```
In [1]: N
    import torch
    from torch import nn
    import torch.optim as optim
    from torch.utils.data import Dataset, DataLoader
    from torchvision import datasets
    from torchvision.transforms import ToTensor
    import time
    from sklearn.model_selection import KFold
    import matplotlib.pyplot as plt
    import numpy as np
    import pandas as pd
    import torch.nn.functional as F
    import gc
    from ptflops import get_model_complexity_info
```

```
In [2]:

    | train_path = r"C:\Users\ccm51\OneDrive\Desktop\ECGR 4106\fashion-mnist_trai

              test path = r"C:\Users\ccm51\OneDrive\Desktop\ECGR 4106\fashion-mnist test.
              train_numpy = np.loadtxt(train_path, dtype = np.float32, delimiter = ",", 
              test numpy = np.loadtxt(test path, dtype = np.float32, delimiter = ",", ski
              print(train numpy.shape)
              (60000, 785)
           data_path = '../data-unversioned/ecgr4106/'
 In [3]:
              fashion MNIST = datasets.FashionMNIST(data path, train=True, download=True,
              fashion MNIST val = datasets.FashionMNIST(data path, train=False, download=
In [123]:
           ▶ len(fashion MNIST val)
   Out[123]: 10000
 In [4]:
           ▶ def try_gpu(i=0):
                  if torch.cuda.device_count() >= i+1:
                      return torch.device(f'cuda:{i}')
                  return torch.device('cpu')
```

```
In [5]:
         def training loop(n epochs, optimizer, model, loss fn, train loader, val lo
                train loss hist = []
                train acc hist = []
                val acc hist = []
                main_tic = time.perf_counter()
                for epoch in range(1, n epochs + 1):
                    tic = time.perf counter()
                    loss train = 0.0
                    correct_train = 0
                    correct val = 0
                    for imgs, labels in train_loader:
                        # if torch.cuda.is available():
                        imgs = imgs.to(device=try gpu())
                        labels = labels.to(device=try_gpu())
                        outputs = model(imgs)
                        loss = loss_fn(outputs, labels)
                        optimizer.zero_grad()
                        loss.backward()
                        optimizer.step()
                        loss_train += loss.item()
                    toc = time.perf counter()
                    with torch.no grad():
                        total = 0
                        for imgs, labels in train loader:
                            outputs = model(imgs.to(device=try gpu()))
                            _, predicted = torch.max(outputs, dim=1)
                            total += labels.shape[0]
                            correct_train += int((predicted == labels.to(device=try_gpu
                        train acc = round(correct train/total, 3)
                        total = 0
                        for imgs, labels in val_loader:
                            outputs = model(imgs.to(device=try_gpu()))
                            _, predicted = torch.max(outputs, dim=1)
                            total += labels.shape[0]
                            correct val += int((predicted == labels.to(device=try gpu())
                        val_acc = round(correct_val/total, 3)
                    train loss hist.append(round(loss train / len(train loader), 5))
                    train acc hist.append(train acc)
                    val acc hist.append(val acc)
                    if epoch == 1 or epoch == n epochs or epoch % update freq == 0:
                        print(f"Epoch {epoch}:\n\tDuration = {round(toc - tic, 3)} sec
                main toc = time.perf counter()
                print(f"\nTotal Training Time = {round(main toc - main tic, 3)} second
                return train loss hist, train acc hist, val acc hist
```

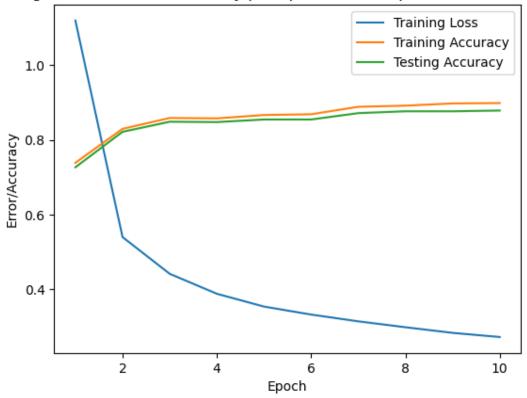
```
In [6]:
          | def plot model(title, fig num, loss hist, train hist, test hist, leg loc):
                 plot = plt.figure(fig num)
                 x = range(1, len(loss hist)+1)
                 plt.plot(x, loss hist)
                 plt.plot(x, train hist)
                 plt.plot(x, test_hist)
                 plt.legend(["Training Loss", "Training Accuracy", "Testing Accuracy"],
                 plt.xlabel('Epoch')
                 plt.ylabel('Error/Accuracy')
                 plt.title(title)
In [14]:
          def init (self):
                     super().__init__()
                     self.conv1 = nn.LazyConv2d(6, kernel_size=5, padding=2)
                     self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                     self.conv2 = nn.LazyConv2d(16, kernel size=5)
                     self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                     self.fc1 = nn.LazyLinear(120)
                     self.fc2 = nn.LazyLinear(84)
                     self.fc3 = nn.LazyLinear(10)
                     self.relu = nn.ReLU()
                     self.flat = nn.Flatten()
                 def forward(self, x):
                     out = self.maxp1(self.relu(self.conv1(x)))
                     out = self.maxp2(self.relu(self.conv2(out)))
                     out = self.flat(out)
                     out = self.relu(self.fc1(out))
                     out = self.relu(self.fc2(out))
                     out = self.fc3(out)
                     return out
             model 1 = alt1 LeNet().to(device=try gpu())
             optimizer 1 = optim.SGD(model 1.parameters(), lr=1e-1)
             model 1.eval()
   Out[14]: alt1 LeNet(
               (conv1): LazyConv2d(0, 6, kernel size=(5, 5), stride=(1, 1), padding=
             (2, 2))
               (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
             mode=False)
               (conv2): LazyConv2d(0, 16, kernel size=(5, 5), stride=(1, 1))
               (maxp2): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
             mode=False)
               (fc1): LazyLinear(in_features=0, out_features=120, bias=True)
               (fc2): LazyLinear(in_features=0, out_features=84, bias=True)
               (fc3): LazyLinear(in features=0, out features=10, bias=True)
               (relu): ReLU()
               (flat): Flatten(start dim=1, end dim=-1)
```

```
Epoch 1:
        Duration = 19.818 seconds
        Training Loss: 1.16621
        Training Accuracy: 0.754
        Validation Accuracy: 0.746
Epoch 2:
        Duration = 20.963 seconds
        Training Loss: 0.52571
        Training Accuracy: 0.819
        Validation Accuracy: 0.809
Epoch 3:
        Duration = 19.146 seconds
        Training Loss: 0.42693
        Training Accuracy: 0.86
        Validation Accuracy: 0.851
Epoch 4:
        Duration = 18.765 seconds
        Training Loss: 0.38071
        Training Accuracy: 0.873
        Validation Accuracy: 0.859
Epoch 5:
        Duration = 18.794 seconds
        Training Loss: 0.35066
        Training Accuracy: 0.858
        Validation Accuracy: 0.845
Epoch 6:
        Duration = 18.724 seconds
        Training Loss: 0.32908
        Training Accuracy: 0.89
        Validation Accuracy: 0.875
Epoch 7:
        Duration = 18.646 seconds
        Training Loss: 0.31203
        Training Accuracy: 0.887
        Validation Accuracy: 0.87
Epoch 8:
        Duration = 18.85 seconds
        Training Loss: 0.29782
        Training Accuracy: 0.888
        Validation Accuracy: 0.875
Epoch 9:
        Duration = 18.87 seconds
        Training Loss: 0.28718
        Training Accuracy: 0.9
        Validation Accuracy: 0.884
Epoch 10:
        Duration = 18.473 seconds
        Training Loss: 0.27909
        Training Accuracy: 0.874
        Validation Accuracy: 0.858
```

