

10. Data Science – Machine Learning – Polynomial Features

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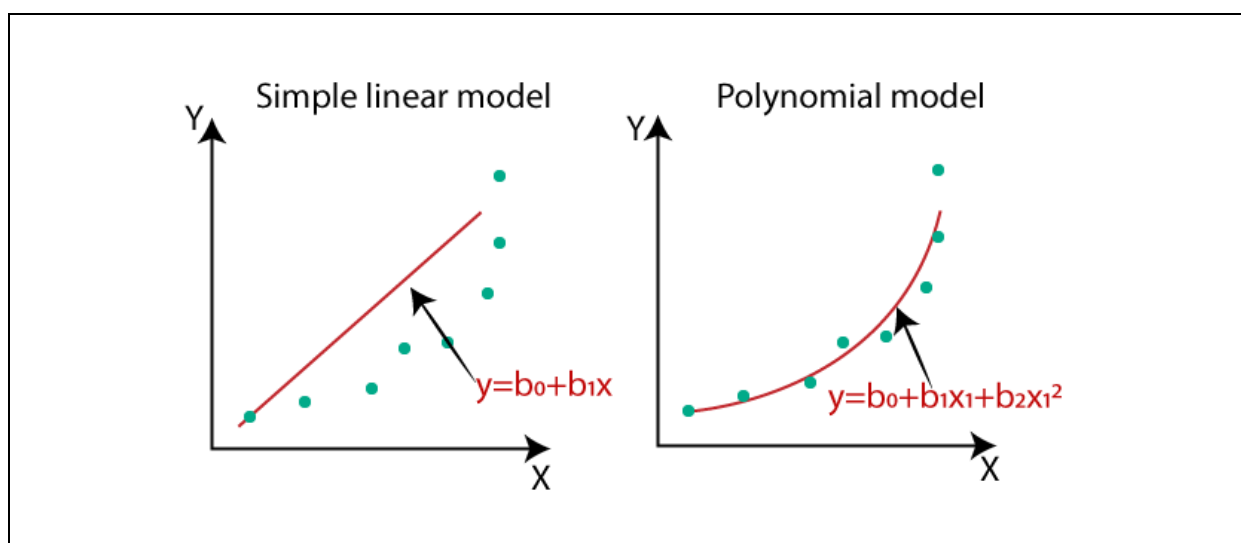
10. Data Science – Machine Learning – Polynomial Features

1. Polynomial Features for machine learning

- ✓ As such, polynomial features are a type of feature engineering means creating new input features based on the existing features.
- ✓ For example, if a dataset had one input feature X , then a polynomial feature would be the addition of a new feature (column) where values were calculated by squaring the values in X , e.g. X^2 .
- ✓ This process can be repeated for each input variable in the dataset, creating a transformed version of each.
- ✓ The "degree" of the polynomial is used to control the number of features added.

2. Need of Polynomial Features

- ✓ If we apply a linear model on a linear dataset, then it provides us a good result as we have seen in Simple Linear Regression.
- ✓ If we apply the same model without any modification on a non-linear dataset, then it will produce wrong results
- ✓ Due to this,
 - The error rate will be high etc



3. Equations

Simple Linear Regression equation

✓ $y = b_0 + b_1x$

Multiple Linear Regression equation

✓ $y = b_0 + b_1x + b_2x^2 + b_3x^3 + \dots + b_nx^n$

Polynomial Regression equation: $y = b_0 + b_1x + b_2x^2 + b_3x^3 + \dots + b_nx^n$

Program Name Creating an array
demo1.py

```
from numpy import asarray
```

```
data = asarray([[2], [3], [4]])  
print(data)
```

