

# Poly\_1\_2\_Salary\_1\_Advertising\_1

February 6, 2025

## 1 Salary

```
[91]: # Importing the libraries
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns
```

```
[92]: url = "https://raw.githubusercontent.com/content-anu/
↳dataset-polynomial-regression/master/Position_Salaries.csv"
df = pd.read_csv(url)
```

```
[93]: df.head()
```

```
[93]:
```

	Position	Level	Salary
0	Business Analyst	1	45000
1	Junior Consultant	2	50000
2	Senior Consultant	3	60000
3	Manager	4	80000
4	Country Manager	5	110000

```
[94]: # Splitting the data to X and Y
X = df.Level.values
y = df.Salary.values
print(f"The shape of X is {X.shape} and the shape of y is {y.shape}")
```

The shape of X is (10,) and the shape of y is (10,)

```
[95]: # Reshaping the data
X = X.reshape(-1, 1)
y = y.reshape(-1, 1)
print(f"The shape of X is {X.shape} and the shape of y is {y.shape}")
```

The shape of X is (10, 1) and the shape of y is (10, 1)

```
[96]: # Implementing the Polynomial Regression
from sklearn.preprocessing import PolynomialFeatures
from sklearn.linear_model import LinearRegression
```

```
# For Simple Linear Regression
lin_reg = LinearRegression()
lin_reg.fit(X, y)
y_pred = lin_reg.predict(X)
```

```
[97]: # Linear Regression on Scatter Plot
plt.scatter(X, y, color = 'red')
plt.plot(X, y_pred, color = 'blue')
plt.title('Salary vs Position Level (Linear Regression)')
plt.xlabel('Position level')
plt.ylabel('Salary')
plt.show()
```



```
[98]: # Evaluating the model
from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y, y_pred)
r2 = r2_score(y, y_pred)
print(f"The Mean Squared Error is {mse} and the R2 Score is {r2}")
```

The Mean Squared Error is 26695878787.878788 and the R2 Score is 0.6690412331929895

```
[99]: from sklearn.preprocessing import PolynomialFeatures
poly_reg = PolynomialFeatures(degree = 3)
X_poly = poly_reg.fit_transform(X)
```

```
[100]: X
```