Total Training Time = 381.495 seconds

Average Training Time per Epoch = 38.149 seconds

Figure 1 - Loss and Accuracy per Epoch for the updated LeNet Model



```
In [164]:
           lass alt2a LeNet(nn.Module):
                  def __init__(self):
                      super(). init ()
                      self.conv1 = nn.LazyConv2d(6, kernel size=3, padding=1)
                      self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv2 = nn.LazyConv2d(16, kernel_size=3)
                      self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.fc1 = nn.LazyLinear(120)
                      self.fc2 = nn.LazyLinear(84)
                      self.fc3 = nn.LazyLinear(10)
                      self.relu = nn.ReLU()
                      self.flat = nn.Flatten()
                  def forward(self, x):
                      out = self.maxp1(self.relu(self.conv1(x)))
                      out = self.maxp2(self.relu(self.conv2(out)))
                      out = self.flat(out)
                      out = self.relu(self.fc1(out))
                      out = self.relu(self.fc2(out))
                      out = self.fc3(out)
                      return out
              model 2a = alt2a LeNet().to(device=try gpu())
              optimizer 2a = optim.SGD(model 2a.parameters(), lr=1e-1)
              model 2a.eval()
   Out[164]: alt2a LeNet(
                (conv1): LazyConv2d(0, 6, kernel size=(3, 3), stride=(1, 1), padding=
              (1, 1))
                (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv2): LazyConv2d(0, 16, kernel size=(3, 3), stride=(1, 1))
                (maxp2): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (fc1): LazyLinear(in_features=0, out_features=120, bias=True)
                (fc2): LazyLinear(in features=0, out features=84, bias=True)
                (fc3): LazyLinear(in features=0, out features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start_dim=1, end_dim=-1)
```

Duration = 8.57 seconds
Training Loss: 1.0938
Training Accuracy: 0.718
Validation Accuracy: 0.708

#### Epoch 5:

Duration = 9.154 seconds Training Loss: 0.35228 Training Accuracy: 0.878 Validation Accuracy: 0.868

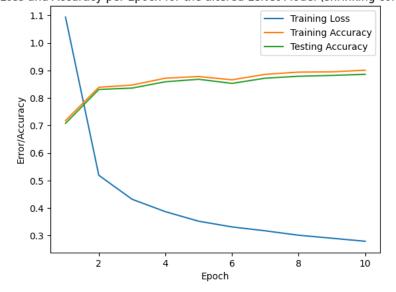
#### Epoch 10:

Duration = 8.199 seconds Training Loss: 0.27944 Training Accuracy: 0.901 Validation Accuracy: 0.886

Total Training Time = 157.307 seconds
Average Training Time per Epoch = 15.731 seconds

In [167]: ► title\_2a = "Figure 2a - Loss and Accuracy per Epoch for the altered LeNet N plot\_model(title\_2a, 2, t\_loss\_hist\_2a, t\_acc\_hist\_2a, v\_acc\_hist\_2a, 'upper loss\_hist\_2a, t\_acc\_hist\_2a, v\_acc\_hist\_2a, 'upper loss\_hist\_2a, 'up

