# **Project 03**

# **Ece 763 Computer Vision**

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## Aim:

 Use Neural Network including FeedForward NN and LeNet5, to do face and nonface images classification

Babysitting the Process and tuning the Hyperparameters

Platform: Google Colabs (GPU), Jupiter Notebook

Data Preparation: Load the dataset in the Google Drive to a folder named "Data".

This folder should contain your "face" folder and "nonface" folder having images for training and testing.

After Mounting the Google Drive in Google Colab's Jupiter Notebook, give the path of the "Data" folder to the Variable named as "Train\_PATH".

I have taken 90% of Data as the Training set and the rest 10% as the test set.

## code and kesuits:

## **MOUNT THE DRIVE**

```
from google.colab import drive
drive.mount('/content/drive')
```

Go to this URL in a browser: <a href="https://accounts.google.com/o/oauth2/auth?client\_id=947318989803-6bn6qk8qdgf4n4g3pfee6491hc0brc4i.a">https://accounts.google.com/o/oauth2/auth?client\_id=947318989803-6bn6qk8qdgf4n4g3pfee6491hc0brc4i.a</a>
Enter your authorization code:
............

Mounted at /content/drive

```
from future import print function
import os
import torch
from PIL import Image
import matplotlib.image as mpimg
import matplotlib.pyplot as plt
import torch
import torch.nn as nn
import torch.nn.functional as F
import torch.optim as optim
import torch.utils.data as data
import torchvision.transforms as transforms
import torchvision.models as models
import torch.utils.data
import torchvision.datasets
from torch.autograd import Variable
```

```
import copy
import time
import numpy as np
import os
```

## **LOAD IMAGES**

```
TRAIN PATH = '/content/drive/My Drive/Data'
loader1=transforms.ToTensor()# if not normalize then in range[0,1]
#normalization for simple feedforward neural network
loader2=transforms.Compose(
    [transforms.ToTensor(),#convert an image to tensor
    transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))])
#without normalization for LeNet5
loader3=transforms.Compose(
        [transforms.Resize((32,32)),
         transforms.ToTensor()])
#normalization for LeNet5
loader4=transforms.Compose(
        [transforms.Resize((32,32)),
         transforms.ToTensor(),
         transforms.Normalize((0.5,0.5,0.5),(0.5,0.5,0.5))
def load images flow(batch size,which trans):
    if(which trans == 1):
         transform = loader1
    if(which trans == 2):
         transform = loader2
    if(which trans == 3):
         transform = loader3
   if(which trans == 4):
         transform = loader4
```

#### **OPTIMIZER**

```
def createLossAndOptimizer(net, learning_rate = 0.001, weight_decay = 0, loss_method = "SGD"):
    #weight_decay: tuning parameter of L2 term
    #Loss function
    loss = torch.nn.CrossEntropyLoss()

#Optimizer
    if loss_method == "SGD":
        optimizer = optim.SGD(net.parameters(), lr=learning_rate, weight_decay=weight_decay)
    if loss_method == "Adam":
        optimizer = optim.Adam(net.parameters(), lr=learning_rate, weight_decay=weight_decay)
    return(loss, optimizer)
```

## **Feedforward Neural Network and LeNet5**

```
input_size = 3*60*60; hidden_size = 50; output_size = 2
class Two_Layers_NN(torch.nn.Module):

    def __init__(self):
       super(Two_Layers_NN,self).__init__()
```

```
self.fc1 = nn.Linear(input_size, hidden_size)
        self.fc2 = nn.Linear(hidden size, output size)
   def forward(self, x):
       out = self.fc1(x)
       out = self.fc2(out)
        return out
class LeNet(torch.nn.Module):
    def init (self):
        super(LeNet,self). init ()
        self.conv1 = torch.nn.Conv2d(3, 6, kernel_size=(5,5), stride=1)#input 3 channels, output 6 channels
        self.maxpool = nn.MaxPool2d(kernel size=(2,2),stride=2)
        # torch.nn.AdaptiveAvgPool2d() can avoid overfitting
        self.conv2 = torch.nn.Conv2d(6, 16, kernel size=(5,5), stride=1)#output may not be the times of input
        self.fc1 = torch.nn.Linear(5*5*16,120)
        self.fc2 = torch.nn.Linear(120,84)
        self.fc3 = torch.nn.Linear(84,2)
   def forward(self, x):
       out = self.conv1(x) #input 32*32*3 output 28*28*6
        #x = self.batchnorm1(x)
       out = self.maxpool(out) #output 14*14*6
```

```
out = self.conv2(out) #output 10*10*16

out = self.maxpool(out) #output 5*5*16

out = out.view(-1, 5 * 5 * 16)#flatten

out = self.fc1(out)

out = self.fc2(out)

out = self.fc3(out)

return(out)
```

#### TRAINING FUNCTION

```
def train net(net, loss method, which model, whether norm, batch size, n epochs, learning rate,
              weight decay, print train process = True, print test process= True, validation = True):
    assert(which model == "NN" or which model == "LeNet")# the input should be reasonable
    #choose right transformation
    if (whether norm == False and which model == "NN"):
       which trans = 1
    elif (whether norm == True and which model == "NN"):
        which trans = 2
    elif (whether norm == False and which model == "LeNet"):
        which trans = 3
    elif (whether norm == True and which model == "LeNet"):
        which trans = 4
    print(which_trans)
    #Get training data and test data
    train_loader,test_loader = load_images_flow(batch_size,which_trans)
    n_batches = len(train_loader)
```

```
loss, optimizer = createLossAndOptimizer(net, learning rate = learning rate,
                                          weight decay = weight_decay,loss method=loss method)
#Time for printing
start time = time.time()
print every = n batches // 10
for epoch in range(n epochs):
    \#epoch = 0
    running loss = 0
    running correct num = 0
    total train loss = 0
    total train num = 0
    total correct train num = 0
    for i, data in enumerate(train loader): # handle every batch size pictures
        (inputs,labels) = data
        if(which model == "NN"):
             inputs = inputs.view(inputs.size()[0],3*60*60)#flatten
             #inputs = inputs.view(batch size,3*60*60) not batch size since
        inputs, labels = Variable(inputs), Variable(labels)
        optimizer.zero grad() # whether zero setting is okay ?
        #print(inputs.size())
        #Forward pass, backward pass, optimize
        outputs = net(inputs) #why ? same as forward
        m, predicted = torch.max(outputs.data,1)
        total train num += labels.size(0)
        running correct num += (predicted == labels).sum().item()
        total_correct_train_num += (predicted == labels).sum().item()
        loss_size = loss(outputs, labels)
        if np.isnan(loss size.data):
```

```
raise ValueError("loss explode due to large regularization or learning rate")
   loss size.backward()
   optimizer.step()
   #print statistics
   running loss += loss size.data
   total train loss += loss size.data
   #print every 10th batch
   if(print train process == True):
         if (i+1) % (print every) == 0:
                print("Epoch {}, {:d}% \t train loss: {:.4f} train accuracy:{:d}% took: {:.2f}s".format(
                     epoch+1, int(100*(i+1)/n batches), running loss/print every/batch size, int(100 * running correct num /p
                     time.time()-start_time)) # loss for currect running loss and running correct num not accumulated ones
                #reset running loss and time
                running loss = 0.0
                running correct num = 0.0
                start time = time.time()
#For validation
if(validation == True):
   total test loss = 0
   total test num = 0
   correct test num = 0
   for inputs, labels in test loader:
        if (which model == "NN"):
            inputs = inputs.view(inputs.size()[0], 3*60*60)
            inputs, labels = Variable(inputs), Variable(labels)
       #Forward pass
       test_outputs = net(inputs)
       #test accuracy rate
       m, predicted = torch.max(test_outputs.data, 1)
       #print(predicted)
```

