Polynomial Regression

October 17, 2021

1 Polynomial Regression

1.1 Importing the Libraries

[2]

```
[2]: import numpy as np
      import pandas as pd
      import matplotlib.pyplot as plt
[11]: from sklearn.model_selection import train_test_split
[13]: from sklearn.linear_model import LinearRegression
[17]: from sklearn.preprocessing import PolynomialFeatures
     1.2 Importing the data set
 [3]: dataset = pd.read_csv("Position_Salaries.csv")
 [5]: print(dataset)
                 Position Level
                                    Salary
                                     45000
     0
         Business Analyst
                                1
        Junior Consultant
                                2
                                     50000
     1
        Senior Consultant
                                3
                                     60000
     3
                   Manager
                                4
                                     80000
     4
          Country Manager
                                5
                                    110000
           Region Manager
     5
                                6
                                    150000
     6
                   Partner
                                7
                                    200000
     7
           Senior Partner
                                8
                                    300000
     8
                   C-level
                                9
                                    500000
     9
                       CE<sub>0</sub>
                               10 1000000
 [6]: x = dataset.iloc[:,1:-1].values
      y = dataset.iloc[:,-1].values
 [7]: print(x)
     [[ 1]
```

```
[ 3]
[ 4]
[ 5]
[ 6]
[ 7]
[ 8]
[ 9]
```

[10]: print(y)

[10]]

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

1.3 Spliting dataset into training and testing set

```
[12]: x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2, □ → random_state = 0)
```

1.4 Training the Linear Regression model on the whole dataset

```
[15]: linear_Regressor = LinearRegression()
linear_Regressor.fit(x,y)
```

[15]: LinearRegression()

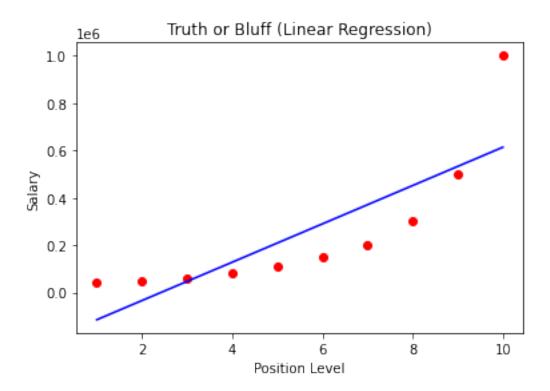
1.5 Training the Polynomial Regression model on the whole dataset

```
[19]: poly_reg = PolynomialFeatures(degree = 2)
x_poly= poly_reg.fit_transform(x)
linear_Reg = LinearRegression()
linear_Reg.fit(x_poly,y)
```

[19]: LinearRegression()

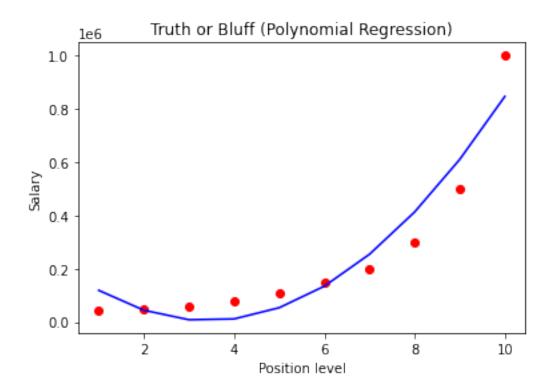
1.6 Visualising the Linear Regression results

```
[23]: plt.scatter(x, y, color = 'red')
  plt.plot(x, linear_Regressor.predict(x), color = 'blue')
  plt.title('Truth or Bluff (Linear Regression)')
  plt.xlabel('Position Level')
  plt.ylabel('Salary')
  plt.show()
```



1.7 Visualising the Polynomial Regression result

```
[24]: plt.scatter(x, y, color = 'red')
   plt.plot(x, linear_Reg.predict(poly_reg.fit_transform(x)), color = 'blue')
   plt.title('Truth or Bluff (Polynomial Regression)')
   plt.xlabel('Position level')
   plt.ylabel('Salary')
   plt.show()
```



1.8 Predicting a new result with Linear Regression

```
[26]: linear_Regressor.predict([[6.5]])
```

[26]: array([330378.78787879])

1.9 Predicting a new result with Polynomial Regression

```
[28]: linear_Reg.predict(poly_reg.fit_transform([[6.5]]))
```

[28]: array([189498.10606061])

thus you can see difference in polynominal regression and linear Regression, prediction using polynominal in more accurate $\frac{1}{2}$