## [W2S2\_example2][HW3\_prob1]\_CNN\_Training\_for\_CIFAR10

## October 18, 2021

```
[1]: import argparse
     import os
     import time
     import shutil
     import torch
     import torch.nn as nn
     import torch.optim as optim
     import torch.nn.functional as F
     import torch.backends.cudnn as cudnn
     import torchvision
     import torchvision.transforms as transforms
     from models import * # bring everything in the folder models
     global best_prec
     use_gpu = torch.cuda.is_available()
     print('=> Building model...')
     batch_size = 128
     model_name = "resnet"
     model = resnet20_cifar()
     normalize = transforms.Normalize(mean=[0.491, 0.482, 0.447], std=[0.247, 0.243,
     \rightarrow 0.262])
     train_dataset = torchvision.datasets.CIFAR10(
         root='./data',
         train=True,
```

```
download=True,
    transform=transforms.Compose([
        transforms.RandomCrop(32, padding=4),
        transforms.RandomHorizontalFlip(),
        transforms.ToTensor(),
        normalize,
    1))
trainloader = torch.utils.data.DataLoader(train_dataset, batch_size=batch_size,_
⇒shuffle=True, num workers=2)
test_dataset = torchvision.datasets.CIFAR10(
    root='./data',
    train=False,
    download=True,
    {\tt transform =} {\tt transforms.Compose([}
       transforms.ToTensor(),
       normalize,
    ]))
testloader = torch.utils.data.DataLoader(test dataset, batch size=batch size,
⇒shuffle=False, num_workers=2)
print_freq = 100 # every 100 batches, accuracy printed. Here, each batch
→ includes "batch_size" data points
# CIFAR10 has 50,000 training data, and 10,000 validation data.
def train(trainloader, model, criterion, optimizer, epoch):
    batch_time = AverageMeter() ## at the begining of each epoch, this should_
→be reset
    data_time = AverageMeter()
    losses = AverageMeter()
    top1 = AverageMeter()
    model.train()
    end = time.time() # measure current time
    for i, (input, target) in enumerate(trainloader):
        # measure data loading time
        data_time.update(time.time() - end) # data loading time
        input, target = input.cuda(), target.cuda()
        # compute output
        output = model(input)
```

```
loss = criterion(output, target)
        # measure accuracy and record loss
        prec = accuracy(output, target)[0]
        losses.update(loss.item(), input.size(0))
        top1.update(prec.item(), input.size(0))
        # compute gradient and do SGD step
        optimizer.zero grad()
        loss.backward()
        optimizer.step()
        # measure elapsed time
        batch_time.update(time.time() - end) # time spent to process one batch
        end = time.time()
        if i % print_freq == 0:
            print('Epoch: [{0}][{1}/{2}]\t'
                  'Time {batch_time.val:.3f} ({batch_time.avg:.3f})\t'
                  'Data {data_time.val:.3f} ({data_time.avg:.3f})\t'
                  'Loss {loss.val:.4f} ({loss.avg:.4f})\t'
                  'Prec {top1.val:.3f}% ({top1.avg:.3f}%)'.format(
                   epoch, i, len(trainloader), batch_time=batch_time,
                   data_time=data_time, loss=losses, top1=top1))
def validate(val_loader, model, criterion ):
    batch_time = AverageMeter()
    losses = AverageMeter()
    top1 = AverageMeter()
    # switch to evaluate mode
    model.eval()
    end = time.time()
    with torch.no_grad():
        for i, (input, target) in enumerate(val_loader):
            input, target = input.cuda(), target.cuda()
            # compute output
            output = model(input)
            loss = criterion(output, target)
            # measure accuracy and record loss
```

```
prec = accuracy(output, target)[0]
            losses.update(loss.item(), input.size(0))
            top1.update(prec.item(), input.size(0))
            # measure elapsed time
            batch_time.update(time.time() - end)
            end = time.time()
            if i % print_freq == 0: # This line shows how frequently print out_
\rightarrow the status. e.g., i%5 => every 5 batch, prints out
                print('Test: [{0}/{1}]\t'
                  'Time {batch_time.val:.3f} ({batch_time.avg:.3f})\t'
                  'Loss {loss.val:.4f} ({loss.avg:.4f})\t'
                  'Prec {top1.val:.3f}% ({top1.avg:.3f}%)'.format(
                   i, len(val_loader), batch_time=batch_time, loss=losses,
                   top1=top1))