Output

```
[[2]  
 [3]  
 [4]]
```

Program Name Creating feature from existing feature
demo2.py

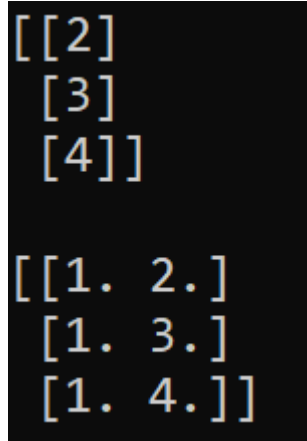
```
from numpy import asarray
from sklearn.preprocessing import PolynomialFeatures

data1 = asarray([[2],[3],[4]])

trans = PolynomialFeatures(degree = 1)
data2 = trans.fit_transform(data1)

print(data1)
print()
print(data2)
```

Output



```
[[2]
 [3]
 [4]]

[[1.  2.]
 [1.  3.]
 [1.  4.]]
```

Program Name Creating feature from existing feature
demo3.py

```
from numpy import asarray
from sklearn.preprocessing import PolynomialFeatures

data1 = asarray([[2],[3],[4]])

trans = PolynomialFeatures(degree = 2)
data2 = trans.fit_transform(data1)

print(data1)
print()
print(data2)
```

Output

```
[[2]
 [3]
 [4]]

[[ 1.  2.  4.]
 [ 1.  3.  9.]
 [ 1.  4. 16.]]
```

Program Name Creating feature from existing feature
demo4.py

```
from numpy import asarray
from sklearn.preprocessing import PolynomialFeatures

data1 = asarray([[2],[3],[4]])

trans = PolynomialFeatures(degree = 3)
data2 = trans.fit_transform(data1)

print(data1)
print()
print(data2)
```

Output

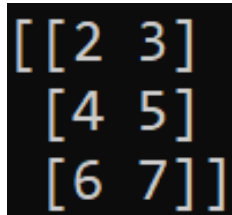
```
[[2]
 [3]
 [4]]

[[ 1.  2.  4.  8.]
 [ 1.  3.  9. 27.]
 [ 1.  4. 16. 64.]
```

Program Name Creating feature from existing feature
demo5.py

```
from numpy import asarray  
  
data1 = asarray([[2, 3],[4, 5],[6, 7]])  
  
print(data1)
```

Output



```
[[2 3]  
 [4 5]  
 [6 7]]
```


Program Name Creating feature from existing feature
demo6.py

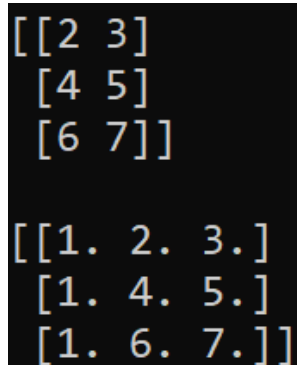
```
from numpy import asarray
from sklearn.preprocessing import PolynomialFeatures

data1 = asarray([[2, 3],[4, 5],[6, 7]])

trans = PolynomialFeatures(degree = 1)
data2 = trans.fit_transform(data1)

print(data1)
print()
print(data2)
```

Output



```
[[2 3]
 [4 5]
 [6 7]]

[[1.  2.  3.]
 [1.  4.  5.]
 [1.  6.  7.]]
```

Program Name Creating feature from existing feature
demo7.py

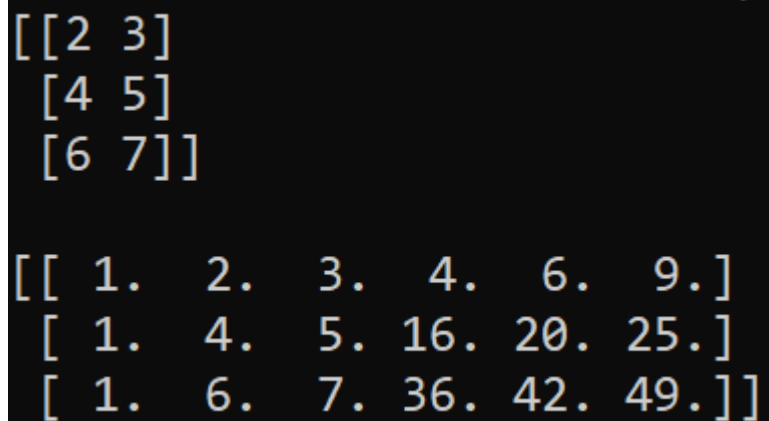
```
from numpy import asarray
from sklearn.preprocessing import PolynomialFeatures

data1 = asarray([[2, 3],[4, 5],[6, 7]])

trans = PolynomialFeatures(degree = 2)
data2 = trans.fit_transform(data1)

print(data1)
print()
print(data2)
```

Output



```
[[2 3]
 [4 5]
 [6 7]]

[[ 1.  2.  3.  4.  6.  9.]
 [ 1.  4.  5. 16. 20. 25.]
 [ 1.  6.  7. 36. 42. 49.]
```

Program Name Creating feature from existing feature
demo8.py

```
from numpy import asarray
from sklearn.preprocessing import PolynomialFeatures

data1 = asarray([[2, 3],[4, 5],[6, 7]])

trans = PolynomialFeatures(degree = 3)
data2 = trans.fit_transform(data1)

print(data1)
print()
print(data2)
```

Output

```
[[2 3]
 [4 5]
 [6 7]]

[[ 1.  2.  3.  4.  6.  9.  8. 12. 18. 27.]
 [ 1.  4.  5. 16. 20. 25. 64. 80. 100. 125.]
 [ 1.  6.  7. 36. 42. 49. 216. 252. 294. 343.]]
```

Program Name Loading dataset
demo9.py

```
import pandas as pd

df = pd.read_csv("poly_dataset.csv")

print(df)
```

Output

	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
5	Region Manager	6	150000
6	Partner	7	200000
7	Senior Partner	8	300000
8	C-level	9	500000
9	CEO	10	1000000

Program Name Data preparation
demo10.py

```
import pandas as pd

df = pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

print(X)
print()
print(y)
```

Output

```
[[ 1]
 [ 2]
 [ 3]
 [ 4]
 [ 5]
 [ 6]
 [ 7]
 [ 8]
 [ 9]
[10]]

[ 45000  50000  60000  80000 110000 150000 200000 300000 500000
1000000]
```

Program Name Plotting the dataset
demo11.py

```
import pandas as pd
import matplotlib.pyplot as plt

df=pd.read_csv("poly_dataset.csv")

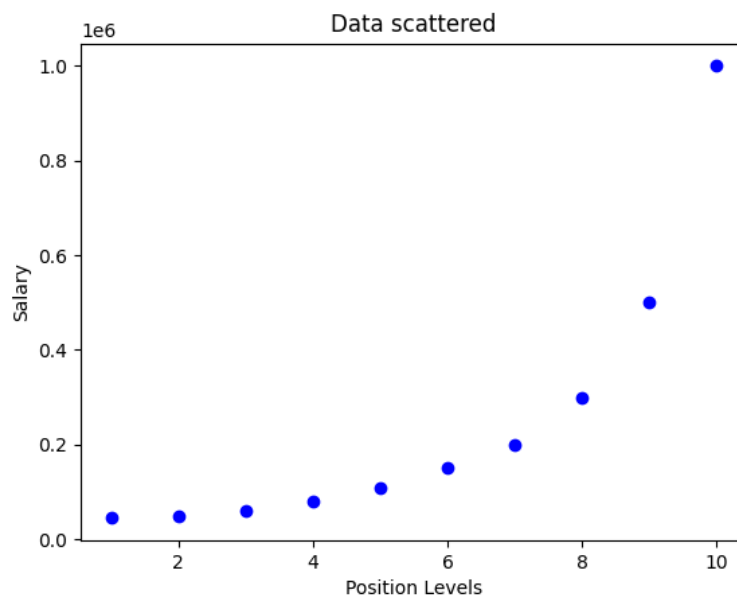
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

plt.scatter(X, y, color="blue")

plt.title("Data scattered")
plt.xlabel("Position Levels")
plt.ylabel("Salary")

plt.show()
```

