Data Augmentation

```
import torch
import torchvision
import numpy as np
import matplotlib.pyplot as plt
import PIL
from torchsummary import summary
# From local helper files
from helper_evaluation import set_all_seeds, set_deterministic, compute_confusion_ma
from helper train import train model
from helper_plotting import plot_training_loss, plot_accuracy, show_examples, plot_c
from helper_dataset import get_dataloaders_cifar10, UnNormalize
RANDOM SEED = 123
BATCH SIZE = 256
NUM EPOCHS = 40
DEVICE = torch.device('cuda:0' if torch.cuda.is_available() else 'cpu')
train transforms = torchvision.transforms.Compose([
    torchvision.transforms.Resize((16, 16)),
    torchvision.transforms.ToTensor(),
    torchvision.transforms.RandomRotation(degrees=30, interpolation=PIL.Image.BILINE
    torchvision.transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))
1)
test transforms = torchvision.transforms.Compose([
      torchvision.transforms.Resize((16, 16)),
    torchvision.transforms.ToTensor(),
    torchvision.transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))
1)
train loader augmented, valid loader augmented, test loader augmented = get dataload
    batch size=BATCH SIZE,
    validation fraction=0.1,
    train_transforms=train_transforms,
    test transforms=test transforms,
    num workers=2)
```

/usr/local/lib/python3.7/dist-packages/torchvision/transforms/transforms.py:1306 "Argument 'interpolation' of type int is deprecated since 0.13 and will be rem Downloading https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz to data/cifa 170498071/170498071 [00:02<00:00, 86044312.77it/s] 100%

```
class CNN1(torch.nn.Module):
  def __init__(self, num_classes):
    super().__init__()
   self.features = torch.nn.Sequential(
           # Conv 1
           torch.nn.Conv2d(3, 16, kernel_size=3, padding="same"), # output 16 - 3 +
                           # , stride=4, padding=2),
           torch.nn.ReLU(inplace=True),
           torch.nn.MaxPool2d(kernel size=2), # 16 / 2 => output 8
           # Conv 2
           torch.nn.Conv2d(16, 32, kernel size=2, padding="same"), # output 7 - 2 +
                           # , padding=2),
           torch.nn.ReLU(inplace=True),
           torch.nn.MaxPool2d(kernel_size=2) #output 8 / 2 => output 4
    )
    self.classifier = torch.nn.Sequential(
        torch.nn.Linear(32*4*4, 100),
         torch.nn.ReLU(inplace=True),
         torch.nn.Linear(100, num classes),
    )
 def forward(self, x):
   x = self.features(x)
   x = torch.flatten(x, 1)
   # print(x.size())
    logits = self.classifier(x)
    return logits
model1 augmented = CNN1(num classes=10)
model1 augmented = model1 augmented.to(DEVICE)
print(summary(model1_augmented, (3, 16, 16)))
            Layer (type)
                                      Output Shape
                                                          Param #
    ______
                                  [-1, 16, 16, 16]
                                                              448
                Conv2d-1
                                  [-1, 16, 16, 16]
                  ReLU-2
                                                                0
```

[-1, 16, 8, 8]

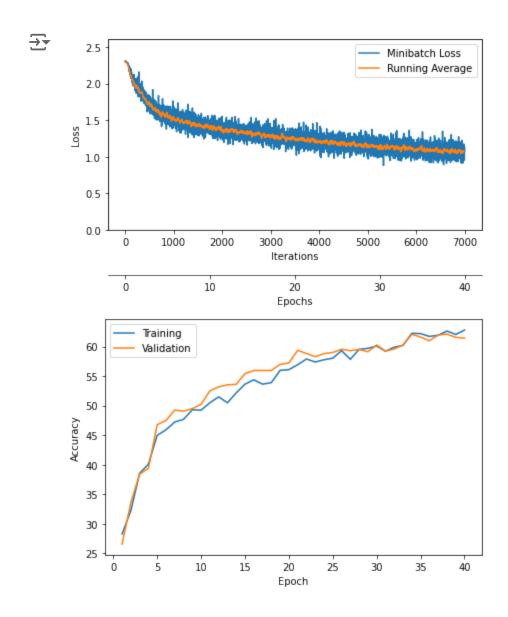
MaxPool2d-3

0

```
Conv2d-4
                                      [-1, 32, 8, 8]
                                                                2,080
                                      [-1, 32, 8, 8]
                   ReLU-5
                                      [-1, 32, 4, 4]
             MaxPool2d-6
                                                                    0
                                           [-1, 100]
                                                               51,300
                 Linear-7
                                           [-1, 100]
                   ReLU-8
                 Linear-9
                                            [-1, 10]
                                                                1,010
    Total params: 54,838
    Trainable params: 54,838
    Non-trainable params: 0
    Input size (MB): 0.00
    Forward/backward pass size (MB): 0.11
    Params size (MB): 0.21
    Estimated Total Size (MB): 0.32
    None
    /usr/local/lib/python3.7/dist-packages/torch/nn/modules/conv.py:454: UserWarning
      self.padding, self.dilation, self.groups)
optimizer_augmented = torch.optim.SGD(model1_augmented.parameters(), lr=0.1)
scheduler augmented = torch.optim.lr scheduler.ReduceLROnPlateau(optimizer augmented
                                                        factor=0.1,
                                                        mode='max'.
                                                        verbose=True)
minibatch_loss_list_augmented1, train_acc_list_augmented1, valid_acc_list_augmented1
    model=model1 augmented,
    num epochs=NUM EPOCHS,
    train loader=train loader augmented,
    valid_loader=valid_loader_augmented,
    test loader=test loader augmented,
    optimizer=optimizer augmented,
    device=DEVICE,
    scheduler=None,
    scheduler_on='valid_acc',
    logging interval=100)
plot_training_loss(minibatch_loss_list=minibatch_loss_list_augmented1,
                   num epochs=NUM EPOCHS,
                   iter_per_epoch=len(train_loader_augmented),
                   results dir=None,
                   averaging iterations=20)
plt.show()
plot accuracy(train acc list=train acc list augmented1,
              valid_acc_list=valid_acc_list_augmented1,
```

```
results_dir=None)
```

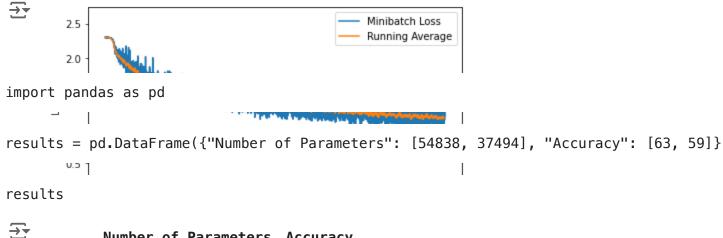
```
# plt.ylim([80, 100])
plt.show()
```

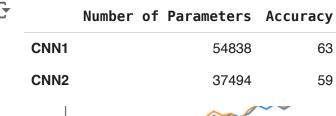


```
# , stride=4, padding=2),
            torch.nn.ReLU(inplace=True),
            torch.nn.MaxPool2d(kernel size=2), # 16 / 2 => output 8
            # Conv 2
            torch.nn.Conv2d(16, 32, kernel_size=2, padding="same"), # output 7 - 2 +
                            # , padding=2),
            torch.nn.ReLU(inplace=True),
            torch.nn.MaxPool2d(kernel_size=2), #output 8 / 2 => output 4
            # Conv 3
            torch.nn.Conv2d(32, 64, kernel_size=2, padding="same"), # output 7 - 2 +
                            # , padding=2),
            torch.nn.ReLU(inplace=True),
            torch.nn.MaxPool2d(kernel_size=2) #output 4 / 2 => output 2
    )
    self.classifier = torch.nn.Sequential(
         torch.nn.Linear(64*2*2, 100),
          torch.nn.ReLU(inplace=True),
          torch.nn.Linear(100, num_classes),
    )
  def forward(self, x):
    x = self.features(x)
   x = torch.flatten(x, 1)
    # print(x.size())
    logits = self.classifier(x)
    return logits
model2 augmented = CNN2(num classes=10)
model2 augmented = model2 augmented.to(DEVICE)
print(summary(model2 augmented, (3, 16, 16)))
```

<u> </u>			
ث	Layer (type)	Output Shape	Param #
	Conv2d-1 ReLU-2	[-1, 16, 16, 16] [-1, 16, 16, 16]	448 0
	MaxPool2d-3 Conv2d-4	[-1, 16, 8, 8] [-1, 32, 8, 8]	0 2,080
	ReLU-5 MaxPool2d-6	[-1, 32, 8, 8] [-1, 32, 4, 4]	0
	Conv2d-7	[-1, 64, 4, 4]	8,256
	ReLU-8 MaxPool2d-9	[-1, 64, 4, 4] [-1, 64, 2, 2]	0 0
	Linear-10	[-1, 100]	25 , 700

```
ReLU-11
                                         [-1, 100]
                                                                0
               Linear-12
                                          [-1, 10]
                                                            1,010
    ______
    Total params: 37,494
    Trainable params: 37,494
    Non-trainable params: 0
    Input size (MB): 0.00
    Forward/backward pass size (MB): 0.12
    Params size (MB): 0.14
    Estimated Total Size (MB): 0.27
    None
optimizer_augmented2 = torch.optim.SGD(model2_augmented.parameters(), lr=0.1)
scheduler_augmented2 = torch.optim.lr_scheduler.ReduceLROnPlateau(optimizer_augmente
                                                     factor=0.1,
                                                     mode='max',
                                                     verbose=True)
minibatch loss list augmented2, train acc list augmented2, valid acc list augmented2
   model=model2_augmented,
    num epochs=NUM EPOCHS,
    train loader=train loader augmented,
   valid_loader=valid_loader_augmented,
    test loader=test loader augmented,
    optimizer=optimizer_augmented2,
   device=DEVICE,
    scheduler=None,
    scheduler_on='valid_acc',
    logging interval=100)
plot training loss(minibatch loss list=minibatch loss list augmented2,
                  num epochs=NUM EPOCHS,
                  iter_per_epoch=len(train_loader_augmented),
                  results_dir=None,
                  averaging iterations=20)
plt.show()
plot_accuracy(train_acc_list=train_acc_list_augmented2,
             valid_acc_list=valid_acc_list_augmented2,
             results dir=None)
# plt.ylim([80, 100])
plt.show()
```





Even with data augmentation the simpler CNN works better as there are more pixels to work with.

