```
In [1]:
            print("""
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In [1]:
            import torch
            import torchvision
            from torchvision import transforms, datasets
            import torch.nn as nn
            import torch.nn.functional as F
            import torch.optim as optim
            from torch.autograd import Variable
            from torch.utils.data import DataLoader
            from torch.utils.data import sampler
            from torch.utils.data import random split
            from torch.utils.data import SubsetRandomSampler
            from random import sample
            import matplotlib.pyplot as plt
            import glob
            from PIL import Image as I
            import pandas as pd
            import os
            import numpy as np
            import timeit
```

In [2]: ▶ pwd

```
In [3]:
            #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/mmahade
            test_dataset_path = "/mnt/batch/tasks/shared/LS_root/mounts/clusters/mmahadev
            val_dataset_path = "/mnt/batch/tasks/shared/LS_root/mounts/clusters/mmahadeva
            mean = [0.3124, 0.3782, 0.3708] # found by caliculation net.batch_mean_and_sq
            std = [0.2778, 0.3213, 0.3222]
            #data argumentaion resizing, normalizing, converting to tensor
            train_transforms = transforms.Compose([
                transforms.Resize((224,224)),
                transforms.ToTensor(),
                transforms.Normalize(torch.Tensor(mean),
                torch.Tensor(std))
            ])
            test_transforms = transforms.Compose([
                transforms.Resize((224,224)),
                transforms.ToTensor(),
                transforms.Normalize(torch.Tensor(mean),
                torch.Tensor(std))
            ])
            val transforms = transforms.Compose([
                transforms.Resize((224,224)),
                transforms.ToTensor(),
                transforms.Normalize(torch.Tensor(mean),
                torch.Tensor(std))
            ])
            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
```

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

tensor([1, 6, 9, 6, 0, 5]) torch.Size([18, 50176])



In [5]: ▶ torch.cuda.is_available()

```
In [6]:

    def my_plot(epochs, loss,title1,x_label1,y_label1):

                plt.plot(epochs, loss)
                plt.title(title1)
                plt.xlabel(x_label1)
                plt.ylabel(y_label1)
                return plt.show()
            def just_plot(accu,accu1,title1,y_label1):
                epochs = range(1,4)
                plt.plot(epochs, accu, 'g', label='Training accuracy')
                plt.plot(epochs, accu1, 'b', label='validation accuracy')
                plt.title(title1)
                plt.xlabel('Epochs')
                plt.ylabel(y_label1)
                plt.legend()
                return plt.show()
```

```
In [7]: N

num_classes = 10
net = torchvision.models.alexnet()
net.fc = nn.Linear(100, num_classes)
gpu_net = net.to('cuda')

import torch.optim as optim
from sklearn.metrics import accuracy_score

criterion = nn.CrossEntropyLoss()
optimizer = optim.SGD(gpu_net.parameters(), lr=0.001, momentum=0.9, nesterov=
```

```
In [8]:
            train_loss= []
            train_accuracy=[]
            num_epochs = 3
            for epoch in range(num_epochs): # Loop over the dataset multiple times
                epoch_loss= []
                running_train_loss = 0.0
                total=0
                correct = 0
                print('train load')
                for i, data in enumerate(train_loader):
                    # get the inputs; data is a list of [inputs, labels]
                    inputs, labels = data
                    inputs = inputs.to('cuda')
                    labels = labels.to('cuda')
                    # zero the parameter gradients
                    optimizer.zero_grad()
                    # forward + backward + optimize
                    outputs = gpu_net(inputs)
                    loss = criterion(outputs, labels)
                    epoch_loss.append(loss.item())
                    loss.backward()
                    optimizer.step()
                    for idx, k in enumerate(outputs):
                      if torch.argmax(k) == labels[idx]:
                        correct += 1
                      total += 1
                    # print statistics
                    running_train_loss += loss.item()
                    if i % 50 == 49:
                                       # print every 50 mini-batches
                        print('[%d, %5d] Train loss: %.3f' %
                              (epoch + 1, i + 1, running_train_loss / 50))
                        running_train_loss = 0.0
                train_accuracy.append(round(correct/total,3))
                train_loss.append(sum(epoch_loss)/len(epoch_loss))
                running_val_loss = 0.0
            print('Finished Training')
            [1, 1200] Train loss: 1.400
            train load
                  50] Train loss: 1.273
            [2,
```

Γ2.

