pil_img_right = pil_img_right.crop((int(img_right_pos[0]),int(img_right_pos[1]), 6 int(img_right_pos[2]),int(img_right_pos[3]))) pil_img_top = Image.open(os.path.join(DataDir, random_apple_info.at['top'])) pil_img_top = pil_img_top.crop((int(img_top_pos[0]),int(img_top_pos[1]), 9 int(img_top_pos[2]),int(img_top_pos[3]))) 10 11 # pil im = Image.open('1.jpg').convert('L') #灰度操作 12 13 plt.figure(figsize=(5 * 3 , 5)) 14 15 ax1 = plt.subplot('131')plt.imshow(pil img left) 16 17 ax1.set title("Left") 18 19 ax2 = plt.subplot(132)plt.imshow(pil img top) 20 21 ax2.set_title("top") 22 23 ax3 = plt.subplot(133)24 plt.imshow(pil img right) 25 ax3.set_title("right") 26 27 plt.tight_layout() 28 plt.show() right top Left 0 50 50 50 -100 100 100 150 150 200 200 200 250 250 200 250 100 150 250 print('Labels: ', sorted(set(data['label']))) In [5]: Labels: ['80+', '80-', '90+', 'bad'] In [6]: # data[["id", "label"]].groupby(['label'], as_index=False) print("Count the amount of each label:") pd.value_counts(data["label"]) Count the amount of each label: 2907 Out[6]: 90+ 80+ 1967 80-441396 bad Name: label, dtype: int64 Get mean and std of every channal This is only a example, just show how to get mean and std from one picture. Notice: In training phase, your mean and std must calculate from the whole training set. In [7]: pic = np.array(pil_img_left) print("pic.shape:",pic.shape) pic_mean = np.mean(pic/255, axis = (0,1),keepdims = True) print("pic_mean:",pic_mean) pic_mean = pic_mean.reshape(3,) print("pic_mean:", pic_mean) pic_std = np.std(pic,axis = (0,1),keepdims = True) 10 11 print("pic_std:",pic_std) 12 13 pic_std = pic_std.reshape(3,) print("pic_std:",pic_std) 15 pic.shape: (262, 305, 3) pic_mean: [[[0.4864052 0.38608327 0.34705215]]] pic_mean: [0.4864052 0.38608327 0.34705215] pic_std: [[[60.17899054 61.28892634 65.43322674]]] pic_std: [60.17899054 61.28892634 65.43322674] def line_chart(x,y,x_label,y_labal,tital,color): In [8]: #在当前绘图对象绘图 (X轴, Y轴, 蓝色虚线, 线宽度) plt.plot(x,y,color,linewidth=1) 2 #X轴标签 plt.xlabel(x_label) #Y轴标签 plt.ylabel(y_labal) 4 #图标题 5 plt.title(tital) 6 plt.show() start In [9]: class AppleSet(Dataset): 3 def __init__(self, root, train = True, transform = None,): self.train = train 5 self.root = root 6 self.transform = transform 7 classes = ['90+','80+','80-','bad'] 8 if self.train: 9 self.data = pd.read_csv(os.path.join(self.root, 'train.csv')) 10 print(pd.value_counts(self.data["label"])) 11 else: 12 self.data = pd.read_csv(os.path.join(self.root, 'test.csv')) 13 print(pd.value_counts(self.data["label"])) 14 15 def len (self): 16 return self.data.index.size 17 18 def __getitem__(self, index): 19 self.classes = ['90+','80+','80-','bad'] 20 21 data item = self.data.iloc[index] 22 apple_item_path = data_item['left'] 23 apple_item_img = Image.open(os.path.join(self.root, apple_item_path)) apple_img_box_list = data_item['left_box'][2:-2].split("', '") 24 25 apple_item_img = apple_item_img.crop((int(apple_img_box_list[0]), 26 int(apple_img_box_list[1]), 27 int(apple_img_box_list[2]), 28 int(apple_img_box_list[3]))) 29 30 if self.transform: 31 apple_item_img = self.transform(apple_item_img) 32 target = data_item['label'] 33 34 return apple_item_img, self.classes.index(target),data_item['id'],data_item['label'] 35 N_CLASSES = 4 In [10]: $IMAGE_SIZE = 227$ 3 BATCH_SIZE = 16 use_cuda = torch.cuda.is_available() if (use_cuda): 6 print("Great, you have a GPU!") 8 else: 9 print("Life is short -- consider a GPU!") device = torch.device("cuda:2" if use_cuda else "cpu")

print('Your device will be: ', device)

transforms.CenterCrop(480), #???