## First We do for FEED Forward Neural Network and then LeNet5

## Compare the Results with and without normalization

C→

```
weight decay:0.0000 learning rate:0.0010
Training Examples - 3003
Validation Examples - 334
               train loss: 0.1263 train accuracy:66% took: 0.50s
Epoch 1, 9%
Epoch 1, 19% train loss: 0.1034 train accuracy:77% took: 0.45s
              train loss: 0.1013 train accuracy:78% took: 0.43s
Epoch 1, 29%
               train loss: 0.0852 train accuracy:84% took: 0.42s
Epoch 1, 39%
Epoch 1, 49%
               train loss: 0.0790 train accuracy:86% took: 0.44s
                train loss: 0.0855 train accuracy:83% took: 0.43s
Epoch 1, 59%
                train loss: 0.0730 train accuracy:89% took: 0.42s
Epoch 1, 69%
               train loss: 0.0818 train accuracy:83% took: 0.44s
Epoch 1, 79%
Epoch 1, 89%
                train loss: 0.0748 train accuracy:85% took: 0.43s
Epoch 1, 99%
                train loss: 0.0690 train accuracy:86% took: 0.43s
```

Without normalization, optimization converge slow.

\*Sanity Check : \*

Varying Regularisation and keeping learning rate same

C→

```
weight decay:0.0000 learning rate:0.0010
Training Examples - 3003
Validation Examples - 334
Epoch 1, 9%
                train loss: 0.1070 train accuracy:74% took: 0.53s
Epoch 1, 19%
                train loss: 0.0989 train accuracy:78% took: 0.46s
Epoch 1, 29%
                train loss: 0.0840 train accuracy:82% took: 0.47s
                train loss: 0.0940 train accuracy:80% took: 0.48s
Epoch 1, 39%
Epoch 1, 49%
                train loss: 0.0765 train accuracy:84% took: 0.46s
                train loss: 0.0869 train accuracy:79% took: 0.47s
Epoch 1, 59%
                train loss: 0.0802 train accuracy:81% took: 0.45s
Epoch 1, 69%
                train loss: 0.0847 train accuracy:80% took: 0.46s
Epoch 1, 79%
Epoch 1, 89%
                train loss: 0.0743 train accuracy:84% took: 0.47s
                train loss: 0.0767 train accuracy:81% took: 0.49s
Epoch 1, 99%
weight decay:1000.0000 learning rate:0.0010
Training Examples - 3003
Validation Examples - 334
Epoch 1, 9%
               train loss: 0.1387 train accuracy:51% took: 0.59s
                train loss: 0.1386 train accuracy:51% took: 0.45s
Epoch 1, 19%
Epoch 1, 29%
                train loss: 0.1386 train accuracy:44% took: 0.45s
                train loss: 0.1386 train accuracy:49% took: 0.45s
Epoch 1, 39%
                train loss: 0.1386 train accuracy:48% took: 0.47s
Epoch 1, 49%
Epoch 1, 59%
                train loss: 0.1386 train accuracy:50% took: 0.47s
                train loss: 0.1386 train accuracy:44% took: 0.45s
Epoch 1, 69%
                train loss: 0.1386 train accuracy:52% took: 0.49s
Epoch 1, 79%
Epoch 1, 89%
               train loss: 0.1386 train accuracy:52% took: 0.55s
                train loss: 0.1386 train accuracy:51% took: 0.51s
Epoch 1, 99%
weight decay:10000.0000 learning rate:0.0010
Training Examples - 3003
Validation Examples - 334
ValueError
                                         Traceback (most recent call last)
<ipython-input-67-d241aabf54ac> in <module>()
    22 NN = Two Layers NN()
    23 train net(NN, which model = "NN", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
---> 24
                     weight decay = weight decay, print train process = True, print test process = False, validation = False, 1
    25
    26 del NN
```

<invthon-input-65-6a79802d62e0> in train net(net. loss method. which model. whether norm. hatch size. n enochs. learning rate. which model. whether norm. hatch size. n enochs. learning rate. which model. whether norm. hatch size. n enochs. learning rate. which model. whether norm. hatch size. n enochs. learning rate. which model.

Case 1: As seen, when we increase the weight decay / regularization from 0 to 1000, Loss also increases.

Case 2: when Regularization is 10000, Loss EXPLODES.

