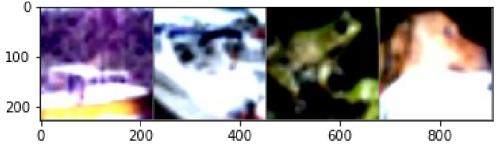
```
import torch
import torchvision
import torchvision.transforms as transforms
transform = transforms.Compose([
    transforms.Resize(256).
    transforms.CenterCrop(224),
    transforms.ToTensor(),
    transforms.Normalize(mean=[0.485, 0.456, 0.406], std=[0.229,
0.224, 0.225]),
train data = torchvision.datasets.CIFAR10(root='./data', train=True,
download=True, transform=transform)
trainloader = torch.utils.data.DataLoader(train data, batch size=4,
shuffle=True, num workers=2)
test data = torchvision.datasets.CIFAR10(root='./data', train=False,
download=True, transform=transform)
testloader = torch.utils.data.DataLoader(test data, batch size=4,
shuffle=False, num workers=2)
classes = ('Airplane', 'Car', 'Bird', 'Cat', 'Deer', 'Dog', 'Frog',
'Horse', 'Ship', 'Truck')
import matplotlib.pyplot as plt
import numpy as np
# functions to show an image
def imshow(img):
    img = img / 2 + 0.5 # unnormalize
    npimq = imq.numpy()
    plt.imshow(np.transpose(npimg, (1, 2, 0)))
    plt.show()
# get some random training images
dataiter = iter(trainloader)
images, labels = dataiter.next()
# show images
imshow(torchvision.utils.make grid(images))
# print labels
print(' '.join('%5s' % classes[labels[j]] for j in range(4)))
```



```
Ship Ship Froq
                    Dog
# using the AlexNet
AlexNet Model = torch.hub.load('pytorch/vision:v0.6.0', 'alexnet',
pretrained=True)
AlexNet Model.eval()
import torch.nn as nn
AlexNet Model.classifier[1] = nn.Linear(9216,4096)
AlexNet Model.classifier[4] = nn.Linear(4096,1024)
AlexNet Model.classifier[6] = nn.Linear(1024,10)
AlexNet Model.eval()
AlexNet(
  (features): Sequential(
    (0): Conv2d(3, 64, kernel size=(11, 11), stride=(4, 4),
padding=(2, 2)
    (1): ReLU(inplace=True)
    (2): MaxPool2d(kernel size=3, stride=2, padding=0, dilation=1,
ceil mode=False)
    (3): Conv2d(64, 192, kernel size=(5, 5), stride=(1, 1),
padding=(2, 2)
    (4): ReLU(inplace=True)
    (5): MaxPool2d(kernel size=3, stride=2, padding=0, dilation=1,
ceil mode=False)
    (6): Conv2d(192, 384, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1)
    (7): ReLU(inplace=True)
    (8): Conv2d(384, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1)
    (9): ReLU(inplace=True)
    (10): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1)
    (11): ReLU(inplace=True)
    (12): MaxPool2d(kernel size=3, stride=2, padding=0, dilation=1,
ceil mode=False)
  (avgpool): AdaptiveAvgPool2d(output size=(6, 6))
  (classifier): Sequential(
    (0): Dropout(p=0.5, inplace=False)
```

```
(1): Linear(in features=9216, out features=4096, bias=True)
    (2): ReLU(inplace=True)
    (3): Dropout(p=0.5, inplace=False)
    (4): Linear(in features=4096, out features=1024, bias=True)
    (5): ReLU(inplace=True)
    (6): Linear(in features=1024, out features=10, bias=True)
  )
)
# move the input and model to GPU for speed if available
device = torch.device("cuda:0" if torch.cuda.is available() else
"cpu")
print(device)
cuda:0
AlexNet Model.to(device)
AlexNet(
  (features): Sequential(
    (0): Conv2d(3, 64, kernel size=(11, 11), stride=(4, 4),
padding=(2, 2)
    (1): ReLU(inplace=True)
    (2): MaxPool2d(kernel size=3, stride=2, padding=0, dilation=1,
ceil mode=False)
    (3): Conv2d(64, 192, kernel size=(5, 5), stride=(1, 1),
padding=(2, 2)
    (4): ReLU(inplace=True)
    (5): MaxPool2d(kernel size=3, stride=2, padding=0, dilation=1,
ceil mode=False)
    (6): Conv2d(192, 384, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1)
    (7): ReLU(inplace=True)
    (8): Conv2d(384, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1)
    (9): ReLU(inplace=True)
    (10): Conv2d(256, 256, \text{kernel size}=(3, 3), \text{stride}=(1, 1),
padding=(1, 1)
    (11): ReLU(inplace=True)
    (12): MaxPool2d(kernel size=3, stride=2, padding=0, dilation=1,
ceil mode=False)
  (avgpool): AdaptiveAvgPool2d(output size=(6, 6))
  (classifier): Sequential(
    (0): Dropout(p=0.5, inplace=False)
    (1): Linear(in features=9216, out features=4096, bias=True)
    (2): ReLU(inplace=True)
    (3): Dropout(p=0.5, inplace=False)
    (4): Linear(in features=4096, out features=1024, bias=True)
    (5): ReLU(inplace=True)
```

```
(6): Linear(in features=1024, out features=10, bias=True)
  )
)
import torch.optim as optim
criterion = nn.CrossEntropyLoss()
optimizer = optim.SGD(AlexNet Model.parameters(), lr=0.001,
momentum=0.9)
import time
for epoch in range(7): # loop over the dataset multiple times