Figure 2a - Loss and Accuracy per Epoch for the altered LeNet Model (shrinking convolution window)



```
In [168]:
           class alt2b LeNet(nn.Module):
                  def __init__(self):
                      super(). init ()
                      self.conv1 = nn.LazyConv2d(6, kernel size=7, padding=3)
                      self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv2 = nn.LazyConv2d(16, kernel_size=7)
                      self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.fc1 = nn.LazyLinear(120)
                      self.fc2 = nn.LazyLinear(84)
                      self.fc3 = nn.LazyLinear(10)
                      self.relu = nn.ReLU()
                      self.flat = nn.Flatten()
                  def forward(self, x):
                      out = self.maxp1(self.relu(self.conv1(x)))
                      out = self.maxp2(self.relu(self.conv2(out)))
                      out = self.flat(out)
                      out = self.relu(self.fc1(out))
                      out = self.relu(self.fc2(out))
                      out = self.fc3(out)
                      return out
              model 2b = alt2b LeNet().to(device=try gpu())
              optimizer 2b = optim.SGD(model 2b.parameters(), lr=1e-1)
              model 2b.eval()
   Out[168]: alt2b LeNet(
                (conv1): LazyConv2d(0, 6, kernel size=(7, 7), stride=(1, 1), padding=
              (3, 3))
                (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv2): LazyConv2d(0, 16, kernel size=(7, 7), stride=(1, 1))
                (maxp2): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (fc1): LazyLinear(in_features=0, out_features=120, bias=True)
                (fc2): LazyLinear(in features=0, out features=84, bias=True)
                (fc3): LazyLinear(in features=0, out features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start_dim=1, end_dim=-1)
```

Duration = 9.395 seconds Training Loss: 1.08711 Training Accuracy: 0.739 Validation Accuracy: 0.73

#### Epoch 5:

Duration = 9.349 seconds Training Loss: 0.34582 Training Accuracy: 0.875 Validation Accuracy: 0.86

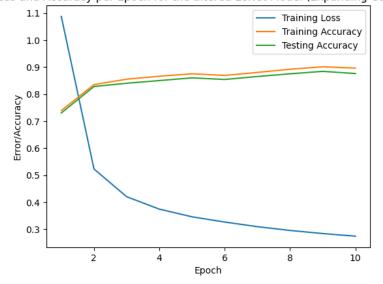
#### Epoch 10:

Duration = 10.162 seconds Training Loss: 0.27412 Training Accuracy: 0.896 Validation Accuracy: 0.876

Total Training Time = 186.064 seconds Average Training Time per Epoch = 18.606 seconds

In [170]: ► title\_2b = "Figure 2b - Loss and Accuracy per Epoch for the altered LeNet N plot\_model(title\_2b, 3, t\_loss\_hist\_2b, t\_acc\_hist\_2b, v\_acc\_hist\_2b, 'upper loss and Accuracy per Epoch for the altered LeNet N plot\_model(title\_2b, 3, t\_loss\_hist\_2b, t\_acc\_hist\_2b, v\_acc\_hist\_2b, 'upper loss and Accuracy per Epoch for the altered LeNet N plot\_model(title\_2b, 3, t\_loss\_hist\_2b, t\_acc\_hist\_2b, v\_acc\_hist\_2b, 'upper loss and Accuracy per Epoch for the altered LeNet N plot\_model(title\_2b, 3, t\_loss\_hist\_2b, t\_acc\_hist\_2b, v\_acc\_hist\_2b, 'upper loss and Accuracy per Epoch for the altered LeNet N plot\_model(title\_2b, 3, t\_loss\_hist\_2b, t\_acc\_hist\_2b, v\_acc\_hist\_2b, v\_acc\_hist\_2b, t\_acc\_hist\_2b, t\_acc\_

Figure 2b - Loss and Accuracy per Epoch for the altered LeNet Model (Expanding Convolution Window)



```
In [181]:
           class alt3 LeNet(nn.Module):
                  def __init__(self):
                      super(). init ()
                      self.conv1 = nn.LazyConv2d(12, kernel size=5, padding=2)
                      self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv2 = nn.LazyConv2d(32, kernel size=5)
                      self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.fc1 = nn.LazyLinear(120)
                      self.fc2 = nn.LazyLinear(84)
                      self.fc3 = nn.LazyLinear(10)
                      self.relu = nn.ReLU()
                      self.flat = nn.Flatten()
                  def forward(self, x):
                      out = self.maxp1(self.relu(self.conv1(x)))
                      out = self.maxp2(self.relu(self.conv2(out)))
                      out = self.flat(out)
                      out = self.relu(self.fc1(out))
                      out = self.relu(self.fc2(out))
                      out = self.fc3(out)
                      return out
              model 3 = alt3 LeNet().to(device=try gpu())
              optimizer 3 = optim.SGD(model 3.parameters(), lr=1e-1)
              model 3.eval()
   Out[181]: alt3 LeNet(
                (conv1): LazyConv2d(0, 12, kernel size=(5, 5), stride=(1, 1), padding=
              (2, 2))
                (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv2): LazyConv2d(0, 32, kernel size=(5, 5), stride=(1, 1))
                (maxp2): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (fc1): LazyLinear(in_features=0, out_features=120, bias=True)
                (fc2): LazyLinear(in features=0, out features=84, bias=True)
                (fc3): LazyLinear(in features=0, out features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start_dim=1, end_dim=-1)
```

Duration = 11.009 seconds Training Loss: 1.02504 Training Accuracy: 0.789 Validation Accuracy: 0.78

#### Epoch 5:

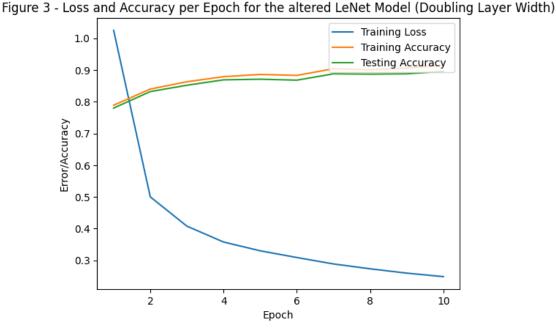
Duration = 10.568 seconds Training Loss: 0.32999 Training Accuracy: 0.886 Validation Accuracy: 0.871

#### Epoch 10:

Duration = 11.317 seconds Training Loss: 0.24853 Training Accuracy: 0.915 Validation Accuracy: 0.896

Total Training Time = 190.937 seconds Average Training Time per Epoch = 19.094 seconds

