

Employee Salary Prediction using Linear Regression

```
In [2]: import pandas as pd
import numpy as np
import seaborn as sb
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
%matplotlib inline
```

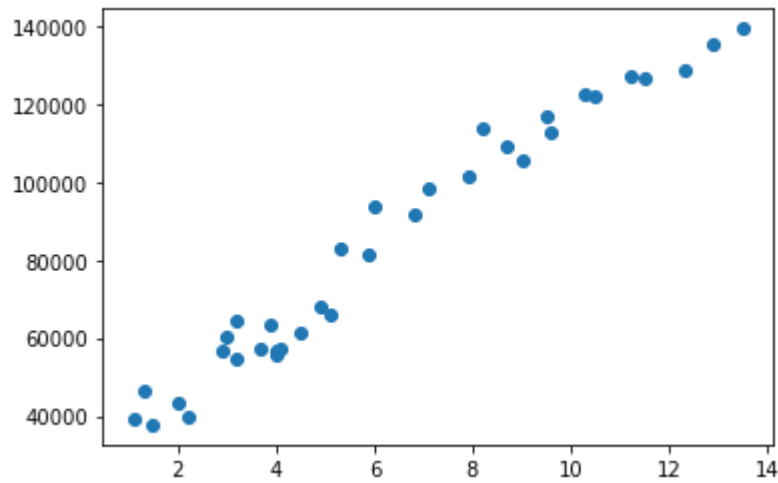
```
In [4]: #reading the data
data = pd.read_csv('Salary.csv')
```

```
In [52]: data.head()
```

	YearsExperience	Salary
0	1.1	39343
1	1.3	46205
2	1.5	37731
3	2.0	43525
4	2.2	39891

```
In [13]: #visualising the data inorder to find out about the trends in it
plt.scatter(data['YearsExperience'],data['Salary'])
```

Out[13]: <matplotlib.collections.PathCollection at 0x1907a3230d0>



```
In [14]: #storing the experience values and the salaries in particular variables
X =data.iloc[:, :-1].values
y=data.iloc[:, -1].values
```

```
In [15]: X
```

```
Out[15]: array([[ 1.1],
 [ 1.3],
 [ 1.5],
 [ 2. ],
 [ 2.2],
 [ 2.9],
 [ 3. ],
 [ 3.2],
 [ 3.2],
 [ 3.7],
 [ 3.9],
 [ 4. ],
 [ 4. ],
 [ 4.1],
 [ 4.5],
 [ 4.9],
 [ 5.1],
 [ 5.3],
 [ 5.9],
 [ 6. ],
 [ 6.8],
 [ 7.1],
 [ 7.9],
 [ 8.2],
 [ 8.7],
 [ 9. ],
 [ 9.5],
 [ 9.6],
 [10.3],
 [10.5],
 [11.2],
 [11.5],
 [12.3],
 [12.9],
 [13.5]])
```

```
In [16]: y
```

```
Out[16]: array([ 39343,  46205,  37731,  43525,  39891,  56642,  60150,  54445,
        64445,  57189,  63218,  55794,  56957,  57081,  61111,  67938,
        66029,  83088,  81363,  93940,  91738,  98273, 101302, 113812,
       109431, 105582, 116969, 112635, 122391, 121872, 127345, 126756,
       128765, 135675, 139465], dtype=int64)
```

```
In [17]: #using sklearn's train_test_split in order to divide the data into training and testing set and derive insights
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0)
```

```
In [18]: from sklearn.linear_model import LinearRegression
#applying the linear regression method in the data in order to find out the optimal fit
linReg = LinearRegression()
linReg.fit(X_train,y_train)
```

Out[18]: LinearRegression()

```
In [19]: linReg
```

Out[19]: LinearRegression()

```
In [20]: linAns =linReg.predict(X_test)
```

```
In [21]: linAns
```

```
Out[21]: array([[120197.8256403 ,  88644.21802942,  74146.61453254, 118492.2252289 ,
        98025.02029212,  72441.01412114,  63913.01206415,  43445.80712736,
        64765.81226984, 112522.623789 , 107405.82255481])
```

```
In [22]: X_test
```

```
Out[22]: array([[10.5],
 [ 6.8],
 [ 5.1],
 [10.3],
 [ 7.9],
 [ 4.9],
 [ 3.9],
 [ 1.5],
 [ 4. ],
 [ 9.6],
 [ 9. ]])
```

```
In [23]: linAns
```

```
Out[23]: array([[120197.8256403 ,  88644.21802942,  74146.61453254, 118492.2252289 ,
        98025.02029212,  72441.01412114,  63913.01206415,  43445.80712736,
        64765.81226984, 112522.623789 , 107405.82255481])
```

```
In [24]: y_test
```

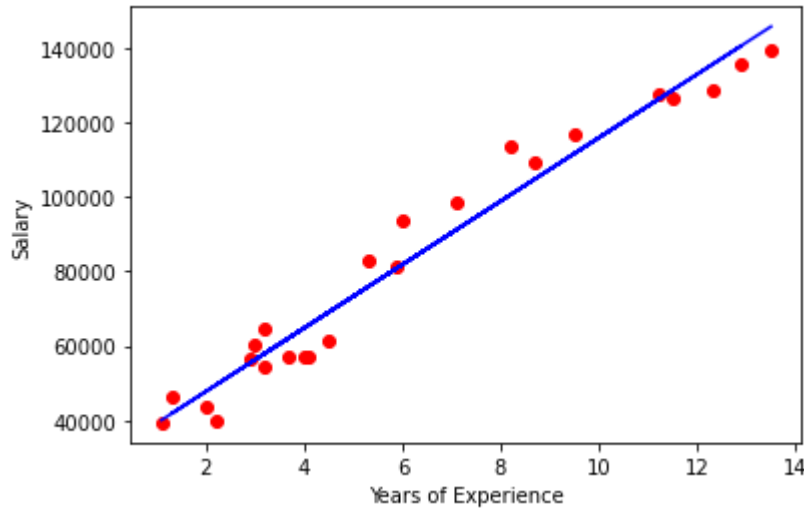
```
Out[24]: array([121872,  91738,  66029, 122391, 101302,  67938,  63218,  37731,
        55794, 112635, 105582], dtype=int64)
```

```
In [25]: #Finding the difference between the predicted salary and the actual salary
#the values which have the smallest positive value will be closest to the actual output
difference = y_test - linAns
```

```
In [26]: difference
```

```
Out[26]: array([ 1674.1743597 ,  3093.78197058, -8117.61453254,  3898.7747711 ,
        3276.97970788, -4503.01412114,  -695.01206415, -5714.80712736,
       -8971.81226984,  112.376211 , -1823.82255481])
```

```
In [28]: #Data visualization with training dataset
plt.scatter(X_train, y_train, color = 'red')
plt.plot(X_train, linReg.predict(X_train), color = 'blue')
plt.xlabel('Years of Experience')
plt.ylabel('Salary')
plt.show()
```



```
In [29]: #data visualization with the test data
plt.scatter(X_test,y_test,color='green')
plt.plot(X_train,linReg.predict(X_train),color='yellow')
plt.xlabel('Years of Experience')
plt.ylabel('Salary')
plt.show()
```

