

Types of machine Learning

- supervised learning
 - Regression
 - depends on continuous data
 1. +ve Linearity(Linear Regression)
 - Linear regression with one feature/one input
 - Linear regression with multiple features/multiple inputs
 2. -ve Linearity(Polynomial Regression)
 - Polynomial with one feature
 - Polynomial with multiple features
 - Classification
 - depends on categorical(0,1,yes,no,true,False,goog,bad)
 - KNN Algorithm - K-Nearest Neighbours
 - Logistic Regression
 - SVM - Support Vector Machine Regressor/Classifier
 - Decision Tree Regressor/Classifier
 - Random Forest Regressor/Classifier
- unsupervised learning
 - Clustering
 - K-Means Algorithm
 - Dimensionality Reduction PCA - Principal Component Analysis
- reinforcement Learning

Linear Regression

- Linear model is sum weighted predict data to target values
- linear regression with one feature formula
 - $Y = mx + c$
 - y is a Target value or output value
 - m is a Slope
 - x is a input variable
 - c is a Coefficient/Intercept

Slope Formula

$$(x - x_{\text{mean}}) * (y - y_{\text{mean}}) / (x - x_{\text{mean}})^2$$

Coefficient/Intercept Formula

$$c = y_{\text{mean}} - (m * x_{\text{mean}})$$

Machine Learning Steps

1.get or load the data 2.preprocessing the data 3.Define input and output data 4.applying model or algorithm 5.train the data 6.calculate the score

```
In [1]: 1  ## getting the libraries
        2  import numpy as np
        3  import pandas as pd
        4  import matplotlib.pyplot as plt
```

```
In [4]: 1  # 1. getting the data
        2  df = pd.read_csv("https://raw.githubusercontent.com/AP-State-Skill-Developme
```



```
In [5]: 1  df
```

...

```
In [6]: 1  df = pd.read_csv('Salary_Data.csv')
```

In [7]:

1	df
---	----

Out[7]:

	YearsExperience	Salary
0	1.1	39343.0
1	1.3	46205.0
2	1.5	37731.0
3	2.0	43525.0
4	2.2	39891.0
5	2.9	56642.0
6	3.0	60150.0
7	3.2	54445.0
8	3.2	64445.0
9	3.7	57189.0
10	3.9	63218.0
11	4.0	55794.0
12	4.0	56957.0
13	4.1	57081.0
14	4.5	61111.0
15	4.9	67938.0
16	5.1	66029.0
17	5.3	83088.0
18	5.9	81363.0
19	6.0	93940.0
20	6.8	91738.0
21	7.1	98273.0
22	7.9	101302.0
23	8.2	113812.0
24	8.7	109431.0
25	9.0	105582.0
26	9.5	116969.0
27	9.6	112635.0
28	10.3	122391.0
29	10.5	121872.0

```
In [8]: 1 # Data preprocessing
        2 df.shape
```

Out[8]: (30, 2)

```
In [9]: 1 df.isnull().sum()
```

Out[9]: YearsExperience 0
Salary 0
dtype: int64

```
In [10]: 1 df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 30 entries, 0 to 29
Data columns (total 2 columns):
#   Column          Non-Null Count  Dtype
---  -
0   YearsExperience  30 non-null    float64
1   Salary          30 non-null    float64
dtypes: float64(2)
memory usage: 608.0 bytes
```

```
In [11]: 1 df.describe()
```

Out[11]:

	YearsExperience	Salary
count	30.000000	30.000000
mean	5.313333	76003.000000
std	2.837888	27414.429785
min	1.100000	37731.000000
25%	3.200000	56720.750000
50%	4.700000	65237.000000
75%	7.700000	100544.750000
max	10.500000	122391.000000

```
In [13]: 1 # Define input and output data
        2 X = df[['YearsExperience']]# input data
        3 y = df['Salary']# output data
```

```
In [14]: 1 # applying the model
        2 from sklearn.linear_model import LinearRegression
        3
```

```
In [15]: 1 # create object
        2 model = LinearRegression()
```

```
In [17]: 1 # fit the data
        2 model.fit(X,y)
```

