[HW_Code6]_Memory_Write_Read

November 26, 2025

[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

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([
```

```python
        transforms.RandomCrop(32, padding=4),
        transforms.RandomHorizontalFlip(),
        transforms.ToTensor(),
        normalize,
    ]))
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))
```

```python
            # measure elapsed time
            batch_time.update(time.time() - end)
            end = time.time()

            if i % print_freq == 0:  # This line shows how frequently print out
 ↪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
```

```python
        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␣
 ↪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…
VGG_quant(
  (features): Sequential(
    (0): QuantConv2d(
      3, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (2): ReLU(inplace=True)
    (3): QuantConv2d(
      64, 64, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (4): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (5): ReLU(inplace=True)
    (6): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
    (7): QuantConv2d(
      64, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
```

```
    )
    (8): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (9): ReLU(inplace=True)
    (10): QuantConv2d(
      128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (11): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (12): ReLU(inplace=True)
    (13): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
    (14): QuantConv2d(
      128, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (15): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (16): ReLU(inplace=True)
    (17): QuantConv2d(
      256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (18): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (19): ReLU(inplace=True)
    (20): QuantConv2d(
      256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (21): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (22): ReLU(inplace=True)
    (23): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
    (24): QuantConv2d(
      256, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (25): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (26): ReLU(inplace=True)
    (27): QuantConv2d(
      512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (28): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
```

```
track_running_stats=True)
    (29): ReLU(inplace=True)
    (30): QuantConv2d(
      512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (31): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (32): ReLU(inplace=True)
    (33): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
    (34): QuantConv2d(
      512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (35): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (36): ReLU(inplace=True)
    (37): QuantConv2d(
      512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (38): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (39): ReLU(inplace=True)
    (40): QuantConv2d(
      512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1), bias=False
      (weight_quant): weight_quantize_fn()
    )
    (41): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    (42): ReLU(inplace=True)
    (43): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1,
ceil_mode=False)
    (44): AvgPool2d(kernel_size=1, stride=1, padding=0)
  )
  (classifier): Linear(in_features=512, out_features=10, bias=True)
)
Files already downloaded and verified
Files already downloaded and verified
```

[2]:
```
PATH = "result/VGG16_quant/model_best.pth.tar"
checkpoint = torch.load(PATH)
model.load_state_dict(checkpoint['state_dict'])
device = torch.device("cuda")

model.cuda()
```

```python
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)))
```

```
Test set: Accuracy: 8745/10000 (87%)
```

[3]:
```python
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)
```

```
3 -th layer prehooked
7 -th layer prehooked
```

```
12 -th layer prehooked
16 -th layer prehooked
21 -th layer prehooked
25 -th layer prehooked
29 -th layer prehooked
34 -th layer prehooked
38 -th layer prehooked
42 -th layer prehooked
47 -th layer prehooked
51 -th layer prehooked
55 -th layer prehooked
```

[4]:
```
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)
```

[5]:
```
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)
```

[6]:
```
## 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 /␣
 ↪(2**(w_bit-1)-1))
# print(output_recovered)
```

[7]:
```
## 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())
```

```
tensor(2.2279, device='cuda:0', grad_fn=<MeanBackward0>)
```

[13]:

```
# 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])  <- output_ch, input_ch,␣
↪ki, kj
w_int = torch.reshape(weight_int, (weight_int.size(0), weight_int.size(1), -1))␣
↪ # 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,␣
↪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,:,:,:] =␣
↪w_int[oc_tile*array_size:(oc_tile+1)*array_size, ic_tile*array_size:
↪(ic_tile+1)*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),␣
↪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   ␣
↪
            for nij in p_nijg:       # time domain, sequentially given input
                    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()
```

[14]:
```
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 +␣
↪kij%3)
```

[15]:
```
out_2D = torch.reshape(out, (out.size(0), o_ni_dim, -1))
difference = (out_2D - output_int[0,:,:,:])
print(difference.sum())
```

```
tensor(0.0088, device='cuda:0', grad_fn=<SumBackward0>)
```

[16]:
```
out_2D.shape
```

[16]:
```
torch.Size([64, 32, 32])
```

[17]:
```
### 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.txt', 'w') #write to file
file.write('#time0row7[msb-lsb],time0row6[msb-lst],....,time0row0[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(round(X[7-j,i].item()))
        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
```

[48]:
```
### Complete this cell ###
tile_id = 0
kij = 0
W = w_tile[tile_id,:,:,kij]  # w_tile[tile_num, array col num, array row num,␣
↪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(1)): # time step (col)
    for j in range(W.size(0)): # row
        weight = round(W[7-j,i].item())
        if weight < 0:
            weight = 16 + weight
        W_bin = '{0:04b}'.format(weight)
        for k in range(bit_precision):
            file.write(W_bin[k])
        file.write(' ')  # for visibility with blank between words, you can use
    file.write('\n')
file.close() #close file
```

[46]:
```
W[0,:] # check this number with your 2nd line in weight.txt
```

[46]:
```
tensor([ 3.0000, -4.0000,  2.0000, -4.0000,  2.0000, -3.0000,  2.0000,  2.0000],
       device='cuda:0', grad_fn=<SliceBackward0>)
```

[58]:
```
### 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.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
    for j in range(psum_tile.size(0)): # col
        psum_dec = round(psum_tile[7-j,i].item())
        if psum_dec < 0:
            psum_dec = 2 ** 16 + psum_dec
        psum_bin = '{0:016b}'.format(psum_dec)
        for k in range(bit_precision):
            file.write(psum_bin[k])
        file.write(' ')
    file.write('\n')
file.close()
```

[ ]:
```

```

[ ]:
```

```

[ ]:
```

```

[ ]:
```

```

[ ]:
```

```

[ ]:
```

```