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In [1]: #how to use google colab: https://pytorch.org/tutorials/beginner/co  
#pytorch tutorial: https://pytorch.org/tutorials/beginner/deep_lear  
import torch  
import torch.nn.functional as F  
from torchvision import datasets, transforms  
import numpy as np  
import matplotlib.pyplot as plt
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In [2]: # define the network structure  
class fc_net(torch.nn.Module):  
    def __init__(self, num_in, num_out):  
        super(fc_net, self).__init__()  
        # Initialize two linear neural networks, one as the input layer &  
        self.h1 = torch.nn.Linear(in_features=num_in, out_features=256)  
        self.h2 = torch.nn.Linear(in_features=256, out_features=num_out)  
    def forward(self, inputs):  
        # We use relu as the activation function as the first layer  
        a1 = F.relu(self.h1(inputs))  
        # We use softmax as the activation function as the second layer  
        a2 = F.softmax(self.h2(a1),dim=-1)  
        return a2
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In [3]: # use data_loader to load_in data  
train_data = datasets.MNIST('./', train=True, download=True, transf  
# We use DataLoader to load the train data.  
# We specify the batch_size and the DataLoader will return the spli  
# We use shuffle = False so we won't shuffle the data.  
train_loader = torch.utils.data.DataLoader(train_data, batch_size=1  
batch_size = 10  
# Model is a fully-connected net.  
model = fc_net(num_in=28*28, num_out=10)  
# We use Cross Entropy as our loss.  
loss = torch.nn.CrossEntropyLoss()  
# We use SGD as our optimizer  
optimizer = torch.optim.SGD(model.parameters(), lr=0.1)
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In [4]: epoch = 10
epoch_accuracies = []
for j in range (epoch):
    loader = iter(train_loader)
    epoch_accuracy = 0.0
    print (j)
    for i in range (1, len(train_loader)):
        # cur_x, cur_y gets the data and label for the next iteration
        cur_x, cur_y = next(loader)
        # We unsqueeze the data for one batch.
        cur_x = torch.reshape(cur_x, (10, 28*28))
        # We get the predictions, which are probabilities, after the forward pass
        preds = model.forward(cur_x)
        # Use the predictions and labels to compute loss.
        cur_loss = loss(preds, cur_y)
        optimizer.zero_grad()
        # Backward
        cur_loss.backward()
        optimizer.step()
        preds_numpy = preds.detach().numpy()
        preds_label = np.argmax(preds_numpy, axis=1)
        cur_y_numpy = cur_y.detach().numpy()
        acc_iter = np.sum(1*(preds_label==(cur_y_numpy))/batch_size)
        epoch_accuracy += acc_iter
    epoch_accuracy = epoch_accuracy/len(train_loader)
    epoch_accuracies.append(epoch_accuracy)
    print (epoch_accuracy)
#         new_preds = model.forward(cur_x)
#         print(new_preds[0])

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0
0.86046666666666702
1
0.93426666666666593
2
0.94923333333333249
3
0.95863333333333246
4
0.9654499999999991
5
0.97028333333333245
6
0.97433333333333243
7
0.97716666666666581
8
0.97981666666666586
9
[5. 0. 4. 1. 0. 0. 1. 0. 1. 4.]

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