Now, Varying the Learning Rate and keeping the Regularisation constant at 0.001.

```
weight decay = 0.001; learning rate = 0.000001
print("weight decay:{:.4f} learning rate:{:.8f}".format(weight decay, learning rate))
NN = Two Layers NN()
train net(NN, which model = "NN", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
              weight decay = weight decay, print train process = True, print test process = False, validation = False, loss method =
del NN
weight decay = 0.001; learning rate = 0.001
print("weight decay:{:.4f} learning rate:{:.8f}".format(weight decay, learning rate))
NN = Two Layers NN()
train net(NN, which model = "NN", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
              weight_decay = weight_decay, print_train_process = True, print test process = False, validation = False, loss method =
del NN
weight decay = 0.001; learning rate = 0.1
print("weight decay:{:.4f} learning rate:{:.8f}".format(weight_decay, learning_rate))
NN = Two_Layers_NN()
train net(NN, which model = "NN", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
```

weight\_aceay - weight\_aceay, print\_crain\_process - rrac, print\_cest\_process - raise, variaacion - raise, ioss\_methoa del NN ₽

```
weight decay:0.0010 learning rate:0.00000100
Training Examples - 3003
Validation Examples - 334
                train loss: 0.1407 train accuracy:49% took: 0.50s
Epoch 1, 9%
Epoch 1, 19%
                train loss: 0.1408 train accuracy:50% took: 0.46s
                train loss: 0.1406 train accuracy:48% took: 0.46s
Epoch 1, 29%
Epoch 1, 39%
                train loss: 0.1407 train accuracy:49% took: 0.47s
Epoch 1, 49%
                 train loss: 0.1386 train accuracy:52% took: 0.44s
                train loss: 0.1368 train accuracy:56% took: 0.48s
Epoch 1, 59%
                train loss: 0.1381 train accuracy:52% took: 0.46s
Epoch 1, 69%
                train loss: 0.1385 train accuracy:52% took: 0.46s
Epoch 1, 79%
Epoch 1, 89%
                train loss: 0.1373 train accuracy:55% took: 0.47s
               train loss: 0.1336 train accuracy:60% took: 0.45s
Epoch 1, 99%
weight decay:0.0010 learning rate:0.00100000
Training Examples - 3003
Validation Examples - 334
Epoch 1, 9%
               train loss: 0.1035 train accuracy:77% took: 0.49s
                train loss: 0.0921 train accuracy:79% took: 0.43s
Epoch 1, 19%
Epoch 1, 29%
                 train loss: 0.0851 train accuracy:81% took: 0.44s
                train loss: 0.0778 train accuracy:83% took: 0.47s
Epoch 1, 39%
                train loss: 0.0889 train accuracy:79% took: 0.50s
Epoch 1, 49%
Epoch 1, 59%
                train loss: 0.0711 train accuracy:86% took: 0.51s
                train loss: 0.0656 train accuracy:86% took: 0.47s
Epoch 1, 69%
                train loss: 0.0731 train accuracy:84% took: 0.44s
Epoch 1, 79%
Epoch 1, 89%
               train loss: 0.0653 train accuracy:85% took: 0.48s
                train loss: 0.0629 train accuracy:86% took: 0.45s
Epoch 1, 99%
weight decay:0.0010 learning rate:0.10000000
Training Examples - 3003
Validation Examples - 334
Epoch 1, 9%
             train loss: 4606958574909423059699256174575616.0000 train accuracy:44% took: 0.47s
ValueError
                                         Traceback (most recent call last)
<ipython-input-68-64bd5c402aad> in <module>()
     18 NN = Two Layers NN()
    19 train net(NN, which model = "NN", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
---> 20
                     weight decay = weight decay, print train process = True, print test process = False, validation = False, 1
     21
     22 del NN
```

Case 1: When learning rate is 0.000001(too small), Loss barely changes.

Case 2: When learning rate is 0.001, Loss changes reasonably.

**Case 3: When learning rate is 0.1, Loss EXPLODES.** 

\*After this, we will do \*

Hyperparameter optimization (random search)

```
Training Examples - 3003
Validation Examples - 334
Test loss = 0.5192 Test Accuracy = 80%
train loss: 0.10345407 train accuracy:79% learning rate:0.00000521 regularization:0.00275003 running time:25.9373
Training Examples - 3003
Validation Examples - 334
Test loss = 0.2479 Test Accuracy = 90%
train loss: 0.05355413 train accuracy:89% learning rate:0.00564785 regularization:0.21250154 running time:26.1331
Training Examples - 3003
Validation Examples - 334
Test loss = 0.1921 Test Accuracy = 93%
train loss: 0.03042419 train accuracy:94% learning rate:0.00264795 regularization:0.00263606 running time:25.3011
Training Examples - 3003
Validation Examples - 334
Test loss = 0.1610 Test Accuracy = 93%
train loss: 0.02487139 train accuracy:95% learning rate:0.00553499 regularization:0.00002068 running time:25.4065
Training Examples - 3003
Validation Examples - 334
Test loss = 0.5057 Test Accuracy = 78%
train loss: 0.10146198 train accuracy:79% learning rate:0.00000875 regularization:0.03200406 running time:25.8074
Training Examples - 3003
Validation Examples - 334
Test loss = 0.1772 Test Accuracy = 91%
train loss: 0.03561469 train_accuracy:93% learning rate:0.00478636 regularization:0.06362139 running time:25.3221
Training Examples - 3003
Validation Examples - 334
Test loss = 0.2202 Test Accuracy = 90%
train loss: 0.03931881 train accuracy:93% learning rate:0.00145240 regularization:0.00028463 running time:25.2512
Training Examples - 3003
Validation Examples - 334
Test loss = 0.2494 Test Accuracy = 88%
train loss: 0.02351312 train_accuracy:95% learning rate:0.02027592 regularization:0.00003782 running time:25.4330
Training Examples - 3003
```

```
Validation Examples - 334
Test loss = 0.1624 Test Accuracy = 95%
train_loss: 0.03653247 train_accuracy:94% learning rate:0.03080337 regularization:0.00443945 running time:25.4879
2
Training Examples - 3003
Validation Examples - 334
Test loss = 0.3335 Test Accuracy = 86%
train_loss: 0.05732502 train_accuracy:88% learning rate:0.00033360 regularization:0.00914354 running time:25.7065
```

When Learning Rate: 0.03080337 and Regularization: 0.00443945,

then train loss is the smallest and accuracy is largest

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```
Training Examples - 3003
Validation Examples - 334
                train_loss: 0.1209 train_accuracy:71% took: 0.54s
Epoch 1, 9%
Epoch 1, 19%
                 train loss: 0.0883 train accuracy:80% took: 0.45s
Epoch 1, 29%
                 train loss: 0.0822 train accuracy:82% took: 0.47s
Epoch 1, 39%
                 train loss: 0.0725 train accuracy:85% took: 0.50s
                 train loss: 0.0618 train accuracy:86% took: 0.48s
Epoch 1, 49%
Epoch 1, 59%
                 train loss: 0.0592 train accuracy:87% took: 0.47s
                 train loss: 0.0637 train accuracy:87% took: 0.46s
Epoch 1, 69%
                 train loss: 0.0506 train accuracy:91% took: 0.46s
Epoch 1, 79%
                 train loss: 0.0358 train accuracy:93% took: 0.46s
Epoch 1, 89%
Epoch 1, 99%
                 train loss: 0.0477 train accuracy:89% took: 0.48s
Test loss = 0.1688 Test Accuracy = 94%
                 train loss: 0.0412 train accuracy:92% took: 0.99s
Epoch 2, 9%
                 train loss: 0.0345 train accuracy:92% took: 0.44s
Epoch 2, 19%
Epoch 2, 29%
                 train loss: 0.0540 train accuracy:90% took: 0.48s
                 train loss: 0.0391 train accuracy:94% took: 0.48s
Epoch 2, 39%
Epoch 2, 49%
                 train loss: 0.0532 train accuracy:91% took: 0.47s
                 train loss: 0.0432 train accuracy:92% took: 0.47s
Epoch 2, 59%
                 train loss: 0.0732 train accuracy:88% took: 0.47s
Epoch 2, 69%
                 train loss: 0.0443 train accuracy:91% took: 0.47s
Epoch 2, 79%
Epoch 2, 89%
                 train loss: 0.0421 train accuracy:93% took: 0.48s
Epoch 2, 99%
                 train loss: 0.0372 train accuracy:92% took: 0.47s
Test loss = 0.1781 Test Accuracy = 93%
                 train loss: 0.0291 train accuracy:94% took: 1.00s
Epoch 3, 9%
Epoch 3, 19%
                 train loss: 0.0320 train accuracy:93% took: 0.50s
Epoch 3, 29%
                 train loss: 0.0666 train accuracy:89% took: 0.46s
                 train loss: 0.0515 train accuracy:89% took: 0.45s
Epoch 3, 39%
                 train loss: 0.0240 train accuracy:96% took: 0.47s
Epoch 3, 49%
Epoch 3, 59%
                 train loss: 0.0887 train accuracy:87% took: 0.47s
                 train loss: 0.0601 train accuracy:89% took: 0.46s
Epoch 3, 69%
Epoch 3, 79%
                 train loss: 0.0325 train accuracy:95% took: 0.52s
                 train loss: 0.0356 train accuracy:93% took: 0.53s
Epoch 3, 89%
                 train loss: 0.0234 train accuracy:95% took: 0.47s
Epoch 3, 99%
Test loss = 0.1506 Test Accuracy = 94%
Epoch 4, 9%
                 train loss: 0.0257 train accuracy:95% took: 1.00s
Epoch 4, 19%
                train loss: 0.0213 train accuracy:95% took: 0.46s
Epoch 4, 29%
                 train loss: 0.0313 train accuracy:93% took: 0.46s
Epoch 4, 39%
                 train loss: 0.0234 train accuracy:95% took: 0.48s
Epoch 4, 49%
                 train loss: 0.3182 train accuracy:84% took: 0.49s
Epoch 4, 59%
                 train loss: 0.0260 train accuracy:94% took: 0.52s
```

```
train loss: 0.0306 train accuracy:94% took: 0.49s
Epoch 4, 69%
Epoch 4, 79%
                 train loss: 0.1254 train accuracy:85% took: 0.45s
                train loss: 0.0299 train accuracy:93% took: 0.48s
Epoch 4, 89%
                 train loss: 0.0273 train accuracy:95% took: 0.47s
Epoch 4, 99%
Test loss = 0.1659 Test Accuracy = 93%
Epoch 5, 9%
                 train loss: 0.0226 train accuracy:95% took: 0.98s
Epoch 5, 19%
                 train loss: 0.0202 train accuracy:95% took: 0.45s
Epoch 5, 29%
                 train loss: 0.0290 train accuracy:94% took: 0.45s
Epoch 5, 39%
                 train loss: 0.0260 train accuracy:94% took: 0.46s
Epoch 5, 49%
                 train loss: 0.0335 train accuracy:94% took: 0.49s
Epoch 5, 59%
                 train loss: 0.0256 train accuracy:94% took: 0.50s
Epoch 5, 69%
                train loss: 0.0307 train accuracy:93% took: 0.47s
                 train loss: 0.0171 train accuracy:98% took: 0.46s
Epoch 5, 79%
Epoch 5, 89%
                 train loss: 0.0291 train accuracy:95% took: 0.48s
Epoch 5, 99%
                 train loss: 0.0291 train accuracy:94% took: 0.45s
Test loss = 0.1645 Test Accuracy = 95%
                 train loss: 0.0209 train accuracy:96% took: 0.97s
Epoch 6, 9%
                 train loss: 0.0223 train accuracy:96% took: 0.43s
Epoch 6, 19%
Epoch 6, 29%
                 train loss: 0.0262 train accuracy:94% took: 0.48s
                 train loss: 0.0262 train accuracy:95% took: 0.46s
Epoch 6, 39%
Epoch 6, 49%
                 train loss: 0.0248 train accuracy:95% took: 0.46s
                 train loss: 0.0341 train accuracy:94% took: 0.45s
Epoch 6, 59%
Epoch 6, 69%
                 train loss: 0.0260 train accuracy:94% took: 0.48s
Epoch 6, 79%
                 train loss: 0.0486 train accuracy:90% took: 0.48s
                 train loss: 0.0270 train accuracy:95% took: 0.50s
Epoch 6, 89%
Epoch 6, 99%
                 train loss: 0.0209 train accuracy:96% took: 0.50s
Test loss = 0.1476 Test Accuracy = 95%
Epoch 7, 9%
                 train loss: 0.0147 train accuracy:98% took: 0.96s
                train loss: 0.0155 train_accuracy:97% took: 0.47s
Epoch 7, 19%
                 train loss: 0.0284 train accuracy:93% took: 0.47s
Epoch 7, 29%
                 train loss: 0.0362 train accuracy:91% took: 0.46s
Epoch 7, 39%
                 train loss: 0.0187 train accuracy:97% took: 0.47s
Epoch 7, 49%
Epoch 7, 59%
                 train loss: 0.0163 train accuracy:96% took: 0.47s
Epoch 7, 69%
                 train loss: 0.0297 train accuracy:95% took: 0.52s
                 train loss: 0.0268 train accuracy:95% took: 0.46s
Epoch 7, 79%
Epoch 7, 89%
                 train loss: 0.0184 train accuracy:96% took: 0.52s
Epoch 7, 99%
                 train loss: 0.0362 train accuracy:93% took: 0.47s
Test loss = 0.1555 Test Accuracy = 94%
Epoch 8, 9%
                 train loss: 0.0231 train accuracy:96% took: 0.98s
Epoch 8, 19%
                train loss: 0.0229 train accuracy:96% took: 0.44s
Epoch 8, 29%
                 train loss: 0.0281 train accuracy:94% took: 0.47s
Epoch 8, 39%
                 train loss: 0.0159 train accuracy:97% took: 0.54s
```

