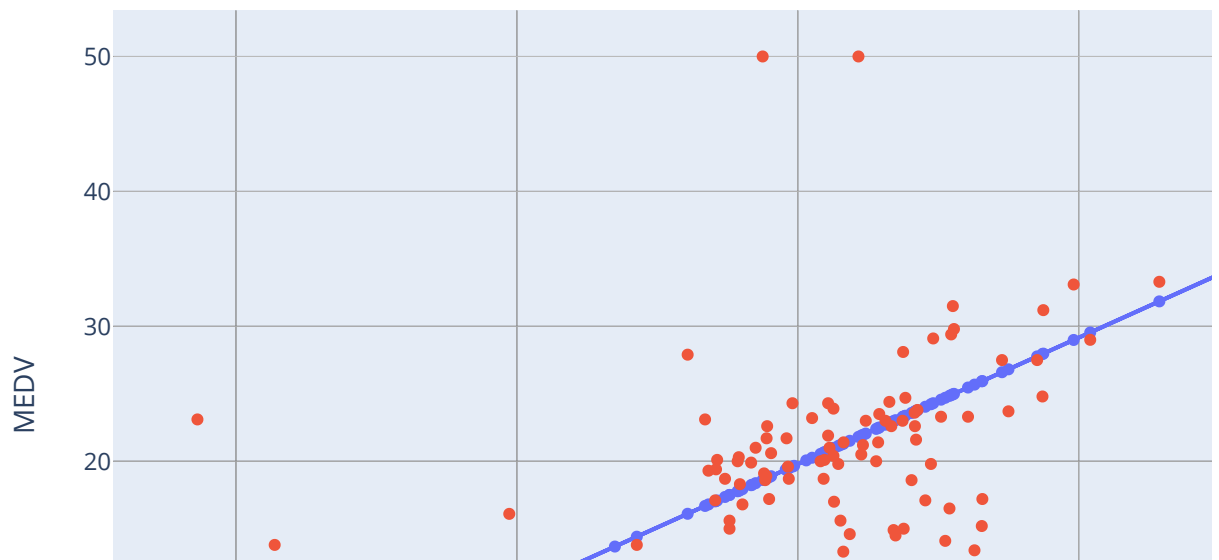
Out[13]:

	y_pred
0	22.904452
1	21.807415
2	23.279508
3	13.678090
4	21.957437
...	...
97	23.373271
98	24.704718
99	17.353633
100	23.298260
101	24.226522

102 rows × 1 columns

```
In [17]: import plotly.express as px
fig = px.scatter()
fig.add_scatter(x = X_test['RM'], y=y_predict['y_pred'],mode = 'lines+markers',name='PREDICTED')
fig.add_scatter(x = X_test['RM'], y=y_test['MEDV'],mode = 'markers',name = 'OBSERVED')
fig.update_layout(title='PREDICTED v/s OBSERVED',xaxis_title='RM', yaxis_title='MEDV')
fig.show()
```

PREDICTED v/s OBSERVED



In []: