```
In [ ]: # Reference: Thanks to
    # Moon, T. (2018, November 23). Tjmoon0104/pytorch-tiny-imagenet. Retrie
    ved June 14,
        # 2020, from https://github.com/tjmoon0104/pytorch-tiny-imagenet

#Zakka, K. (2017). Train, Validation and Test Split for torchvision Data
    sets. Retrieved June
        #14, 2020, from https://gist.github.com/kevinzakka/d33bf8d6c7f06a9d8
        c76d97a7879f5cb

#Luo, Z. (2018, September 21). Icpm/pytorch-cifar10. Retrieved June 14,
        2020, from
        #https://github.com/icpm/pytorch-cifar10/tree/master/models
```

Setup

In [2]: pip install livelossplot

```
Requirement already satisfied: livelossplot in ./.local/lib/python3.7/s
ite-packages (0.5.1)
Requirement already satisfied: bokeh; python version >= "3.6" in /opt/c
onda/lib/python3.7/site-packages (from livelossplot) (1.3.4)
Requirement already satisfied: matplotlib; python version >= "3.6" in /
opt/conda/lib/python3.7/site-packages (from livelossplot) (3.1.1)
Requirement already satisfied: ipython in /opt/conda/lib/python3.7/site
-packages (from livelossplot) (7.7.0)
Requirement already satisfied: numpy>=1.7.1 in /opt/conda/lib/python3.
7/site-packages (from bokeh; python version >= "3.6"->livelossplot) (1.
16.4)
Requirement already satisfied: PyYAML>=3.10 in /opt/conda/lib/python3.
7/site-packages (from bokeh; python version >= "3.6"->livelossplot) (5.
Requirement already satisfied: tornado>=4.3 in /opt/conda/lib/python3.
7/site-packages (from bokeh; python_version >= "3.6"->livelossplot) (6.
0.3)
Requirement already satisfied: Jinja2>=2.7 in /opt/conda/lib/python3.7/
site-packages (from bokeh; python_version >= "3.6"->livelossplot) (2.1
0.1)
Requirement already satisfied: six>=1.5.2 in /opt/conda/lib/python3.7/s
ite-packages (from bokeh; python_version >= "3.6"->livelossplot) (1.12.
0)
Requirement already satisfied: packaging>=16.8 in /opt/conda/lib/python
3.7/site-packages (from bokeh; python_version >= "3.6"->livelossplot)
(19.0)
Requirement already satisfied: python-dateutil>=2.1 in /opt/conda/lib/p
ython3.7/site-packages (from bokeh; python_version >= "3.6"->livelosspl
ot) (2.8.0)
Requirement already satisfied: pillow>=4.0 in /opt/conda/lib/python3.7/
site-packages (from bokeh; python_version >= "3.6"->livelossplot) (6.1.
Requirement already satisfied: cycler>=0.10 in /opt/conda/lib/python3.
7/site-packages (from matplotlib; python_version >= "3.6"->livelossplo
t) (0.10.0)
Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1
in /opt/conda/lib/python3.7/site-packages (from matplotlib; python vers
ion \geq "3.6"->livelossplot) (2.4.2)
Requirement already satisfied: kiwisolver>=1.0.1 in /opt/conda/lib/pyth
on3.7/site-packages (from matplotlib; python_version >= "3.6"->liveloss
plot) (1.1.0)
Requirement already satisfied: prompt-toolkit<2.1.0,>=2.0.0 in /opt/con
da/lib/python3.7/site-packages (from ipython->livelossplot) (2.0.9)
Requirement already satisfied: pickleshare in /opt/conda/lib/python3.7/
site-packages (from ipython->livelossplot) (0.7.5)
Requirement already satisfied: traitlets>=4.2 in /opt/conda/lib/python
3.7/site-packages (from ipython->livelossplot) (4.3.2)
Requirement already satisfied: backcall in /opt/conda/lib/python3.7/sit
e-packages (from ipython->livelossplot) (0.1.0)
Requirement already satisfied: decorator in /opt/conda/lib/python3.7/si
te-packages (from ipython->livelossplot) (4.4.0)
Requirement already satisfied: setuptools>=18.5 in /opt/conda/lib/pytho
n3.7/site-packages (from ipython->livelossplot) (41.0.1)
Requirement already satisfied: pexpect; sys platform != "win32" in /op
t/conda/lib/python3.7/site-packages (from ipython->livelossplot) (4.7.
0)
```

```
Requirement already satisfied: jedi>=0.10 in /opt/conda/lib/python3.7/s
ite-packages (from ipython->livelossplot) (0.14.1)
Requirement already satisfied: pygments in /opt/conda/lib/python3.7/sit
e-packages (from ipython->livelossplot) (2.4.2)
Requirement already satisfied: MarkupSafe>=0.23 in /opt/conda/lib/pytho
n3.7/site-packages (from Jinja2>=2.7->bokeh; python_version >= "3.6"->1
ivelossplot) (1.1.1)
Requirement already satisfied: wcwidth in /opt/conda/lib/python3.7/site
-packages (from prompt-toolkit<2.1.0,>=2.0.0->ipython->livelossplot)
(0.1.7)
Requirement already satisfied: ipython genutils in /opt/conda/lib/pytho
n3.7/site-packages (from traitlets>=4.2->ipython->livelossplot) (0.2.0)
Requirement already satisfied: ptyprocess>=0.5 in /opt/conda/lib/python
3.7/site-packages (from pexpect; sys platform != "win32"->ipython->live
lossplot) (0.6.0)
Requirement already satisfied: parso>=0.5.0 in /opt/conda/lib/python3.
7/site-packages (from jedi>=0.10->ipython->livelossplot) (0.5.1)
Note: you may need to restart the kernel to use updated packages.
```

```
In [3]: pip install --user opency-python
```