```
Epoch 8, 49%
                 train loss: 0.0247 train accuracy:94% took: 0.55s
                 train loss: 0.0255 train accuracy:96% took: 0.51s
Epoch 8, 59%
Epoch 8, 69%
                 train loss: 0.0197 train accuracy:95% took: 0.50s
Epoch 8, 79%
                 train loss: 0.0181 train accuracy:96% took: 0.46s
                train loss: 0.0315 train accuracy:93% took: 0.45s
Epoch 8, 89%
                 train loss: 0.0163 train accuracy:97% took: 0.48s
Epoch 8, 99%
Test loss = 0.1663 Test Accuracy = 95%
Epoch 9, 9%
                 train loss: 0.0161 train accuracy:96% took: 0.97s
                train loss: 0.0227 train accuracy:96% took: 0.47s
Epoch 9, 19%
                 train loss: 0.0183 train accuracy:97% took: 0.46s
Epoch 9, 29%
                 train loss: 0.0164 train accuracy:96% took: 0.47s
Epoch 9, 39%
Epoch 9, 49%
                train loss: 0.0269 train accuracy:95% took: 0.48s
Epoch 9, 59%
                 train loss: 0.0217 train accuracy:94% took: 0.47s
                train loss: 0.0222 train accuracy:96% took: 0.50s
Epoch 9, 69%
                train loss: 0.0332 train accuracy:94% took: 0.50s
Epoch 9, 79%
Epoch 9, 89%
                train loss: 0.0310 train accuracy:93% took: 0.47s
                 train loss: 0.0176 train accuracy:96% took: 0.48s
Epoch 9, 99%
Test loss = 0.1589 Test Accuracy = 95%
                train loss: 0.0209 train accuracy:95% took: 1.03s
Epoch 10, 9%
                 train loss: 0.0295 train accuracy:94% took: 0.47s
Epoch 10, 19%
                 train loss: 0.0178 train accuracy:96% took: 0.45s
Epoch 10, 29%
                 train loss: 0.0214 train accuracy:95% took: 0.48s
Epoch 10, 39%
Epoch 10, 49%
                 train loss: 0.0146 train accuracy:97% took: 0.46s
                 train loss: 0.0161 train accuracy:97% took: 0.50s
Epoch 10, 59%
                 train loss: 0.0193 train accuracy:97% took: 0.47s
Epoch 10, 69%
                train loss: 0.0156 train accuracy:96% took: 0.47s
Epoch 10, 79%
                train loss: 0.0302 train accuracy:96% took: 0.47s
Epoch 10, 89%
Epoch 10, 99%
               train loss: 0.0274 train accuracy:95% took: 0.47s
Test loss = 0.1383 Test Accuracy = 95%
```

## **Sanity Check to LENET5**

С→

```
weight decay:0.0000 learning rate:0.00100000
Training Examples - 3003
Validation Examples - 334
                train loss: 0.1388 train accuracy:46% took: 0.62s
Epoch 1, 9%
Epoch 1, 19%
                train loss: 0.1375 train accuracy:56% took: 0.55s
                train loss: 0.1363 train accuracy:78% took: 0.53s
Epoch 1, 29%
                train loss: 0.1345 train accuracy:82% took: 0.53s
Epoch 1, 39%
Epoch 1, 49%
                train loss: 0.1332 train accuracy:80% took: 0.53s
                train loss: 0.1303 train accuracy:83% took: 0.54s
Epoch 1, 59%
                train loss: 0.1286 train accuracy:80% took: 0.54s
Epoch 1, 69%
                train loss: 0.1263 train accuracy:78% took: 0.56s
Epoch 1, 79%
Epoch 1, 89%
                train loss: 0.1194 train accuracy:80% took: 0.56s
               train loss: 0.1137 train accuracy:78% took: 0.63s
Epoch 1, 99%
weight decay:0.0010 learning rate:0.00100000
Training Examples - 3003
Validation Examples - 334
Epoch 1, 9%
               train loss: 0.1397 train accuracy:53% took: 0.60s
Epoch 1, 19%
                train loss: 0.1391 train accuracy:49% took: 0.52s
Epoch 1, 29%
                train loss: 0.1374 train accuracy:55% took: 0.54s
                train loss: 0.1368 train accuracy:59% took: 0.56s
Epoch 1, 39%
                train loss: 0.1355 train accuracy:72% took: 0.58s
Epoch 1, 49%
Epoch 1, 59%
                train loss: 0.1339 train accuracy:75% took: 0.57s
                train loss: 0.1310 train accuracy:82% took: 0.57s
Epoch 1, 69%
                train loss: 0.1308 train accuracy:77% took: 0.53s
Epoch 1, 79%
               train loss: 0.1282 train accuracy:76% took: 0.53s
Epoch 1, 89%
                train loss: 0.1264 train accuracy:74% took: 0.53s
Epoch 1, 99%
weight decay:10000.0000 learning rate:0.00100000
Training Examples - 3003
Validation Examples - 334
ValueError
                                         Traceback (most recent call last)
<ipython-input-71-891dac2a2eb1> in <module>()
    19 LN = LeNet()
    20 train net(LN, which model = "LeNet", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
---> 21
                     weight decay = weight decay, print train process = True, print test process = False, validation = False, lo
    22
    23 del LN
```

<invthon-input-65-6a79802d62e0> in train net(net. loss method. which model. whether norm. batch size. n enochs. learning rate. μ

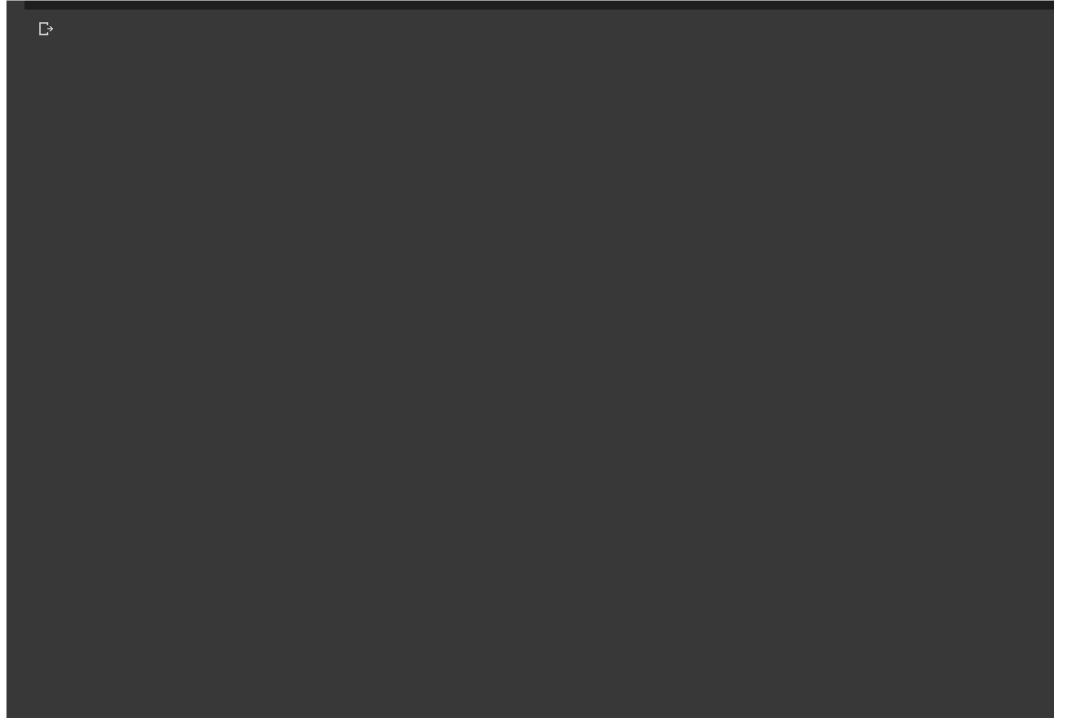
```
56 loss_size = loss(outputs, labels)
57 if np.isnan(loss_size.data):
---> 58 raise ValueError("loss explode due to large regularization or learning rate")
59
60 loss_size.backward()