In [183]: ► title\_3 = "Figure 3 - Loss and Accuracy per Epoch for the altered LeNet Mod
plot\_model(title\_3, 4, t\_loss\_hist\_3, t\_acc\_hist\_3, v\_acc\_hist\_3, 'upper ri



```
In [196]:
           class alt4 LeNet(nn.Module):
                  def __init__(self):
                      super(). init ()
                      self.conv1 = nn.LazyConv2d(6, kernel size=5, padding=2)
                      self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv2 = nn.LazyConv2d(16, kernel size=5)
                      self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv3 = nn.LazyConv2d(8, kernel size=5, padding=2)
                      self.maxp3 = nn.MaxPool2d(kernel size=2, stride=1)
                      self.fc1 = nn.LazyLinear(120)
                      self.fc2 = nn.LazyLinear(84)
                      self.fc3 = nn.LazyLinear(10)
                      self.relu = nn.ReLU()
                      self.flat = nn.Flatten()
                  def forward(self, x):
                      out = self.maxp1(self.relu(self.conv1(x)))
                      out = self.maxp2(self.relu(self.conv2(out)))
                      out = self.maxp3(self.relu(self.conv3(out)))
                      out = self.flat(out)
                      out = self.relu(self.fc1(out))
                      out = self.relu(self.fc2(out))
                      out = self.fc3(out)
                      return out
              model 4 = alt4 LeNet().to(device=try_gpu())
              optimizer 4 = optim.SGD(model 4.parameters(), lr=1e-1)
              model 4.eval()
   Out[196]: alt4 LeNet(
                (conv1): LazyConv2d(0, 6, kernel_size=(5, 5), stride=(1, 1), padding=
                (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv2): LazyConv2d(0, 16, kernel size=(5, 5), stride=(1, 1))
                (maxp2): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              _mode=False)
                (conv3): LazyConv2d(0, 8, kernel size=(5, 5), stride=(1, 1), padding=
              (2, 2))
                (maxp3): MaxPool2d(kernel_size=2, stride=1, padding=0, dilation=1, ceil
              mode=False)
                (fc1): LazyLinear(in_features=0, out_features=120, bias=True)
                (fc2): LazyLinear(in features=0, out features=84, bias=True)
                (fc3): LazyLinear(in features=0, out features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start_dim=1, end_dim=-1)
```

Duration = 9.411 seconds Training Loss: 1.42292 Training Accuracy: 0.636 Validation Accuracy: 0.627

#### Epoch 5:

Duration = 9.522 seconds Training Loss: 0.37345 Training Accuracy: 0.869 Validation Accuracy: 0.856

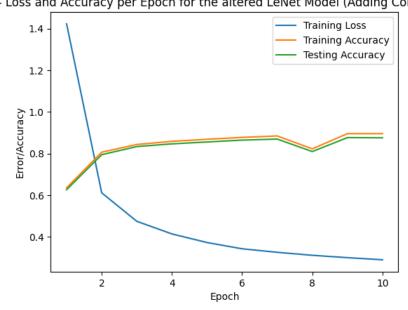
#### Epoch 10:

Duration = 9.648 seconds Training Loss: 0.29061 Training Accuracy: 0.896 Validation Accuracy: 0.876

Total Training Time = 172.741 seconds
Average Training Time per Epoch = 17.274 seconds

In [198]: ► title\_4 = "Figure 4 - Loss and Accuracy per Epoch for the altered LeNet Mod
plot\_model(title\_4, 5, t\_loss\_hist\_4, t\_acc\_hist\_4, v\_acc\_hist\_4, 'upper ri

Figure 4 - Loss and Accuracy per Epoch for the altered LeNet Model (Adding Convolution Layer)



```
In [199]:
           class alt5 LeNet(nn.Module):
                  def __init__(self):
                      super(). init ()
                      self.conv1 = nn.LazyConv2d(6, kernel size=5, padding=2)
                      self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv2 = nn.LazyConv2d(16, kernel_size=5)
                      self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.fc1 = nn.LazyLinear(256)
                      self.fc2 = nn.LazyLinear(120)
                      self.fc3 = nn.LazyLinear(84)
                      self.fc4 = nn.LazyLinear(10)
                      self.relu = nn.ReLU()
                      self.flat = nn.Flatten()
                  def forward(self, x):
                      out = self.maxp1(self.relu(self.conv1(x)))
                      out = self.maxp2(self.relu(self.conv2(out)))
                      out = self.flat(out)
                      out = self.relu(self.fc1(out))
                      out = self.relu(self.fc2(out))
                      out = self.relu(self.fc3(out))
                      out = self.fc4(out)
                      return out
              model 5 = alt5 LeNet().to(device=try gpu())
              optimizer 5 = optim.SGD(model 5.parameters(), lr=1e-1)
              model 5.eval()
   Out[199]: alt5 LeNet(
                (conv1): LazyConv2d(0, 6, kernel_size=(5, 5), stride=(1, 1), padding=
                (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv2): LazyConv2d(0, 16, kernel size=(5, 5), stride=(1, 1))
                (maxp2): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (fc1): LazyLinear(in features=0, out features=256, bias=True)
                (fc2): LazyLinear(in features=0, out features=120, bias=True)
                (fc3): LazyLinear(in_features=0, out_features=84, bias=True)
                (fc4): LazyLinear(in_features=0, out_features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start dim=1, end dim=-1)
```

Duration = 9.023 seconds Training Loss: 1.3352 Training Accuracy: 0.717 Validation Accuracy: 0.709

#### Epoch 5:

Duration = 9.593 seconds Training Loss: 0.36617 Training Accuracy: 0.875 Validation Accuracy: 0.863

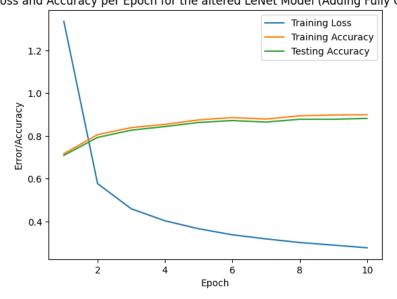
#### Epoch 10:

Duration = 11.919 seconds Training Loss: 0.27715 Training Accuracy: 0.899 Validation Accuracy: 0.882

Total Training Time = 195.642 seconds Average Training Time per Epoch = 19.564 seconds

In [201]: ► title\_5 = "Figure 5 - Loss and Accuracy per Epoch for the altered LeNet Mod
plot\_model(title\_5, 6, t\_loss\_hist\_5, t\_acc\_hist\_5, v\_acc\_hist\_5, 'upper ri

Figure 5 - Loss and Accuracy per Epoch for the altered LeNet Model (Adding Fully Connected Layer)



```
In [204]:
           def __init__(self):
                      super(). init ()
                      self.conv1 = nn.LazyConv2d(6, kernel size=5, padding=2)
                      self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv2 = nn.LazyConv2d(16, kernel_size=5)
                      self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.fc1 = nn.LazyLinear(120)
                      self.fc2 = nn.LazyLinear(84)
                      self.fc3 = nn.LazyLinear(10)
                      self.relu = nn.ReLU()
                      self.flat = nn.Flatten()
                  def forward(self, x):
                      out = self.maxp1(self.relu(self.conv1(x)))
                      out = self.maxp2(self.relu(self.conv2(out)))
                      out = self.flat(out)
                      out = self.relu(self.fc1(out))
                      out = self.relu(self.fc2(out))
                      out = self.fc3(out)
                      return out
              model 6a = alt6a LeNet().to(device=try gpu())
              optimizer 6a = optim.SGD(model 6a.parameters(), lr=2e-1)
              model 6a.eval()
   Out[204]: alt6a LeNet(
                (conv1): LazyConv2d(0, 6, kernel size=(5, 5), stride=(1, 1), padding=
              (2, 2))
                (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv2): LazyConv2d(0, 16, kernel size=(5, 5), stride=(1, 1))
                (maxp2): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (fc1): LazyLinear(in_features=0, out_features=120, bias=True)
                (fc2): LazyLinear(in features=0, out features=84, bias=True)
                (fc3): LazyLinear(in features=0, out features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start_dim=1, end_dim=-1)
```

Duration = 8.659 seconds Training Loss: 0.93004 Training Accuracy: 0.826 Validation Accuracy: 0.813

#### Epoch 5:

Duration = 11.621 seconds Training Loss: 0.30714 Training Accuracy: 0.893 Validation Accuracy: 0.878

#### Epoch 10:

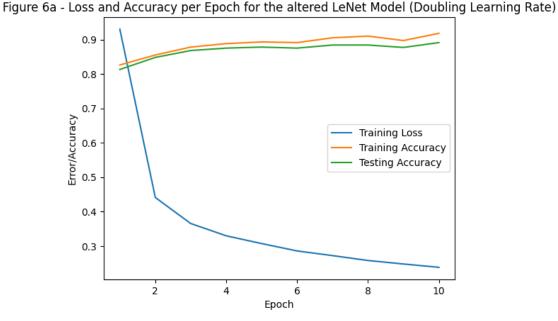
Duration = 10.245 seconds Training Loss: 0.23812 Training Accuracy: 0.918 Validation Accuracy: 0.891

Total Training Time = 190.071 seconds

Average Training Time per Epoch = 19.007 seconds

In [289]: ► title\_6a = "Figure 6a - Loss and Accuracy per Epoch for the altered LeNet N plot\_model(title\_6a, 7, t\_loss\_hist\_6a, t\_acc\_hist\_6a, v\_acc\_hist\_6a, 'right')

office moder(title\_oa, /, t\_loss\_mist\_oa, t\_act\_mist\_oa, v\_act\_mist\_oa, rig



```
In [207]:
           def __init__(self):
                      super(). init ()
                      self.conv1 = nn.LazyConv2d(6, kernel size=5, padding=2)
                      self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv2 = nn.LazyConv2d(16, kernel_size=5)
                      self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.fc1 = nn.LazyLinear(120)
                      self.fc2 = nn.LazyLinear(84)
                      self.fc3 = nn.LazyLinear(10)
                      self.relu = nn.ReLU()
                      self.flat = nn.Flatten()
                  def forward(self, x):
                      out = self.maxp1(self.relu(self.conv1(x)))
                      out = self.maxp2(self.relu(self.conv2(out)))
                      out = self.flat(out)
                      out = self.relu(self.fc1(out))
                      out = self.relu(self.fc2(out))
                      out = self.fc3(out)
                      return out
              model 6b = alt6b LeNet().to(device=try gpu())
              optimizer 6b = optim.SGD(model 6b.parameters(), lr=5e-2)
              model 6b.eval()
   Out[207]: alt6b LeNet(
                (conv1): LazyConv2d(0, 6, kernel size=(5, 5), stride=(1, 1), padding=
              (2, 2))
                (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv2): LazyConv2d(0, 16, kernel size=(5, 5), stride=(1, 1))
                (maxp2): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (fc1): LazyLinear(in_features=0, out_features=120, bias=True)
                (fc2): LazyLinear(in features=0, out features=84, bias=True)
                (fc3): LazyLinear(in features=0, out features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start_dim=1, end_dim=-1)
```

Duration = 10.28 seconds Training Loss: 1.54733 Training Accuracy: 0.702 Validation Accuracy: 0.698

#### Epoch 5:

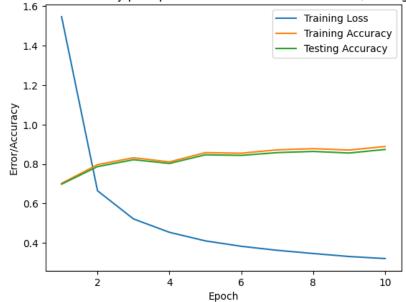
Duration = 8.634 seconds Training Loss: 0.41036 Training Accuracy: 0.858 Validation Accuracy: 0.847