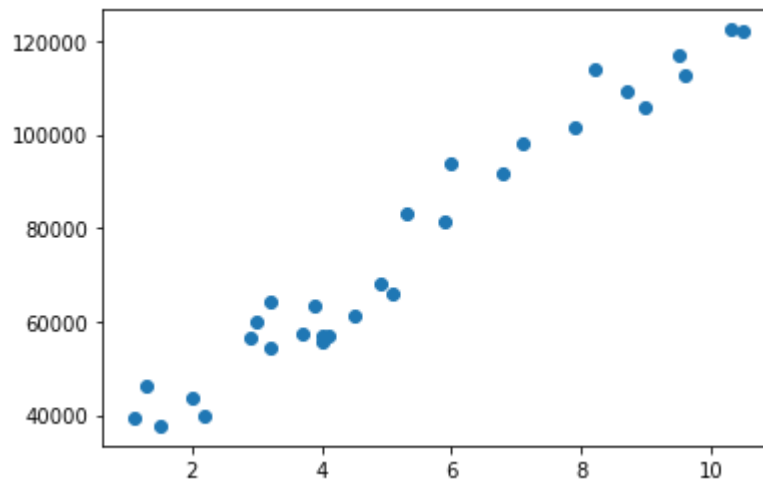
Out[17]: LinearRegression()

```
In [18]: 1 y_pred = model.predict(X)
```

```
In [19]: 1 y_pred
```

Out[19]: array([36187.15875227, 38077.15121656, 39967.14368085, 44692.12484158,
 46582.11730587, 53197.09093089, 54142.08716303, 56032.07962732,
 56032.07962732, 60757.06078805, 62647.05325234, 63592.04948449,
 63592.04948449, 64537.04571663, 68317.03064522, 72097.0155738 ,
 73987.00803809, 75877.00050238, 81546.97789525, 82491.9741274 ,
 90051.94398456, 92886.932681 , 100446.90253816, 103281.8912346 ,
 108006.87239533, 110841.86109176, 115566.84225249, 116511.83848464,
 123126.81210966, 125016.80457395])

```
In [20]: 1 # Visualize the data to relationship of input and output
        2 plt.scatter(df['YearsExperience'],df['Salary'])
        3 plt.show()
```



```
In [22]: 1 # identify the score
        2 model.score(X,y)*100
```

Out[22]: 95.69566641435085

```
In [23]: 1 plt.scatter(df['YearsExperience'],df['Salary'],c='g',label='Actual values')
2         plt.plot(X,y_pred,c='r',label = 'predicted values')
3         plt.title('yearsExperience vs Salary')
4         plt.xlabel('yearsExperience')
5         plt.ylabel('Salary')
6         plt.legend()
7         plt.show()
```



Predict the data jio subscribers

```
In [24]: 1 df = pd.read_csv("https://raw.githubusercontent.com/AP-State-Skill-Developme")
```

In [25]:

1 df

Out[25]:

	Revenue	JioSubscribers
0	8136	160.1
1	8421	186.6
2	10023	215.3
3	11416	252.3
4	12893	280.1
5	14328	306.7
6	15741	331.3
7	16534	355.2
8	17555	370.0
9	18632	387.5

In [26]:

1 df.isnull().sum()

Out[26]:

```
Revenue      0
JioSubscribers  0
dtype: int64
```

In [27]:

1 df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10 entries, 0 to 9
Data columns (total 2 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Revenue         10 non-null    int64
1   JioSubscribers  10 non-null    float64
dtypes: float64(1), int64(1)
memory usage: 288.0 bytes
```

In [28]:

```
1 # define input and output
2 X = df[['Revenue']]
3
```

In [29]:

1 y= df['JioSubscribers']

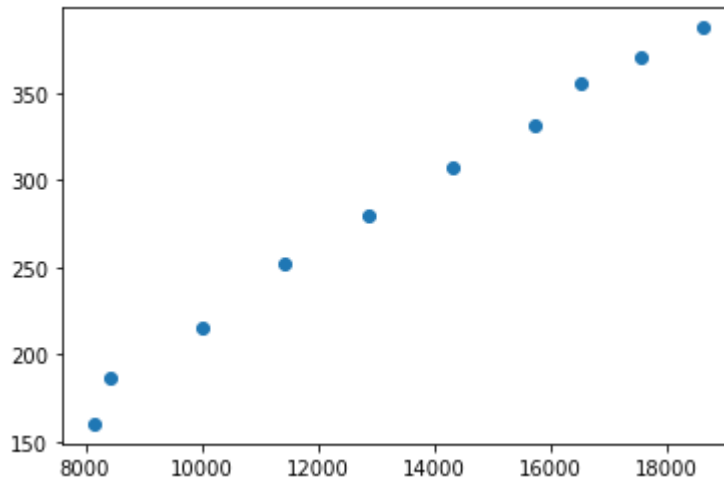
In [30]:

```
1 from sklearn.linear_model import LinearRegression
2 model = LinearRegression()
```

```
In [31]: 1 model.fit(X,y)
          2
          3
```

Out[31]: LinearRegression()

```
In [32]: 1 plt.scatter(df['Revenue'],df['JioSubscribers'])
          2 plt.show()
```



```
In [33]: 1 model.score(X,y)
```

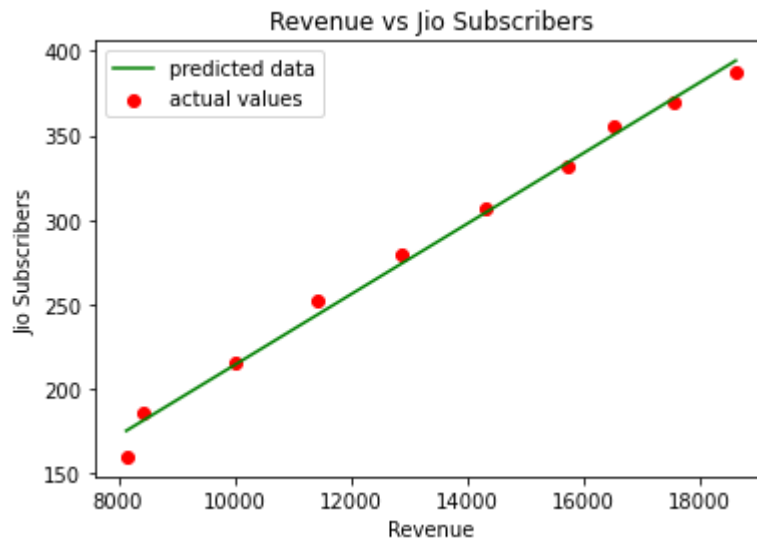
Out[33]: 0.9920788182785107

```
In [34]: 1 y_pred =model.predict(X)
```

```
In [35]: 1 y_pred
```

Out[35]: array([175.39851552, 181.34220196, 214.75197625, 243.80304716,
274.60594144, 304.53292403, 334.00109574, 350.53914257,
371.83213855, 394.29301678])


```
In [36]: 1 plt.scatter(df['Revenue'],df['JioSubscribers'],c='r',label = 'actual values'
2          plt.plot(X,y_pred,c='g',label = 'predicted data')
3          plt.title('Revenue vs Jio Subscribers')
4          plt.xlabel('Revenue')
5          plt.ylabel('Jio Subscribers')
6          plt.legend()
7          plt.show()
8
9
```



Linear Regression with Multiple variables

- input = more than one feature
- output = single target
- $y = ax^2 + bx + c$ (degree = 2)
- $y = ax^3 + bx^2 + cx + 1$ (degree = 3)