ValueError: loss explode due to large regularization or learning rate

SEARCH STACK OVERFLOW
```

Case 1: As Regularization increases, loss also increases. Case 2: When regularization is too large i.e. 10000, LOSS EXPLODES.

```
weight decay = 0.001; learning rate = 0.00001
print("weight decay:{:.4f} learning rate:{:.8f}".format(weight decay, learning rate))
LN = LeNet()
train net(LN, which model = "LeNet", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
              weight decay = weight decay, print train process = True, print test process = False, validation = False, loss method =
del LN
weight decay = 0.001; learning_rate = 0.001
print("weight decay:{:.4f} learning rate:{:.8f}".format(weight decay, learning rate))
LN = LeNet()
train net(LN, which model = "LeNet", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
              weight decay = weight decay, print train process = True, print test process = False, validation = False, loss method =
del LN
weight decay = 0.001; learning rate = 0.125
print("weight decay:{:.4f} learning rate:{:.8f}".format(weight decay, learning rate))
LN = LeNet()
train net(LN, which model = "LeNet", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
              weight decay = weight decay, print train process = True, print test process = False, validation = False, loss method =
del LN
```



```
weight decay:0.0010 learning rate:0.00001000
Training Examples - 3003
Validation Examples - 334
               train loss: 0.1389 train accuracy:52% took: 0.53s
Epoch 1, 9%
Epoch 1, 19%
                train loss: 0.1407 train accuracy:47% took: 0.51s
                train loss: 0.1385 train accuracy:54% took: 0.50s
Epoch 1, 29%
                train loss: 0.1398 train accuracy:50% took: 0.52s
Epoch 1, 39%
Epoch 1, 49%
                train loss: 0.1407 train accuracy:48% took: 0.56s
                train loss: 0.1392 train accuracy:51% took: 0.54s
Epoch 1, 59%
                train loss: 0.1388 train accuracy:53% took: 0.51s
Epoch 1, 69%
                train loss: 0.1406 train accuracy:48% took: 0.54s
Epoch 1, 79%
Epoch 1, 89%
                train loss: 0.1412 train accuracy:46% took: 0.52s
              train loss: 0.1404 train accuracy:49% took: 0.54s
Epoch 1, 99%
weight decay:0.0010 learning rate:0.00100000
Training Examples - 3003
Validation Examples - 334
Epoch 1, 9%
               train loss: 0.1400 train accuracy:47% took: 0.54s
Epoch 1, 19%
                train loss: 0.1381 train accuracy:52% took: 0.52s
Epoch 1, 29%
                train loss: 0.1375 train accuracy:44% took: 0.50s
Epoch 1, 39%
                train loss: 0.1358 train accuracy:52% took: 0.50s
                train loss: 0.1336 train accuracy:64% took: 0.52s
Epoch 1, 49%
Epoch 1, 59%
                train loss: 0.1314 train accuracy:65% took: 0.55s
                train loss: 0.1297 train accuracy:70% took: 0.54s
Epoch 1, 69%
                train loss: 0.1273 train accuracy:70% took: 0.52s
Epoch 1, 79%
Epoch 1, 89%
               train loss: 0.1223 train accuracy:77% took: 0.50s
                train loss: 0.1188 train accuracy:74% took: 0.50s
Epoch 1, 99%
weight decay:0.0010 learning rate:0.12500000
Training Examples - 3003
Validation Examples - 334
Epoch 1, 9% train loss: 0.1257 train accuracy:68% took: 0.53s
Epoch 1, 19% train loss: 0.0971 train accuracy:82% took: 0.46s
ValueError
                                         Traceback (most recent call last)
<ipython-input-73-53ee84d8bd14> in <module>()
    19 LN = LeNet()
    20 train net(LN, which model = "LeNet", whether norm = True, batch size=5, n epochs=1, learning rate = learning rate,
                     weight decay = weight decay, print train process = True, print test process = False, validation = False, lo
---> 21
     22
     23 del LN
```

Case 1: When learning rate is too small (0.00001), Loss barely changes.

Case 2: When learning rate is 0.001, Loss changes reasonably.

Case 3: When learning rate is too large 0.125 here, loss explodes.

**Hyperparameter Optimization(random search) LENET5** 

