```
print("""
In [4]: ▶
            Immadi Shyam Prasad
                                   (AM.EN.U4CSE19164)
            D B R S Praneeth Varma (AM.EN.U4CSE19265)
            D S K Phani Chyavan
                                   (AM.EN.U4CSE19163)
            Immadi Shyam Prasad
                                    (AM.EN.U4CSE19164)
            D B R S Praneeth Varma (AM.EN.U4CSE19265)
            D S K Phani Chyavan
                                   (AM.EN.U4CSE19163)
In [1]:
         M
            import torch
            import torchvision
            from torchvision import transforms, datasets
            import os
            import matplotlib.pyplot as plt
            import numpy as np
            #https://www.kaggle.com/c/state-farm-distracted-driver-detection/data
            #path of test,train,validation
            train_dataset_path = "/mnt/batch/tasks/shared/LS_root/mounts/clusters/teslage
            test dataset path = "/mnt/batch/tasks/shared/LS root/mounts/clusters/teslagpu
            val dataset path = "/mnt/batch/tasks/shared/LS root/mounts/clusters/teslagpu/
            mean = [0.3124, 0.3782, 0.3708] # found by caliculation net.batch_mean_and_sa
            std = [0.2778, 0.3213, 0.3222]
            #data argumentaion resizing, normalizing, converting to tensor
            train transforms = transforms.Compose([transforms.Resize((80,80)),transforms.
            test_transforms = transforms.Compose([transforms.Resize((80,80)),transforms.T
            val_transforms = transforms.Compose([transforms.Resize((80,80)),transforms.Tompose())
            train_dataset = datasets.ImageFolder(root=train_dataset_path,transform=train_
            test dataset = datasets.ImageFolder(root=test dataset path,transform=test tra
            val_dataset = datasets.ImageFolder(root=val_dataset_path,transform=val_transf
            #data Loading
            train loader = torch.utils.data.DataLoader(dataset = train dataset,batch size
            test_loader = torch.utils.data.DataLoader(dataset = test_dataset,batch_size=1
            val loader = torch.utils.data.DataLoader(dataset = val dataset,batch size=10
```

```
In [2]: #function for printing images in grid
def show_transformed_images(dataset):
    loader = torch.utils.data.DataLoader(dataset, batch_size=6,shuffle=True)
    for data in loader:
        images , labels = data
        break
    print(labels)
    print(images.view(-1,80*80).shape)
    grid = torchvision.utils.make_grid(images,nrow=3)
    plt.figure(figsize=(11,11))
    plt.imshow(np.transpose(grid,(1,2,0)))
    plt.show()
```

Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

```
tensor([5, 6, 3, 0, 6, 1])
torch.Size([18, 6400])
```



