# Advanced Intelligence Systems

Assignment 03: Part 02 Saurabh Jawahar Kakade sk2354@nau.edu

#### Solution 01:

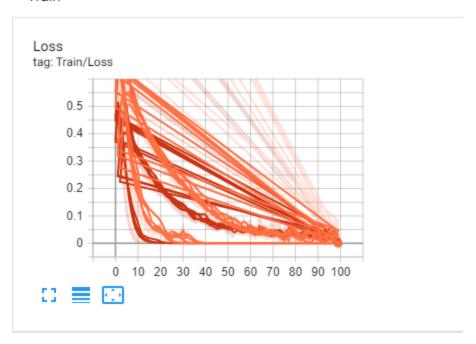
- Data sets used: FASHION-MNIST and KMINST
- Three train/splits are used.

#### Solution 02:

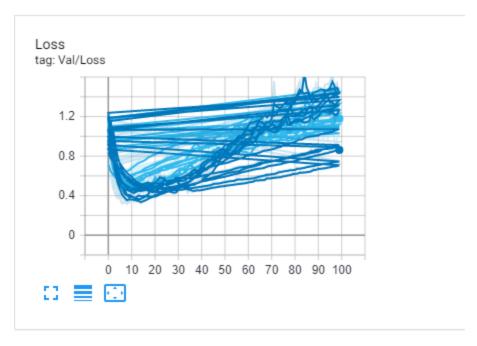
Output: Epoch = 100 (on local machine)

```
E:\Spring 2021\Subjects\AI\Assignments\Assignment 03\PART 2>python AI A03 PART2 sk2354.py && tensorboard --logdir=runs
Test Error: Accuracy: 10.4%, Avg loss: 0.072578
DATA=FashionMNIST model=fullyConnected splitNum=0 selectedEpochs=9
Test Error: Accuracy: 10.3%, Avg loss: 0.072655
DATA=FashionMNIST model=convolutional splitNum=0 selectedEpochs=9
Test Error: Accuracy: 9.7%, Avg loss: 0.072411
Test Error: Accuracy: 12.5%, Avg loss: 0.072613
DATA=FashionMNIST model=fullyConnected splitNum=1 selectedEpochs=9
Test Error: Accuracy: 13.5%, Avg loss: 0.072382
DATA=FashionMNIST model=convolutional splitNum=1 selectedEpochs=7
Test Error: Accuracy: 13.2%, Avg loss: 0.072571
Test Error: Accuracy: 9.5%, Avg loss: 0.072807
DATA=FashionMNIST model=fullyConnected splitNum=2 selectedEpochs=14
Test Error: Accuracy: 10.5%, Avg loss: 0.072718
DATA=FashionMNIST model=convolutional splitNum=2 selectedEpochs=8
Test Error: Accuracy: 10.0%, Avg loss: 0.072740
Test Error: Accuracy: 7.1%, Avg loss: 0.072672
DATA=KMNIST model=fullyConnected splitNum=0 selectedEpochs=3
Test Error: Accuracy: 9.0%, Avg loss: 0.072586
DATA=KMNIST model=convolutional splitNum=0 selectedEpochs=5
Test Error: Accuracy: 11.9%, Avg loss: 0.072470
Test Error: Accuracy: 7.9%, Avg loss: 0.072748
DATA-KMNIST model=fullyConnected splitNum=1 selectedEpochs=4
Test Error: Accuracy: 10.7%, Avg loss: 0.072558
DATA=KMNIST model=convolutional splitNum=1 selectedEpochs=4
Test Error: Accuracy: 8.4%, Avg loss: 0.072705
Test Error: Accuracy: 11.3%, Avg loss: 0.072491
DATA=KMNIST model=fullyConnected splitNum=2 selectedEpochs=5
Test Error: Accuracy: 10.7%, Avg loss: 0.072481
DATA=KMNIST model=convolutional splitNum=2 selectedEpochs=4
Test Error: Accuracy: 9.8%, Avg loss: 0.072589
TensorFlow installation not found - running with reduced feature set.
Serving TensorBoard on localhost; to expose to the network, use a proxy or pass --bind_all
TensorBoard 2.2.1 at http://localhost:6006/ (Press CTRL+C to quit)
```

# Train



# Val



• Batch size: 32

Learning rate: 3 sec (avg)Max Epochs used: 100.

• The number of epochs that minimized the validation loss, the test error percent for every data set / model / split:

(selectedEpochs below is the number of epochs that minimized the validation loss.)