```
Training Examples - 3003
Validation Examples - 334
Test loss = 0.1174 Test Accuracy = 96%
train loss: 0.02429818 train accuracy:95% learning rate:0.00515260 regularization:0.00002367 running time:28.5694
Training Examples - 3003
Validation Examples - 334
Test loss = 0.6808 Test Accuracy = 54%
train loss: 0.13652195 train accuracy:55% learning rate:0.00003117 regularization:0.00000141 running time:28.2623
Training Examples - 3003
Validation Examples - 334
Test loss = 0.6186 Test Accuracy = 80%
train loss: 0.12761182 train accuracy:77% learning rate:0.00010018 regularization:0.00087610 running time:28.6384
Training Examples - 3003
Validation Examples - 334
Test loss = 0.6745 Test Accuracy = 74%
train loss: 0.13559210 train accuracy:73% learning rate:0.00003850 regularization:0.00799900 running time:28.4200
Training Examples - 3003
Validation Examples - 334
Test loss = 0.6765 Test Accuracy = 48%
train loss: 0.13595977 train accuracy:47% learning rate:0.00001101 regularization:0.00000662 running time:28.7284
Training Examples - 3003
Validation Examples - 334
Test loss = 0.5562 Test Accuracy = 76%
train loss: 0.10825501 train_accuracy:79% learning rate:0.00017152 regularization:0.00157937 running time:28.4043
Training Examples - 3003
Validation Examples - 334
Test loss = 0.6897 Test Accuracy = 53%
train loss: 0.13809760 train accuracy:55% learning rate:0.00001328 regularization:0.01522049 running time:28.3302
Training Examples - 3003
Validation Examples - 334
Test loss = 0.3677 Test Accuracy = 90%
train loss: 0.07652792 train_accuracy:86% learning rate:0.00086848 regularization:0.00003098 running time:29.6632
Training Examples - 3003
```

```
Validation Examples - 334
Test loss = 0.1491 Test Accuracy = 94%
train_loss: 0.03695062 train_accuracy:94% learning rate:0.00601909 regularization:0.09616404 running time:28.2185
4
Training Examples - 3003
Validation Examples - 334
Test loss = 0.1310 Test Accuracy = 94%
train_loss: 0.02450075 train_accuracy:95% learning rate:0.00654510 regularization:0.00009792 running time:28.4715
```

Choose the pair with highest accuracy rate and smallest loss on LENET5

Learning Rate: 0.00515260 and Regularization: 0.00002367.

Г⇒

```
Training Examples - 3003
Validation Examples - 334
                train_loss: 0.1338 train_accuracy:64% took: 0.54s
Epoch 1, 9%
Epoch 1, 19%
                 train loss: 0.1149 train accuracy:80% took: 0.54s
Epoch 1, 29%
                 train loss: 0.1029 train accuracy:80% took: 0.51s
Epoch 1, 39%
                 train loss: 0.0821 train accuracy:83% took: 0.52s
                 train loss: 0.0943 train accuracy:80% took: 0.50s
Epoch 1, 49%
Epoch 1, 59%
                 train loss: 0.0734 train accuracy:86% took: 0.52s
                 train loss: 0.0794 train accuracy:86% took: 0.53s
Epoch 1, 69%
                 train loss: 0.0713 train accuracy:88% took: 0.53s
Epoch 1, 79%
                 train loss: 0.0659 train accuracy:88% took: 0.53s
Epoch 1, 89%
Epoch 1, 99%
                 train loss: 0.0668 train accuracy:91% took: 0.55s
Test loss = 0.2463 Test Accuracy = 93%
                 train loss: 0.0546 train accuracy:91% took: 1.07s
Epoch 2, 9%
                 train loss: 0.0524 train accuracy:91% took: 0.48s
Epoch 2, 19%
Epoch 2, 29%
                 train loss: 0.0440 train accuracy:92% took: 0.52s
                 train loss: 0.0361 train accuracy:94% took: 0.53s
Epoch 2, 39%
                 train loss: 0.0281 train accuracy:97% took: 0.55s
Epoch 2, 49%
                 train loss: 0.0391 train accuracy:94% took: 0.59s
Epoch 2, 59%
Epoch 2, 69%
                 train loss: 0.0476 train accuracy:91% took: 0.52s
                train loss: 0.0399 train accuracy:92% took: 0.52s
Epoch 2, 79%
Epoch 2, 89%
                 train loss: 0.0363 train accuracy:93% took: 0.52s
Epoch 2, 99%
                 train loss: 0.0265 train accuracy:95% took: 0.52s
Test loss = 0.1349 Test Accuracy = 94%
                 train loss: 0.0374 train accuracy:94% took: 1.13s
Epoch 3, 9%
Epoch 3, 19%
                 train loss: 0.0400 train accuracy:93% took: 0.49s
Epoch 3, 29%
                 train loss: 0.0309 train accuracy:95% took: 0.49s
                 train loss: 0.0288 train accuracy:95% took: 0.53s
Epoch 3, 39%
                 train loss: 0.0379 train accuracy:94% took: 0.55s
Epoch 3, 49%
Epoch 3, 59%
                 train loss: 0.0414 train accuracy:93% took: 0.52s
                 train loss: 0.0353 train accuracy:94% took: 0.54s
Epoch 3, 69%
Epoch 3, 79%
                 train loss: 0.0281 train accuracy:94% took: 0.53s
                 train loss: 0.0235 train accuracy:96% took: 0.54s
Epoch 3, 89%
                 train loss: 0.0221 train accuracy:95% took: 0.52s
Epoch 3, 99%
Test loss = 0.1109 Test Accuracy = 95%
Epoch 4, 9%
                 train loss: 0.0194 train accuracy:97% took: 1.07s
Epoch 4, 19%
                train loss: 0.0367 train accuracy:95% took: 0.47s
Epoch 4, 29%
                 train loss: 0.0264 train accuracy:93% took: 0.54s
Epoch 4, 39%
                 train loss: 0.0331 train accuracy:95% took: 0.52s
                 train loss: 0.0386 train accuracy:94% took: 0.53s
Epoch 4, 49%
Epoch 4, 59%
                 train loss: 0.0223 train accuracy:95% took: 0.53s
```

