

```
In [1]: # Working upon Salary Data set
```

```
In [5]: import numpy as np
import pandas as pd
```

```
In [9]: # importing the salary data set
salary_df = pd.read_csv(r'F:\Prthon Programming\Udemy\Part 2 - Regression\Section 4 - Simple Linear Regression\Salary_Data.csv')
```

```
In [10]: salary_df.head() # by default first 5 obv it shows
```

Out[10]:

| | YearsExperience | Salary |
|---|-----------------|--------|
| 0 | 1 | 39343 |
| 1 | 1 | 46205 |
| 2 | 2 | 37731 |
| 3 | 2 | 43525 |
| 4 | 2 | 39891 |

```
In [11]: salary_df.tail() # by default last 5 obv it shows
```

Out[11]:

| | YearsExperience | Salary |
|----|-----------------|--------|
| 25 | 9 | 105582 |
| 26 | 10 | 116969 |
| 27 | 10 | 112635 |
| 28 | 10 | 122391 |
| 29 | 11 | 121872 |

```
In [12]: salary_df.shape # try to see number of obv + number of columns
```

Out[12]: (30, 2)

```
In [13]: salary_df.isnull().sum() # checking if there are any missing values in data set
```

```
Out[13]: YearsExperience    0
Salary                    0
dtype: int64
```

```
In [19]: X = salary_df.values[:, :-1] # -1 means that i m not taking the salary column i
n (X variable) is IV
Y = salary_df.values[:, -1] # (Y variable) salary is my dependent variable
```

```
In [20]: print(X) # now my independent variable is ready
```

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```
In [29]: print(Y) # now my dependent variable is ready
```

```
[ 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]
```

```
In [41]: #from sklearn.cross_validation import train_test_split
from sklearn.model_selection import train_test_split
```

```
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 1/3, random_state = 0)
# I have split the data set in 80:20 ratio with the help of model_selection
```

In [42]: `print(X_train)`

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```

In [43]: `print(Y_test)`

```
[ 37731 122391  57081  63218 116969 109431 112635  55794  83088 101302]
```

In []: `#=====END OF THE DATA PREPROCESSING=====`

In [44]: `from sklearn.linear_model import LinearRegression # taken class as LinearRegression`

```
regressor = LinearRegression()
regressor.fit(X_train, Y_train) # fitting X_train and Y_train by using fit function
```

Out[44]: `LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False)`

In []: `#=====WE HAVE TRAIN THE MACHINE=====`

In [45]: `Y_pred = regressor.predict(X_test) # predicting test set results`

In [48]: `print(Y_pred)`

```
[ 46517.38475499 117804.99274047  64339.28675136  64339.28675136
 117804.99274047 108894.04174229 117804.99274047  64339.28675136
 73250.23774955  99983.0907441 ]
```

In [47]: `import matplotlib.pyplot as plt # data visualisation`

```
In [50]: # visualisation training set results
plt.scatter(X_train, Y_train, Color = 'red') # data points
plt.plot(X_train, regressor.predict(X_train), color = 'blue') # slope line
plt.title('Salary VS YearsExperience(training set)')
plt.xlabel('YearsExperience')
plt.ylabel('Salary')
plt.show
# real salary with predicted salary
```

Out[50]: <function matplotlib.pyplot.show(*args, **kw)>



```
In [51]: # visualisation testing set results
plt.scatter(X_test, Y_test, Color = 'red')
plt.plot(X_train, regressor.predict(X_train), color = 'blue')
plt.title('Salary VS YearsExperience(testing set)')
plt.xlabel('YearsExperience')
plt.ylabel('Salary')
plt.show
```

Out[51]: <function matplotlib.pyplot.show(*args, **kw)>



```
In [ ]: #=====END WITH A GOOD RESULT=====
=====
```