1001 Train loss: 1.289

```
100] 110111 1000. 1.200
ر ڪ ي
[2,
      150] Train loss: 1.207
      200] Train loss: 1.220
[2,
      250] Train loss: 1.226
[2,
      300] Train loss: 1.022
[2,
[2,
      350] Train loss: 0.918
      400] Train loss: 0.987
[2,
[2,
      450] Train loss: 0.996
      500] Train loss: 0.982
[2,
      550] Train loss: 0.944
[2,
[2,
      600] Train loss: 0.829
[2,
      650] Train loss: 0.823
[2,
      700] Train loss: 0.743
      750] Train loss: 0.725
[2,
      800] Train loss: 0.606
[2,
      850] Train loss: 0.734
[2,
      900] Train loss: 0.701
[2,
```

```
In [9]:
         N val_loss= []
            val_accuracy=[]
            num epochs = 3
            for epoch in range(num_epochs):
                print('val load')
                epoch_loss= []
                running_val_loss = 0.0
                correct = 0
                total=0
                for i, data in enumerate(val_loader):
                    # get the inputs; data is a list of [inputs, labels]
                    inputs, labels = data
                    inputs = inputs.to('cuda')
                    labels = labels.to('cuda')
                    # zero the parameter gradients
                    optimizer.zero_grad()
                    # forward + backward + optimize
                    outputs = gpu_net(inputs)
                    loss = criterion(outputs, labels)
                    epoch_loss.append(loss.item())
                    loss.backward()
                    optimizer.step()
                    for idx, j in enumerate(outputs):
                      if torch.argmax(j) == labels[idx]:
                        correct += 1
                      total += 1
                    running_train_loss += loss.item()
                    if i % 50 == 49:
                                       # print every 50 mini-batches
                        print('[%d, %5d] Train loss: %.3f' %
                              (epoch + 1, i + 1, running_train_loss / 50))
                        running_train_loss = 0.0
                val_accuracy.append(round(correct/total,3))
                val_loss.append(sum(epoch_loss)/len(epoch_loss))
                running_val_loss = 0.0
            print('validation')
                TOO 11 0TH TO33. 0.70T
            LΤν
            [1,
                  150] Train loss: 0.279
                  200] Train loss: 0.326
            [1,
            [1,
                  250] Train loss: 0.265
            [1,
                  300] Train loss: 0.339
            val load
            [2,
                  50] Train loss: 0.448
            [2,
                  100] Train loss: 0.150
            [2, 150] Train loss: 0.250
            [2,
                200] Train loss: 0.225
            [2,
                  250] Train loss: 0.197
            Γ2,
                  300] Train loss: 0.208
            val load
                   50] Train loss: 0.425
            [3,
```

```
[3, 100] Train loss: 0.190

[3, 150] Train loss: 0.138

[3, 200] Train loss: 0.150

[3, 250] Train loss: 0.192

[3, 300] Train loss: 0.192

validation
```

In [11]: print(train_accuracy,val_accuracy)

[0.232, 0.692, 0.864] [0.906, 0.935, 0.946]

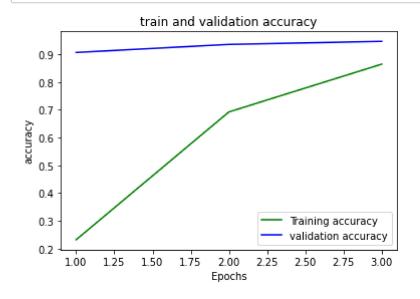
In [12]: print(train_loss,val_loss)

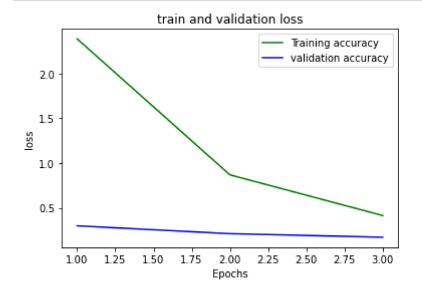
In [14]:

[2.385819746556337, 0.8672663599325865, 0.41273387137468165] [0.29770814210 98237, 0.21012411083836632, 0.17033227367585466]

In [13]: Ŋ just_plot(train_accuracy, val_accuracy, "train and validation accuracy", "accur

just_plot(train_loss, val_loss, "train and validation loss", "loss")