print('Train Size: ', train_data.data.size)

transforms.RandomHorizontalFlip(), #???

transforms.Resize((IMAGE SIZE, IMAGE SIZE)),

transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5)),

train_loader = torch.utils.data.DataLoader(dataset = train_data,

test loader = torch.utils.data.DataLoader(dataset = test data,

def imshow(batch, class names=None, num images=4):

grid = torchvision.utils.make_grid(img[:img_num],

grid = grid.cpu().numpy().transpose((1, 2, 0))

titles = [class_names[x] for x in classes[:img_num]]

def __init__(self, num_classes = 4, init_weights = True, pretrained = True):

mean = np.array([0.485, 0.456, 0.406])

std = np.array([0.229, 0.224, 0.225])

imshow(next(iter(train_loader)),num_images = 3)

tensor([0, 3, 0, 0, 2, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0])

plt.figure(figsize=(4 * num images, 4))

img_num = min(num_images, img.shape[0])

img, classes,ids,labels = batch

grid = std * grid + mean

grid = np.clip(grid, 0, 1)

plt.axis('off')

plt.title(titles)

plt.imshow(grid)

plt.pause(0.001)

torch.manual_seed(117)

Out[16]: <torch._C.Generator at 0x2ab89bd4ea70>

class Net(nn.Module):

]))

]))

if init_weights:

x = self.conv(x)

x = self.dense(x)

x = x.view(x.size(0), -1)

def _initialize_weights(self):

for m in self.modules():

def forward(self, x):

return x

super(Net, self).__init__()

self.conv = torch.nn.Sequential(OrderedDict([

('ReLU_1',nn.ReLU(inplace = True)),

('ReLU_2',nn.ReLU(inplace=True)),

('ReLU_3',nn.ReLU(inplace = True)),

('ReLU_4',nn.ReLU(inplace=True)),

('ReLU_5',nn.ReLU(inplace=True)),

('Dropout_1',nn.Dropout()),

('Dropout_2',nn.Dropout()),

self._initialize_weights()

if isinstance(m, nn.Conv2d):

if m.bias is not None:

elif isinstance(m, nn.Linear):

m.bias.data.zero_()

Create a function to return a pre-trained model

print("Try to use pre-trained model")

model = torch.load(filePath)

def train(model, device, train_loader, optimizer, epoch):

train_list = {'loss':[],'acc':[],'idx':[]}

indx_target = targets.clone() # clone labels

correct = pred.cpu().eq(indx_target).sum()

if batch_idx % 5 == 0 and batch_idx >= 0:

train_list['acc'].append(acc)

format(epoch,

filePath = os.path.join(DataDir, 'AppleModel.pth')

print("Failed! Create a new model instead.")

tqdm_loader = tqdm_notebook(train_loader,total=len(train_loader))

loss = F.cross_entropy(outputs, targets) # calculate loss

total_loss += loss.data.item() # add loss to total_loss

acc = float(correct) * 1.0 / len(inputs) * 100

loss.data.item(),

100 * float(total acc)/len(train loader.dataset),

train_list_global['acc'].append(100 * float(total_acc)/len(train_loader.dataset))

train_list['loss'].append(loss.data.item())

total_acc += correct # the amount of corrent items

train_list['idx'].append(batch_idx)

acc,

print("Train Info: Loss: {:.6f}, Acc: {:.6f}, lr: {:.2e}".

optimizer.param groups[0]['lr']))

train_list_global['lr'].append(optimizer.param_groups[0]['lr'])

line_chart(train_list['idx'],train_list['loss'],"iterations","loss",

line_chart(train_list['idx'],train_list['acc'],"iterations","loss",

tqdm_loader = tqdm_notebook(test_loader,total=len(test_loader))

for batch_idx, (inputs, target,ids,labels) in enumerate(tqdm_loader):

inputs, target = inputs.to(device), target.to(device)

pred = output.data.max(1)[1] # get the index of the max log-probability

if the acc of the model better than older one weo times, than update it.

print("Best acc is {},lower than current acc {}. Patience set to {}.".

new file = os.path.join(DataDir, 'AppleModel.pth')

print("Save a new model to {}.".format(new_file))

model = AppleModel(pretrained = True, num_classes = N_CLASSES, init_weights = True)

(Conv2d_1): Conv2d(3, 64, kernel_size=(11, 11), stride=(4, 4), padding=(2, 2))