## In [3]: \| torch.cuda.is\_available()

Out[3]: True

```
In [4]:
            import torch.nn as nn
            import torch.nn.functional as F
            #Net inheriting from nn. module
            class Net(nn.Module):
                  def __init__(self):
                    super().__init__()
                    self.fc1 = nn.Linear(3*80*80,100)
                    self.fc2 = nn.Linear(100,100)
                    self.fc3 = nn.Linear(100,60)
                    self.fc4 = nn.Linear(60,10)
                  def forward(self,x):
                    x = F.relu(self.fc1(x))
                    x = F.relu(self.fc2(x))
                    x = F.relu(self.fc3(x))
                    x = self.fc4(x)
                    return F.log_softmax(x,dim=1)
                  def batch_mean_and_sd(self,loader):
                    cnt = 0
                    fst moment = torch.empty(3)
                    snd moment = torch.empty(3)
                    for images, _ in loader:
                        b, c, h, w = images.shape
                        nb pixels = b * h * w
                        sum = torch.sum(images, dim=[0, 2, 3])
                        sum of square = torch.sum(images ** 2,
                                                   dim=[0, 2, 3])
                        fst_moment = (cnt * fst_moment + sum_) / (cnt + nb_pixels)
                        snd_moment = (cnt * snd_moment + sum_of_square) / (cnt + nb_pixel
                        cnt += nb pixels
                    mean, std = fst_moment, torch.sqrt(snd_moment - fst_moment ** 2)
                    print("mean and std: \n", mean, std)
            net = Net()
            print(net)
            #qpu usage
            device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
            net.to(device)
            Net(
              (fc1): Linear(in_features=19200, out_features=100, bias=True)
              (fc2): Linear(in_features=100, out_features=100, bias=True)
              (fc3): Linear(in features=100, out features=60, bias=True)
              (fc4): Linear(in_features=60, out_features=10, bias=True)
            )
   Out[4]: Net(
```

```
(fc1): Linear(in features=19200, out features=100, bias=True)
              (fc2): Linear(in_features=100, out_features=100, bias=True)
              (fc3): Linear(in_features=100, out_features=60, bias=True)
              (fc4): Linear(in_features=60, out_features=10, bias=True)
            )
        # net.batch mean and sd(train loader)
In [5]:
         x1,y1 = next(iter(train_loader)) #checking the tensor values of 1st image in
In [6]:
            x1[0]
   Out[6]: tensor([[[-0.0940, 0.0189, 0.0612,
                                                       1.0917, 1.2188, 0.4283],
                                                  . . . ,
                     [ 0.8094, 1.1764, 1.5294,
                                                       1.0917, 1.2188, 0.9788],
                                                  . . . ,
                     [ 1.7835, 1.6988, 1.6423,
                                                  . . . ,
                                                       1.1482, 1.1623, 1.2470],
                     [-1.0399, -1.0399, -1.0399, \dots, -0.6022, -0.8705, -0.9128],
                     [-1.0399, -1.0399, -1.0399, \dots, -0.4893, -0.8563, -0.8987],
                     [-1.0399, -1.0399, -1.0540,
                                                  \dots, -0.7293, -0.9128, -0.9128]],
                    [-0.1396, -0.0420, -0.0176,
                                                  \dots, 1.7522, 1.7034, 0.6781],
                    [ 0.7269, 1.0199, 1.2762,
                                                  ..., 1.7034, 1.7766, 1.3616],
                                                  ...,
                     [ 1.8742, 1.7522, 1.6667,
                                                       1.7400, 1.8132, 1.7888],
                     [-1.0917, -1.0917, -1.0917, \dots, -0.6523, -0.9940, -1.0550],
                     [-1.0917, -1.0917, -1.0917,
                                                 ..., -0.5790, -0.9696, -1.0306],
                     [-1.0917, -1.0917, -1.1039,
                                                  ..., -0.8475, -0.9940, -1.0062]],
                    [[-0.3962, -0.2989, -0.2867, \ldots, 1.8798, 1.7824, 0.8087],
                                                       1.8068, 1.8555, 1.4660],
                     [ 0.3949, 0.6870, 0.9913,
                                                  . . . ,
                     [ 1.5025, 1.4051, 1.3686,
                                                  ..., 1.8920, 1.8920, 1.8433],
                     [-1.1265, -1.1265, -1.1265, ..., -0.6031, -0.9926, -1.0535],
                     [-1.1265, -1.1265, -1.1265, ..., -0.5179, -0.9439, -1.0291],
                     [-1.1265, -1.1265, -1.1387, \ldots, -0.7614, -0.9439, -0.9804]]])
In [7]:
         #training model
```

```
#printing loss at each epoch
```

```
In [13]:
             import torch.optim as optim
             optimizer = optim.Adam(net.parameters(), lr = 0.001)
             EPOCHS = 3
             for i in range(EPOCHS):
                 for data in train_loader:
                     x1, y1 = data
                     y1 = y1.to(device)
                     net.zero_grad()
                     output = net(x1.view(-1,3*80*80).to(device))