    print(' * Prec {top1.avg:.3f}% '.format(top1=top1))
    return top1.avg
def accuracy(output, target, topk=(1,)):
    """Computes the precision@k for the specified values of k"""
    \max k = \max(\text{topk})
    batch_size = target.size(0)
    _, pred = output.topk(maxk, 1, True, True)
    pred = pred.t()
    correct = pred.eq(target.view(1, -1).expand_as(pred))
   res = []
    for k in topk:
        correct_k = correct[:k].view(-1).float().sum(0)
        res.append(correct_k.mul_(100.0 / batch_size))
    return res
class AverageMeter(object):
    """Computes and stores the average and current value"""
    def __init__(self):
        self.reset()
    def reset(self):
       self.val = 0
        self.avg = 0
        self.sum = 0
        self.count = 0
```

```
def update(self, val, n=1):
        self.val = val
        self.sum += val * n ## n is impact factor
        self.count += n
        self.avg = self.sum / self.count
def save checkpoint(state, is best, fdir):
    filepath = os.path.join(fdir, 'checkpoint.pth')
    torch.save(state, filepath)
    if is best:
        shutil.copyfile(filepath, os.path.join(fdir, 'model_best.pth.tar'))
def adjust_learning_rate(optimizer, epoch):
    """For resnet, the lr starts from 0.1, and is divided by 10 at 80 and 120_{\sqcup}
⇔epochs"""
    adjust_list = [150, 225]
    if epoch in adjust_list:
        for param group in optimizer.param groups:
            param_group['lr'] = param_group['lr'] * 0.1
#model = nn.DataParallel(model).cuda()
#all_params = checkpoint['state_dict']
#model.load_state_dict(all_params, strict=False)
#criterion = nn.CrossEntropyLoss().cuda()
#validate(testloader, model, criterion)
```

## => Building model...

Files already downloaded and verified Files already downloaded and verified

```
[]: import matplotlib.pyplot as plt
import numpy as np

# functions to show an image

def imshow(img):
    img = img / 2 + 0.5  # unnormalize
    npimg = img.numpy()
    plt.imshow(np.transpose(npimg, (1, 2, 0)))
    plt.show()

# get some random training images
dataiter = iter(testloader)
```

```
images, labels = dataiter.next() ## If you run this line, the next data batch
    →is called subsequently.

# show images
imshow(torchvision.utils.make_grid(images))
```

```
[]: # This cell is from the website
     lr = 4e-2
     weight_decay = 1e-4
     epochs = 300
     best_prec = 0
     model = model.cuda()
     criterion = nn.CrossEntropyLoss().cuda()
     optimizer = torch.optim.SGD(model.parameters(), lr=lr, momentum=0.9, __
     →weight_decay=weight_decay)
     # weight decay: for regularization to prevent overfitting
     if not os.path.exists('result'):
        os.makedirs('result')
     fdir = 'result/'+str(model_name)
     if not os.path.exists(fdir):
         os.makedirs(fdir)
     for epoch in range(0, epochs):
         adjust_learning_rate(optimizer, epoch)
         train(trainloader, model, criterion, optimizer, epoch)
         # evaluate on test set
         print("Validation starts")
         prec = validate(testloader, model, criterion)
         # remember best precision and save checkpoint
         is_best = prec > best_prec
         best_prec = max(prec,best_prec)
         print('best acc: {:1f}'.format(best_prec))
         save_checkpoint({
             'epoch': epoch + 1,
             'state_dict': model.state_dict(),
             'best_prec': best_prec,
             'optimizer': optimizer.state_dict(),
```

```
}, is_best, fdir)
[2]: fdir = 'result/'+str(model name)+'/model best.pth.tar'
     checkpoint = torch.load(fdir)
     model.load_state_dict(checkpoint['state_dict'])
     criterion = nn.CrossEntropyLoss().cuda()
     model.eval()
     model.cuda()
    prec = validate(testloader, model, criterion)
    Test: [0/79]
                    Time 0.781 (0.781)
                                        Loss 0.1425 (0.1425)
                                                                    Prec 96.094%
    (96.094\%)
     * Prec 91.520%
[ ]: # HW
     # 1. train resnet20 and vgg16 to achieve >90% accuracy
     # 2. save your trained model in the result folder
     # 3. Restart your jupyter notebook by "Kernel - Restart & Clear Output"
     # 4. Load your saved model for vgq16 and validate to see the accuracy
     # 5. such as the last part of "[W2S2_example2]_CNN_for_MNIST.ipynb", prehook_
     → the input layers of all the conv layers.