Requirement already satisfied: opency-python in ./.local/lib/python3.7/site-packages (4.2.0.34)
Requirement already satisfied: numpy>=1.14.5 in /opt/conda/lib/python3.
7/site-packages (from opency-python) (1.16.4)
Note: you may need to restart the kernel to use updated packages.

```
In [4]: import torch, os
    import torch.nn as nn
    import torch.optim as optim
    import torchvision.datasets as datasets
    import torchvision.transforms as transforms
    import torchvision.models as models
    from train_model import train_model
    from test_model import test_model
    %matplotlib inline
    import torchvision
    import torch.utils.model_zoo as model_zoo
    from torch.utils.data.sampler import SubsetRandomSampler
    from data_loader import get_train_valid_loader
    from data_loader import get_test_loader
```

Load data

```
In [5]: transform = transforms.Compose(
            [transforms.ToTensor(),
             transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))])
        trainloader, valloader = get_train_valid_loader('./data', batch_size=4,
        augment=0, random seed=34567,
                                                         valid_size=0.02, shuffle
        =True, num workers=2, pin memory=True)
        testset = torchvision.datasets.CIFAR10(root='./data', train=False, downl
        oad=True, transform=transform)
        testloader = torch.utils.data.DataLoader(testset, batch size=4, shuffle=
        False, num_workers=2)
        dataloaders = { 'train': trainloader, 'val': valloader, 'test': testloade
        r}
        dataset sizes = {'train': 49000, 'val': 1000, 'test': len(testset)}
        classes = ('plane', 'car', 'bird', 'cat',
                   'deer', 'dog', 'frog', 'horse', 'ship', 'truck')
        Files already downloaded and verified
        Files already downloaded and verified
        Files already downloaded and verified
In [6]: print(dataloaders)
        print(dataset_sizes)
        {'train': <torch.utils.data.dataloader.DataLoader object at 0x7fccb095a
        c50>, 'val': <torch.utils.data.dataloader.DataLoader object at 0x7fccb0
        95aba8>, 'test': <torch.utils.data.dataloader.DataLoader object at 0x7f
        ccb095ae48>}
```

Alexnet

```
In [7]: __all__ = ['AlexNet', 'alexnet']

model_urls = {
        'alexnet': 'https://download.pytorch.org/models/alexnet-owt-4df8aa7
        1.pth',
     }
```

{'train': 49000, 'val': 1000, 'test': 10000}

Alexnet-Relu

```
In [8]: class AlexNet(nn.Module):
             def __init__(self, num_classes=10):
                 super(AlexNet, self).__init__()
                 self.features = nn.Sequential(
                     nn.Conv2d(3, 64, kernel_size=3, stride=2, padding=1),
                     nn.ReLU(inplace=True),
                     nn.MaxPool2d(kernel size=2),
                     nn.Conv2d(64, 192, kernel_size=3, padding=1),
                     nn.ReLU(inplace=True),
                     nn.MaxPool2d(kernel_size=2),
                     nn.Conv2d(192, 384, kernel_size=3, padding=1),
                     nn.ReLU(inplace=True),
                     nn.Conv2d(384, 256, kernel size=3, padding=1),
                     nn.ReLU(inplace=True),
                     nn.Conv2d(256, 256, kernel_size=3, padding=1),
                     nn.ReLU(inplace=True),
                     nn.MaxPool2d(kernel_size=2),
                 self.classifier = nn.Sequential(
                     nn.Dropout(),
                     nn.Linear(256 * 2 * 2, 4096),
                     nn.ReLU(inplace=True),
                     nn.Dropout(),
                     nn.Linear(4096, 4096),
                     nn.ReLU(inplace=True),
                     nn.Linear(4096, num_classes),
                 )
             def forward(self, x):
                 x = self.features(x)
                 x = x.view(x.size(0), 256 * 2 * 2)
                 x = self.classifier(x)
                 return x
In [9]: def alexnet(pretrained=False, **kwargs):
             r"""AlexNet model architecture from the
             "One weird trick..." <a href="https://arxiv.org/abs/1404.5997">https://arxiv.org/abs/1404.5997</a> paper.
             Args:
                 pretrained (bool): If True, returns a model pre-trained on Image
        Net
```

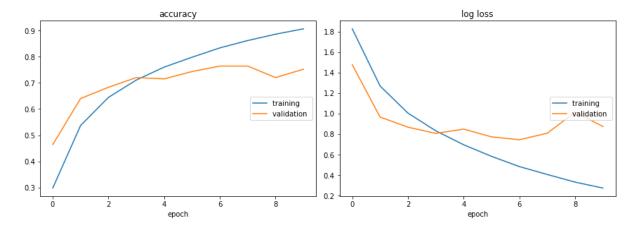
```
model = AlexNet(**kwargs)
if pretrained:
    model.load_state_dict(model_zoo.load_url(model_urls['alexnet']))
model.classifier[1] = nn.Linear(256 * 2 * 2, 4096)
model.classifier[6] = nn.Linear(4096, 10)
return model
```

optimizer: SGD

```
In [10]: #Load AlexNet
    model_ft = alexnet()

device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
```



accuracy

training (min: 0.298, max: 0.906, cur:

0.906)

validation (min: 0.464, max: 0.764, cur:

0.752)

log loss

training (min: 0.271, max: 1.826, cur:

0.271)

validation (min: 0.744, max: 1.476, cur:

0.872)

Train Loss: 0.2715 Acc: 0.9062 Val Loss: 0.8724 Acc: 0.7520

Training complete in 23m 22s

Best Validation Accuracy: 0.764, Epoch: 7

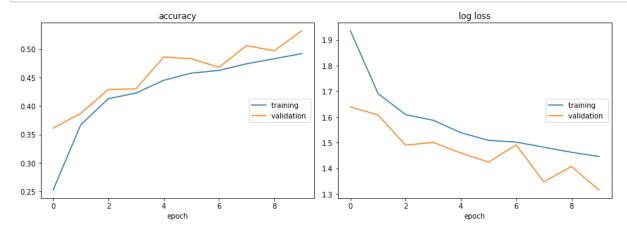
```
In [13]: #Load AlexNet
          model ft = alexnet()
          device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
          model ft.load state dict(torch.load('models/Alex_sgd/model 7 epoch.pt'))
          model ft = model ft.to(device)
          #Loss Function
          criterion = nn.CrossEntropyLoss()
          # Observe that all parameters are being optimized
          optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
          #Test
          test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft
          Iteration: 2500/2500, Loss: 17.335664749145508...
          Test Loss: 1.9669 Acc: 0.4743
          Test complete in 0m 12s
In [334]: #Load AlexNet
          model ft = alexnet()
          device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
          model ft.load state dict(torch.load('models/Alex sgd/model_11_epoch.pt'
          ))
          model_ft = model_ft.to(device)
          #Loss Function
          criterion = nn.CrossEntropyLoss()
          # Observe that all parameters are being optimized
          optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
          #Test
          test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft
          Iteration: 2500/2500, Loss: 21.763010025024414.5.
          Test Loss: 2.8026 Acc: 0.4463
          Test complete in 0m 9s
```

Alexnet-Relu - optimizer:Adam

```
In [304]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.Adam(model_ft.parameters(), lr=0.001)
```



accuracy

training (min: 0.252, max: 0.492, cur: 0.492) validation (min: 0.361, max: 0.532, cur:

0.532) log loss

training (min: 1.446, max: 1.935, cur:

1.446)

validation (min: 1.316, max: 1.639, cur:

1.316)

Train Loss: 1.4463 Acc: 0.4919 Val Loss: 1.3156 Acc: 0.5320

Training complete in 24m 48s

Best Validation Accuracy: 0.532, Epoch: 10

```
In [306]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft.load_state_dict(torch.load('models/Alex_adam/model_10_epoch.pt'
    ))
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.Adam(model_ft.parameters(), lr=0.001)

#Test
    test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft
)

Iteration: 2500/2500, Loss: 15.240313529968262.
Test Loss: 2.4420 Acc: 0.2305
```

Alexnet- Tanh

Test complete in 0m 10s

```
In [288]: class AlexNet(nn.Module):
              def __init__(self, num_classes=10):
                   super(AlexNet, self).__init__()
                   self.features = nn.Sequential(
                       nn.Conv2d(3, 64, kernel_size=3, stride=2, padding=1),
                       nn.Tanh(),
                       nn.MaxPool2d(kernel size=2),
                       nn.Conv2d(64, 192, kernel_size=3, padding=1),
                       nn.Tanh(),
                       nn.MaxPool2d(kernel_size=2),
                       nn.Conv2d(192, 384, kernel_size=3, padding=1),
                       nn.Tanh(),
                       nn.Conv2d(384, 256, kernel size=3, padding=1),
                       nn.Tanh(),
                       nn.Conv2d(256, 256, kernel_size=3, padding=1),
                       nn.Tanh(),
                       nn.MaxPool2d(kernel_size=2),
                   self.classifier = nn.Sequential(
                      nn.Dropout(),
                       nn.Linear(256 * 2 * 2, 4096),
                      nn.Tanh(),
                      nn.Dropout(),
                      nn.Linear(4096, 4096),
                      nn.Tanh(),
                      nn.Linear(4096, num_classes),
                   )
              def forward(self, x):
                  x = self.features(x)
                  x = x.view(x.size(0), 256 * 2 * 2)
                  x = self.classifier(x)
                  return x
```

```
In [273]: #Load AlexNet
    model_ft = alexnet()

device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
```

```
accuracy
                                                                                                                        log loss
                                                                                   2.0
0.65
                                                                                   1.8
0.60
0.55
                                                                                   1.6
                                                                     training
                                                                                                                                                       training
                                                                     validation
0.50
                                                                                   1.4
0.45
                                                                                   1.2
0.40
```

```
accuracy
                                             0.405, max:
                                                             0.564, cur:
        training
                                   (min:
0.557)
                                             0.562, max:
        validation
                                   (min:
                                                            0.650, cur:
0.573)
log loss
        training
                                   (min:
                                             1.310, max:
                                                            1.664, cur:
1.664)
                                             1.077, max:
        validation
                                   (min:
                                                            1.973, cur:
1.973)
Train Loss: 1.6640 Acc: 0.5574
```

Val Loss: 1.9726 Acc: 0.5730

Training complete in 15m 49s
Best Validation Accuracy: 0.65, Epoch: 5

```
In [289]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft.load_state_dict(torch.load('models/Alex_tanh/model_5_epoch.pt'
    ))
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

#Test
    test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft
    )
```

Iteration: 2500/2500, Loss: 11.6156644821167... Test Loss: 2.2321 Acc: 0.3937

Test complete in 0m 10s

Alexnet- LeakyRelu

```
In [280]: class AlexNet(nn.Module):
              def __init__(self, num_classes=10):
                  super(AlexNet, self).__init__()
                  self.features = nn.Sequential(
                       nn.Conv2d(3, 64, kernel_size=3, stride=2, padding=1),
                      nn.LeakyReLU(inplace = True),
                      nn.MaxPool2d(kernel size=2),
                       nn.Conv2d(64, 192, kernel_size=3, padding=1),
                      nn.LeakyReLU(inplace = True),
                      nn.MaxPool2d(kernel size=2),
                      nn.Conv2d(192, 384, kernel size=3, padding=1),
                      nn.LeakyReLU(inplace = True),
                      nn.Conv2d(384, 256, kernel_size=3, padding=1),
                      nn.LeakyReLU(inplace = True),
                      nn.Conv2d(256, 256, kernel_size=3, padding=1),
                      nn.LeakyReLU(inplace = True),
                      nn.MaxPool2d(kernel_size=2),
                  self.classifier = nn.Sequential(
                      nn.Dropout(),
                      nn.Linear(256 * 2 * 2, 4096),
                      nn.LeakyReLU(inplace = True),
                      nn.Dropout(),
                      nn.Linear(4096, 4096),
                      nn.LeakyReLU(inplace = True),
                      nn.Linear(4096, num_classes),
                   )
              def forward(self, x):
                  x = self.features(x)
                  x = x.view(x.size(0), 256 * 2 * 2)
                  x = self.classifier(x)
                  return x
```

```
In [281]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
```

```
0.9
                                                                                         1.8
                                                                                         1.6
0.8
                                                                                         1.4
0.7
                                                                                         1.2
                                                                         training
                                                                                                                                                                  training
0.6
                                                                                        1.0
                                                                                                                                                                  validation
                                                                                         0.8
0.5
                                                                                         0.6
0.4
                                                                                         0.4
0.3
```

```
accuracy
                                            0.302, max:
                                                            0.908, cur:
        training
                                   (min:
0.908)
                                            0.474, max:
        validation
                                   (min:
                                                            0.749, cur:
0.727)
log loss
        training
                                   (min:
                                            0.262, max:
                                                            1.821, cur:
0.262)
                                            0.759, max:
        validation
                                   (min:
                                                            1.373, cur:
1.041)
Train Loss: 0.2623 Acc: 0.9085
```

Training complete in 15m 52s
Best Validation Accuracy: 0.749, Epoch: 6

Val Loss: 1.0414 Acc: 0.7270

```
In [286]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft.load_state_dict(torch.load('models/Alex_leaky/model_6_epoch.pt'
    ))
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

#Test
    test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft
)
```

Iteration: 2500/2500, Loss: 12.675323486328125.. Test Loss: 1.7513 Acc: 0.4848

Test complete in 0m 9s

Alexnet-Sigmoid

```
In [291]: class AlexNet(nn.Module):
              def __init__(self, num_classes=10):
                   super(AlexNet, self).__init__()
                   self.features = nn.Sequential(
                       nn.Conv2d(3, 64, kernel_size=3, stride=2, padding=1),
                       nn.Sigmoid(),
                      nn.MaxPool2d(kernel size=2),
                       nn.Conv2d(64, 192, kernel_size=3, padding=1),
                       nn.Sigmoid(),
                       nn.MaxPool2d(kernel size=2),
                       nn.Conv2d(192, 384, kernel size=3, padding=1),
                       nn.Sigmoid(),
                       nn.Conv2d(384, 256, kernel size=3, padding=1),
                       nn.Sigmoid(),
                       nn.Conv2d(256, 256, kernel_size=3, padding=1),
                       nn.Sigmoid(),
                       nn.MaxPool2d(kernel size=2),
                  self.classifier = nn.Sequential(
                       nn.Dropout(),
                       nn.Linear(256 * 2 * 2, 4096),
                      nn.Sigmoid(),
                       nn.Dropout(),
                      nn.Linear(4096, 4096),
                      nn.Sigmoid(),
                      nn.Linear(4096, num_classes),
                   )
              def forward(self, x):
                  x = self.features(x)
                  x = x.view(x.size(0), 256 * 2 * 2)
                  x = self.classifier(x)
                  return x
```

```
In [292]: #Load AlexNet
    model_ft = alexnet()

device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
```

```
In [293]: train_model("Alex_sigmoid", model_ft, dataloaders, dataset_sizes, criteri
    on, optimizer_ft, num_epochs=20)
```

```
accuracy
                                                                      log loss
0.1100
                                                2.36
0.1075
                                                2.35
0.1050
                                                2.34
0.1025
                                        training
                                                                                      training
                                        validation
                                                2.33
                                                                                       validation
0.1000
                                                2.32
0.0975
                                                2.31
0.0950
0.0925
                                                2.30
                                        8
                        epoch
accuracy
                                                       0.099, max:
          training
                                           (min:
                                                                           0.101, cur:
0.099)
                                                       0.093, max:
          validation
                                           (min:
                                                                           0.110, cur:
0.094)
log loss
          training
                                           (min:
                                                       2.304, max:
                                                                           2.361, cur:
2.304)
                                                       2.303, max:
          validation
                                           (min:
                                                                           2.307, cur:
2.303)
Train Loss: 2.3036 Acc: 0.0989
Val Loss: 2.3031 Acc: 0.0940
```

Training complete in 16m 8s

Best Validation Accuracy: 0.11, Epoch: 3

```
In [294]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft.load_state_dict(torch.load('models/Alex_sigmoid/model_3_epoch.p
    t'))
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

#Test
    test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft
)
```

Iteration: 2500/2500, Loss: 9.132913589477539.
Test Loss: 2.3040 Acc: 0.1000

Test complete in 0m 10s

Alexnet- Elu

```
In [308]: class AlexNet(nn.Module):
              def __init__(self, num_classes=10):
                   super(AlexNet, self).__init__()
                   self.features = nn.Sequential(
                       nn.Conv2d(3, 64, kernel_size=3, stride=2, padding=1),
                       nn.ELU(),
                       nn.MaxPool2d(kernel size=2),
                       nn.Conv2d(64, 192, kernel_size=3, padding=1),
                       nn.ELU(),
                       nn.MaxPool2d(kernel size=2),
                       nn.Conv2d(192, 384, kernel size=3, padding=1),
                       nn.Conv2d(384, 256, kernel size=3, padding=1),
                       nn.ELU(),
                       nn.Conv2d(256, 256, kernel_size=3, padding=1),
                       nn.ELU(),
                       nn.MaxPool2d(kernel size=2),
                  self.classifier = nn.Sequential(
                       nn.Dropout(),
                       nn.Linear(256 * 2 * 2, 4096),
                      nn.ELU(),
                       nn.Dropout(),
                      nn.Linear(4096, 4096),
                      nn.ELU(),
                      nn.Linear(4096, num_classes),
                   )
              def forward(self, x):
                  x = self.features(x)
                  x = x.view(x.size(0), 256 * 2 * 2)
                  x = self.classifier(x)
                  return x
```

```
In [309]: #Load AlexNet
    model_ft = alexnet()

device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
# Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
```

```
log loss
                                      accuracy
0.675
0.650
0.625
0.600
0.575
                                                                   training
                                                                                                                                                training
                                                                   validation
                                                                                                                                                validation
0.550
                                                                                 3 -
0.525
                                                                                 2
0.500
0.475
                                        epoch
```

```
accuracy
                                                            0.652, cur:
                                            0.462, max:
        training
                                   (min:
0.582)
                                            0.510, max:
        validation
                                   (min:
                                                            0.666, cur:
0.510)
log loss
        training
                                   (min:
                                            1.075, max:
                                                            2.823, cur:
2.823)
                                            0.994, max:
        validation
                                   (min:
                                                            5.730, cur:
5.730)
Train Loss: 2.8231 Acc: 0.5818
Val Loss: 5.7301 Acc: 0.5100
```

Training complete in 15m 59s
Best Validation Accuracy: 0.666, Epoch: 5

```
In [311]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft.load_state_dict(torch.load('models/Alex_elu/model_5_epoch.pt'))
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

#Test
    test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft
)
```

Iteration: 2500/2500, Loss: 9.334567070007324..
Test Loss: 2.0060 Acc: 0.3979

Test complete in 0m 10s

Alexnet- average pooling

```
In [295]: class AlexNet(nn.Module):
              def __init__(self, num_classes=10):
                  super(AlexNet, self). init ()
                  self.features = nn.Sequential(
                       nn.Conv2d(3, 64, kernel_size=3, stride=2, padding=1),
                      nn.ReLU(inplace=True),
                       nn.AvgPool2d(kernel_size=2),
                      nn.Conv2d(64, 192, kernel_size=3, padding=1),
                      nn.ReLU(inplace=True),
                       nn.AvgPool2d(kernel size=2),
                      nn.Conv2d(192, 384, kernel_size=3, padding=1),
                      nn.ReLU(inplace=True),
                      nn.Conv2d(384, 256, kernel_size=3, padding=1),
                      nn.ReLU(inplace=True),
                      nn.Conv2d(256, 256, kernel size=3, padding=1),
                      nn.ReLU(inplace=True),
                      nn.AvgPool2d(kernel_size=2),
                  self.classifier = nn.Sequential(
                      nn.Dropout(),
                      nn.Linear(256 * 2 * 2, 4096),
                      nn.ReLU(inplace=True),
                      nn.Dropout(),
                      nn.Linear(4096, 4096),
                      nn.ReLU(inplace=True),
                      nn.Linear(4096, num_classes),
                   )
              def forward(self, x):
                  x = self.features(x)
                  x = x.view(x.size(0), 256 * 2 * 2)
                  x = self.classifier(x)
                  return x
```

```
In [296]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
```

```
In [297]: train_model("Alexnet_avgpool", model_ft, dataloaders, dataset_sizes, crit
    erion, optimizer_ft, num_epochs=10)
```

```
accuracy
                                                                                                                             log loss
                                                                                      2.0
0.8
                                                                                      1.8
0.7
                                                                                      1.6
0.6
                                                                                      1.4
                                                                       training
                                                                                                                                                             training
0.5
                                                                                                                                                             validation
                                                                                      1.2
                                                                                      1.0
0.4
                                                                                      0.8
0.3
                                                                                      0.6
0.2
                                                                                      0.4
                                                                                                                               epoch
```

```
accuracy
                                             0.200, max:
                                                             0.834, cur:
        training
                                   (min:
0.834)
                                             0.358, max:
        validation
                                   (min:
                                                             0.767, cur:
0.767)
log loss
                                             0.478, max:
        training
                                   (min:
                                                             2.061, cur:
0.478)
                                             0.777, max:
        validation
                                   (min:
                                                             1.678, cur:
0.777)
```

Train Loss: 0.4777 Acc: 0.8338 Val Loss: 0.7768 Acc: 0.7670

Training complete in 15m 55s
Best Validation Accuracy: 0.767, Epoch: 10

Iteration: 2500/2500, Loss: 15.447891235351562..
Test Loss: 1.9702 Acc: 0.4470

Test complete in 0m 10s

Fine tune- RandomHorizontalFlip

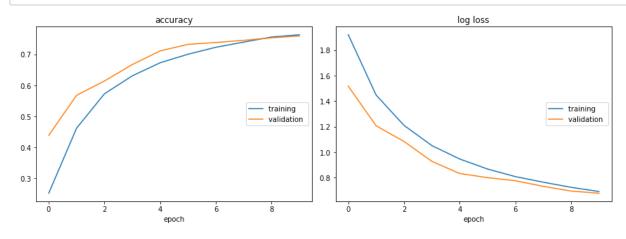
```
In [221]: class AlexNet(nn.Module):
              def __init__(self, num_classes=10):
                  super(AlexNet, self).__init__()
                  self.features = nn.Sequential(
                       nn.Conv2d(3, 64, kernel_size=3, stride=2, padding=1),
                       nn.ReLU(inplace=True),
                      nn.MaxPool2d(kernel size=2),
                       nn.Conv2d(64, 192, kernel_size=3, padding=1),
                       nn.ReLU(inplace=True),
                       nn.MaxPool2d(kernel size=2),
                       nn.Conv2d(192, 384, kernel size=3, padding=1),
                       nn.ReLU(inplace=True),
                       nn.Conv2d(384, 256, kernel size=3, padding=1),
                       nn.ReLU(inplace=True),
                       nn.Conv2d(256, 256, kernel_size=3, padding=1),
                       nn.ReLU(inplace=True),
                       nn.MaxPool2d(kernel size=2),
                  self.classifier = nn.Sequential(
                       nn.Dropout(),
                       nn.Linear(256 * 2 * 2, 4096),
                      nn.ReLU(inplace=True),
                       nn.Dropout(),
                      nn.Linear(4096, 4096),
                      nn.ReLU(inplace=True),
                      nn.Linear(4096, num_classes),
                   )
              def forward(self, x):
                  x = self.features(x)
                  x = x.view(x.size(0), 256 * 2 * 2)
                  x = self.classifier(x)
                  return x
```

Files already downloaded and verified Files already downloaded and verified

In [223]: #Load AlexNet model_ft = alexnet() device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu") model_ft = model_ft.to(device) #Loss Function criterion = nn.CrossEntropyLoss() # Observe that all parameters are being optimized optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

In [224]: #Train

train_model("aug1",model_ft, dataloaders, dataset_sizes, criterion, opti
mizer_ft, num_epochs=10)



accuracy

training (min: 0.252, max: 0.763, cur:

0.763)

validation (min: 0.439, max: 0.759, cur:

0.759)

log loss

training (min: 0.690, max: 1.920, cur:

0.690)

validation (min: 0.677, max: 1.517, cur:

0.677)

Train Loss: 0.6902 Acc: 0.7630 Val Loss: 0.6765 Acc: 0.7590

Training complete in 15m 44s

Best Validation Accuracy: 0.759, Epoch: 10

```
In [262]: #Load AlexNet
          model ft = alexnet()
          device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
          model ft.load state dict(torch.load('models/aug1/model_10 epoch.pt'))
          model ft = model ft.to(device)
          #Loss Function
          criterion = nn.CrossEntropyLoss()
          # Observe that all parameters are being optimized
          optimizer ft = optim.SGD(model ft.parameters(), lr=0.001, momentum=0.9)
          #Test
          test model (model ft, dataloaders, dataset sizes, criterion, optimizer ft
          Iteration: 2500/2500, Loss: 13.029035568237305...
          Test Loss: 1.9625 Acc: 0.4534
          Test complete in 0m 9s
```

Finetune- Random Vertical Flip

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```
In [263]:
          trainloader, valloader = get_train_valid_loader('./data', batch_size=4,
          augment=2, random seed=34567,
                                                           valid_size=0.02, shuffle
          =True, num_workers=2, pin_memory=True)
          dataloaders = { 'train': trainloader, 'val': valloader, 'test': testloade
          r}
          Files already downloaded and verified
```

```
In [226]: #Load AlexNet
          model ft = alexnet()
          device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
          model_ft = model_ft.to(device)
          #Loss Function
          criterion = nn.CrossEntropyLoss()
          # Observe that all parameters are being optimized
          optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)
```

```
In [227]: #Train
    train_model("aug2",model_ft, dataloaders, dataset_sizes, criterion, opti
    mizer_ft, num_epochs=10)
```

