Imports & private packages

```
[ ] 🖟 8 cells hidden
```

Training routine

```
1 def epoch(net, train_loader, loss_func, optimizer, batch_size, seq_len, measurement='G'
       epoch_losses = np.zeros([len(train_loader), ])
 3
       correct = 0
 4
      incorrect = 0
 5
      for i, d in enumerate(train_loader):
 6
 7
           # Scale inputs and get predictions
 8
           inp = d[:, :, :-1] \#.squeeze(1)
 9
           inp[:, :, 0] /= 1000
10
           # Add random noise to heat capacity values to simulate real measurements
11
           d[:, :, 1] += torch.normal(mean=0, std=0.75, size=(d.shape[0], d.shape[1], ))
12
13
           predictions = net(inp.float().to(device))
14
15
           # Each batch consists of measurement batches, where seq_len measurements are pu
16
           # measurement batch, every measurement has the same label as it needs to be fro
17
18
           # This leads to a target array where the target class is contained for each mea
19
           # batch. With this, CrossEntropyLoss would not work as the predictions are made
           # batch and CEL therefore expects only on class label per measurement batch. Th
20
           # element of the last dimension of d is considered as target (all the entries i
21
           # same anyways so it could be any entry)
22
           targets = d[:, :, -1][:, 0].long().to(device)
23
24
           correct += (predictions.argmax(dim=-1) == targets).sum().item()
25
           incorrect += len(targets) - (predictions.argmax(dim=-1) == targets).sum().item(
26
27
           # Calculate the loss
28
           loss = loss_func(predictions, targets)
29
           epoch losses[i] = loss
30
31
32
          # Backward step
33
          net.zero grad()
           loss.backward()
34
           optimizer.step()
35
36
       accuracy = correct/(correct + incorrect)
37
       return epoch_losses.mean(), accuracy
38
 1 from torch.nn.modules.activation import Softmax
 2 # Training and testing routines
```

```
4 def train(net, optimizer, train loader, test loader, batch size, seq len, measurement='
       loss func = nn.CrossEntropyLoss()
 6
 7
       losses = np.zeros([nr_epochs, ])
       accuracies = np.zeros([nr_epochs, ])
 8
 9
       best_loss, _ = epoch(net, train_loader, loss_func, optimizer, batch_size, seq_len,
10
11
      best_net = net
12
13
      for i in range(nr_epochs):
           losses[i], accuracies[i] = epoch(net, train_loader, loss_func, optimizer, batch
14
15
           if losses[i] < best loss:</pre>
               best net = net
16
               best_loss = losses[i]
17
18
19
           if i % 10 == 0:
               print('Epoch ', i)
20
               print('Loss: ', losses[i])
21
22
               print('Training accuracy: ', accuracies[i])
23
               #test(best_net, test_loader)
24
       return best net, losses, accuracies
25
26
27 def test(net, test_loader):
28
       correct = 0
       incorrect = 0
29
30
      for d in test_loader:
31
32
           inp = d[:, :, :-1] #.squeeze(1)
33
           inp[:, :, 0] /= 1000
34
           predictions = net(inp.float().to(device)).squeeze()#.argmax()
           targets = d[:, :, -1][:, 0].long().to(device)
35
36
37
           correct += (predictions.argmax(dim=-1) == targets).sum().item()
           incorrect += len(targets) - (predictions.argmax(dim=-1) == targets).sum().item(
38
           #print(predictions.argmax(dim=-1), targets, Softmax()(predictions).amax(dim=-1)
39
40
       accuracy = correct/(correct + incorrect)
41
42
       print('Test accuracy: ', accuracy)
43
44
      return accuracy
```