#### FOLD 0:

- o FeatureLess Model: Test Error: Accuracy: 10.4%, Avg loss: 0.072578
- DATA=FashionMNIST model=fullyConnected splitNum=0 selectedEpochs=9
- Test Error: Accuracy: 10.3%, Avg loss: 0.072655
- o DATA=FashionMNIST model=convolutional splitNum=0 selectedEpochs=9
- Test Error: Accuracy: 9.7%, Avg loss: 0.072411

#### FOLD 1:

- FeatureLess Model: Test Error: Accuracy: 12.5%, Avg loss: 0.072613
- DATA=FashionMNIST model=fullyConnected splitNum=1 selectedEpochs=9
- Test Error: Accuracy: 13.5%, Avg loss: 0.072382
- DATA=FashionMNIST model=convolutional splitNum=1 selectedEpochs=7
- o Test Error: Accuracy: 13.2%, Avg loss: 0.072571

## FOLD 2:

- FeatureLess Model: Test Error: Accuracy: 9.5%, Avg loss: 0.072807
- DATA=FashionMNIST model=fullyConnected splitNum=2 selectedEpochs=14
- Test Error: Accuracy: 10.5%, Avg loss: 0.072718
- o DATA=FashionMNIST model=convolutional splitNum=2 selectedEpochs=8
- Test Error: Accuracy: 10.0%, Avg loss: 0.072740

#### FOLD 0:

- o FeatureLess Model: Test Error: Accuracy: 7.1%, Avg loss: 0.072672
- DATA=KMNIST model=fullyConnected splitNum=0 selectedEpochs=3
- Test Error: Accuracy: 9.0%, Avg loss: 0.072586
- DATA=KMNIST model=convolutional splitNum=0 selectedEpochs=5
- Test Error: Accuracy: 11.9%, Avg loss: 0.072470

### FOLD 1:

- FeatureLess Model: Test Error: Accuracy: 7.9%, Avg loss: 0.072748
- DATA=KMNIST model=fullyConnected splitNum=1 selectedEpochs=4
- Test Error: Accuracy: 10.7%, Avg loss: 0.072558
- DATA=KMNIST model=convolutional splitNum=1 selectedEpochs=4
- Test Error: Accuracy: 8.4%, Avg loss: 0.072705

### FOLD 2:

- FeatureLess Model: Test Error: Accuracy: 11.3%, Avg loss: 0.072491
- DATA=KMNIST model=fullyConnected splitNum=2 selectedEpochs=5
- Test Error: Accuracy: 10.7%, Avg loss: 0.072481
- DATA=KMNIST model=convolutional splitNum=2 selectedEpochs=4