Output



Program Name Model training
demo12.py

```
import pandas as pd
from sklearn.linear_model import LinearRegression

# Loading the dataset
df = pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

# Model training
lin_regs= LinearRegression()
lin_regs.fit(X, y)

print("Model got trained")
```

Output

Model got trained

Program Name Model prediction
demo13.py

```
import pandas as pd
from sklearn.linear_model import LinearRegression

# Loading the dataset
df = pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

# Model training
lin_regs= LinearRegression()
lin_regs.fit(X, y)

print("Model got trained")
print(lin_regs.predict([[6.5]]) )
```

Output

```
[330378.78787879]
```


Program Name Plotting the dataset
demo14.py

```
import pandas as pd
from sklearn.linear_model import LinearRegression
import matplotlib.pyplot as plt

# Loading the dataset
df=pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

# Model training
lin_regs= LinearRegression()
lin_regs.fit(X, y)

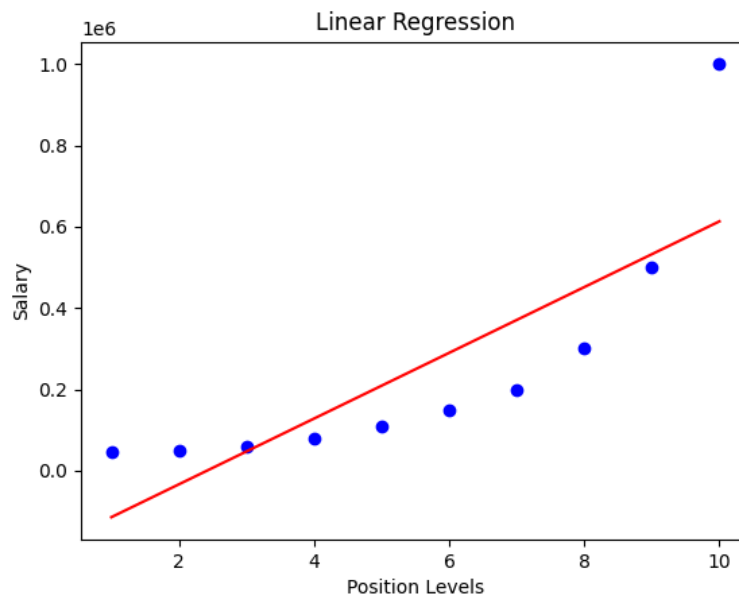
plt.scatter(X, y, color="blue")

plt.plot(X, lin_regs.predict(X), color = "red")

plt.title("Linear Regression")
plt.xlabel("Position Levels")
plt.ylabel("Salary")

plt.show()
```

Output



4. Lets create Polynomial features

- ✓ We need to use PolynomialFeatures class to get polynomial features.

```
Program Name    Fitting the Polynomial regression to the dataset
demo15.py

import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures

# Loading the dataset
df=pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

poly_regs= PolynomialFeatures(degree = 1)

x_poly= poly_regs.fit_transform(X)

print(x_poly)
```

Output

```
      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
5  Region Manager      6  150000
6           Partner      7  200000
7  Senior Partner      8  300000
8           C-level      9  500000
9           CEO     10 1000000

[[ 1.  1.]
 [ 1.  2.]
 [ 1.  3.]
 [ 1.  4.]
 [ 1.  5.]
 [ 1.  6.]
 [ 1.  7.]
 [ 1.  8.]
 [ 1.  9.]
 [ 1. 10.]]
```

Program Name Fitting the Polynomial regression to the dataset
demo16.py

```
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures

# Loading the dataset
df=pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

poly_regs= PolynomialFeatures(degree = 2)

x_poly= poly_regs.fit_transform(X)

print(x_poly)
```

Output

	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
5	Region Manager	6	150000
6	Partner	7	200000
7	Senior Partner	8	300000
8	C-level	9	500000
9	CEO	10	1000000


```
[[ 1.  1.  1.]  
[ 1.  2.  4.]  
[ 1.  3.  9.]  
[ 1.  4. 16.]  
[ 1.  5. 25.]  
[ 1.  6. 36.]  
[ 1.  7. 49.]  
[ 1.  8. 64.]  
[ 1.  9. 81.]  
[ 1. 10. 100.]]
```