Run

Network definition

```
1 # Hyperparameters
2 seq_len = 5
3 measurement = 'C'
4 hatch size = 256
```

```
T DUCCH_312C - 270
 5 \text{ nr epochs} = 500
 6 lr = 0.001
 7 hidden size linear⋅=⋅128
 8 hidden layers⋅=⋅2
 9 \text{ step} = 0.05
 1 def create data loaders(elements, splits, seq len, validation=False, measurement='G', u
       dc = DatasetCreator(elements=elements, splits=splits, validation=validation, seq_1
 2
       train_dataset, test_dataset, val_dataset = dc.get_datasets()
 3
 4
 5
      # Create the DataLoaders
      train_loader = DataLoader(train_dataset, batch_size=batch_size, shuffle=shuffle)
 6
 7
      test_loader = DataLoader(test_dataset, batch_size=1, shuffle=shuffle)
      if val dataset:
 8
           val_loader = DataLoader(val_dataset)
 9
10
      else:
           val loader = None
11
12
13
       return train_loader, test_loader, val_loader
 1 # Create the network
 2 t = True
 3 net = ElementClassifier(train, in_features=seq_len * 2, hidden_size_linear=hidden_size_
 5 # Check if cuda is available and send net to device
 6 device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
 7
 8 net = net.to(device)
10 optimizer = Adam(net.parameters(), lr=lr)
```

Train

```
1 elements = None
3 train loader, test loader, val loader = create data loaders(elements, (0.8, 0.2), seq 1
   Dataset shape: (411319, 5, 3)
1 best net, losses, accuracies = train(net, optimizer, train loader, test loader, batch s
2 test(best_net, test_loader)
   Loss: 0.6527893115032067
☐ Training accuracy: 0.7493988850502894
   Epoch 320
   Loss: 0.6489172258414908
   Training accuracy: 0.749306499335066
   Epoch 330
   Loss: 0.6513613111848926
   Training accuracy: 0.7489199380529468
    Epoch
          340
    1000
          A 6/00/E/070176777
```

LUSS. 0.04004040/0120232 Training accuracy: 0.7508721940878005 Epoch 350 Loss: 0.6497220686117017 Training accuracy: 0.7504686143844559 Epoch 360 Loss: 0.6475774214163135 Training accuracy: 0.7512757737911451 Epoch 370 Loss: 0.6476132421387452 Training accuracy: 0.7504637519783914 Epoch 380 Loss: 0.6497419062577647 Training accuracy: 0.749739253474797 Epoch 390 Loss: 0.6512461012112256 Training accuracy: 0.749430490689708 Epoch 400 Loss: 0.6452652581393088 Training accuracy: 0.7515602245459121 Epoch 410 Loss: 0.6463633147024856 Training accuracy: 0.7512757737911451 Epoch 420 Loss: 0.6464975365456692 Training accuracy: 0.7510861399546338 Epoch 430 Loss: 0.6462330725078784 Training accuracy: 0.7514581140185598 Epoch 440 Loss: 0.6446551240669991 Training accuracy: 0.7511104519849557 Epoch 450 Loss: 0.6480033389143716 Training accuracy: 0.7504977888208422 Epoch 460 Loss: 0.6454645953771114 Training accuracy: 0.7513997651457871 Epoch 470 Loss: 0.642114242301198 Training accuracy: 0.7529557350863928 Epoch 480 Loss: 0.6420811612968054 Training accuracy: 0.7531988553896124 Epoch 490 Loss: 0.6418265373479081 Training accuracy: 0.7522020621464119 Test accuracy: 0.891244983041953 0.891244983041953 1 test(best net, test loader)

1 torch.save(best_net, 'ElementClassifier_var_8912.pth')

Test on Barin data

```
1 net = torch.load('/content/ElementClassifier 9782 3.pth').to(device)
1 inp = torch.tensor([[[ .3,     11.403],
                     [.7, 22.25],
                     [ .8, 23.364],
3
4
                     [.9, 24.248],
5
                     [1.0, 24.979]]).to(device)
1 out = net(inp)
1 print(Encoder()(out.argmax(dim=-1).item()))
2 print(Softmax(dim=-1)(out).amax(dim=-1))
   tensor([1.], grad_fn=<AmaxBackward0>)
1 test(net, test_loader)
   Test accuracy: 0.9772111058201572
   0.9772111058201572
```