- Test Error: Accuracy: 9.8%, Avg loss: 0.072589
- Neural network accuracy:
  - The two neural networks such as Fully Connect Network and Convolution Network shows **better** accuracy than the baseline featureless model.
  - o Fully Connected Network showed more accuracy than convolution network.
  - Fully Connected Model shows better performance on an average than rest of the two networks.
  - Other data analyzed: KMNIST
    - KMNIST found to be difficult for the neural network to learn.
    - Test error accuracy by KMINST on both the neural networks (on an average) showed little poor performance as compared to Fashion-MINST data set.
    - The accuracy for KMNIST was lesser than Fashion-MNIST data set.
- Analysis of more than one train/split:
  - The getTrainTestData(data\_name, splitNum) function in the program was called for 3 folds to split everytime.
  - o For every call, the data was split in train= 6667 and test=3333
  - Here is the actual function:
    - def getTrainTestData(data\_name, splitNum):

```
train_data, test_data =
  torch.utils.data.random_split(full_data_dict[data_name],[6667,3333])
return train_data, test_data
```



```
DATA_SETS = ["FashionMNIST","KMNIST"]
loss_function = nn.CrossEntropyLoss()
results = {}
writer_dict = {}
subtrainResults = {}
def FeatureLess(train_set):
device = 'cuda' if torch.cuda.is_available() else 'cpu'
 pred_model = FullyConnected().to(device)
x1 = torch.rand(1, 28, 28, device=device)
logits = pred_model(x1)
 pred_probab = nn.Softmax(dim=1)(logits)
y_pred = pred_probab.argmax(1)
return y_pred
```

```
class FullyConnected(torch.nn.Module):
def __init__(self, input_size=784, h1=300, h2=100, output_size=10):
  super().__init__()
  self.layer_1 = torch.nn.Linear(input_size, h1)
  self.layer_2 = torch.nn.Linear(h1, h2)
  self.layer_3 = torch.nn.Linear(h2, output_size)
def forward(self, x):
  x = torch.flatten(x, start_dim=1)
  x = F.relu(self.layer_1(x))
  x = F.relu(self.layer_2(x))
  x = self.layer_3(x)
  return x
class LeNet(torch.nn.Module):
def __init__(self):
  super().__init__()
  self.conv1 = torch.nn.Conv2d(in_channels=1, out_channels= 6, kernel_size=5)
  self.conv2 = torch.nn.Conv2d(6, 16, 5)
  self.fc1 = torch.nn.Linear(4*4*16, 120)
  self.fc2 = torch.nn.Linear(120, 84)
  self.output = torch.nn.Linear(84, 10)
 def forward(self, x):
 x = F.relu(self.conv1(x))
```

```
# use x.shape to check the current size
  # print (x.shape)
 x = F.max_pool2d(x, 2, 2)
 x = F.relu(self.conv2(x))
 x = F.max_pool2d(x, 2, 2)
 x = x.view(-1, 4*4*16)
 x = F.relu(self.fc1(x))
 x = F.relu(self.fc2(x))
 x = self.output(x)
  return x
def reset_weights(m):
 Try resetting model weights to avoid
 weight leakage.
for layer in m.children():
 if hasattr(layer, 'reset_parameters'):
 # print(f'Reset trainable parameters of layer = {layer}')
  layer.reset_parameters()
def newModel(model):
if model == "fullyConnected":
  return FullyConnected()
```

```
if model == "convolutional":
 return LeNet()
for data_name in DATA_SETS: # For Extra Credit add another data set.
 data_loader = getattr(datasets, data_name)
 full_data_dict[data_name] = data_loader(
 root="data",
 train=False,
 download=True,
 transform=ToTensor()
 )
def getTrainTestData(data_name, splitNum):
train_data, test_data = torch.utils.data.random_split(full_data_dict[data_name],[6667,3333])
return train_data, test_data
def getMostFrequentLabel(train_set):
return FeatureLess(train_set)
```

```
def splitData(train_set):
transform_data = transforms.Compose([transforms.ToTensor(), transforms.Resize((28,28))])
train_dataset, val_dataset = torch.utils.data.random_split(train_set,[5557,1110])
train_loader = torch.utils.data.DataLoader(train_dataset, batch_size=32, shuffle=True)
validation_loader = torch.utils.data.DataLoader(val_dataset, batch_size=32, shuffle=True)
return train_loader, validation_loader
#***********************
def learn(subtrain_net, subtrain_set, MAX_EPOCHS, validation):
torch.nn.CrossEntropyLoss()
lenet model = LeNet()
fully_connected_model = FullyConnected()
for mm in 'LeNet', 'fully_connected_model':
 for ss in "subtrain", 'validation':
  writer_dict[mm + '_' + ss] = SummaryWriter("runs/" + ss + '_' + 'with' + '_' + mm)
```

```
d_loader = {'subtrain': subtrain_set, 'validation': validation}
neural_networks = {'LeNet':lenet_model, 'fully_connected_model':fully_connected_model}
for pattern in "LeNet", "fully_connected_model":
 optimizer = optim.Adam(neural_networks[pattern].parameters(),lr = 0.001)
 #Make an instance of the loss function
 criterion = torch.nn.CrossEntropyLoss()
 #use a for loop over epochs of learning
 for epoch in range(MAX_EPOCHS):
  loss_dict = {"epoch":epoch, "pattern":subtrain_net}