#### Epoch 10:

Duration = 11.12 seconds Training Loss: 0.32063 Training Accuracy: 0.889 Validation Accuracy: 0.874

Total Training Time = 193.704 seconds Average Training Time per Epoch = 19.37 seconds

Figure 6b - Loss and Accuracy per Epoch for the altered LeNet Model (Halving Learning Rate)



```
In [236]:
          ▶ class alt7 LeNet(nn.Module):
                  def __init__(self):
                      super(). init ()
                      self.conv1 = nn.LazyConv2d(10, kernel size=3, padding=1)
                      self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv2 = nn.LazyConv2d(25, kernel_size=3)
                      self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.fc1 = nn.LazyLinear(200)
                      self.fc2 = nn.LazyLinear(120)
                      self.fc3 = nn.LazyLinear(84)
                      self.fc4 = nn.LazyLinear(10)
                      self.relu = nn.ReLU()
                      self.flat = nn.Flatten()
                  def forward(self, x):
                      out = self.maxp1(self.relu(self.conv1(x)))
                      out = self.maxp2(self.relu(self.conv2(out)))
                      out = self.flat(out)
                      out = self.relu(self.fc1(out))
                      out = self.relu(self.fc2(out))
                      out = self.relu(self.fc3(out))
                      out = self.fc4(out)
                      return out
              model 7 = alt7 LeNet().to(device=try gpu())
              optimizer 7 = optim.SGD(model 7.parameters(), lr=2e-1, weight decay=1e-4)
              model 7.eval()
   Out[236]: alt7 LeNet(
                (conv1): LazyConv2d(0, 10, kernel_size=(3, 3), stride=(1, 1), padding=
              (1, 1))
                (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv2): LazyConv2d(0, 25, kernel size=(3, 3), stride=(1, 1))
                (maxp2): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (fc1): LazyLinear(in features=0, out features=200, bias=True)
                (fc2): LazyLinear(in features=0, out features=120, bias=True)
                (fc3): LazyLinear(in_features=0, out_features=84, bias=True)
                (fc4): LazyLinear(in_features=0, out_features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start_dim=1, end_dim=-1)
In [237]:
           train loader 2 = DataLoader(fashion MNIST, batch size=32, shuffle=True)
              val_loader_2 = DataLoader(fashion_MNIST_val, batch_size=int(len(fashion_MN]
              # Had to reload the loaders with smaller batch sizes because of running low
```

```
In [239]:

★ t_loss_hist_7, t_acc_hist_7, v_acc_hist_7 = training_loop(10,)

                                                                          optimizer_7,
                                                                          model 7,
                                                                          nn.CrossEntropyLoss
                                                                          train_loader_2,
                                                                          val_loader_2,
                                                                          5)
```

Duration = 13.155 seconds Training Loss: 0.67564 Training Accuracy: 0.86 Validation Accuracy: 0.854

Epoch 5:

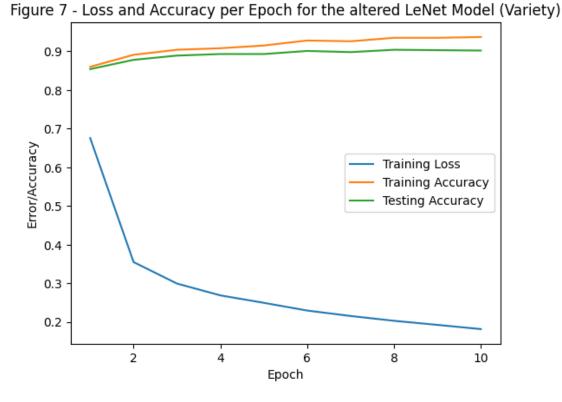
Duration = 13.621 seconds Training Loss: 0.24963 Training Accuracy: 0.915 Validation Accuracy: 0.893

Epoch 10:

Duration = 13.613 seconds Training Loss: 0.18169 Training Accuracy: 0.937 Validation Accuracy: 0.902

Total Training Time = 224.505 seconds Average Training Time per Epoch = 22.45 seconds

In [240]: ▶ title 7 = "Figure 7 - Loss and Accuracy per Epoch for the altered LeNet Mod plot\_model(title\_7, 9, t\_loss\_hist\_7, t\_acc\_hist\_7, v\_acc\_hist\_7, 'right')



```
In [243]:
                  def __init__(self):
                      super(). init ()
                      self.conv1 = nn.LazyConv2d(10, kernel size=3, padding=1)
                      self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv1 drop = nn.Dropout2d(p=0.4)
                      self.conv2 = nn.LazyConv2d(25, kernel_size=3)
                      self.maxp2 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv2 drop = nn.Dropout2d(p=0.4)
                      self.fc1 = nn.LazyLinear(200)
                      self.fc1 drop = nn.Dropout(p=0.3)
                      self.fc2 = nn.LazyLinear(120)
                      self.fc2 drop = nn.Dropout(p=0.2)
                      self.fc3 = nn.LazyLinear(84)
                      self.fc4 = nn.LazyLinear(10)
                      self.relu = nn.ReLU()
                      self.flat = nn.Flatten()
                  def forward(self, x):
                      out = self.maxp1(self.relu(self.conv1(x)))
                      out = self.conv1 drop(out)
                      out = self.maxp2(self.relu(self.conv2(out)))
                      out = self.conv2 drop(out)
                      out = self.flat(out)
                      out = self.relu(self.fc1(out))
                      out = self.fc1 drop(out)
                      out = self.relu(self.fc2(out))
                      out = self.fc2 drop(out)
                      out = self.relu(self.fc3(out))
                      out = self.fc4(out)
                      return out
              model 8 = drop LeNet().to(device=try gpu())
              optimizer_8 = optim.SGD(model_8.parameters(), lr=2e-1, weight_decay=1e-4)
              model_8.eval()
              drop LeNet(
   Out[243]:
                (conv1): LazyConv2d(0, 10, kernel size=(3, 3), stride=(1, 1), padding=
              (1, 1))
                (maxp1): MaxPool2d(kernel_size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv1 drop): Dropout2d(p=0.4, inplace=False)
                (conv2): LazyConv2d(0, 25, kernel size=(3, 3), stride=(1, 1))
                (maxp2): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              mode=False)
                (conv2_drop): Dropout2d(p=0.4, inplace=False)
                (fc1): LazyLinear(in features=0, out features=200, bias=True)
                (fc1 drop): Dropout(p=0.3, inplace=False)
                (fc2): LazyLinear(in features=0, out features=120, bias=True)
                (fc2 drop): Dropout(p=0.2, inplace=False)
                (fc3): LazyLinear(in features=0, out features=84, bias=True)
                (fc4): LazyLinear(in_features=0, out_features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start dim=1, end dim=-1)
              )
```