(Conv2d_2): Conv2d(64, 192, kernel_size=(5, 5), stride=(1, 1), padding=(2, 2))

(Conv2d_3): Conv2d(192, 384, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))

(Conv2d_4): Conv2d(384, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))

(Conv2d_5): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))

(Linear_1): Linear(in_features=9216, out_features=4096, bias=True)

(Linear 2): Linear(in features=4096, out features=4096, bias=True)

(Linear_3): Linear(in_features=4096, out_features=4, bias=True)

train_list_global = {'loss':[],'acc':[],'idx':[],'lr':[]} 5 test list global = {'loss':[], 'acc':[], 'idx':[], 'lr':[]}

lr = 0.001 # init learning rate

print("Best Result: {:.3f}%".format(best_acc))

"Total Loss Analysis", "r")

"Total Loss Analysis", "r")

"Total Accuracy Analysis", "g")

"Total Accuracy Analysis", "g")

"Total Learning Rate Analysis", "b")

print("Leaning Rate set to {}".format(lr))

train(model,device,train_loader,optimizer,epoch)

test(model,device,test_loader,optimizer,epoch)

optimizer = optim.Adam(model.parameters(), lr = lr, weight_decay=5e-4)

line_chart(train_list_global['idx'],train_list_global['loss'],"iterations","loss",

line_chart(train_list_global['idx'],train_list_global['acc'],"iterations","acc",

line_chart(test_list_global['idx'],test_list_global['loss'],"iterations","loss",

line_chart(test_list_global['idx'],test_list_global['acc'],"iterations","acc",

line_chart(train_list_global['idx'],train_list_global['lr'],"iterations","lr",

(MaxPool2d_1): MaxPool2d(kernel_size=3, stride=2, padding=0, dilation=1, ceil_mode=False)

(MaxPool2d 2): MaxPool2d(kernel size=3, stride=2, padding=0, dilation=1, ceil mode=False)

(MaxPool2d_3): MaxPool2d(kernel_size=3, stride=2, padding=0, dilation=1, ceil_mode=False)

test loss percent = test loss / len(test loader) # average over number of mini-batch

test_loss += F.cross_entropy(output, target).data.item()

acc = 100. * float(correct) / len(test_loader.dataset)

test_list_global['lr'].append(optimizer.param_groups[0]['lr'])

save file = True # config. If need to save model

print("Patience already to 2,set to 0.")

print("Leaning Rate set to {}.".format(lr))

format(best_acc,acc,patience))

torch.save(model, new_file)

if save_file: # model save

print('Test set: Loss: {:.4f}, Accuracy: {}/{} ({:.2f}%)'.format(

correct += pred.cpu().eq(indx_target).sum()

train_list_global['loss'].append(total_loss)

train_list_global['idx'].append(epoch)

train list['loss'], "b", linewidth=1,

"Lost Analysis", 'r')

train_list['acc'],

"r", linewidth=1,

label = "epoch " + str(epoch))

label = "epoch_" + str(epoch))

"Accuracy Analysis", "b")

def test(model, device, test_loader, optimizer, epoch):

inputs, targets = Variable(inputs), Variable(targets)

for batch_idx, (inputs, targets,ids,labels) in enumerate(tqdm_loader):

inputs, targets = inputs.to(device), targets.to(device)

optimizer.zero_grad() # Clears the gradients of all optimized outputs = model(inputs) # output will be batch_size * class_size

pred = outputs.data.max(1)[1] # get the index of the max log-probability

batch idx * len(inputs),

len(train loader.dataset),

optimizer.param groups[0]['lr']))

#X轴标签 #Y轴标签

#图标题

#X轴标签

#Y轴标签

#图标题

print('Train Epoch: {} [{:5}/{:5}] Loss: {:.6f} Acc: {:.4f} lr: {:.2e}'.

#在当前绘图对象绘图(X轴,Y轴,蓝色虚线,线宽度)

#在当前绘图对象绘图(X轴,Y轴,蓝色虚线,线宽度)

def AppleModel(pretrained = False, **kwargs):

if os.path.isfile(filePath):

print("Loading...")

print("Success!")

print("Start training...")

global train_list_global

if use cuda:

loss.backward()

optimizer.step()

format(total_loss,

plt.plot(train_list['idx'],

plt.xlabel("iterations")

plt.title("Lost Analysis")

plt.plot(train_list['idx'],

plt.xlabel("iterations")

plt.title("Accuracy Analysis")

plt.ylabel("acc")

plt.show()

