

```
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
```

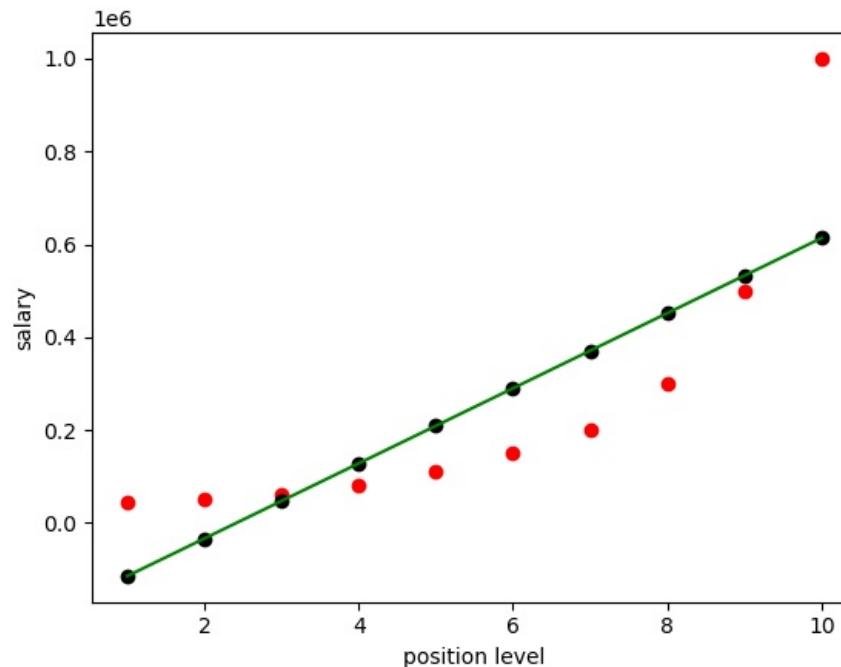
```
Out[8]: array([-114454.55, -33575.76, 47303.03, 128181.82, 209060.61,
 289939.39, 370818.18, 451696.97, 532575.76, 613454.55])
```

```
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
```

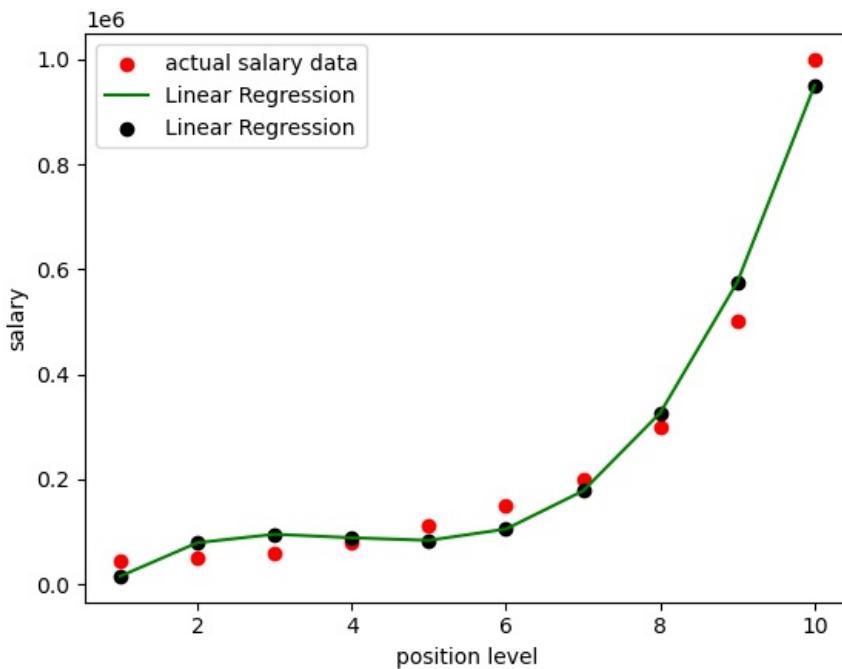
```
Out[12]: array([[ 1.,    1.,    1.,    1.],
 [ 1.,    2.,    4.,    8.],
 [ 1.,    3.,    9.,   27.],
 [ 1.,    4.,   16.,   64.],
 [ 1.,    5.,   25.,  125.],
 [ 1.,    6.,   36.,  216.],
 [ 1.,    7.,   49.,  343.],
 [ 1.,    8.,   64.,  512.],
 [ 1.,    9.,   81.,  729.],
 [ 1.,   10.,  100.,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