```
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 = 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
# 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))
```

```
# 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)
\supseteq
    0
           6.1101
            5.5277
           8.5186
     3
           7.0032
           5.8598
    4
           5.8707
    92
     93
           5.3054
     94
           8.2934
          13.3940
     96
           5.4369
    Name: 0, Length: 97, dtype: float64
     [[ 1.
               6.1101]
     [ 1.
               5.5277]
               8.5186]
      [ 1.
               7.0032]
               5.8598]
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               8.3829]
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              8.5781]
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      Г1.
      [ 1.
               5.0546]
               5.7107]
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              14.164
      [ 1.
              5.734 ]
8.4084]
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      [ 1.
               5.6407]
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               5.3794
      [ 1.
               6.3654]
      [ 1.
               5.1301]
               6.4296]
      [ 1.
               7.0708]
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               6.1891]
      ſ 1.
      [ 1.
              20.27
               5.4901]
      [ 1.
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               6.3261
      [ 1.
               5.5649
      [ 1.
              18.945
              12.828
      [ 1.
              10.957
              13.176
      [ 1.
      [ 1.
              22.203
               5.2524]
      [ 1.
      [ 1.
               6.5894]
               9.2482]
      [ 1.
               5.8918]
      [ 1.
      [ 1.
               8.2111]
      [ 1.
               7.9334]
               8.0959]
               5.6063]
      [ 1.
              12.836 ]
               6.3534]
      [ 1.
               5.4069]
      Γ1.
               6.8825]
       1.
              11.708
      ſ 1.
               5.7737]
7.8247]
      1.
      [ 1.
Double-click (or enter) to edit
     ι т.
               J.0014]
```

https://colab.research.google.com/drive/1v3M-xnKcwF3Fq7kz2roSSxAkuY0KgRWi?authuser=1#printMode=true

```
# Step 2 : preprocessing the data
x = df.iloc[:,0]
print(x)
m = x.shape[0]
print(m)
type(x)
x= x.reshape(m,1)
x.shape
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))
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)
\label{eq:theta} 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+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)
  \verb|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()
  return Theta
```

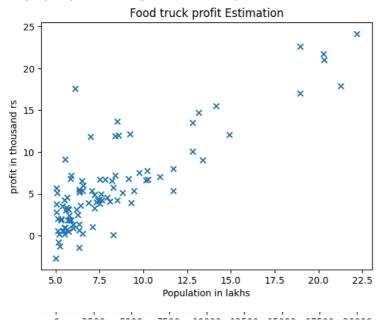
[0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0. [0.] [0.] [0.] [0. [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0. [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.] [0.]] 32.072733877455676 iteration = 0cost = 32.072733877455676 iteration = 1000 cost = 5.480269332020323 iteration = 2000 cost = 5.176562563777922iteration = 3000 cost = 4.964790400326137 iteration = 4000 cost = 4.817123460031757iteration = 5000 cost = 4.714156550214695 iteration = 6000 cost = 4.64235859397521iteration = 7000 cost = 4.592294485716297 iteration = 8000 cost = 4.557385206023154iteration = 9000 cost = 4.533043260244523iteration = 10000 cost = 4.516069827120197 iteration = 11000 cost = 4.5042343956125475 iteration = 12000 cost = 4.495981649220805 iteration = 13000 cost = 4.490227078889981 iteration = 14000cost = 4.486214465619138 iteration = 15000cost = 4.483416504288282 iteration = 16000 cost = 4.481465509492482 iteration = 17000 cost = 4.48010509730549iteration = 18000 cost = 4.479156493381733iteration = 19000 cost = 4.4784950398805305

profit made by a foodtruck

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

```
6.1101
5.5277
0
1
       8.5186
2
3
       7.0032
4
       5.8598
       5.8707
92
93
       5.3054
94
       8.2934
95
      13.3940
96
       5.4369
Name: 0, Length: 97, dtype: float64
```

Text(0.5, 1.0, '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]
{\tt def trainLinearRegression} ({\tt 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()
  return Theta
Theta = trainLinearRegression(x,y,0.001,20000,1000)
```

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) $\label{thm:continuous} Theta[1]=Theta[1]-(alpha/m)*np.sum((h-y)*(x[:,1].reshape(m,1)))$ return [j,Theta] def trainLinearRegression(x,y,alpha,noIter,printIter): Theta = np.zeros((2,1))jHistory =[] for i in range(noIter+1): j=gradientDescent(x,y,Theta,alpha)[0] jHistory.append(j) if(i % printIter==0): print("iteration =",i) print("cost =",j) nlo+1 - nl+ figura/1)