    running loss = 0.0
    start time = time.time()
    for i, data in enumerate(trainloader, 0):
        # get the inputs; data is a list of [inputs, labels]
        inputs, labels = data[0].to(device), data[1].to(device)
        # zero the parameter gradients
        optimizer.zero grad()
        # forward + backward + optimize
        output = AlexNet Model(inputs)
        loss = criterion(output, labels)
        loss.backward()
        optimizer.step()
        #Time
        end time = time.time()
        time taken = end time - start time
        # print statistics
        running loss += loss.item()
        if i % 2000 == 1999: # print every 2000 mini-batches
            print('[%d, %5d] loss: %.3f' % (epoch + 1, i + 1,
running loss / 2000))
            print('Time:',time_taken)
            running loss = 0.0
print('Finished Training of AlexNet')
[1, 2000] loss: 1.202
Time: 26.783158779144287
[1, 4000] loss: 0.892
Time: 52.17550587654114
[1, 6000] loss: 0.831
Time: 76.89599442481995
[1, 8000] loss: 0.741
```

Time: 102.43446254730225 [1, 10000] loss: 0.703 Time: 127.73311996459961 [1, 12000] loss: 0.652 Time: 153.21423721313477 2000] loss: 0.517 [2. Time: 25.728044033050537 4000] loss: 0.518 [2. Time: 51.18855333328247 6000] loss: 0.503 [2, Time: 76.28134536743164 [2. 8000] loss: 0.503 Time: 102.24490404129028 [2, 10000] loss: 0.498 Time: 128.30848574638367 [2, 12000] loss: 0.479 Time: 154.10634303092957 2000] loss: 0.344 Time: 25.733392238616943 40001 loss: 0.345 [3. Time: 51.15397524833679 [3, 6000] loss: 0.361 Time: 76.16496968269348 [3, 8000] loss: 0.353 Time: 101.2636661529541 [3, 10000] loss: 0.362 Time: 126.73455238342285 [3, 12000] loss: 0.367 Time: 152.39756155014038 2000] loss: 0.228 Time: 25.683215379714966 [4. 4000] loss: 0.247 Time: 50.77170968055725 [4, 6000] loss: 0.271 Time: 76.3136351108551 8000] loss: 0.266 Time: 101.3867175579071 [4. 10000] loss: 0.279 Time: 126.63472771644592 [4, 12000] loss: 0.286 Time: 152.09565353393555 [5, 2000] loss: 0.159 Time: 25.56282687187195 [5, 4000] loss: 0.189 Time: 51.04423999786377 [5, 6000] loss: 0.188 Time: 76.4984700679779 [5, 80001 loss: 0.216 Time: 101.08033537864685 [5, 10000] loss: 0.219

```
Time: 126.45636343955994
[5, 12000] loss: 0.234
Time: 151.6725754737854
[6, 2000] loss: 0.134
Time: 25.55684208869934
[6. 4000] loss: 0.154
Time: 51.152565240859985
[6, 6000] loss: 0.162
Time: 76.65935111045837
[6. 8000] loss: 0.166
Time: 101.64705610275269
[6, 10000] loss: 0.162
Time: 127.05616068840027
[6, 12000] loss: 0.171
Time: 152.73403024673462
[7, 2000] loss: 0.102
Time: 25.616260290145874
[7, 4000] loss: 0.132
Time: 51.10871887207031
[7. 6000] loss: 0.131
Time: 76.59881019592285
[7, 8000] loss: 0.131
Time: 101.46219182014465
[7, 10000] loss: 0.135
Time: 126.87981843948364
[7, 12000] loss: 0.121
Time: 152.46708822250366
Finished Training of AlexNet
#Testing Accuracy
correct = 0
total = 0
with torch.no grad():
    for data in testloader:
        images, labels = data[0].to(device), data[1].to(device)
        outputs = AlexNet Model(images)
        , predicted = torch.max(outputs.data, 1)
        total += labels.size(0)
        correct += (predicted == labels).sum().item()
print('Accuracy of the network on the 10000 test images: %.2f %%' %
(100 * correct / total))
Accuracy of the network on the 10000 test images: 83.02 %
class correct = list(0. for i in range(10))
class_total = list(0. for i in range(10))
with torch.no grad():
    for data in testloader:
        images, labels = data[0].to(device), data[1].to(device)
        outputs = AlexNet Model(images)
```

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_, predicted = torch.max(outputs, 1)
       c = (predicted == labels).squeeze()
       for i in range (4):
           label = labels[i]
           class correct[label] += c[i].item()
           class total[label] += 1
for i in range(10):
    print('Accuracy of %5s: %2d %%' % (
       classes[i], 100 * class correct[i] / class total[i]))
Accuracy of Airplane : 83 %
Accuracy of Car: 93 %
Accuracy of Bird: 82 %
Accuracy of Cat: 70 %
Accuracy of Deer: 82 %
           Dog : 78 %
Accuracy of
Accuracy of Frog: 88%
Accuracy of Horse: 84 %
Accuracy of Ship: 75%
Accuracy of Truck: 90 %
#Verifying average accuracy of the network
avg = 0
for i in range(10):
 temp = (100 * class correct[i] / class total[i])
  avg = avg + temp
avg = avg/10
print('Average accuracy = ', avg)
Average accuracy = 83.0200000000001
```