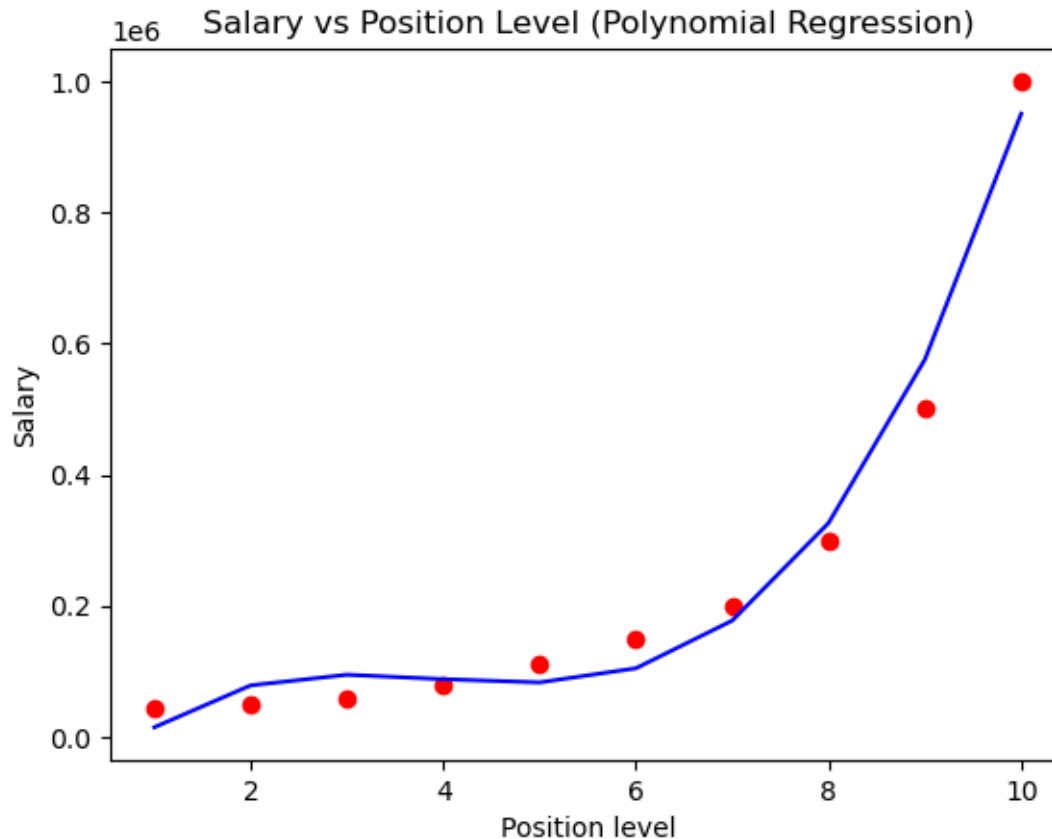
```
[100]: array([[ 1],
              [ 2],
              [ 3],
              [ 4],
              [ 5],
              [ 6],
              [ 7],
              [ 8],
              [ 9],
              [10]], dtype=int64)
```

```
[101]: X_poly
```

```
[101]: array([[ 1.,   1.,   1.,   1.],
              [ 1.,   2.,   4.,   8.],
              [ 1.,   3.,   9.,  27.],
              [ 1.,   4.,  16.,  64.],
              [ 1.,   5.,  25., 125.],
              [ 1.,   6.,  36., 216.],
              [ 1.,   7.,  49., 343.],
              [ 1.,   8.,  64., 512.],
              [ 1.,   9.,  81., 729.],
              [ 1.,  10., 100.,1000.]])
```

```
[102]: lin_reg_2 = LinearRegression()
lin_reg_2.fit(X_poly, y)
y_pred_2 = lin_reg_2.predict(X_poly)
```

```
[103]: # Poly Regression on Scatter Plot
plt.scatter(X, y, color = 'red')
plt.plot(X, y_pred_2, color = 'blue')
plt.title('Salary vs Position Level (Polynomial Regression)')
plt.xlabel('Position level')
plt.ylabel('Salary')
plt.show()
```



```
[104]: # Evaluating the model
from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y, y_pred_2)
r2 = r2_score(y, y_pred_2)
print(f"The Mean Squared Error is {mse} and the R2 Score is {r2}")
```

The Mean Squared Error is 1515662004.6620033 and the R2 Score is 0.9812097727913367

## 2 Advertising

```
[105]: # Practice on Advertisemnt Dataset
url = r"D:\Supervised Machine Learning lab (SMLL)\2\Assignment 2 Advertising.
      ↪CSV"
dataset = pd.read_csv(url)
dataset.head()
```

```
[105]: Unnamed: 0    TV    Radio    Newspaper    Sales
0          1  230.1    37.8         69.2    22.1
```