Create Test Function

model.eval()

test loss = 0

global patience

global test_list_global

if use_cuda:

test_loss,

if acc > best acc:

correct,

acc))

indx target = target.clone()

inputs = Variable(inputs)

target = Variable(target)

with torch.no_grad():

output = model(inputs)

len(test_loader.dataset),

test_list_global['acc'].append(acc)

patience = patience + 1

if patience == 2:

lr = lr/10

patience = 0

best_acc = acc

test_list_global['idx'].append(epoch)

test_list_global['loss'].append(test_loss)

correct = 0 global best acc

global lr

plt.ylabel("loss")

losss pic

plt.show()

#acc pic

model = Net(**kwargs)

if pretrained:

else:

return model

Create Train Function

SAPARATE()

correct = 0

total_loss = 0

total_acc = 0

m.bias.data.zero_()

m.weight.data.normal_(0, 0.01)

('MaxPool2d_3',nn.MaxPool2d(3, 2, 0))

self.dense = torch.nn.Sequential(OrderedDict([

('Linear_1',nn.Linear(9216, 4096)),

('Linear_2',nn.Linear(4096, 4096)),

('Linear_3',nn.Linear(4096, num_classes))

n = m.kernel_size[0] * m.kernel_size[1] * m.out_channels

print('epoch: {}, Learning rate: {}'.format(epoch,optimizer.param_groups[0]['lr']))

m.weight.data.normal_(0, math.sqrt(2. / n))

('ReLU_2',nn.ReLU(inplace=True)),

('ReLU_1',nn.ReLU(inplace=True)),

('Conv2d_1',nn.Conv2d(3, 64, 11, 4, 2)),

('Conv2d_2',nn.Conv2d(64, 192, 5, 1, 2)),

('Conv2d_3',nn.Conv2d(192, 384, 3, 1, 1)),

('Conv2d_4',nn.Conv2d(384, 256, 3, 1, 1)),

('Conv2d_5',nn.Conv2d(256,256,3,1,1)),

('MaxPool2d_1',nn.MaxPool2d(3, 2, 0)),

('MaxPool2d_2',nn.MaxPool2d(3, 2, 0)),

Create Model

if class names:

batch_size = BATCH_SIZE,

batch_size = BATCH_SIZE)

shuffle = True)

nrow=img_num,

padding = 1, pad_value = 1)

train_data = AppleSet(root = DataDir, train = True, transform = transform) 10 | test_data = AppleSet(root = DataDir, train = False, transform = transform)

transform = transforms.Compose([

transforms.ToTensor(),

Great, you have a GPU!

In [11]:

2

3

4 5

6 7

8 9

90+

80+

80-

bad

90+

+08

80bad

5711 1428

6

3

4 5

6 7

8

9

10

11 12

13

14 15

16

17 18

19

20

21

22

23

50

100

150

200

3

4

5

6

7

8

9

10

11

12

13

14

15

16

17

18

19 20

21

22 23

24 25

26

27

28

29 30

31

32 33

34

35

36 37

38 39

40

41 42

43

44

45

46

47

48

49

3

4

5

6 7

8

9

10

11

12

3

8

9

10

11

12 13

14

15

16

17

18

19 20

21 22 23

24

25

26 27

28 29

30 31

32 33

34

35

36

37

38

39

43 #

44

45 46

47

48

49 50

51

52

53

54 55 #

56 #

57

58

59

60

61

62

63 64

65 66

67

68 # #

69 70 #

71

72

73

74

75

76

3 4

5

6 7

8

9

10 11

12

13 14 15

16

17

18 19

20

21 22

23

24 25

26 27

28 29 30

31

32

33

34

35 36

37

38

39

40 41

42

43 44

45 46

47

48 49

50

51

52

53 54 55

56

57

58 59

60

2

3

))

7

8

9

10

11 12

13 14

15

16 17

18

19

20

21 22

23 24

25

26

27

28

29

30 31

32

33

34 35

In [22]:

Loading... Success! Net(

In [21]:

Create model

if use cuda:

print(model)

model.to(device)

(ReLU 1): ReLU(inplace)

(ReLU 2): ReLU(inplace)

(ReLU_3): ReLU(inplace)

(ReLU 4): ReLU(inplace)

(ReLU 5): ReLU(inplace)

(ReLU_1): ReLU(inplace) (Dropout 2): Dropout(p=0.5)

(ReLU 2): ReLU(inplace)

epoch_size = 30

except Exception as e:

import traceback

traceback.print_exc()

if epoch == 0:

for epoch in range(epoch_size):

torch.cuda.empty_cache()