Program Name Fitting the Polynomial regression to the dataset
demo17.py

```
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures

# Loading the dataset
df=pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

poly_regs= PolynomialFeatures(degree = 2)
x_poly= poly_regs.fit_transform(X)

model = LinearRegression()
model.fit(x_poly, y)

print("Fitting the Polynomial regression to the dataset ")
```

Output

Fitting the Polynomial regression to the dataset

Program Name Plotting Polynomial Regression features
demo18.py

```
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
import matplotlib.pyplot as plt

# Loading the dataset
df=pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

#Fitting the Polynomial regression to the dataset
poly_regs= PolynomialFeatures(degree = 2)
x_poly = poly_regs.fit_transform(X)
model =LinearRegression()
model.fit(x_poly, y)

# Plotting Polynomial Regression

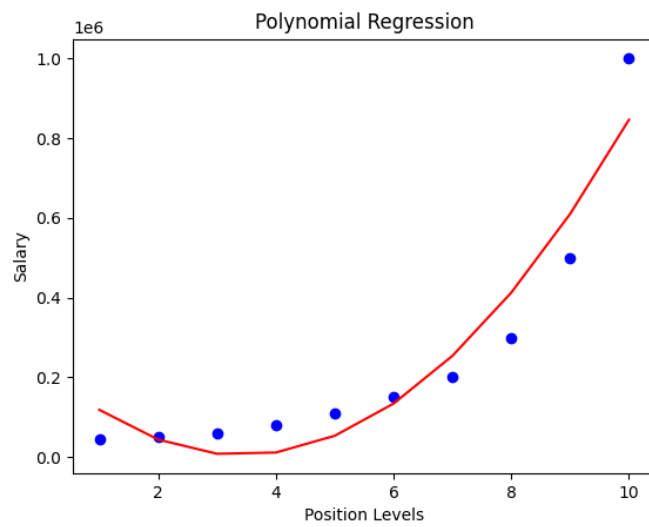
plt.scatter(X, y, color="blue")

plt.plot(X, model.predict(x_poly), color="red")

plt.title("Polynomial Regression")
plt.xlabel("Position Levels")
plt.ylabel("Salary")

plt.show()
```


Output



Program Name Plotting Polynomial Regression features
demo19.py

```
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
import matplotlib.pyplot as plt

# Loading the dataset
df=pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

#Fitting the Polynomial regression to the dataset
poly_regs= PolynomialFeatures(degree = 3)
x_poly = poly_regs.fit_transform(X)
model =LinearRegression()
model.fit(x_poly, y)

# Plotting Polynomial Regression

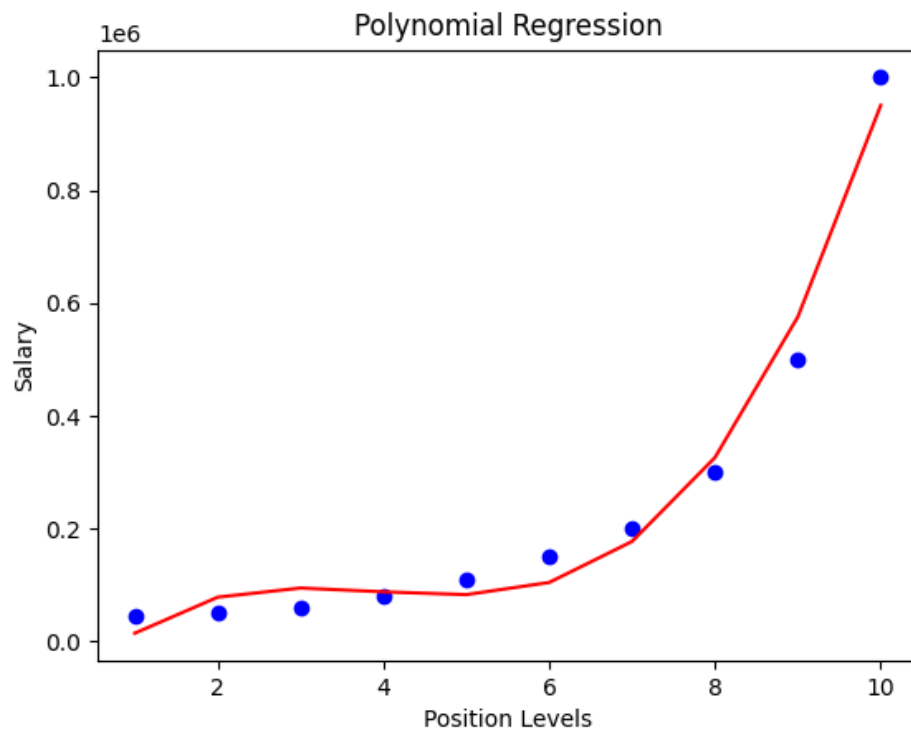
plt.scatter(X, y, color="blue")

plt.plot(X, model.predict(x_poly), color="red")

plt.title("Polynomial Regression")
plt.xlabel("Position Levels")
plt.ylabel("Salary")

plt.show()
```