                     loss = F.nll_loss(output,y1)
                     loss.backward()
                     optimizer.step()
                 print(loss)
             tensor(0.0002, device='cuda:0', grad_fn=<NllLossBackward>)
             tensor(0.0041, device='cuda:0', grad_fn=<NllLossBackward>)
             tensor(0.0006, device='cuda:0', grad fn=<NllLossBackward>)
In [19]: ▶ #validating model
             correct = 0
             total = 0
             for data in val_loader:
                 x, y = data
                 output = net(x.view(-1,3*80*80).to(device))
                 for idx, i in enumerate(output):
                     if torch.argmax(i) == y[idx]:
                         correct += 1
                     total += 1
             print('accuracy:\t',round(correct/total,3))
                              0.951
             accuracy:
In [20]:
          #testing model
             correct = 0
             total = 0
             with torch.no_grad():
                 for data in test_loader:
                     x, y = data
                     output = net(x.view(-1,3*80*80).to(device))
                     for idx, i in enumerate(output):
                       if torch.argmax(i) == y[idx]:
                         correct += 1
                       total += 1
                 print('accuracy:\t',round(correct/total,3))
                              0.936
             accuracy:
          ▶ | x, y = next(iter(test_loader))
In [21]:
```

```
In [22]: ► x[0]
```

```
Out[22]: tensor([[[-0.5740, -0.4752, -0.1646,
                                                 . . . ,
                                                       1.4729,
                                                                1.9246,
                                                                          2.2069],
                   [-0.2634, 0.4283, -0.2211,
                                                       1.6846,
                                                                 1.9952,
                                                                          2.1222],
                                                 . . . ,
                   [-0.5740, 0.0330, -0.7011,
                                                  . . . ,
                                                       1.8964,
                                                                 1.9529,
                                                                          2.0234],
                   . . . ,
                   [-0.6305, -0.4893, -0.4470,
                                                 \dots, -0.8846, -0.8705, -0.8705],
                   [-0.5175, -0.4611, -0.4187,
                                                 \dots, -0.8987, -0.8846, -0.8563],
                   [-0.4328, -0.4470, -0.4187,
                                                  ..., -0.8846, -0.8705, -0.8563]],
                  [-0.6034, -0.4936, -0.2129,
                                                       1.8376, 1.9109,
                                                 . . . ,
                                                                          1.9231],
                   [-0.3471, 0.2387, -0.3593,
                                                                          1.9109],
                                                 . . . ,
                                                       1.8742,
                                                                1.9353,
                   [-0.6156, -0.1030, -0.7133,
                                                 . . . ,
                                                       1.9353, 1.9231,
                                                                          1.9109],
                   [-0.5180, -0.4204, -0.4082,
                                                 \dots, -0.9940, -0.9818, -0.9818],
                   [-0.4570, -0.3960, -0.3715,
                                                 \dots, -1.0062, -0.9940, -0.9696],
                   [-0.4204, -0.3715, -0.3471,
                                                 \dots, -0.9940, -0.9818, -0.9696]],
                  [-0.7370, -0.6153, -0.3719,
                                                       1.8920,
                                                                1.9163, 1.9407],
                                                 . . . ,
                   [-0.3232, 0.2367, -0.4327,
                                                 . . . ,
                                                       1.9041, 1.9528,
                                                                          1.9407],
                   [-0.6031, -0.1406, -0.8587,
                                                  . . . ,
                                                       1.9528,
                                                                 1.9285,
                                                                           1.9407],
                   [-0.5666, -0.4449, -0.3962,
                                                 \dots, -0.9074, -0.9074, -0.9074],
                   [-0.4692, -0.3962, -0.3597, \dots, -0.9196, -0.9318, -0.9074],
                   [-0.3962, -0.3719, -0.3232, \ldots, -0.9196, -0.9196, -0.9196]]]
```

## 

```
print(y[8])
plt.imshow(np.transpose(x[8],(1,2,0)))
plt.show()
```

Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

## tensor(0)