(Dropout_1): Dropout(p=0.5)

(dense): Sequential(

Strart Train model

best acc = 0

3 patience = 0

2 | 1r = 0

Try to use pre-trained model

(conv): Sequential(

In [20]:

#

#

#

#

40 # 41

42

#

In [18]:

In [19]:

In [16]:

In [17]:

#

In [12]:

In [13]:

In [14]:

In [15]:

])

Your device will be: cuda:2

11 | print(len(train_data)) print(len(test data))

2907

1967

441

396

738

497 118

75

Train Size: 51399

Name: label, dtype: int64

Name: label, dtype: int64

train_data.data

print(classes)

In [1]:

In [2]:

In [3]:

3

9

10

11 12

13

14 15

16

17

18

19 20

21

22

Label:

Left:

Top:

3

In [4]:

bad

Data/04.bad/00105_left.jpg

Data/04.bad/00105 top.jpg

Right: Data/04.bad/00105 right.jpg

7

0

7

12

%matplotlib inline

5 import torch.nn as nn

13 import matplotlib as mpl

15 | import seaborn as sns 16 | import numpy as np 17 | import seaborn as sns 18 | import pandas as pd

22 **from** PIL **import** Image

20 | import math 21 **import** random

23 import os

Preprocess

SAPARATE()

The head of 'train.csv': Unnamed: 0 id

1 Data/01.90+/00001 top.jpg 2 Data/01.90+/00002_top.jpg

3 Data/01.90+/00003 top.jpg

4 Data/01.90+/00006_top.jpg

DataDir = './'

8 import torchvision

6 import torch.nn.functional as F import torch.optim as optim

9 **from** torch.utils.data **import** Dataset

11 **from** torch.autograd **import** Variable

24 | from collections import OrderedDict

3 | SAPARATE = lambda: print(SAPARATELINE)

4 data = pd.read csv('train.csv')

The shape of 'train.csv': (5711, 9) _____

14 | import matplotlib.pyplot as plt

10 | from torchvision import datasets, transforms

19 **from** tqdm **import** tnrange, tqdm_notebook,tqdm

2 | SAPARATELINE = '-' * 98 # the lens is same as github

top label 0 Data/01.90+/00000_top.jpg 90+ ['142', '37', '563', '431']

0 ['104', '22', '567', '442'] ['125', '76', '499', '408'] 1 ['120', '45', '522', '420'] ['174', '101', '533', '427'] 2 ['119', '70', '546', '426'] ['167', '94', '498', '441'] 3 ['142', '66', '535', '428'] ['173', '108', '526', '456'] 4 ['118', '19', '589', '445'] ['105', '95', '496', '444']

random_index = np.random.randint(0,data.index.size) # print('Label:\t',data['label'][random index])

img_left_pos = random_apple_info['left_box'][2:-2].split("', '")

img right pos = random apple info['right box'][2:-2].split("', '")

img top pos = random apple info['top box'][2:-2].split("',

img_left_pos[0],img_left_pos[1],

img left pos[2],img left pos[3]))

img top pos[0], img top pos[1],

img_top_pos[2],img_top_pos[3]))

img_right_pos[0],img_right_pos[1],

img_right_pos[2],img_right_pos[3]))

right_box

random apple info = data.loc[random index]

print('Label:\t',data['label'][random_index])

print('Left:\t{:30}\tPos: ({:3},{:3}),({:3},{:3})'

.format(random apple info.at['left'],

print('Top:\t{:30}\tPos: ({:3},{:3}),({:3},{:3})'

.format(random apple info.at['top'],

print('Right:\t{:30}\tPos: ({:3},{:3}),({:3},{:3})' .format(random apple info.at['right'],

left

90+ ['93', '35', '555', '442']

90+ ['100', '59', '548', '437'] 90+ ['89', '67', '520', '433']

90+ ['116', '77', '520', '426']

6 Data/01.90+/00006_left.jpg Data/01.90+/00006_right.jpg

right \

left box \

top_box

Pos: (98,83),(403,345)

Pos: (157,199),(437,444)

Pos: (199,125),(510,405)

int(img_left_pos[2]),int(img_left_pos[3])))

pil img left = Image.open(os.path.join(DataDir, random apple info.at['left']))

pil_img_right = Image.open(os.path.join(DataDir, random_apple_info.at['right']))

pil img left = pil img left.crop((int(img left pos[0]),int(img left pos[1]),

6 print("The shape of 'train.csv': ",data.shape)