```
train loss: 0.0211 train accuracy:95% took: 0.53s
Epoch 4, 69%
Epoch 4, 79%
                 train loss: 0.0419 train accuracy:93% took: 0.52s
                train loss: 0.0247 train accuracy:96% took: 0.51s
Epoch 4, 89%
                train loss: 0.0272 train accuracy:95% took: 0.53s
Epoch 4, 99%
Test loss = 0.1057 Test Accuracy = 95%
Epoch 5, 9%
                 train loss: 0.0257 train accuracy:96% took: 1.07s
Epoch 5, 19%
                train loss: 0.0333 train accuracy:94% took: 0.50s
                train loss: 0.0350 train accuracy:95% took: 0.51s
Epoch 5, 29%
Epoch 5, 39%
                 train loss: 0.0271 train accuracy:95% took: 0.54s
Epoch 5, 49%
                 train loss: 0.0214 train accuracy:96% took: 0.51s
Epoch 5, 59%
                 train loss: 0.0215 train accuracy:96% took: 0.52s
Epoch 5, 69%
                train loss: 0.0303 train accuracy:94% took: 0.52s
                train loss: 0.0219 train accuracy:95% took: 0.54s
Epoch 5, 79%
Epoch 5, 89%
                train loss: 0.0186 train accuracy:95% took: 0.52s
Epoch 5, 99%
                 train loss: 0.0203 train accuracy:96% took: 0.53s
Test loss = 0.0951 Test Accuracy = 96%
                 train loss: 0.0202 train accuracy:94% took: 1.08s
Epoch 6, 9%
                 train loss: 0.0285 train accuracy:95% took: 0.53s
Epoch 6, 19%
Epoch 6, 29%
                 train loss: 0.0272 train accuracy:95% took: 0.51s
                 train loss: 0.0182 train accuracy:97% took: 0.52s
Epoch 6, 39%
Epoch 6, 49%
                 train loss: 0.0212 train accuracy:97% took: 0.55s
                 train loss: 0.0213 train accuracy:96% took: 0.52s
Epoch 6, 59%
Epoch 6, 69%
                 train loss: 0.0231 train accuracy:96% took: 0.53s
Epoch 6, 79%
                train loss: 0.0311 train accuracy:95% took: 0.53s
                 train loss: 0.0258 train accuracy:94% took: 0.54s
Epoch 6, 89%
Epoch 6, 99%
                train loss: 0.0223 train accuracy:96% took: 0.51s
Test loss = 0.0962 Test Accuracy = 96%
Epoch 7, 9%
                 train loss: 0.0183 train accuracy:95% took: 1.08s
Epoch 7, 19%
                train loss: 0.0193 train accuracy:96% took: 0.53s
                 train loss: 0.0210 train accuracy:96% took: 0.48s
Epoch 7, 29%
                 train loss: 0.0275 train accuracy:95% took: 0.54s
Epoch 7, 39%
                 train loss: 0.0161 train accuracy:98% took: 0.52s
Epoch 7, 49%
Epoch 7, 59%
                 train loss: 0.0151 train accuracy:97% took: 0.53s
Epoch 7, 69%
                 train loss: 0.0222 train accuracy:95% took: 0.56s
                 train loss: 0.0195 train accuracy:96% took: 0.56s
Epoch 7, 79%
Epoch 7, 89%
                 train loss: 0.0289 train accuracy:94% took: 0.52s
Epoch 7, 99%
                 train loss: 0.0241 train accuracy:96% took: 0.51s
Test loss = 0.0884 Test Accuracy = 97%
Epoch 8, 9%
                train loss: 0.0199 train accuracy:96% took: 1.08s
Epoch 8, 19%
                train loss: 0.0288 train accuracy:95% took: 0.51s
Epoch 8, 29%
                train loss: 0.0226 train accuracy:96% took: 0.51s
Epoch 8, 39%
                 train loss: 0.0181 train accuracy:96% took: 0.52s
```

```
Epoch 8, 49%
                train loss: 0.0171 train accuracy:97% took: 0.51s
                train loss: 0.0162 train accuracy:98% took: 0.52s
Epoch 8, 59%
                train loss: 0.0174 train accuracy:96% took: 0.53s
Epoch 8, 69%
                train loss: 0.0165 train accuracy:97% took: 0.56s
Epoch 8, 79%
                train loss: 0.0179 train accuracy:97% took: 0.52s
Epoch 8, 89%
                train loss: 0.0244 train accuracy:97% took: 0.55s
Epoch 8, 99%
Test loss = 0.0719 Test Accuracy = 96%
                train loss: 0.0246 train accuracy:96% took: 1.14s
Epoch 9, 9%
                train loss: 0.0193 train accuracy:96% took: 0.49s
Epoch 9, 19%
                train loss: 0.0174 train accuracy:96% took: 0.52s
Epoch 9, 29%
                train loss: 0.0238 train accuracy:97% took: 0.52s
Epoch 9, 39%
                train loss: 0.0136 train accuracy:98% took: 0.55s
Epoch 9, 49%
Epoch 9, 59%
                train loss: 0.0187 train accuracy:96% took: 0.52s
                train loss: 0.0131 train accuracy:97% took: 0.51s
Epoch 9, 69%
                train loss: 0.0100 train accuracy:97% took: 0.51s
Epoch 9, 79%
                train loss: 0.0262 train accuracy:95% took: 0.53s
Epoch 9, 89%
                train loss: 0.0145 train accuracy:97% took: 0.52s
Epoch 9, 99%
Test loss = 0.0672 Test Accuracy = 97%
                train loss: 0.0153 train accuracy:97% took: 1.05s
Epoch 10, 9%
                train loss: 0.0111 train accuracy:98% took: 0.53s
Epoch 10, 19%
                train loss: 0.0196 train accuracy:97% took: 0.51s
Epoch 10, 29%
                 train loss: 0.0205 train accuracy:96% took: 0.53s
Epoch 10, 39%
Epoch 10, 49%
                train loss: 0.0209 train accuracy:97% took: 0.56s
                 train loss: 0.0116 train accuracy:97% took: 0.56s
Epoch 10, 59%
                train loss: 0.0182 train accuracy:96% took: 0.57s
Epoch 10, 69%
                train loss: 0.0168 train accuracy:96% took: 0.55s
Epoch 10, 79%
                train loss: 0.0205 train accuracy:96% took: 0.55s
Epoch 10, 89%
Epoch 10, 99%
                train loss: 0.0163 train accuracy:97% took: 0.56s
Test loss = 0.0696 Test Accuracy = 97%
```

