```
# import all the lib
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np

# read the dataset using pandas
data = pd.read_csv('/content/Salary_Data.csv')

# This displays the top 5 rows of the data
data.head()
```

	YearsExperience	Salary	
0	1.1	39343.0	ıl.
1	1.3	46205.0	
2	1.5	37731.0	
3	2.0	43525.0	
4	2.2	39891.0	

Provides some information regarding the columns in the data data.info()

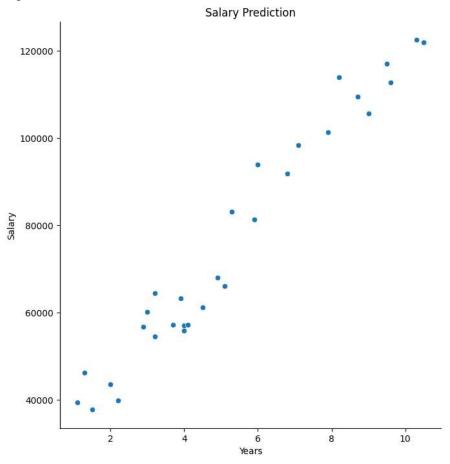
this describes the basic stat behind the dataset used
data.describe()

	YearsExperience	Salary	Ħ
count	30.000000	30.000000	ılı
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	

These Plots help to explain the values and how they are scattered

```
plt.figure(figsize=(12,6))
sns.pairplot(data,x_vars=['YearsExperience'],y_vars=['Salary'],size=7,kind='scatter')
plt.xlabel('Years')
plt.ylabel('Salary')
plt.title('Salary Prediction')
plt.show()
```

/usr/local/lib/python3.10/dist-packages/seaborn/axisgrid.py:2100: UserWarning: The `siz warnings.warn(msg, UserWarning)
<Figure size 1200x600 with 0 Axes>



```
# Cooking the data
X = data['YearsExperience']
X.head()
     0
          1.1
     1
          1.3
     2
          1.5
     3
          2.0
     4
          2.2
     Name: YearsExperience, dtype: float64
# Cooking the data
y = data['Salary']
y.head()
     0
          39343.0
          46205.0
     1
          37731.0
     2
     3
          43525.0
          39891.0
     Name: Salary, dtype: float64
# Import Segregating data from scikit learn
from sklearn.model_selection import train_test_split
# Split the data for train and test
X_train,X_test,y_train,y_test = train_test_split(X,y,train_size=0.7,random_state=100)
\mbox{\tt\#} Create new axis for x column
X_train = X_train[:,np.newaxis]
X_test = X_test[:,np.newaxis]
```

```
# Predicting the Salary for the Test values
y_pred = lr.predict(X_test)

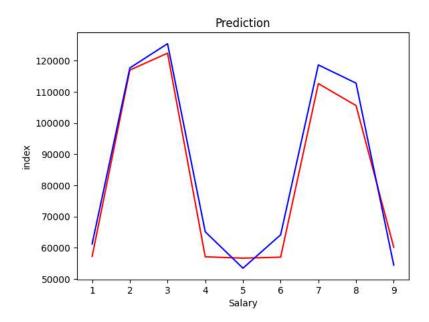
# Plotting the actual and predicted values

c = [i for i in range (1,len(y_test)+1,1)]
plt.plot(c,y_test,color='r',linestyle='-')
plt.plot(c,y_pred,color='b',linestyle='-')
plt.xlabel('Salary')
plt.ylabel('index')
```

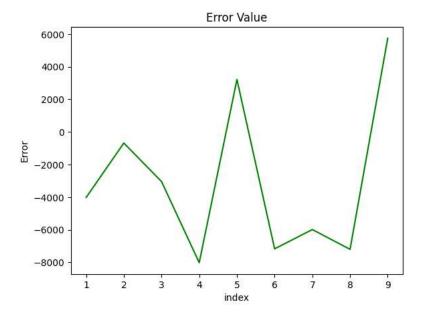
* LinearRegression
LinearRegression()

plt.title('Prediction')

plt.show()



```
# plotting the error
c = [i for i in range(1,len(y_test)+1,1)]
plt.plot(c,y_test-y_pred,color='green',linestyle='-')
plt.xlabel('index')
plt.ylabel('Error')
plt.title('Error Value')
plt.show()
```



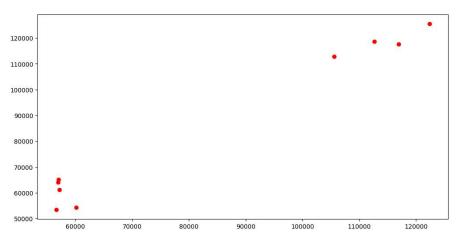
Importing metrics for the evaluation of the model from sklearn.metrics import r2_score,mean_squared_error

```
# calculate Mean square error
mse = mean_squared_error(y_test,y_pred)

# Calculate R square vale
rsq = r2_score(y_test,y_pred)

print('mean squared error :',mse)
print('r square :',rsq)
    mean squared error : 30310299.043402452
    r square : 0.9627668685473267

# Just plot actual and predicted values for more insights
plt.figure(figsize=(12,6))
plt.scatter(y_test,y_pred,color='r',linestyle='-')
plt.show()
```



print('Intercept of the model:',lr.intercept_)
print('Coefficient of the line:',lr.coef_)

Intercept of the model: 25202.887786154883
Coefficient of the line: [9731.20383825]

Then it is said to form a line with

$$y = 25202.8 + 9731.2x$$