8 print("The head of 'train.csv':\n",data.head())

3 | import argparse 4 import torch

from __future__ import print_function

ss 0.3 0.2 0.0 50 100 150 200 250 300 iterations Accuracy Analysis 100 95 90 85 80 75 50 100 200 250 300 350 iterations 100% 90/90 [00:09<00:00, 9.43it/s] Test set: Loss: 14.6386, Accuracy: 1357/1428 (95.03%) Best acc is 0, lower than current acc 95.0280112044818. Patience set to 1. Start training... epoch: 1, Learning rate: 0.001 100% 357/357 [00:46<00:00, 7.95it/s] Train Info: Loss: 50.001444, Acc: 96.340396, lr: 1.00e-03 Lost Analysis 0.8 0.7 0.6 0.5 S 0.4 0.3 0.2 0.1 0.0 50 250 100 150 200 300 350 iterations Accuracy Analysis 100 95 90 85 80 75 50 100 250 0 150 200 300 350 iterations 100% 90/90 [00:09<00:00, 9.21it/s] Test set: Loss: 11.9997, Accuracy: 1370/1428 (95.94%) Best acc is 0, lower than current acc 95.93837535014005. Patience set to 2. Patience already to 2, set to 0. Leaning Rate set to 0.0001. /lustre/wangdan/anaconda2/envs/py3/lib/python3.7/site-packages/torch/serialization.py:250: UserWarning: Couldn 't retrieve source code for container of type Net. It won't be checked for correctness upon loading. "type " + obj.__name__ + ". It won't be checked " Save a new model to ./AppleModel.pth. Start training... epoch: 2, Learning rate: 0.0001 100% 357/357 [00:45<00:00, 6.97it/s] Train Info: Loss: 24.782798, Acc: 97.811241, lr: 1.00e-04 Lost Analysis 0.7 0.6 0.5 0.4 0.3 0.2 0.1 0.0 250 100 150 200 300 350 Accuracy Analysis 100 98 96 94 92 90 88 50 100 150 200 250 0 300 350 iterations 100% 90/90 [00:09<00:00, 9.18it/s] Test set: Loss: 8.4261, Accuracy: 1395/1428 (97.69%) Best acc is 95.93837535014005, lower than current acc 97.6890756302521. Patience set to 1. Start training... epoch: 3, Learning rate: 0.0001 100% 357/357 [00:46<00:00, 8.13it/s] Train Info: Loss: 17.983536, Acc: 98.476624, lr: 1.00e-04 Lost Analysis 0.30 0.25 0.20 [∞] 0.15 0.10 0.05 0.00 50 100 150 200 250 300 350 iterations Accuracy Analysis 100 98 96 94 92 90 88 50 100 150 200 350 300 iterations 100% 90/90 [00:09<00:00, 9.12it/s] Test set: Loss: 10.3177, Accuracy: 1396/1428 (97.76%) Best acc is 95.93837535014005, lower than current acc 97.75910364145658. Patience set to 2. Patience already to 2, set to 0. Leaning Rate set to 1e-05. /lustre/wangdan/anaconda2/envs/py3/lib/python3.7/site-packages/torch/serialization.py:250: UserWarning: Couldn 't retrieve source code for container of type Net. It won't be checked for correctness upon loading. "type " + obj.__name__ + ". It won't be checked " Save a new model to ./AppleModel.pth. Start training... epoch: 4, Learning rate: 1e-05 100% 357/357 [00:45<00:00, 8.09it/s] Train Info: Loss: 13.625235, Acc: 98.861846, lr: 1.00e-05 Lost Analysis 0.40 0.35 0.30 0.25 S 0.20 0.15 0.10 0.05 0.00 100 150 200 250 300 350 iterations Accuracy Analysis 100 99 98 97 96 95 94 150 0 100 200 300 50 350 iterations 100% 90/90 [00:10<00:00, 8.82it/s] Test set: Loss: 10.5685, Accuracy: 1396/1428 (97.76%) Start training... epoch: 5, Learning rate: 1e-05 100% 357/357 [00:45<00:00, 8.03it/s] Train Info: Loss: 13.775540, Acc: 98.879356, lr: 1.00e-05 Lost Analysis 0.25 0.20 0.15 0.10 0.05 0.00 50 100 150 200 250 300 350 iterations Accuracy Analysis 100 99 98 97 96 95 94 50 100 150 200 350 250 