Epoch 5:

Duration = 13.294 seconds Training Loss: 0.2448 Training Accuracy: 0.925 Validation Accuracy: 0.904

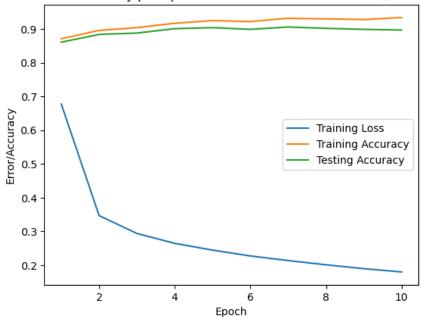
Epoch 10:

Duration = 14.133 seconds Training Loss: 0.18009 Training Accuracy: 0.934 Validation Accuracy: 0.897

Total Training Time = 228.715 seconds Average Training Time per Epoch = 22.871 seconds

```
In [246]: ► title_8 = "Figure 8 - Loss and Accuracy per Epoch for the altered LeNet Mod plot_model(title_8, 10, t_loss_hist_8, t_acc_hist_8, v_acc_hist_8, 'right')
```

Figure 8 - Loss and Accuracy per Epoch for the altered LeNet Model (Variety w/ Dropout)



```
In [279]:
          def __init__(self):
                     super(). init ()
                     self.conv1 = nn.LazyConv2d(8, kernel size=3, stride=2)
                     self.maxp1 = nn.MaxPool2d(kernel size=2, stride=2)
                      self.conv1 drop = nn.Dropout2d(p=0.3)
                      self.fc1 = nn.LazyLinear(40)
                     self.fc2 = nn.LazyLinear(10)
                     self.relu = nn.ReLU()
                     self.flat = nn.Flatten()
                 def forward(self, x):
                     out = self.maxp1(self.relu(self.conv1(x)))
                     out = self.conv1_drop(out)
                     out = self.flat(out)
                     out = self.relu(self.fc1(out))
                     out = self.fc2(out)
                     return out
             model 9 = drop LeNet().to(device=try gpu())
              optimizer 9 = optim.SGD(model 9.parameters(), lr=2e-1)
             model_9.eval()
   Out[279]: drop LeNet(
                (conv1): LazyConv2d(0, 8, kernel size=(3, 3), stride=(2, 2))
                (maxp1): MaxPool2d(kernel size=2, stride=2, padding=0, dilation=1, ceil
              _mode=False)
                (conv1 drop): Dropout2d(p=0.3, inplace=False)
                (fc1): LazyLinear(in features=0, out features=40, bias=True)
                (fc2): LazyLinear(in_features=0, out_features=10, bias=True)
                (relu): ReLU()
                (flat): Flatten(start dim=1, end dim=-1)
```

```
In [280]:

★ t_loss_hist_9, t_acc_hist_9, v_acc_hist_9 = training_loop(10,)

                                                                          optimizer 9,
                                                                          model 9,
                                                                          nn.CrossEntropyLoss
                                                                          train_loader_1,
                                                                          val_loader_1,
                                                                          5)
```

Duration = 7.261 seconds Training Loss: 0.66293 Training Accuracy: 0.829 Validation Accuracy: 0.819

Epoch 5:

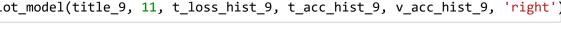
Duration = 7.352 seconds Training Loss: 0.36155 Training Accuracy: 0.859 Validation Accuracy: 0.849

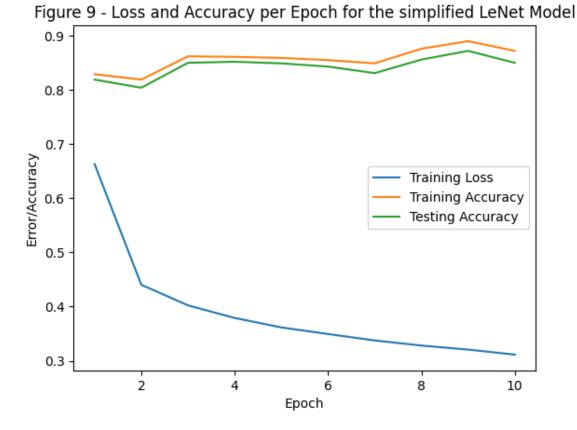
Epoch 10:

Duration = 7.29 seconds Training Loss: 0.31137 Training Accuracy: 0.872 Validation Accuracy: 0.85

Total Training Time = 143.395 seconds Average Training Time per Epoch = 14.34 seconds