  for s in "subtrain", "validation":
    #************************
    if s == "subtrain":
      train_loss = 0
      #writing a for loop over batches using your DataLoader (for x,y in data_loader).
      for batch_index, (data,target) in enumerate(d_loader[s]):
```

```
pred = neural_networks[pattern](data) #predictions
    loss = criterion(pred, target) #loss
    optimizer.zero_grad() #zero the gradient
    loss.backward() #compute gradient
    optimizer.step() #Update the neural network weights using gradient
    train_loss += loss.item()
  #Compute loss
  train_loss /= len(d_loader[s])
  loss = train_loss
  writer_dict[pattern + "_" + s].add_scalar('Train/Loss',loss,epoch)
  # print("Subtrain: {}".format(pred))
  # print(max(pred))
else:
  train_loss = 0
  for data,target in d_loader[s]:
    pred = neural_networks[pattern](data)
    train_loss += criterion(pred,target).item()
  #Compute loss
  train_loss /= len(d_loader[s])
  loss = train_loss
  subtrainResults[epoch] = loss
  writer_dict[pattern + "_" + s].add_scalar('Val/Loss',loss,epoch)
loss_dict[s] = loss
```

```
#print these values on the screen and log these loss values to the tensorboard writer
  # print("pattern=%(pattern)s epoch=%(epoch)d subtrain=%(subtrain)f validation=%(validation)f" %
loss_dict)
return subtrainResults
#************************
def getBestEpochs(subtrain_result):
min_loss = min(subtrain_result.values())
for keyValue in subtrain_result:
   if subtrain_result[keyValue] == min_loss:
     min_losskey = keyValue
     # print(min_losskey)
return min_losskey
def PredictOnSet(train_net, test_set, model):
device = 'cuda' if torch.cuda.is_available() else 'cpu'
if model == "fullyConnected":
```

```
pred_model = FullyConnected().to(device)
  x1 = torch.rand(1, 28, 28, device=device)
  logits = pred_model(x1)
  pred_probab = nn.Softmax(dim=1)(logits)
 y_pred = pred_probab.argmax(1)
  return y_pred
else:
  pred_model = LeNet().to(device)
  x1 = torch.rand(1, 1, 28, 28, device=device)
  logits = pred_model(x1)
  pred_probab = nn.Softmax(dim=1)(logits)
  y_pred = pred_probab.argmax(1)
  return y_pred
def PercentError(test_predictions, test_set):
```

```
network = FullyConnected()
network.apply(reset_weights)
criterion = torch.nn.CrossEntropyLoss()
test_loader = torch.utils.data.DataLoader(test_set, batch_size=32, shuffle=True)
size = len(test_loader.dataset)
test_loss, correct = 0, 0
with torch.no_grad():
  for X, y in test_loader:
    pred = network(X)
    test_loss += criterion(pred, y).item()
    correct += (pred.argmax(1) == y).type(torch.float).sum().item()
test_loss /= size
correct /= size
print(f"Test Error: Accuracy: {(100*correct):>0.1f}%, Avg loss: {test_loss:>8f}")
```

```
def PercentError1(test_predictions, test_set):
# a = PredictOnSet(FullyConnected(), test_set)
# print(f"percent error {a}")
# print(f"test prediction: {test_predictions}")
network = FullyConnected()
network.apply(reset_weights)
optimizer = torch.optim.Adam(network.parameters(), Ir=1e-4)
test_loader = torch.utils.data.DataLoader(test_set, batch_size=32, shuffle=True)
# Evaluationfor this fold
correct, total = 0, 0
# for folds in range(MAX_EPOCHS):
with torch.no_grad():
 # Iterate over the test data and generate predictions
 for i, data in enumerate(test_loader, 0):
  # Get inputs
  inputs, targets = data
  # Generate outputs
```

```
outputs = network(inputs)
  # Set total and correct
  # _, predicted = torch.max(outputs.data, 1)
  total += targets.size(0)
  correct += (test_predictions == targets).sum().item()
 # Print accuracy
 # print('Accuracy for fold %d: %d %%' % (splitNum,100.0 * correct / total))
 # print('----')
 results[splitNum] = 100.0 * (correct / total)
# Print fold results
print(f'K-FOLD CROSS VALIDATION RESULTS FOR {splitNum} FOLDS')
print('----')
sum = 0.0
for key, value in results.items():
 print(f'Fold {key}: {value} %')
 sum += value
print(f'Average: {sum/len(results.items())} %')
return None
```

```
def TestErrorOneSplit(data_name, splitNum, model):
train_set, test_set = getTrainTestData(data_name, splitNum)
if model == "featureless":
  test_predictions = getMostFrequentLabel(train_set)
 else:
  subtrain_set, validation_set = splitData(train_set)
  subtrain_net = newModel(model)
  subtrain_result = learn(subtrain_net, subtrain_set, MAX_EPOCHS, validation=validation_set)
  selected_epochs = getBestEpochs(subtrain_result)
  print("DATA=%s model=%s splitNum=%d selectedEpochs=%d" % (data_name, model, splitNum,
selected_epochs))
  train_net = newModel(model)
  learn(train_net, train_set, selected_epochs, validation=validation_set)
  test_predictions = PredictOnSet(train_net, test_set, model)
```

