Linear Regression

May 14, 2019

Linear Regression with one variable

Suppose you are the CEO of a restaurant franchise and are considering different cities for opening a new outlet. The chain already has trucks in various cities and you have data for profits and populations from the cities. You would like to use this data to help you select which city to expand to next.

```
[1]: import numpy as np
import matplotlib.pyplot as plt
from numpy.linalg import pinv
%matplotlib inline
```

Let us define our hypothesis, costFunction and implementation of gradient Descent in a vectorised way..

```
[2]: def hypothesis(Theta,X):
    return X@Theta #Vectorised Implementation

[3]: def costFunction(Theta,X,Y):
    m,n = X.shape
    return ((0.5/m)*((hypothesis(Theta,X) - Y).T @ (hypothesis(Theta,X) - Y))).
    →item() #Vectorised Implementation

[4]: def gradientDescent(Theta,X,Y,alpha,iterations):
    m,n = X.shape
    for count in range(0,iterations):
        Theta = Theta - (alpha * (1/m)*(X.T @ (hypothesis(Theta,X)-Y)))
    →#Vectorised Implementation
    return Theta
```

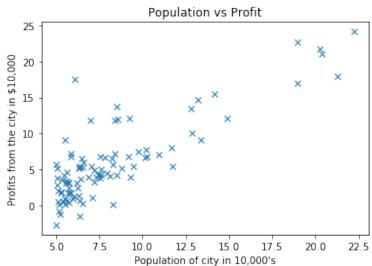
We have the data in csv format as population, profit

```
[5]: data = np.loadtxt("Data/ex1data1.txt", delimiter=",")
X = data[:,:-1]
Y = data[:,-1:]
```

Visualizing data: Population vs Profit

```
[6]: fig, ax = plt.subplots()
ax.plot(X,Y,ls='', marker = 'x', label='Population vs Profit')
ax.set(xlabel="Population of city in 10,000's", ylabel="Profits from the city
→in $10,000", title='Population vs Profit')
```

[6]: [Text(0, 0.5, 'Profits from the city in \$10,000'), Text(0.5, 0, "Population of city in 10,000's"), Text(0.5, 1.0, 'Population vs Profit')]



 $Regression_files/LinearRegression_10_1.bb$

Converting data into matrices as adding column of ones to X

```
[7]: X = np.matrix(X)
Y = np.matrix(Y)
m,n = X.shape
X_new = np.hstack((np.ones((m,1)),X))
n = n+1
```

Let us take, theta initially as zeros and set learning rate (alpha) and number of iterations

```
[8]: initialTheta = np.zeros((n,1))
alpha = 0.01  #Learning rate
iterations = 1500
```

Cost when initial parameters are zero.

```
[9]: print("Cost when theta is\n",initialTheta,"\nis:⊔

→",costFunction(initialTheta,X_new,Y))
```

```
Cost when theta is [[0.] [0.]] is: 32.072733877455676
```

Running Gradient Descent:

```
[10]: finalTheta = gradientDescent(initialTheta, X_new, Y, alpha, iterations) print("Optimized parameters after running gradient Descent are: \n", finalTheta)
```

```
Optimized parameters after running gradient Descent are: [[-3.63029144] [ 1.16636235]]
```

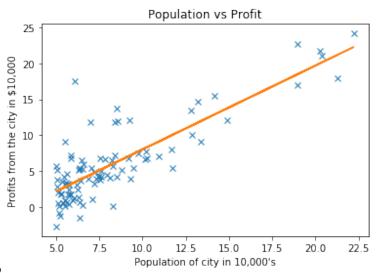
Cost after running Gradient Descent:

```
[11]: print("Cost after running gradient Descent is:

→",costFunction(finalTheta,X_new,Y))
```

Cost after running gradient Descent is: 4.483388256587725

Linear fit after obtaining the values of theta:



 $Regression_{f} iles/Linear Regression_{2} 2_{1}.bb \\$

Predicting profits for given population using our model:

```
[13]: print("Profits in $10,000's when the city population is 35,000: ",[1, 3.

→5]*finalTheta)
print("Profits in $10,000's when the city population is 70,000: ",[1,

→7]*finalTheta)
```

Profits in \$10,000's when the city population is 35,000: [[0.45197679]] Profits in \$10,000's when the city population is 70,000: [[4.53424501]]

Linear Regreesion with multivariable

Suppose you are selling your house and you want to know what a good market price would be. One way to do this is to first collect information on recent houses sold and make a model of housing prices.

Note: hypothesis, costFunction, gradientDescent are same as above.

Feature Normalization:

```
[14]: def featureNormalization(X):
         m, n = X.shape
         means = []
         stds = []
         for column in range(1,n):
             mean = X[:,column].mean()
             std = X[:,column].std()
             means.append(mean)
             stds.append(std)
             X[:,column] = (X[:,column] - mean)/std
         return X, means, stds
[15]: def predict(X, theta, means, stds):
         X = np.matrix(X)
         m,n = X.shape
         for column in range(1,n):
             X[0:,column] = (X[0:,column] - means[column-1])/stds[column-1]
         return X@theta
```

Loading Data:

```
[16]: data = np.loadtxt("Data/ex1data2.txt", delimiter=',')
X = data[:,:-1]
Y = data[:,-1:]
```

Converting data into matrices, adding column of ones to X and normalizing features:

```
[17]: X = np.matrix(X)
Y = np.matrix(Y)
m,n = X.shape
X_new = np.hstack((np.ones((m,1)),X))
X_new,means,stds = featureNormalization(X_new)
n = n+1
```

Intializing theta, learning rate and number of iterations:

```
[18]: initialTheta = np.zeros((n,1))
alpha = 0.01  #Learning rate
iterations = 400
```

Cost when initial parameters are zero.

```
[19]: print("Cost when theta is\n",initialTheta,"\nis:

→",costFunction(initialTheta,X_new,Y))
```

```
Cost when theta is \lceil \lceil 0. \rceil
```

```
[0.]
     [0.1]
    is: 65591548106.45744
       Running Gradient Descent:
[20]: finalTheta = gradientDescent(initialTheta, X_new, Y, alpha, iterations)
     print("Optimized parameters after running gradient Descent are: \n", finalTheta)
    Optimized parameters after running gradient Descent are:
     [[334302.06399328]
     [ 99411.44947359]
     [ 3267.01285407]]
       Cost after running Gradient Descent:
[21]: print("Cost after running gradient Descent is:
      →",costFunction(finalTheta,X_new,Y))
    Cost after running gradient Descent is: 2105448288.6292472
       Predicting cost of the houses using our model:
[22]: print("Predicted price of a 1650 sq-ft, 3 br house (using gradient descent):
      →",predict([1.,1650.,3.],finalTheta,means,stds).item())
    Predicted price of a 1650 sq-ft, 3 br house (using gradient descent):
    289221.5473712181
       Normal equation Method
[23]: def normalEqn(X,y):
         return pinv(X.T@X)@X.T@y
       Loading Data
[24]: data = np.loadtxt("Data/ex1data2.txt", delimiter=',')
     X = data[:,:-1]
     Y = data[:,-1:]
[25]: X = np.matrix(X)
     Y = np.matrix(Y)
     m,n = X.shape
     X_new = np.hstack((np.ones((m,1)),X))
       Applying normal equation to get theta
[26]: theta = normalEqn(X_new,Y)
       Predicting values:
[27]: print("Predicted price of a 1650 sq-ft, 3 br house (using normal eqn): ",[1.
      \rightarrow,1650.,3.]*theta)
    Predicted price of a 1650 sq-ft, 3 br house (using normal eqn):
```

[[293081.46433497]]