     # 6. from the first prehooked input, compute to get the second prehooked input.
     # 7. Compare your computed second input vs. the prehooked second input.
[5]: model = resnet20_cifar()
    print(model)
    ResNet_Cifar(
      (conv1): Conv2d(3, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
    bias=False)
      (bn1): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True,
    track_running_stats=True)
      (relu): ReLU(inplace=True)
      (layer1): Sequential(
        (0): BasicBlock(
          (conv1): Conv2d(16, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
    bias=False)
          (conv2): Conv2d(16, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
```

```
bias=False)
      (bn1): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (bn2): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(16, 16, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (conv2): Conv2d(16, 16, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (bn2): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (2): BasicBlock(
      (conv1): Conv2d(16, 16, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (conv2): Conv2d(16, 16, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (bn2): BatchNorm2d(16, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
  )
  (layer2): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(16, 32, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1),
bias=False)
      (conv2): Conv2d(32, 32, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (bn2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (downsample): Sequential(
        (0): Conv2d(16, 32, kernel_size=(1, 1), stride=(2, 2), bias=False)
        (1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (1): BasicBlock(
```

```
(conv1): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (conv2): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (bn2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (2): BasicBlock(
      (conv1): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (conv2): Conv2d(32, 32, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (bn2): BatchNorm2d(32, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    )
  )
  (layer3): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(32, 64, kernel size=(3, 3), stride=(2, 2), padding=(1, 1),
bias=False)
      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (downsample): Sequential(
        (0): Conv2d(32, 64, kernel size=(1, 1), stride=(2, 2), bias=False)
        (1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
```

```
track_running_stats=True)
        )
        (2): BasicBlock(
          (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
    bias=False)
          (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1),
    bias=False)
          (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
    track_running_stats=True)
          (relu): ReLU(inplace=True)
          (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
    track_running_stats=True)
        )
     )
      (avgpool): AvgPool2d(kernel_size=8, stride=1, padding=0)
      (fc): Linear(in_features=64, out_features=10, bias=True)
[6]: device = torch.device("cuda" if use gpu else "cpu")
    class SaveOutput:
        def __init__(self):
            self.outputs = []
        def call (self, module, module in):
            self.outputs.append(module_in)
        def clear(self):
            self.outputs = []
    ####### Save inputs from selected layer ########
    save_output = SaveOutput()
    for layer in model.modules():
        if isinstance(layer, torch.nn.Conv2d):
            print("prehooked")
            layer.register_forward_pre_hook(save_output)
                                                            ## Input for the
     →module will be grapped
    dataiter = iter(trainloader)
    images, labels = dataiter.next()
    images = images.to(device)
    out = model(images)
    prehooked
    prehooked
    prehooked
    prehooked
```

prehooked

```
prehooked
     prehooked
[24]: my_input = save_output.outputs[1][0]
      conv1 = model.layer1[0].conv1
      conv2 = model.layer1[0].conv2
      bn1 = model.layer1[0].bn1
      relu = F.relu
      bn2 = model.layer1[0].bn2
     my_output = relu(bn2(conv2(r(bn1(conv1(my_input))))) + my_input)
[25]: (my_output - save_output.outputs[3][0]).sum()
[25]: tensor(0., device='cuda:0', grad_fn=<SumBackward0>)
 []:
```