```
log loss
                                      accuracy
0.8
                                                                                    1.8
0.7
                                                                                    1.6
0.6
                                                                                    1.4
                                                                     training
                                                                                                                                                         training
                                                                     validation
                                                                                                                                                         validation
0.5
                                                                                   1.2
                                                                                    1.0
0.4
                                                                                    0.8
0.3
                                        epoch
                                                                                                                            epoch
```

```
accuracy
                                             0.270, max:
        training
                                   (min:
                                                             0.764, cur:
0.764)
        validation
                                   (min:
                                             0.423, max:
                                                             0.777, cur:
0.777)
log loss
                                             0.685, max:
        training
                                   (min:
                                                             1.902, cur:
0.685)
        validation
                                             0.629, max:
                                                             1.498, cur:
                                   (min:
0.629)
Train Loss: 0.6850 Acc: 0.7644
```

Training complete in 15m 49s
Best Validation Accuracy: 0.777, Epoch: 10

Val Loss: 0.6290 Acc: 0.7770

```
In [264]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft.load_state_dict(torch.load('models/aug2/model_10_epoch.pt'))
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

#Test
    test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft
)
```

Iteration: 2500/2500, Loss: 14.188934326171875.
Test Loss: 1.8152 Acc: 0.4939

Test complete in 0m 9s

Fine tune- RandomRotation

optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

Observe that all parameters are being optimized

```
log loss
                                      accuracy
0.8
                                                                                     1.8
0.7
                                                                                     1.6
0.6
                                                                                     1.4
                                                                      training
                                                                                                                                                           training
                                                                      validation
                                                                                                                                                           validation
0.5
0.4
                                                                                     1.0
                                                                                     0.8
0.3
                                                                                     0.6
                                        epoch
                                                                                                                             epoch
```

```
accuracy
                                             0.257, max:
        training
                                   (min:
                                                             0.769, cur:
0.769)
        validation
                                   (min:
                                             0.457, max:
                                                             0.778, cur:
0.778)
log loss
                                             0.681, max:
        training
                                   (min:
                                                             1.922, cur:
0.681)
        validation
                                             0.659, max:
                                   (min:
                                                             1.446, cur:
0.672)
Train Loss: 0.6812 Acc: 0.7686
```

Val Loss: 0.6720 Acc: 0.7780

Training complete in 15m 58s
Best Validation Accuracy: 0.778, Epoch: 10

```
In [266]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft.load_state_dict(torch.load('models/aug3/model_10_epoch.pt'))
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

#Test
    test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft)
```

Iteration: 2500/2500, Loss: 13.035179138183594..
Test Loss: 2.1031 Acc: 0.4322

Test complete in 0m 9s

Fine tune - Combination

criterion = nn.CrossEntropyLoss()

Observe that all parameters are being optimized

optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

```
In [233]: #Train
    train_model("aug4",model_ft, dataloaders, dataset_sizes, criterion, opti
    mizer_ft, num_epochs=10)
```

```
log loss
                                     accuracy
0.8
                                                                                   1.8
0.7
                                                                                   1.6
0.6
                                                                                   1.4
                                                                    training
                                                                                                                                                        training
0.5
                                                                    validation
                                                                                                                                                        validation
0.4
                                                                                   1.0
0.3
                                        epoch
                                                                                                                           epoch
```

```
accuracy
                                             0.243, max:
        training
                                   (min:
                                                            0.760, cur:
0.760)
        validation
                                   (min:
                                             0.347, max:
                                                            0.775, cur:
0.775)
log loss
                                             0.703, max:
        training
                                   (min:
                                                            1.941, cur:
0.703)
        validation
                                             0.656, max:
                                                            1.604, cur:
                                   (min:
0.656)
Train Loss: 0.7026 Acc: 0.7599
```

Train Loss: 0.7026 Acc: 0.7599 Val Loss: 0.6564 Acc: 0.7750

Training complete in 16m 2s
Best Validation Accuracy: 0.775, Epoch: 10

```
In [268]: #Load AlexNet
    model_ft = alexnet()

    device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
    model_ft.load_state_dict(torch.load('models/aug4/model_10_epoch.pt'))
    model_ft = model_ft.to(device)

#Loss Function
    criterion = nn.CrossEntropyLoss()
    # Observe that all parameters are being optimized
    optimizer_ft = optim.SGD(model_ft.parameters(), lr=0.001, momentum=0.9)

#Test
    test_model(model_ft, dataloaders, dataset_sizes, criterion, optimizer_ft
)
```

Iteration: 2500/2500, Loss: 15.442499160766602..
Test Loss: 1.6911 Acc: 0.4895

Test complete in 0m 9s