Output



Program Name Plotting Polynomial Regression features
demo20.py

```
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
import matplotlib.pyplot as plt

# Loading the dataset
df=pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

#Fitting the Polynomial regression to the dataset
poly_regs= PolynomialFeatures(degree = 4)
x_poly = poly_regs.fit_transform(X)
model =LinearRegression()
model.fit(x_poly, y)

# Plotting Polynomial Regression

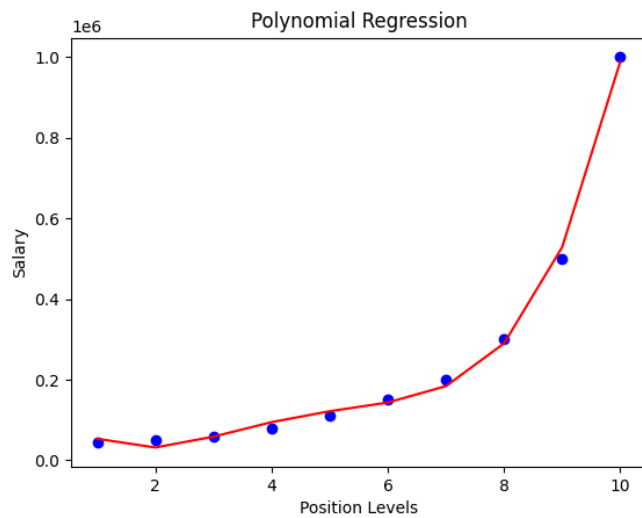
plt.scatter(X, y, color="blue")

plt.plot(X, model.predict(x_poly), color="red")

plt.title("Polynomial Regression")
plt.xlabel("Position Levels")
plt.ylabel("Salary")

plt.show()
```

Output



Program Name Predicting result
demo21.py

```
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
import matplotlib.pyplot as plt

# Loading the dataset
df=pd.read_csv("poly_dataset.csv")

# Data preparation
X = df.iloc[:, 1:2].values
y = df.iloc[:, 2].values

#Fitting the Polynomial regression to the dataset
poly_regs= PolynomialFeatures(degree = 4)
x_poly= poly_regs.fit_transform(X)
model =LinearRegression()
model.fit(x_poly, y)

# Prediction with Polynomial Regression
poly_pred = model.predict(poly_regs.fit_transform([[6.5]]))
print(poly_pred)
```

Output [158862.45265155]

Note

- ✓ LinearRegression predicted output is : [330378.78787879]
- ✓ Polynomial Regression predicted output is : [158862.45265155]

Conclusion

- ✓ Polynomial Regression predicted output is the **accurate** one according to the discussion