```
In [37]: 1 # 1.read the data and get the data
2          df = pd.read_csv("https://raw.githubusercontent.com/AP-State-Skill-Developme
3
```

In [38]: 1 df

Out[38]:

	Gender	Age Range	Head Size(cm^3)	Brain Weight(grams)
0	1	1	4512	1530
1	1	1	3738	1297
2	1	1	4261	1335
3	1	1	3777	1282
4	1	1	4177	1590
...
232	2	2	3214	1110
233	2	2	3394	1215
234	2	2	3233	1104
235	2	2	3352	1170
236	2	2	3391	1120

237 rows × 4 columns

In [39]: 1 df.head()

Out[39]:

	Gender	Age Range	Head Size(cm^3)	Brain Weight(grams)
0	1	1	4512	1530
1	1	1	3738	1297
2	1	1	4261	1335
3	1	1	3777	1282
4	1	1	4177	1590

In [40]: 1 df.columns

Out[40]: Index(['Gender', 'Age Range', 'Head Size(cm^3)', 'Brain Weight(grams)'], dtype='object')

In [41]: 1 df.Gender.unique()

Out[41]: array([1, 2], dtype=int64)

In [42]: 1 df['Gender'].value_counts()

Out[42]: 1 134
2 103
Name: Gender, dtype: int64

In [43]:

```
1 df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 237 entries, 0 to 236
Data columns (total 4 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Gender                237 non-null   int64
1   Age Range             237 non-null   int64
2   Head Size(cm^3)       237 non-null   int64
3   Brain Weight(grams)   237 non-null   int64
dtypes: int64(4)
memory usage: 7.5 KB
```

In [44]:

```
1 df.describe()
```

Out[44]:

	Gender	Age Range	Head Size(cm^3)	Brain Weight(grams)
count	237.000000	237.000000	237.000000	237.000000
mean	1.434599	1.535865	3633.991561	1282.873418
std	0.496753	0.499768	365.261422	120.340446
min	1.000000	1.000000	2720.000000	955.000000
25%	1.000000	1.000000	3389.000000	1207.000000
50%	1.000000	2.000000	3614.000000	1280.000000
75%	2.000000	2.000000	3876.000000	1350.000000
max	2.000000	2.000000	4747.000000	1635.000000

In [45]:

```
1 df.isnull().sum()
```

Out[45]:

Gender	0
Age Range	0
Head Size(cm^3)	0
Brain Weight(grams)	0

dtype: int64

In [79]:

```
1 # separate input and output
2 inputdata = df[['Head Size(cm^3)']]
3 outputdata = df['Brain Weight(grams)']
4 len(inputdata)
5 len(outputdata)
```

Out[79]: 237

In [80]:

```
1 inputdata.columns
```

Out[80]: Index(['Head Size(cm^3)'], dtype='object')

In [81]:

```
1 # applying the model
2 from sklearn.linear_model import LinearRegression
```

```
In [82]: 1 linear = LinearRegression()
```

```
In [83]: 1 linear.fit(inputdata,outputdata)
```

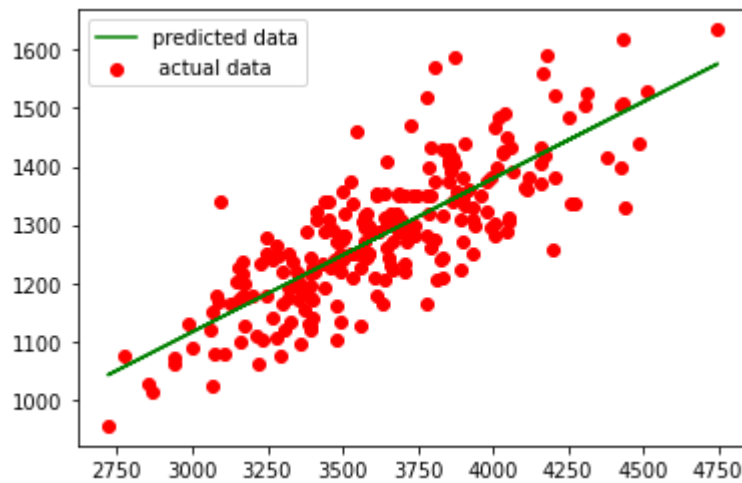
```
Out[83]: LinearRegression()
```

```
In [84]: 1 pred = linear.predict(inputdata)
```

```
In [85]: 1 linear.score(inputdata,outputdata)
```

```
Out[85]: 0.639311719957
```

```
In [87]: 1 plt.scatter(inputdata,outputdata,label= ' actual data',c='red')  
2 plt.plot(inputdata,pred,c='g',label = 'predicted data')  
3 plt.legend()  
4 plt.show()
```



```
In [ ]: 1
```