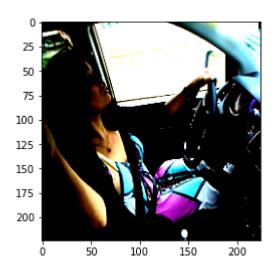




```
In [15]:
          ▶ #testing model
             correct = 0
             total = 0
             with torch.no grad():
                  for data in test_loader:
                      x, y = data
                      output = gpu_net(x.to('cuda'))
                      for idx, i in enumerate(output):
                        if torch.argmax(i) == y[idx]:
                          correct += 1
                        total += 1
                  print('accuracy:\t',round(correct/total,3))
             accuracy:
                               0.937
          | x, y = next(iter(test_loader))
In [16]:
In [17]:
          \mathbf{M} \times [0]
    Out[17]: tensor([[[-0.8987, -0.8987, -0.8987,
                                                           2.1928, 2.2069,
                                                                              1.8964],
                                                     . . . ,
                       [-0.8987, -0.8987, -0.8987,
                                                     . . . ,
                                                           2.0940,
                                                                    2.2493,
                                                                              2.2493],
                       [-0.8987, -0.8987, -0.8987,
                                                     . . . ,
                                                           1.8540, 2.0517,
                                                                              2.1505],
                       . . . ,
                       [-0.7293, -0.7575, -0.7999,
                                                     \dots, -0.9693, -0.9834, -0.9834],
                                                     \dots, -0.9834, -0.9834, -0.9834],
                       [-0.7716, -0.7999, -0.8422,
                       [-0.8563, -0.8846, -0.9128,
                                                     ..., -0.9834, -0.9834, -0.9834]],
                      [[-0.9696, -0.9696, -0.9696,
                                                     \dots, 1.9231, 1.8742, 1.5813],
                       [-0.9696, -0.9696, -0.9696,
                                                           1.8620, 1.9231, 1.9109],
                                                     . . . ,
                       [-0.9696, -0.9696, -0.9696,
                                                           1.7034, 1.8376,
                                                                              1.8986],
                                                     . . . ,
                       [-0.7621, -0.7865, -0.8231,
                                                     ..., -1.0672, -1.0795, -1.0672],
                                                     ..., -1.0795, -1.0795, -1.0672],
                       [-0.7987, -0.8231, -0.8598,
                       [-0.8720, -0.8964, -0.9208,
                                                     ..., -1.0795, -1.0795, -1.0672]],
                      [-1.0048, -1.0048, -1.0048,
                                                          1.6729,
                                                                    1.6972, 1.4538],
                                                     . . . ,
                       [-1.0048, -1.0048, -1.0048,
                                                           1.5999, 1.7581, 1.7703],
                                                     . . . ,
                       [-1.0048, -1.0048, -1.0048,
                                                           1.4295,
                                                                    1.6120,
                                                     . . . ,
                                                                             1.6972],
                       [-0.7370, -0.7614, -0.7979,
                                                     ..., -1.0291, -1.0535, -1.0656],
                       [-0.7735, -0.7979, -0.8344, \ldots, -1.0413, -1.0535, -1.0656],
                       [-0.8466, -0.8709, -0.8952,
                                                     \dots, -1.0413, -1.0535, -1.0656]]])
```


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

tensor(8)



```
In [19]:
          output = gpu_net(x.to('cuda'))
             print(torch.max(output.data,1))
             for idx, i in enumerate(output):
                 if torch.argmax(i) == y[idx]:
                     print(' correct',torch.argmax(i),y[idx])
                 else:
                     print('uncorrect',torch.argmax(i),y[idx])
             torch.return_types.max(
             values=tensor([17.5003, 28.5840, 41.8555, 18.6444, 28.8190, 28.0235, 20.824
             3, 21.7966,
                     25.3960, 29.9994], device='cuda:0'),
             indices=tensor([3, 4, 6, 9, 3, 3, 7, 7, 8, 7], device='cuda:0'))
             uncorrect tensor(3, device='cuda:0') tensor(0)
               correct tensor(4, device='cuda:0') tensor(4)
               correct tensor(6, device='cuda:0') tensor(6)
               correct tensor(9, device='cuda:0') tensor(9)
               correct tensor(3, device='cuda:0') tensor(3)
               correct tensor(3, device='cuda:0') tensor(3)
               correct tensor(7, device='cuda:0') tensor(7)
               correct tensor(7, device='cuda:0') tensor(7)
               correct tensor(8, device='cuda:0') tensor(8)
               correct tensor(7, device='cuda:0') tensor(7)
 In [ ]:
 In [ ]:
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