HW7 2

December 5, 2023

```
[]: import argparse
     import os
     import time
     import shutil
     import tensorboardX
     import torch
     import torch.nn as nn
     import torch.optim as optim
     import torch.nn.functional as F
     import torch.backends.cudnn as cudnn
     from tensorboardX import SummaryWriter
     import torchvision
     import torchvision.transforms as transforms
     from models import *
     global best_prec
     use_gpu = torch.cuda.is_available()
     print('=> Building model...')
     batch_size = 128
     model_name = "VGG16_quant"
     model = VGG16_quant()
     print(model)
     normalize = transforms.Normalize(mean=[0.491, 0.482, 0.447], std=[0.247, 0.243, __
      →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,
    transform=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()
    data_time = AverageMeter()
    losses = AverageMeter()
    top1 = AverageMeter()
    model.train()
    end = time.time()
    for i, (input, target) in enumerate(trainloader):
        # measure data loading time
        data_time.update(time.time() - end)
        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)
        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_
 \hookrightarrow 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"""
    maxk = max(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
             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)
[ ]: # HW
     # 1. Load your saved model and validate
     # 2. Replace your model's all the Conv's weight with quantized weight
     # 3. Apply reasonable alpha
     # 4. Then, try to multiple bit precisions and draw graph of bit precision vs.
      \hookrightarrowaccuracy
[]: PATH = "result/VGG16_quant_4bit_hw7/model_best.pth.tar"
     checkpoint = torch.load(PATH)
     model.load_state_dict(checkpoint['state_dict'])
     device = torch.device("cuda")
     model.cuda()
     model.eval()
     test loss = 0
     correct = 0
     with torch.no_grad():
         for data, target in testloader:
```

```
data, target = data.to(device), target.to(device) # loading to GPU
  output = model(data)
  pred = output.argmax(dim=1, keepdim=True)
  correct += pred.eq(target.view_as(pred)).sum().item()

test_loss /= len(testloader.dataset)

print('\nTest set: Accuracy: {}/{} ({:.0f}%)\n'.format(
  correct, len(testloader.dataset),
  100. * correct / len(testloader.dataset)))
class SaveOutput:
```

```
[]: 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()
    i = 0
    for layer in model.modules():
        i = i+1
        if isinstance(layer, QuantConv2d):
           print(i,"-th layer prehooked")
           layer.register_forward_pre_hook(save_output)
    dataiter = iter(testloader)
    images, labels = next(dataiter)
    images = images.to(device)
    out = model(images)
```

```
[]: weight_q = model.features[3].weight_q
w_alpha = model.features[3].weight_quant.wgt_alpha
w_bit = 4

weight_int = weight_q / (w_alpha / (2**(w_bit-1)-1))
print(weight_int)
```

```
[]: act = save_output.outputs[1][0]
act_alpha = model.features[3].act_alpha
act_bit = 4
act_quant_fn = act_quantization(act_bit)
```

```
act_q = act_quant_fn(act, act_alpha)
            act_int = act_q / (act_alpha / (2**act_bit-1))
            print(act_int)
[]: ## This cell is provided
            conv_int = torch.nn.Conv2d(in_channels = 64, out_channels=64, kernel_size = 3,__
              ⇒padding=1)
            conv_int.weight = torch.nn.parameter.Parameter(weight_int)
            conv int.bias = model.features[3].bias
            output_int = conv_int(act_int)
            output_recovered = output_int * (act_alpha / (2**act_bit-1)) * (w_alpha / __
             \hookrightarrow (2**(w_bit-1)-1))
            print(output_recovered)
[]: ## This cell is provided
            conv_ref = torch.nn.Conv2d(in_channels = 64, out_channels=64, kernel_size = 3,__
              ⇒padding=1)
            conv_ref.weight = model.features[3].weight_q
            conv_ref.bias = model.features[3].bias
            output_ref = conv_ref(act)
            #print(output_ref)
            print(abs((output_ref - output_recovered)).mean())
[]:  # act int.size = torch.Size([128, 64, 32, 32]) <- batch_size, input_ch, ni, nj
            a_int = act_int[0,:,:,:] # pick only one input out of batch
            # a_int.size() = [64, 32, 32]
            \# conv_int.weight.size() = torch.Size([64, 64, 3, 3]) \leftarrow output_ch, input_ch, input_c
            w_int = torch.reshape(weight_int, (weight_int.size(0), weight_int.size(1), -1))_u
             → # merge ki, kj index to kij
            # w_int.weight.size() = torch.Size([64, 64, 9])
            padding = 1
            stride = 1
            array_size = 8 # row and column number
           nig = range(a_int.size(1)) ## ni group
            njg = range(a_int.size(2)) ## nj group
            icg = range(int(w int.size(1))) ## input channel
```

ocg = range(int(w_int.size(0))) ## output channel

```
ic_tileg = range(int(len(icg)/array_size))
    oc_tileg = range(int(len(ocg)/array_size))
    kijg = range(w_int.size(2))
    ki_dim = int(math.sqrt(w_int.size(2))) ## Kernel's 1 dim size
    ####### Padding before Convolution ######
    a_pad = torch.zeros(len(icg), len(nig)+padding*2, len(nig)+padding*2).cuda()
    \# a_pad.size() = [64, 32+2pad, 32+2pad]
    a_pad[:, padding:padding+len(nig), padding:padding+len(njg)] = a_int.cuda()
    a_pad = torch.reshape(a_pad, (a_pad.size(0), -1))
    \# a_pad.size() = [64, (32+2pad)*(32+2pad)]
    a_tile = torch.zeros(len(ic_tileg), array_size, a_pad.size(1)).cuda()
    w_tile = torch.zeros(len(oc_tileg)*len(ic_tileg), array_size, array_size,_u
     →len(kijg)).cuda()
    for ic_tile in ic_tileg:
        a_tile[ic_tile,:,:] = a_pad[ic_tile*array_size:(ic_tile+1)*array_size,:]
    for ic_tile in ic_tileg:
        for oc_tile in oc_tileg:
            w_tile[oc_tile*len(oc_tileg) + ic_tile,:,:,:] =
u
      w_int[oc_tile*array_size:(oc_tile+1)*array_size, ic_tile*array_size:
      p_nijg = range(a_pad.size(1)) ## psum nij group
    psum = torch.zeros(len(ic_tileg), len(oc_tileg), array_size, len(p_nijg),_u
     →len(kijg)).cuda()
    for kij in kijg:
        for ic_tile in ic_tileg: # Tiling into array_sizeXarray_size array
            for oc_tile in oc_tileg: # Tiling into array_sizeXarray_size array
                                      # time domain, sequentially given input
                for nij in p_nijg:
                       m = nn.Linear(array_size, array_size, bias=False)
                       #m.weight = torch.nn.Parameter(w_int[oc_tile*array_size:
      →(oc_tile+1)*array_size, ic_tile*array_size:(ic_tile+1)*array_size, kij])
                       m.weight = torch.nn.
     →Parameter(w_tile[len(oc_tileg)*oc_tile+ic_tile,:,:,kij])
                       psum[ic_tile, oc_tile, :, nij, kij] = m(a_tile[ic_tile,:
      →,nij]).cuda()
```

```
[]: import math
     a_pad_ni_dim = int(math.sqrt(a_pad.size(1))) # 32
     o_ni_dim = int((a_pad_ni_dim - (ki_dim- 1) - 1)/stride + 1)
     o_nijg = range(o_ni_dim**2)
     out = torch.zeros(len(ocg), len(o_nijg)).cuda()
     ### SFP accumulation ###
     for o_nij in o_nijg:
         for kij in kijg:
             for ic_tile in ic_tileg:
                 for oc tile in oc tileg:
                     out[oc_tile*array_size:(oc_tile+1)*array_size, o_nij] = __
      →out[oc_tile*array_size:(oc_tile+1)*array_size, o_nij] + \
                     psum[ic_tile, oc_tile, :, int(o_nij/o_ni_dim)*a_pad_ni_dim +__
      →o_nij%o_ni_dim + int(kij/ki_dim)*a_pad_ni_dim + kij%ki_dim, kij]
                     ## 4th \ index = (int(o_nij/30)*32 + o_nij\%30) + (int(kij/3)*32 + o_nij\%30)
      ⇔kij%3)
[]: out_2D = torch.reshape(out, (out.size(0), o_ni_dim, -1))
     difference = (out_2D - output_int[0,:,:,:])
     print(difference.sum())
[]: ### show this cell partially. The following cells should be printed by students,
     →###
     tile_id = 0
     nij = 200 # just a random number
     X = a_tile[tile_id,:,nij:nij+64] # [tile_num, array row num, time_steps]
     bit precision = 4
     file = open('activation_hw7.txt', 'w') #write to file
     file.write('#timeOrow7[msb-lsb],timeOrow6[msb-lst],...,timeOrow0[msb-lst]#\n')
     file.write('#time1row7[msb-lsb],time1row6[msb-lst],...,time1row0[msb-lst]#\n')
     file.write('#....#\n')
     for i in range(X.size(1)): # time step
         for j in range(X.size(0)): # row #
             X_{bin} = '\{0:04b\}'.format(int(X[7-j,i].item()+0.001))
             for k in range(bit_precision):
                 file.write(X_bin[k])
             #file.write(' ')  # for visibility with blank between words, you can use
         file.write('\n')
```

```
file.close() #close file
```

```
[]: ### Complete this cell ###
     tile id = 0
     kij = 0
     W = w_tile[tile_id,:,:,kij] # w_tile[tile_num, array col num, array row num,_
      \hookrightarrow kij
     bit_precision = 4
     file = open('weight.txt', 'w') #write to file
     file.write('#col0row7[msb-lsb],col0row6[msb-lst],...,col0row0[msb-lst]#\n')
     file.write('#col1row7[msb-lsb],col1row6[msb-lst],...,col1row0[msb-lst]#\n')
     file.write('#....#\n')
     for i in range(W.size(0)): # time step
         for j in range(W.size(1)): # row #
             if (W[7-j,i].item()<0):
                 W_{bin} = {0:04b}'.format(int(W[7-j,i].item()+2**bit_precision+0.
      →001))
             else:
                 W_{bin} = '\{0:04b\}'.format(int(W[7-j,i].item()+0.001))
             for k in range(bit_precision):
                 file.write(W_bin[k])
         file.write('\n')
     file.close()
```

[]: W[0,:] # check this number with your 2nd line in weight.txt

```
[]: ### Complete this cell ###
ic_tile_id = 0
oc_tile_id = 0

kij = 0
nij = 200
psum_tile = psum[ic_tile_id,oc_tile_id,:,nij:nij+64,kij]
# psum[len(ic_tileg), len(oc_tileg), array_size, len(p_nijg), len(kijg)]

bit_precision = 16
file = open('psum_hw7.txt', 'w') #write to file
file.write('#time0col7[msb-lsb],time0col6[msb-lst],...,time0col0[msb-lst]#\n')
file.write('#time1col7[msb-lsb],time1col6[msb-lst],...,time1col0[msb-lst]#\n')
file.write('#...........#\n')

for i in range(psum_tile.size(1)): # time step
```

```
for j in range(psum_tile.size(0)): # row #
             if (psum_tile[7-j,i].item()<0):</pre>
                 psum_tile_bin = '{0:016b}'.format(int(psum_tile[7-j,i].
      ⇔item()+2**bit_precision+0.001))
             else:
                 psum_tile_bin = '{0:016b}'.format(int(psum_tile[7-j,i].item()+0.
      ⇔001))
             for k in range(bit_precision):
                 file.write(psum_tile_bin[k])
             #file.write(' ') # for visibility with blank between words, you can use
         file.write('\n')
     file.close()
[]:
[]:
[]:
[]:
[]:
[]:
[]:
[]:
```