

## Dataset Loading and Understanding

```
import pandas as pd
import numpy as np
df = pd.read_csv('/content/Salary_dataset.csv')
print(df)
```


	Unnamed: 0	YearsExperience	Salary
0	0	1.2	39344.0
1	1	1.4	46206.0
2	2	1.6	37732.0
3	3	2.1	43526.0
4	4	2.3	39892.0
5	5	3.0	56643.0
6	6	3.1	60151.0
7	7	3.3	54446.0
8	8	3.3	64446.0
9	9	3.8	57190.0
10	10	4.0	63219.0
11	11	4.1	55795.0
12	12	4.1	56958.0
13	13	4.2	57082.0
14	14	4.6	61112.0
15	15	5.0	67939.0
16	16	5.2	66030.0
17	17	5.4	83089.0
18	18	6.0	81364.0
19	19	6.1	93941.0
20	20	6.9	91739.0
21	21	7.2	98274.0
22	22	8.0	101303.0
23	23	8.3	113813.0
24	24	8.8	109432.0
25	25	9.1	105583.0
26	26	9.6	116970.0
27	27	9.7	112636.0
28	28	10.4	122392.0
29	29	10.6	121873.0

```
df.head()
```

	Unnamed: 0	YearsExperience	Salary
0	0	1.2	39344.0
1	1	1.4	46206.0
2	2	1.6	37732.0
3	3	2.1	43526.0
4	4	2.3	39892.0

Next steps:

[Generate code with df](#)[New interactive sheet](#)`df.tail()`

	Unnamed: 0	YearsExperience	Salary	
25	25	9.1	105583.0	
26	26	9.6	116970.0	
27	27	9.7	112636.0	
28	28	10.4	122392.0	
29	29	10.6	121873.0	

```
YearsExperience = df[['YearsExperience']]
Salary = df['Salary']

print("Input variable (YearsExperience):")
print(YearsExperience.head())
print("\nOutput variable (Salary):")
print(Salary.head())
```

Input variable (YearsExperience):

```
YearsExperience
0          1.2
1          1.4
2          1.6
3          2.1
4          2.3
```

Output variable (Salary):

```
0    39344.0
1    46206.0
2    37732.0
3    43526.0
4    39892.0
Name: Salary, dtype: float64
```

## Linear Regression using Scikit-learn

```
from sklearn.model_selection import train_test_split
X = df[['YearsExperience']] # Independent variable
y = df['Salary']            # Dependent variable
# Split into 80% training and 20% test data
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
# Create model
model = LinearRegression()
# Train model
model.fit(X_train, y_train)
```

▼ LinearRegression ⓘ ?

```
LinearRegression()
```

```
# Predict salaries for test set
y_pred = model.predict(X_test)
# Compare predictions with actual values
print("Predicted salaries:", y_pred)
print("Actual salaries:", list(y_test))
```

```
Predicted salaries: [115791.21011287  71499.27809463 102597.86866063  75268.8
 55478.79204548  60190.69970699]
Actual salaries: [112636.0, 67939.0, 113813.0, 83089.0, 64446.0, 57190.0]
```

## Performance Evaluation

```
from sklearn.metrics import mean_squared_error, r2_score
# Calculate Mean Squared Error
mse = mean_squared_error(y_test, y_pred)
print("Mean Squared Error:", mse)
# Calculate R-squared value
r2 = r2_score(y_test, y_pred)
print("R-squared:", r2)
```

```
Mean Squared Error: 49830096.855908394
R-squared: 0.9024461774180497
```

```
print("Slope (Coefficient):", model.coef_)
print("Intercept:", model.intercept_)
```

```
Slope (Coefficient): [9423.81532303]
Intercept: 24380.201479473704
```

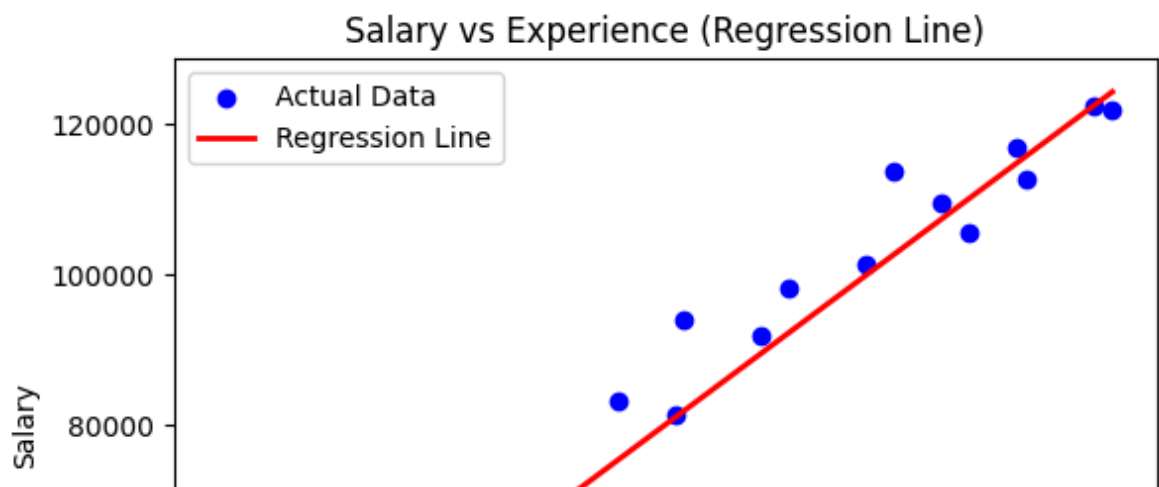
```
print("Predicted salaries:", y_pred)
print("Actual salaries:", list(y_test))
```

```
Predicted salaries: [115791.21011287  71499.27809463 102597.86866063  75268.8
 55478.79204548  60190.69970699]
Actual salaries: [112636.0, 67939.0, 113813.0, 83089.0, 64446.0, 57190.0]
```

```
from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
print("MSE:", mse)
print("R²:", r2)
```

MSE: 49830096.855908394  
R<sup>2</sup>: 0.9024461774180497

```
import matplotlib.pyplot as plt
# Scatter plot of actual data
plt.scatter(X, y, color='blue', label='Actual Data')
# Regression line
plt.plot(X, model.predict(X), color='red', linewidth=2, label='Regression Li
plt.xlabel("Years of Experience")
plt.ylabel("Salary")
plt.title("Salary vs Experience (Regression Line)")
plt.legend()
plt.show()
```



```
plt.scatter(y_test, y_pred, color='green')
plt.plot([y_test.min(), y_test.max()], [
plt.xlabel("Actual Salary")
plt.ylabel("Predicted Salary")
plt.title("Actual vs Predicted Salaries")
plt.show()
```