1	2	44.5	39.3	45.1	10.4
2	3	17.2	45.9	69.3	9.3
3	4	151.5	41.3	58.5	18.5
4	5	180.8	10.8	58.4	12.9

```
[106]: # Using Multiple Linear Regression
X = dataset[['TV', 'Radio', 'Newspaper']]
y = dataset['Sales']

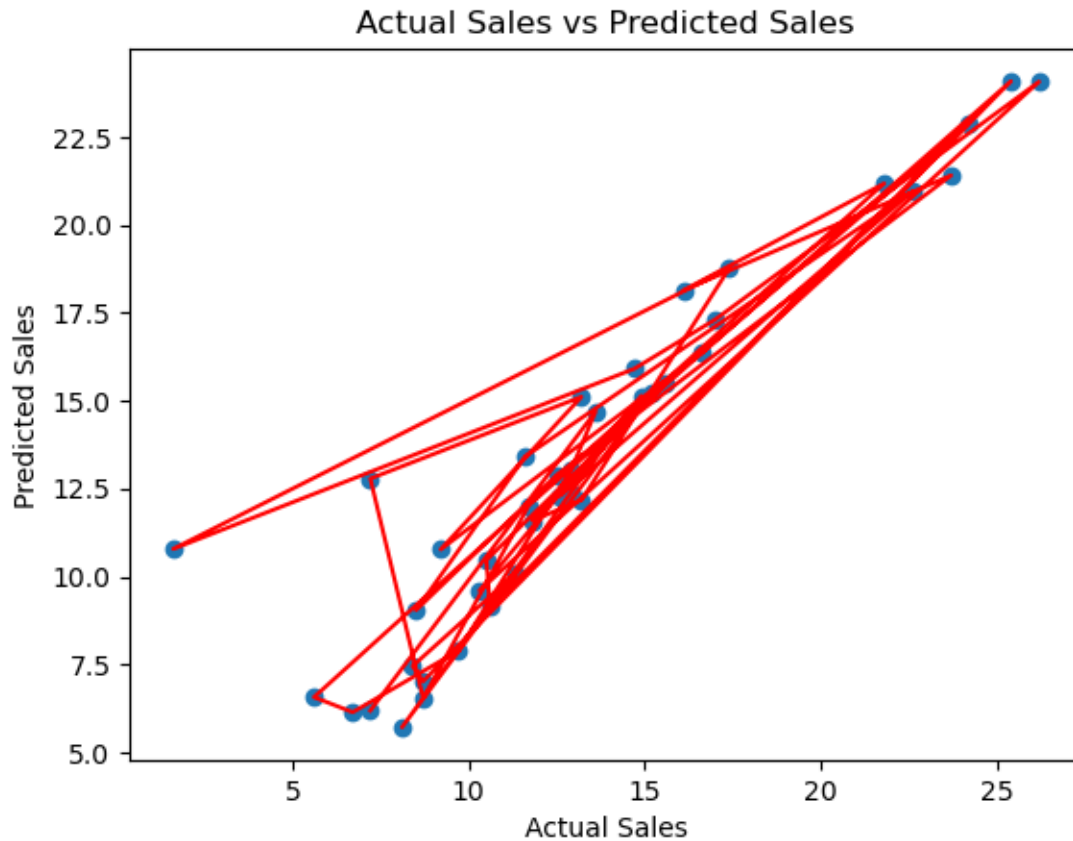
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2,
↳random_state = 0)

from sklearn.linear_model import LinearRegression
lin_reg = LinearRegression()
lin_reg.fit(X_train, y_train)
y_pred = lin_reg.predict(X_test)

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(f"The Mean Squared Error is {mse} and the R2 Score is {r2}")

# Scatter Plot
plt.scatter(y_test, y_pred)
plt.xlabel("Actual Sales")
plt.plot(y_test, y_pred, color = 'red')
plt.ylabel("Predicted Sales")
plt.title("Actual Sales vs Predicted Sales")
plt.show()
```

