assignment-02

April 2, 2020

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[190]: import torch
       import torch.nn as nn
       import torch.nn.functional as F
       import torch.optim as optim
       import matplotlib.pyplot as plt
       import numpy as np
       import random
       random.seed(100) #generate random seed
       torch.manual_seed(1) #generate random seed
       data_number = 50 #number of data
       x_data = [] #space for x data
       y_data = [] #space for y data
       cost_list = [] #space for cost
       for i in range(data_number): #repeat by data_number
           y_data.append(random.randint(1,1000)) #randomly extract y's from 1 to 999
           x_data.append(int(y_data[i]/2)+random.randint(1,10)) #y/2 is randomly added_
       \rightarrow from 1 to 9
       x_train = torch.FloatTensor([sorted(x_data)]) #X_data input
       y_train = torch.FloatTensor([sorted(y_data)]) #Y_data input
       print(x_train) #print x_train
       print(y_train) #print y_train
      tensor([[ 49., 75., 76., 83., 84., 86., 87., 87., 93., 133., 133., 136.,
               162., 176., 182., 196., 214., 231., 236., 255., 275., 284., 285., 289.,
               316., 316., 341., 341., 355., 363., 368., 371., 377., 381., 391., 402.,
               406., 412., 416., 418., 419., 441., 461., 463., 468., 472., 472., 495.,
               497., 505.]])
      tensor([[ 83., 130., 145., 150., 152., 165., 167., 172., 185., 259., 261., 270.,
               317., 344., 347., 379., 413., 444., 466., 496., 546., 556., 568., 569.,
               618., 624., 668., 675., 699., 707., 723., 732., 735., 750., 780., 786.,
               794., 819., 820., 821., 834., 868., 912., 915., 923., 924., 936., 984.,
               987., 994.]])
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[191]: | W = torch.zeros(1, requires_grad=True) #initialize W to O. requires_grad=True_
        →means variable value that keep changing by learning
       b = torch.zeros(1, requires grad=True) #nitialize W to O. requires grad=True<sub>| 1</sub>
        →means variable value that keep changing by learning
[192]: optimizer = optim.SGD([W, b], lr=1e-7) #pytorch optimizer library call
       nb_epochs = 1000 # repeat gradient descent as many as we want.
       for epoch in range(nb_epochs + 1): #repeat nb_epochs times
           \# H(x)
           hypothesis = x_train * W + b
           # cost = objective function
           cost = torch.mean((hypothesis - y train) ** 2)/2
           cost list.append(cost.item())
           # Gradient Descent algorithm
           optimizer.zero_grad()
           cost.backward()
           optimizer.step()
           # print functions every hundred times
           if epoch % 100 == 0:
               print('Epoch {:4d}/{} W: {:.3f}, b: {:.3f} Cost: {:.6f}'.format(
                   epoch, nb_epochs, W.item(), b.item(), cost.item()
               ))
               0/1000 W: 0.021, b: 0.000 Cost: 203117.515625
      Epoch
      Epoch 100/1000 W: 1.289, b: 0.004 Cost: 24733.380859
      Epoch 200/1000 W: 1.732, b: 0.005 Cost: 3028.404053
      Epoch 300/1000 W: 1.886, b: 0.005 Cost: 387.437347
      Epoch 400/1000 W: 1.940, b: 0.005 Cost: 66.095879
      Epoch 500/1000 W: 1.959, b: 0.005 Cost: 26.996191
      Epoch 600/1000 W: 1.966, b: 0.005 Cost: 22.239185
      Epoch 700/1000 W: 1.968, b: 0.005 Cost: 21.660322
      Epoch 800/1000 W: 1.969, b: 0.005 Cost: 21.589867
      Epoch 900/1000 W: 1.969, b: 0.005 Cost: 21.581261
      Epoch 1000/1000 W: 1.969, b: 0.005 Cost: 21.580193
[193]: Hy = [] #value of y that predicted by x_{a} data
       for i in x_data:
           Hy.append(int(W.item()*i+b.item())) #put value in Hy (value of y that ∪
       \rightarrow predicted by x_{data}
       print(x_data,end="\n\n")
       print(y_data,end="\n\n")
       print(Hy)
```

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[83, 236, 368, 381, 231, 412, 275, 49, 136, 341, 176, 162, 495, 76, 463, 196, 497, 214, 289, 468, 84, 93, 316, 86, 87, 87, 472, 505, 416, 75, 285, 316, 377, 406, 371, 363, 182, 355, 284, 255, 391, 419, 461, 341, 133, 441, 402, 418, 472, 133]

[150, 466, 723, 750, 444, 820, 546, 83, 270, 675, 344, 317, 984, 145, 915, 379, 987, 413, 569, 923, 165, 185, 618, 152, 167, 172, 936, 994, 819, 130, 568, 624, 735, 794, 732, 707, 347, 699, 556, 496, 780, 834, 912, 668, 259, 868, 786, 821, 924, 261]

[163, 464, 724, 750, 454, 811, 541, 96, 267, 671, 346, 318, 974, 149, 911, 385, 978, 421, 569, 921, 165, 183, 622, 169, 171, 171, 929, 994, 819, 147, 561, 622, 742, 799, 730, 714, 358, 698, 559, 502, 769, 825, 907, 671, 261, 868, 791, 823, 929, 261]
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