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import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

# Step 1 : Fetching the data

df = pd.read_csv("/content/Profit_Dataset.txt")
df.head()
df = pd.read_csv("/content/Profit_Dataset.txt", header=None);
df.head(10)
df.describe()

# # Step 2 : preprocessing the data

x = df.iloc[:,0]
print(x)
m = x.shape[0]
print(m)
x=x.values
type(x)
x= x.reshape(m,1)
x.shape
x
y= df.iloc[:,1]
y= (y.values).reshape(m,1)
y.shape
plt.scatter(x,y,marker='x')
plt.xlabel('Population in lakhs')
plt.ylabel('profit in thousand rs')
plt.title('Food truck profit Estimation')

# # step 3 : construct a model

col1 = np.ones((m,1))
col1
x= np.hstack((col1,x))
print(x)
Theta = np.zeros((2,1))
j=0
alpha = 0.1
print(Theta)

# #hypothesis
h = np.dot(x,Theta)
print(h)
j = np.sum(np.square(h-y))/(2*m)
print(j)

Theta[0]-(alpha/m)*np.sum(h-y)
Theta[1]-(alpha/m)*np.sum((h-y)*(x[:,1].reshape(m,1)))

def computeCost(x,y,Theta):
    m=y.shape[0]
    h=np.dot(x,Theta)
    j=np.sum(np.square(h-y))/(2*m)
    return [h,j]

def gradientDescent(x,y,Theta,alpha):
    m=y.shape[0]
    h=computeCost(x,y,Theta)[0]
    j=computeCost(x,y,Theta)[1]
    Theta[0]=Theta[0]-(alpha/m)*np.sum(h-y)
    Theta[1]=Theta[1]-(alpha/m)*np.sum((h-y)*(x[:,1].reshape(m,1)))
    return [j,Theta]

def trainLinearRegression(x,y,alpha,noIter,printIter):
    Theta = np.zeros((2,1))
    jHistory =[]
    for i in range(noIter):
        j=gradientDescent(x,y,Theta,alpha)[0]
        jHistory.append(j)
        if(i % printIter==0):
            print("iteration =",i)
            print("cost =",j)

    plot1 = plt.figure(1)
    plt.scatter(x[:,1],y,marker='x')
    plt.plot(x,np.dot(x,Theta))

```

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# plt.xlabel('# iteration')
# plt.ylabel("profit in lakh rs")
# plt.title('profit made by a foodtruck')

# plot2 = plt.figure(2)
# plt.plot(list(range(noIter)),jHistory)
# plt.xlabel("# Iteration")
# plt.ylabel("J")
# plt.title("convergence of cost function")

# plt.show()
# return Theta

# Theta = trainLinearRegression(x,y,0.001,20000,1000)
```

```
0      6.1101
1      5.5277
2      8.5186
3      7.0032
4      5.8598
...
92     5.8707
93     5.3054
94     8.2934
95     13.3940
96     5.4369
Name: 0, Length: 97, dtype: float64
97
[[ 1.      6.1101]
 [ 1.      5.5277]
 [ 1.      8.5186]
 [ 1.      7.0032]
 [ 1.      5.8598]
 [ 1.      8.3829]
 [ 1.      7.4764]
 [ 1.      8.5781]
 [ 1.      6.4862]
 [ 1.      5.0546]
 [ 1.      5.7107]
 [ 1.     14.164 ]
 [ 1.      5.734 ]
 [ 1.      8.4084]
 [ 1.      5.6407]
 [ 1.      5.3794]
 [ 1.      6.3654]
 [ 1.      5.1301]
 [ 1.      6.4296]
 [ 1.      7.0708]
 [ 1.      6.1891]
 [ 1.     20.27 ]
 [ 1.      5.4901]
 [ 1.      6.3261]
 [ 1.      5.5649]
 [ 1.     18.945 ]
 [ 1.     12.828 ]
 [ 1.     10.957 ]
 [ 1.     13.176 ]
 [ 1.     22.203 ]
 [ 1.      5.2524]
 [ 1.      6.5894]
 [ 1.      9.2482]
 [ 1.      5.8918]
 [ 1.      8.2111]
 [ 1.      7.9334]
 [ 1.      8.0959]
 [ 1.      5.6063]
 [ 1.     12.836 ]
 [ 1.      6.3534]
 [ 1.      5.4069]
 [ 1.      6.8825]
 [ 1.     11.708 ]
 [ 1.      5.7737]
 [ 1.      7.8247]
```

Double-click (or enter) to edit

```
[ 1.      5.0014]
```

```

# Step 2 : preprocessing the data

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def gradientDescent(x,y,Theta,alpha):
    m=y.shape[0]
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    Theta[0]=Theta[0]-(alpha/m)*np.sum(h-y)
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    return [j,Theta]

def trainLinearRegression(x,y,alpha,noIter,printIter):
    Theta = np.zeros((2,1))
    jHistory =[]
    for i in range(noIter+1):
        j=gradientDescent(x,y,Theta,alpha)[0]
        jHistory.append(j)
        if(i % printIter==0):
            print("iteration =",i)
            print("cost =",j)

    plot1 = plt.figure(1)
    plt.scatter(x[:,1],y,marker='x')
    plt.plot(x,np.dot(x,Theta))
    plt.xlabel('# iteration')
    plt.ylabel("profit in lakh rs")
    plt.title('profit made by a foodtruck')

    plot2 = plt.figure(2)
    plt.plot(list(range(noIter)),jHistory)
    plt.xlabel("# Iteration")
    plt.ylabel("J")
    plt.title("convergence of cost function")

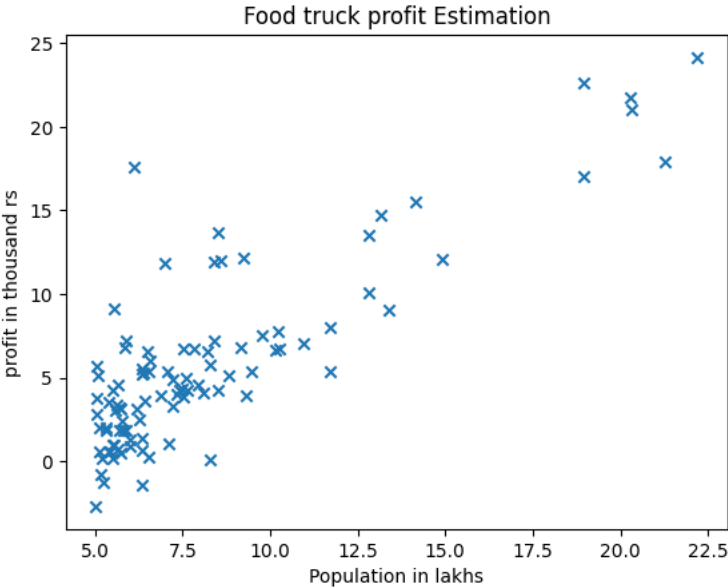
    plt.show()
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```

profit made by a foodtruck

```
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Name: 0, Length: 97, dtype: float64
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Text(0.5, 1.0, 'Food truck profit Estimation')
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