

# LINEAR REGRESSION

```
In [2]: import warnings
import sys
if not sys.warnoptions:
    warnings.simplefilter("ignore")
```

```
In [10]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.stats import norm
```

```
In [24]: # Load the dataset
data=pd.read_excel("LR_Sample1.xlsx")
data
```

Out[24]:

	X	Y
0	1	2
1	2	4
2	3	6
3	4	8
4	5	10
5	6	12
6	7	14
7	8	16
8	9	18
9	10	20
10	11	22
11	12	24
12	13	26
13	14	28
14	15	30
15	16	32
16	17	34
17	18	36
18	19	38
19	20	40

```
In [18]: from sklearn.linear_model import LinearRegression
```

```
In [ ]: # If mean square error is approx. equal to zero, the model performance is good.
```

```
In [20]: from sklearn.metrics import mean_squared_error, r2_score
```

```
In [26]: X = data[['X']].values # X variable should 2D dataframe  
Y = data['Y'].values
```

```
In [28]: model = LinearRegression()
```

```
In [32]: 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 [34]: X_train
```

```
Out[34]: array([[ 9],  
                [ 6],  
                [12],  
                [ 4],  
                [19],  
                [17],  
                [14],  
                [ 3],  
                [10],  
                [20],  
                [ 5],  
                [13],  
                [ 8],  
                [11],  
                [15],  
                [ 7]], dtype=int64)
```

```
In [36]: X_test
```

```
Out[36]: array([[ 1],  
                [18],  
                [16],  
                [ 2]], dtype=int64)
```

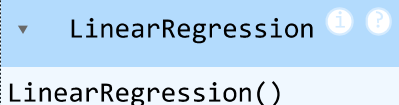
```
In [38]: y_train
```

```
Out[38]: array([18, 12, 24,  8, 38, 34, 28,  6, 20, 40, 10, 26, 16, 22, 30, 14],  
          dtype=int64)
```

```
In [40]: y_test
```

```
Out[40]: array([ 2, 36, 32,  4], dtype=int64)
```

```
In [42]: model.fit(X_train,y_train)
```

```
Out[42]:  LinearRegression()  
A Jupyter Notebook cell output showing a LinearRegression object. The object is represented by a blue box with a dropdown arrow, the text 'LinearRegression', and two small circular icons (one with an 'i' and one with a '?'). Below the box, the text 'LinearRegression()' is displayed.
```

```
In [46]: Y_predict = model.predict(X_test)
```

```
In [48]: Y_predict
```

```
Out[48]: array([ 2., 36., 32.,  4.])
```

```
In [52]: y_test
```

```
Out[52]: array([ 2, 36, 32,  4], dtype=int64)
```

## Model Performance

### Mean Square Error

```
In [71]: print("Mean Squared Error (MSE):", mean_squared_error(y_test, Y_predict))
```

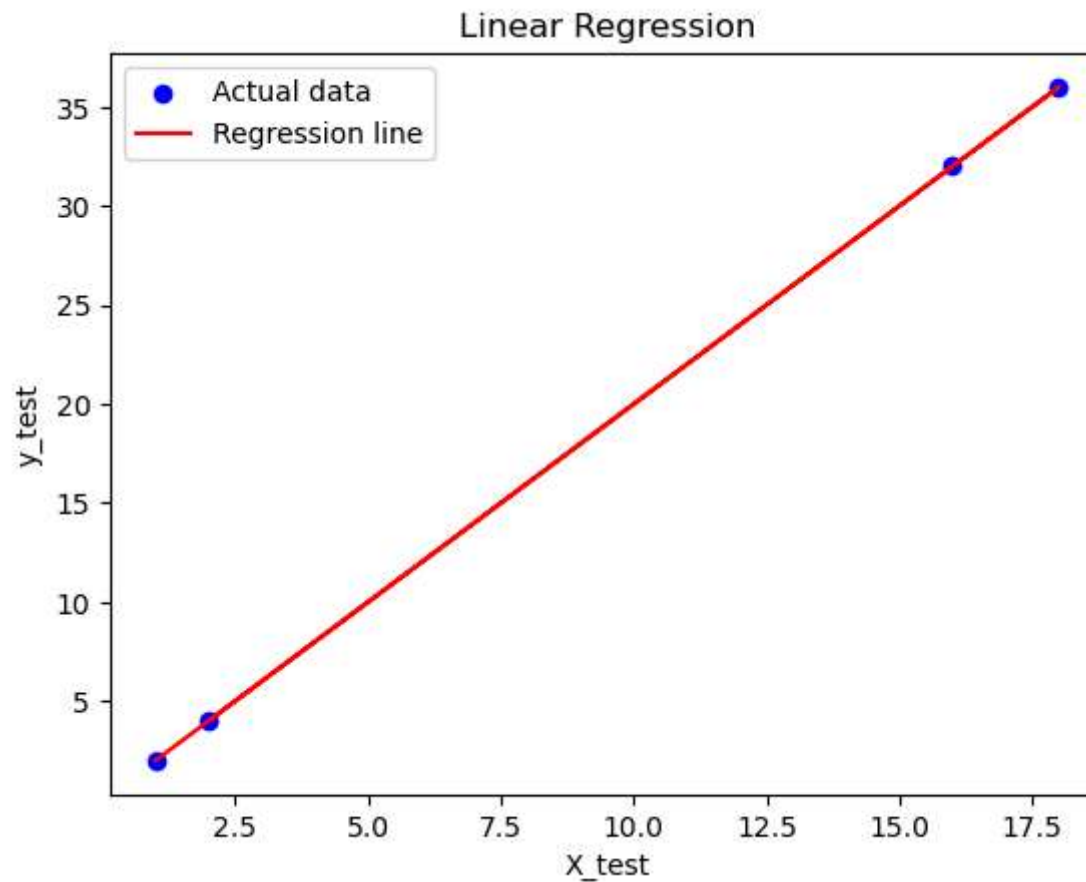
Mean Squared Error (MSE): 3.2540512340366737e-29

### R-Square Error

```
In [59]: print("R-Squared Error (RSE):", r2_score(y_test, Y_predict)) # If RSE = 1, the model is perfect (highly unlikely)
```

R-Squared Error (RSE): 1.0

```
In [73]: plt.scatter(X_test, y_test, color='blue', label='Actual data')
plt.plot(X_test, Y_predict, color='red', label='Regression line')
plt.title('Linear Regression')
plt.xlabel('X_test')
plt.ylabel('y_test')
plt.legend()
plt.show()
```



Mean Square Error is also referred as Cost Function.

The best learning rate (alpha) should be 0.001 to ensure best convergence.