

```
In [3]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
```

```
In [4]: data = {'level':[1,2,3,4,5,6,7,8,9,10],
               'salary':[45000,50000,60000,80000,110000,150000,200000,300000,500000,1000000]}
data
```

```
Out[4]: {'level': [1, 2, 3, 4, 5, 6, 7, 8, 9, 10],
         'salary': [45000,
                    50000,
                    60000,
                    80000,
                    110000,
                    150000,
                    200000,
                    300000,
                    500000,
                    1000000]}
```

```
In [5]: df = pd.DataFrame(data)
df
```

```
Out[5]:
```

	level	salary
0	1	45000
1	2	50000
2	3	60000
3	4	80000
4	5	110000
5	6	150000
6	7	200000
7	8	300000
8	9	500000
9	10	1000000

```
In [6]: x=df[['level']]
x
```

```
Out[6]:
```

	level
0	1
1	2
2	3
3	4
4	5
5	6
6	7
7	8
8	9
9	10

```
In [7]: y=df['salary']
y
```

```
Out[7]:
```

0	45000
1	50000
2	60000
3	80000
4	110000
5	150000
6	200000
7	300000
8	500000
9	1000000

Name: salary, dtype: int64

```
In [8]: lr = LinearRegression()
lr.fit(x,y)
y_pred = lr.predict(x).round(2)
y_pred
```

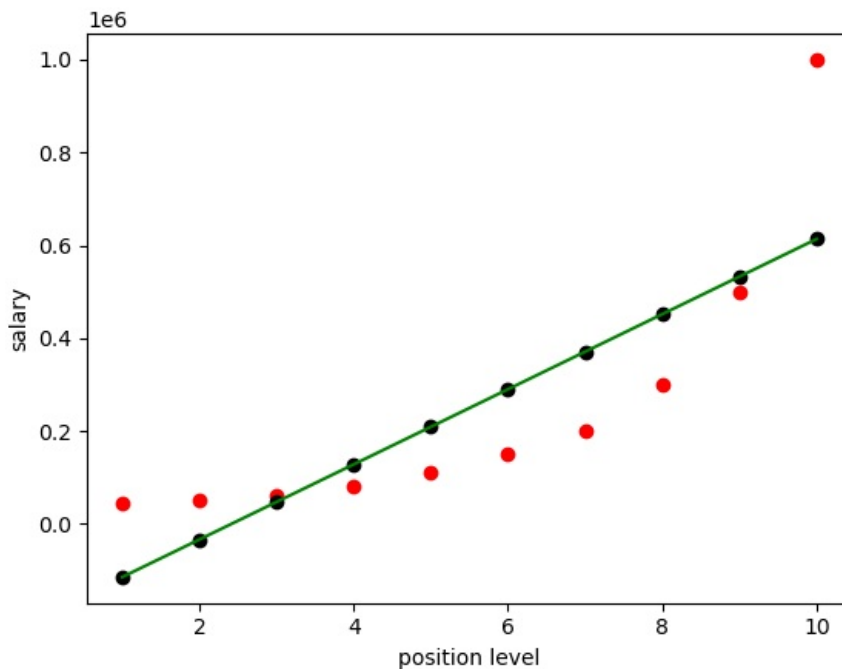
```
Out[8]: array([-114454.55, -33575.76,  47303.03, 128181.82, 209060.61,
        289939.39,  370818.18,  451696.97,  532575.76,  613454.55])
```

```
In [9]: from sklearn.metrics import r2_score,mean_squared_error
```

```
In [10]: print(r2_score(y,y_pred))
```

```
0.6690412358828437
```

```
In [11]: plt.scatter(x,y,color='red',label='actual salary data')
plt.plot(x,y_pred,color='green',label='Linear Regression')
plt.scatter(x,y_pred,color='Black',label='Linear Regression')
plt.xlabel('position level')
plt.ylabel('salary')
plt.show()
```



```
In [12]: poly = PolynomialFeatures(degree=3)
x_poly = poly.fit_transform(x)
x_poly
```

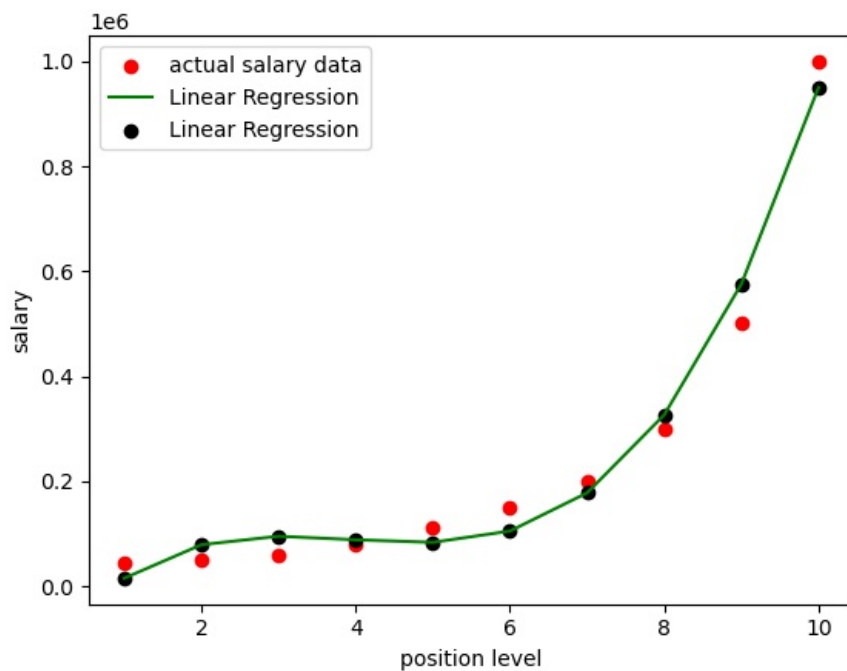
```
Out[12]: array([[ 1.,  1.,  1.,  1.],
        [ 1.,  2.,  4.,  8.],
        [ 1.,  3.,  9., 27.],
        [ 1.,  4., 16., 64.],
        [ 1.,  5., 25., 125.],
        [ 1.,  6., 36., 216.],
        [ 1.,  7., 49., 343.],
        [ 1.,  8., 64., 512.],
        [ 1.,  9., 81., 729.],
        [ 1., 10., 100., 1000.]])
```

```
In [13]: poly_reg = LinearRegression()
poly_reg.fit(x_poly,y)
poly_y = poly_reg.predict(x_poly)
```

```
In [14]: print(r2_score(y,poly_y))
```

```
0.9812097727913367
```

```
In [15]: plt.scatter(x,y,color='red',label='actual salary data')
plt.plot(x,poly_y,color='green',label='Linear Regression')
plt.scatter(x,poly_y,color='black',label='Linear Regression')
plt.xlabel('position level')
plt.ylabel('salary')
plt.legend()
plt.show()
```



```
In [16]: from sklearn.linear_model import Ridge,Lasso
```

```
In [17]: rm = Ridge()
rm.fit(x,y)
y_rd = rm.predict(x).round(2)
y_rd
```

```
Out[17]: array([-110095.81, -30185.63,  49724.55, 129634.73, 209544.91,
                289455.09, 369365.27, 449275.45, 529185.63, 609095.81])
```

```
In [18]: r2_score(y,y_rd)
```

```
Out[18]: 0.6689452755413183
```

```
In [19]: lr = Lasso(alpha=1,max_iter=200)
lr.fit(x,y)
y_pred=lr.predict(x).round()
r2_score(y,y_pred)
```

```
Out[19]: 0.6690412271043766
```