300 iterations 100% 90/90 [00:09<00:00, 9.27it/s] Test set: Loss: 9.9827, Accuracy: 1401/1428 (98.11%) Best acc is 97.75910364145658, lower than current acc 98.10924369747899. Patience set to 1. Start training... epoch: 6, Learning rate: 1e-05 100% 357/357 [00:45<00:00, 7.11it/s] Train Info: Loss: 12.189124, Acc: 98.984416, lr: 1.00e-05 Lost Analysis 0.25 0.20 S 0.15 0.10 0.05 0.00 150 200 iterations Accuracy Analysis 100 99 98 97 96 95 94 50 100 150 200 250 300 350 iterations 100% 90/90 [00:10<00:00, 8.95it/s] Test set: Loss: 11.8516, Accuracy: 1394/1428 (97.62%) Start training... epoch: 7, Learning rate: 1e-05 100% 357/357 [00:45<00:00, 8.09it/s] Train Info: Loss: 12.025548, Acc: 99.089476, lr: 1.00e-05 Lost Analysis 0.25 0.20 0.15 0.10 0.05 0.00 200 50 100 150 250 300 350 iterations **Accuracy Analysis** 100 98 96 94 92 90 88 100 200 iterations 100% 90/90 [00:09<00:00, 9.04it/s] Test set: Loss: 8.7884, Accuracy: 1402/1428 (98.18%) Best acc is 97.75910364145658, lower than current acc 98.17927170868347. Patience set to 2. Patience already to 2, set to 0. Leaning Rate set to 1.0000000000000002e-06. /lustre/wangdan/anaconda2/envs/py3/lib/python3.7/site-packages/torch/serialization.py:250: UserWarning: Couldn 't retrieve source code for container of type Net. It won't be checked for correctness upon loading. "type " + obj.__name__ + ". It won't be checked " Save a new model to ./AppleModel.pth. Start training... epoch: 8, Learning rate: 1.000000000000002e-06 100% 357/357 [00:45<00:00, 8.12it/s] Train Info: Loss: 11.560666, Acc: 99.089476, lr: 1.00e-06 Lost Analysis 0.5 0.4 0.3 0.2 0.1 0.0 50 100 150 200 250 300 350 iterations Accuracy Analysis 100 98 96 92 90 88 50 100 150 200 250 300 350 iterations 100% 90/90 [00:09<00:00, 9.35it/s] Test set: Loss: 10.8159, Accuracy: 1396/1428 (97.76%) Start training... epoch: 9, Learning rate: 1.0000000000000002e-06 100% 357/357 [00:44<00:00, 8.08it/s] Train Info: Loss: 12.092394, Acc: 99.036946, lr: 1.00e-06 Lost Analysis 0.25 0.20 <u>8</u> 0.15 0.10 0.05 0.00 50 100 150 200 250 300 350 iterations Accuracy Analysis 100 98 96 94 92 90 88 50 100 150 200 250 300 350 iterations 100% 90/90 [00:09<00:00, 9.42it/s] Test set: Loss: 11.8045, Accuracy: 1396/1428 (97.76%) Start training... epoch: 10, Learning rate: 1.0000000000000002e-06 100% 357/357 [00:46<00:00, 6.66it/s] Train Info: Loss: 11.156310, Acc: 99.159517, lr: 1.00e-06 Lost Analysis 0.5 0.4 0.3 0.2 0.1 0.0 100 150 200 250 300 350 iterations Accuracy Analysis 100 99 98 97 96 95 94 100 50 150 200 250 300 350 iterations 100% 90/90 [00:09<00:00, 9.20it/s] Test set: Loss: 9.2538, Accuracy: 1400/1428 (98.04%) Start training... epoch: 11, Learning rate: 1.0000000000000002e-06 100% 357/357 [00:45<00:00, 8.05it/s] Train Info: Loss: 11.073025, Acc: 99.177027, lr: 1.00e-06 Lost Analysis 0.35 0.30 0.25 <u>s</u> 0.20 0.15 0.10 0.05 0.00 50 100 200 250 300 350 150 iterations Accuracy Analysis 100 99 98 97 96 95 94 100 350 150 iterations 100% 90/90 [00:09<00:00, 9.41it/s] Test set: Loss: 10.5843, Accuracy: 1399/1428 (97.97%) Start training... epoch: 12, Learning rate: 1.0000000000000002e-06 100% 357/357 [00:46<00:00, 7.95it/s] Train Info: Loss: 10.698438, Acc: 99.194537, lr: 1.00e-06 Lost Analysis 0.30 0.25 0.20 <u>8</u> 0.15 0.10 0.05

