# **Supervised Machine Learning (Regression)**

#### SALARY AND EXPERINCE DATA SET

#### **IMPORTING LIBRARIES**

```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import matplotlib.pyplot as plt # MATLAB-like way of plotting
```

#### sklearn package for machine learning in python

```
In [3]: from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LinearRegression
    from sklearn.metrics import mean_squared_error, r2_score
In [23]: df = pd.read_csv(r"C:\Users\owner\Downloads\salary_data (1).csv")
```

#### **EDA**

```
In [24]: df.head()
```

```
Out[24]:
            YearsExperience Salary
                      1.1 39343.0
         0
                      1.3 46205.0
         1
                      1.5 37731.0
         2
         3
                      2.0 43525.0
                      2.2 39891.0
         4
         df.shape
In [25]:
         (30, 2)
Out[25]:
In [26]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 30 entries, 0 to 29
         Data columns (total 2 columns):
          # Column
                               Non-Null Count Dtype
            YearsExperience 30 non-null
                                               float64
          1 Salary
                               30 non-null
                                               float64
         dtypes: float64(2)
         memory usage: 612.0 bytes
         df.columns
In [27]:
         Index(['YearsExperience', 'Salary'], dtype='object')
Out[27]:
```

In [28]: df.describe()

```
Out[28]:
                 YearsExperience
                                         Salary
                       30.000000
                                      30.000000
           count
           mean
                        5.313333
                                   76003.000000
             std
                        2.837888
                                   27414.429785
                        1.100000
                                   37731.000000
            min
            25%
                         3.200000
                                   56720.750000
            50%
                        4.700000
                                   65237.000000
            75%
                        7.700000 100544.750000
                       10.500000 122391.000000
            max
```

In [29]:

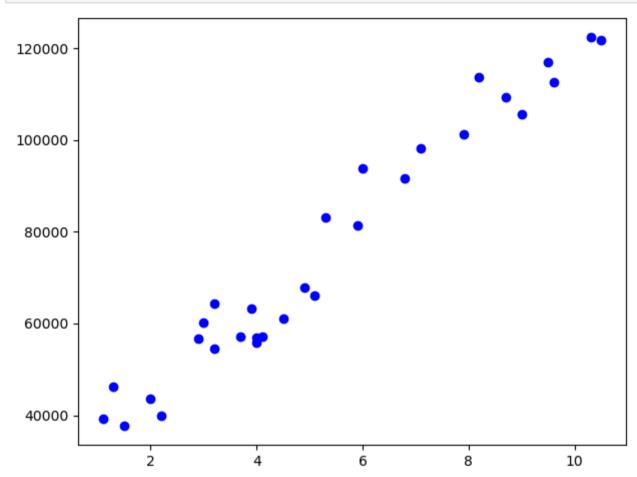
```
df.corr()
                         YearsExperience
Out[29]:
                                          Salary
          YearsExperience
                               1.000000 0.978242
                               0.978242 1.000000
                  Salary
          df.corr(),'\n'
In [32]:
                            YearsExperience
                                                Salary
Out[32]:
           YearsExperience
                                              0.978242
                                    1.000000
                                    0.978242 1.000000,
           Salary
```

## LINEAR REGRESSION

'\n')

```
In [34]: X = df.iloc[:, [0]].values # inputs YearsExperience
         y = df.iloc[:, 1].values # outputs Salary
In [35]: # visualise initial data set
         fig1, ax1 = plt.subplots()
         ax1.scatter(X, y, color = 'blue')
```

```
fig1.tight_layout()
fig1.savefig('LR_initial_plot.png')
```

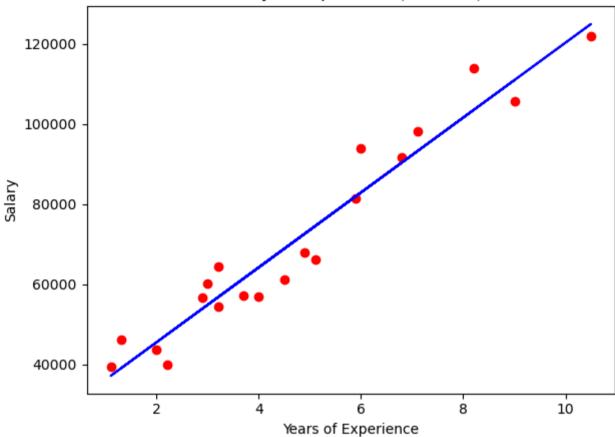


# SPLITTING DATASET INTO TRAIN AND TEST

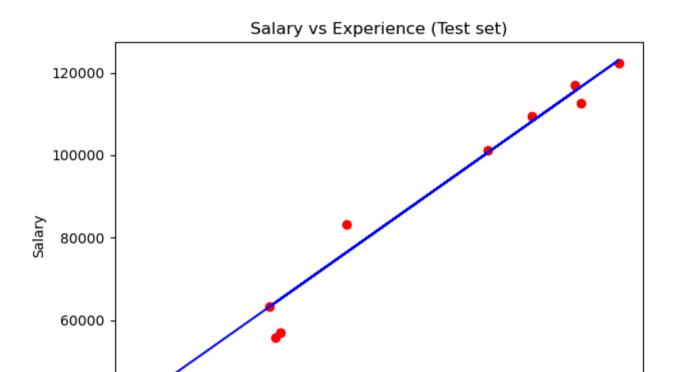
```
In [36]: # split the data into training and test sets:
   X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 1/3,
   random_state = 0)

In [37]: # fit the linear least-squares regression line to the training data:
   regr = LinearRegression()
```

### Salary vs Experience (Train set)



```
In [39]: # visualise test set results
fig3, ax3 = plt.subplots()
    ax3.scatter(X_test, y_test, color = 'red')
    ax3.plot(X_test, regr.predict(X_test), color = 'blue')
    ax3.set_title('Salary vs Experience (Test set)')
    ax3.set_xlabel('Years of Experience')
    ax3.set_ylabel('Salary')
    fig3.tight_layout()
    fig3.savefig('LR_test_plot.png')
```



Years of Experience

# **TESTING MODEL**

```
In [42]: # The coefficients
print('Coefficients: ', regr.coef_)

# The intercept
print('Intercept: ', regr.intercept_)

# The mean squared error
print('Mean squared error: %.8f'
% mean_squared_error(y_test, regr.predict(X_test)))
```

```
# The R^2 value:
print('Coefficient of determination: %.2f'
% r2_score(y_test, regr.predict(X_test)))
```

Coefficients: [9345.94244312] Intercept: 26816.19224403119

Mean squared error: 21026037.32951130 Coefficient of determination: 0.97

#### **PREDICTION**