The Mean Squared Error is 4.402118291449685 and the R2 Score is 0.8601145185017868



```
[107]: # Using Polynomial Regression
from sklearn.preprocessing import PolynomialFeatures
poly = PolynomialFeatures(degree = 3)
X_train_poly = poly.fit_transform(X_train)
X_test_poly = poly.fit_transform(X_test)
y_train = np.array(y_train).reshape(-1, 1)

lin_reg_2 = LinearRegression()
lin_reg_2.fit(X_train_poly, y_train)
y_pred_2 = lin_reg_2.predict(X_test_poly)

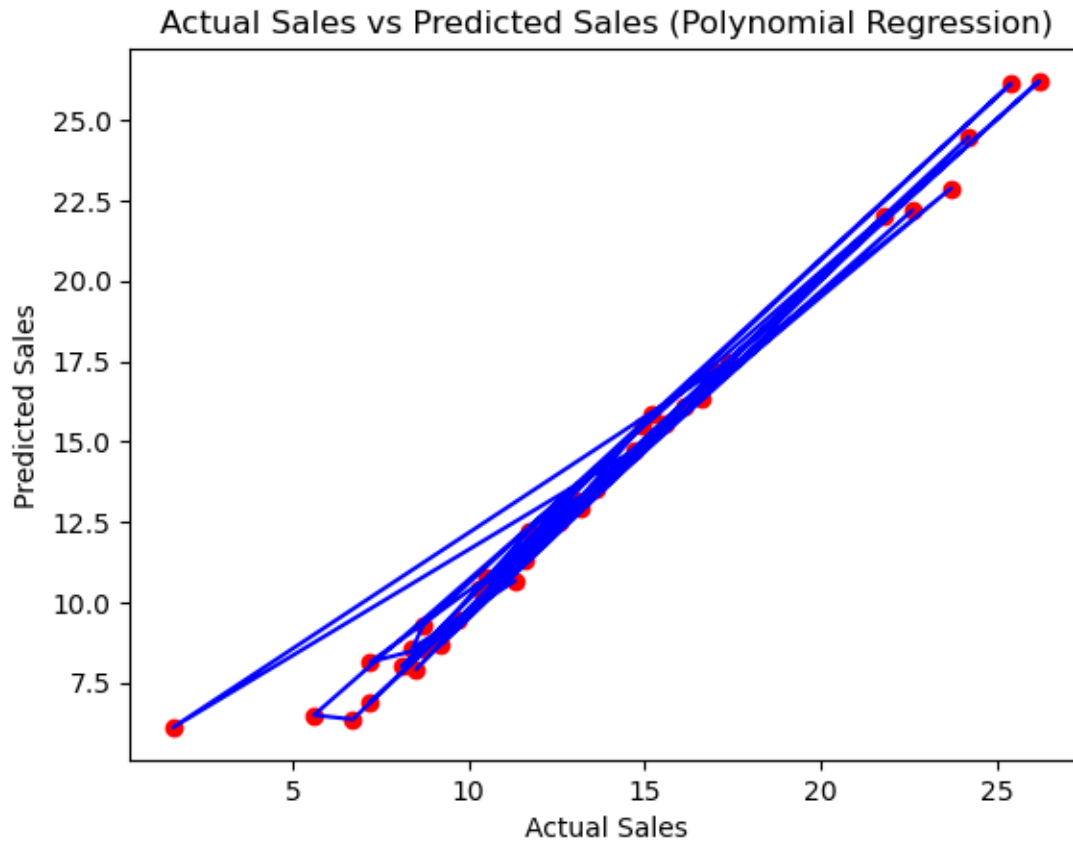
# Poly Regression on Scatter Plot
plt.scatter(y_test, y_pred_2, color = 'red')
plt.plot(y_test, y_pred_2, color = 'blue')
plt.title('Actual Sales vs Predicted Sales (Polynomial Regression)')
plt.xlabel('Actual Sales')
plt.ylabel('Predicted Sales')
plt.show()

# Evaluating the model
```

```

from sklearn.metrics import mean_squared_error, r2_score
mse = mean_squared_error(y_test, y_pred_2)
r2 = r2_score(y_test, y_pred_2)
print(f"The Mean Squared Error is {mse} and the R2 Score is {r2}")

```



The Mean Squared Error is 0.6721344417962923 and the R2 Score is 0.978641680255429

```

[109]: import numpy as np
import matplotlib.pyplot as plt
sorted_idx = np.argsort(y_test)
y_test_sorted = np.array(y_test)[sorted_idx]
y_pred_sorted = y_pred_2[sorted_idx]

# Scatter Plot (Actual vs Predicted)
plt.figure(figsize=(8, 5))
plt.scatter(y_test_sorted, y_pred_sorted, color = 'red', alpha=0.5)
plt.plot(y_test_sorted, y_pred_sorted, color = 'blue')
plt.title('Sorted Actual Sales vs Predicted Sales')
plt.xlabel('Predicted Sales')

```

```
plt.ylabel('Actual Sales')  
plt.show()
```