0.00

100

99

50

100

150

200

iterations

Accuracy Analysis

250

300

350

Leaning Rate set to 0.001

epoch: 0, Learning rate: 0.001

Train Info: Loss: 53.008574, Acc: 96.322886, lr: 1.00e-03

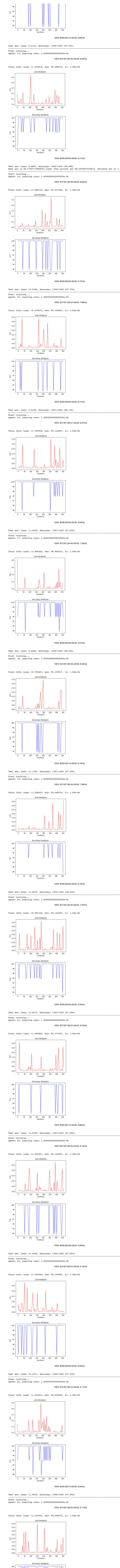
Lost Analysis

100% 357/357 [00:45<00:00, 8.16it/s]

Start training...

0.5

0.4



iterations

ss<u>o</u> 94

0.20 <u>8</u> 0.15 0.10 0.05 0.00 200 250 50 100 150 300 350 iterations Accuracy Analysis 100 99 98 ss 97 96 95 94 50 100 0 150 200 250 300 350 iterations 100% 90/90 [00:09<00:00, 9.39it/s] Test set: Loss: 10.1549, Accuracy: 1395/1428 (97.69%) Start training... epoch: 27, Learning rate: 1.0000000000000002e-06 100% 357/357 [00:45<00:00, 8.26it/s] Train Info: Loss: 11.033789, Acc: 99.071966, lr: 1.00e-06 Lost Analysis 0.40 -0.35 0.30 0.25 S 0.20 0.15 0.10 0.05 0.00 50 100 200 150 250 300 350 iterations Accuracy Analysis 100 99 98 <u>ss</u> 97 96 95 94 50 100 350 Ó 150 200 250 300 iterations 100% 90/90 [00:09<00:00, 9.52it/s] Test set: Loss: 11.9200, Accuracy: 1396/1428 (97.76%) Start training... epoch: 28, Learning rate: 1.0000000000000002e-06 100% 357/357 [00:45<00:00, 8.18it/s] Train Info: Loss: 12.114507, Acc: 98.931886, lr: 1.00e-06 Lost Analysis 0.30 0.25 0.20 <u>8</u> 0.15 0.10 0.05 0.00 50 Ó 100 200 250 300 350 150 iterations Accuracy Analysis 96 94 92 90 88 100 Ò 50 150 200 250 300 350 iterations 100% 90/90 [00:09<00:00, 9.41it/s] Test set: Loss: 12.0962, Accuracy: 1394/1428 (97.62%) Start training... epoch: 29, Learning rate: 1.0000000000000002e-06 100% 357/357 [00:44<00:00, 8.11it/s] Train Info: Loss: 10.783079, Acc: 99.106987, lr: 1.00e-06 Lost Analysis 0.20 0.15 <u>8</u> 0.10 0.05 0.00 50 100 200 250 300 350 150 iterations Accuracy Analysis 100 99 98 97 96 95 94 Ò 50 100 150 200 250 300 350 iterations 100% 90/90 [00:09<00:00, 9.29it/s] Test set: Loss: 10.9479, Accuracy: 1398/1428 (97.90%) Total Loss Analysis 50 40 S 30 20 10 10 15 iterations 20 25 **Total Accuracy Analysis** 99.0 98.5 98.0 97.5 97.0 96.5 25 10 15 iterations ó 5 20 30 Total Learning Rate Analysis 0.0010 0.0008 0.0006 0.0004 0.0002 0.0000 15 iterations 20 25 10 ò 5 30 Total Loss Analysis 14 13 12 · SSO 11 10 10 15 iterations 25 30 **Total Accuracy Analysis** 98.5 98.0 97.5 97.0 96.5 96.0 95.5 95.0 15 iterations 25 20 30

100% 90/90 [00:09<00:00, 9.49it/s]

100% 357/357 [00:44<00:00, 8.14it/s]

Test set: Loss: 11.3734, Accuracy: 1397/1428 (97.83%)

Train Info: Loss: 10.465269, Acc: 99.177027, lr: 1.00e-06

Lost Analysis

epoch: 26, Learning rate: 1.0000000000000002e-06

Start training...

0.30

0.25