In [281]: ▶ title 9 = "Figure 9 - Loss and Accuracy per Epoch for the simplified LeNet plot\_model(title\_9, 11, t\_loss\_hist\_9, t\_acc\_hist\_9, v\_acc\_hist\_9, 'right'





```
In [288]: Number of parameters for Altered LeNet Mc
print('{:<30} {:<8}'.format('Number of parameters for Altered LeNet Mc
print('\n{:<30} {:<8}'.format('Computational complexity for Simple Model)
print('\n{:<30} {:<8}'.format('Computational complexity for Simple Model)
print('\n{:<30} {:<8}'.format('Number of parameters for Altered LeNet Mc
print('\n{:<30} {:<8}'.format('Number of parameters for Simple Model: ',</pre>
```

Computational complexity for LeNet: 435.85 KMac Number of parameters for LeNet: 61.71 k

Computational complexity for Altered LeNet Model: 644.62 KMac Number of parameters for Altered LeNet Model: 217.71 k

Computational complexity for Simple Model: 28.23 KMac Number of parameters for Simple Model: 12.05 k

```
In [20]:
          def __init__(self):
                     super(). init ()
                     self.conv1 = nn.LazyConv2d(6, kernel size=5, padding=2)
                     self.avgp1 = nn.AvgPool2d(kernel size=2, stride=2)
                     self.conv2 = nn.LazyConv2d(16, kernel_size=5)
                     self.avgp2 = nn.AvgPool2d(kernel size=2, stride=2)
                     self.fc1 = nn.LazyLinear(120)
                     self.fc2 = nn.LazyLinear(84)
                     self.fc3 = nn.LazyLinear(10)
                     self.sgmd = nn.Sigmoid()
                     self.flat = nn.Flatten()
                 def forward(self, x):
                     out = self.avgp1(self.sgmd(self.conv1(x)))
                     out = self.avgp2(self.sgmd(self.conv2(out)))
                     out = self.flat(out)
                     out = self.sgmd(self.fc1(out))
                     out = self.sgmd(self.fc2(out))
                     out = self.fc3(out)
                     return out
             model lenet = LeNet().to(device=try gpu())
             optimizer lenet = optim.SGD(model lenet.parameters(), lr=1e-1)
             model lenet.eval()
   Out[20]: LeNet(
               (conv1): LazyConv2d(0, 6, kernel size=(5, 5), stride=(1, 1), padding=
             (2, 2))
               (avgp1): AvgPool2d(kernel_size=2, stride=2, padding=0)
               (conv2): LazyConv2d(0, 16, kernel size=(5, 5), stride=(1, 1))
               (avgp2): AvgPool2d(kernel size=2, stride=2, padding=0)
               (fc1): LazyLinear(in_features=0, out_features=120, bias=True)
               (fc2): LazyLinear(in features=0, out features=84, bias=True)
               (fc3): LazyLinear(in features=0, out features=10, bias=True)
               (sgmd): Sigmoid()
               (flat): Flatten(start dim=1, end dim=-1)
```

```
★ | t_loss_hist_lenet, t_acc_hist_lenet, v_acc_hist_lenet = training_loop(30,)

| t_loss_hist_lenet, t_acc_hist_lenet, v_acc_hist_lenet = training_loop(30,)
| t_loss_hist_lenet, t_acc_hist_lenet, v_acc_hist_lenet = training_loop(30,)
| t_loss_hist_lenet, t_acc_hist_lenet, v_acc_hist_lenet = training_loop(30,)
| t_loss_hist_lenet, t_acc_hist_lenet, v_acc_hist_lenet = training_loop(30,)
| t_loss_hist_lenet, t_acc_hist_lenet, v_acc_hist_lenet, v_acc_hist
In [21]:
                                                                                                                                                                                                                       optimizer lenet,
                                                                                                                                                                                                                       model lenet,
                                                                                                                                                                                                                       nn.CrossEntropyLoss
                                                                                                                                                                                                                       train_loader_1,
                                                                                                                                                                                                                       val_loader_1,
                                                                                                                                                                                                                       5)
                                         Epoch 1:
                                                                 Duration = 19.269 seconds
                                                                 Training Loss: 2.30714
                                                                  Training Accuracy: 0.1
                                                                  Validation Accuracy: 0.1
                                         Epoch 5:
                                                                 Duration = 18.21 seconds
                                                                 Training Loss: 2.3059
                                                                  Training Accuracy: 0.1
                                                                 Validation Accuracy: 0.1
                                         Epoch 10:
                                                                 Duration = 18.659 seconds
                                                                 Training Loss: 2.30354
                                                                  Training Accuracy: 0.1
                                                                 Validation Accuracy: 0.1
                                         Epoch 15:
                                                                 Duration = 19.801 seconds
                                                                 Training Loss: 1.1266
                                                                  Training Accuracy: 0.579
                                                                 Validation Accuracy: 0.576
                                         Epoch 20:
                                                                 Duration = 17.896 seconds
                                                                 Training Loss: 0.79771
                                                                 Training Accuracy: 0.71
                                                                 Validation Accuracy: 0.704
                                         Epoch 25:
                                                                 Duration = 19.274 seconds
                                                                  Training Loss: 0.65109
                                                                  Training Accuracy: 0.756
                                                                 Validation Accuracy: 0.748
                                         Epoch 30:
                                                                 Duration = 18.923 seconds
                                                                  Training Loss: 0.57976
                                                                  Training Accuracy: 0.783
                                                                 Validation Accuracy: 0.774
                                         Total Training Time = 1143.003 seconds
```

Average Training Time per Epoch = 38.1 seconds

localhost:8888/notebooks/Documents/ECGR-4106/Homework 2.ipynb

In [23]: 

title\_10 = "Figure 10 - Loss and Accuracy per Epoch for the original LeNet plot\_model(title\_10, 12, t\_loss\_hist\_lenet, t\_acc\_hist\_lenet, v\_acc\_hist\_lenet)

Figure 10 - Loss and Accuracy per Epoch for the original LeNet Model

