

## Imports & private packages

[ ] ↪ 8 cells hidden

## Training routine

```

1 def epoch(net, train_loader, test_loader, loss_func, optimizer, batch_size, seq_len):
2     epoch_losses = np.zeros([len(train_loader), ])
3     correct = 0
4     incorrect = 0
5
6     for i, d in enumerate(train_loader):
7         # Scale inputs and get predictions
8         inp = d[:, :, :-1]
9         #inp = d[:, :, [0, 2]]
10        inp[:, :, 0] /= 1000
11        predictions = net(inp.float()).squeeze()
12
13        # Each batch consists of measurement batches, where seq_len measurements are pu
14        # measurement batch, every measurement has the same label as it needs to be fro
15        # This leads to a target array where the target class is contained for each mea
16        # batch. With this, CrossEntropyLoss would not work as the predictions are made
17        # batch and CEL therefore expects only on class label per measurement batch. Th
18        # element of the last dimension of d is considered as target (all the entries i
19        # same anyways so it could be any entry)
20        targets = d[:, :, -1][:, 0].long()
21
22        correct += (predictions.argmax(dim=-1) == targets).sum().item()
23        incorrect += len(targets) - (predictions.argmax(dim=-1) == targets).sum().item()
24
25        #print(predictions.argmax(dim=-1), targets)
26
27        # Calculate the loss
28        loss = loss_func(predictions, targets)
29        epoch_losses[i] = loss
30
31        # Backward step
32        net.zero_grad()
33        loss.backward()
34        optimizer.step()
35
36    # Evaluate the network on the test set but DO NOT train on it
37    test_loss, test_accuracy = test(net, test_loader, loss_func)
38
39    train_accuracy = correct/(correct + incorrect)
40    print('Training accuracy: ', train_accuracy)
41    return epoch_losses.mean(), train_accuracy, test_loss, test_accuracy

```

```

1 # Training and testing routines
2
3 def train(net, train_loader, test_loader, batch_size, seq_len):
4     # Hyperparameters
5     nr_epochs = 250
6     lr = 0.001
7
8     loss_func = nn.CrossEntropyLoss()
9     optimizer = Adam(net.parameters(), lr=lr)
10
11     train_losses = np.zeros([nr_epochs, ])
12     train_accuracies = np.zeros([nr_epochs, ])
13     test_losses = np.zeros([nr_epochs, ])
14     test_accuracies = np.zeros([nr_epochs, ])
15
16     # Save the net with the lowest training loss as the best net
17     best_net = net
18     _, _, best_loss, _ = epoch(net, train_loader, test_loader, loss_func, optimizer, ba
19
20     for i in range(nr_epochs):
21         train_losses[i], train_accuracies[i], test_losses[i], test_accuracies[i] = epoc
22
23         if test_losses[i] < best_loss:
24             best_net = net
25
26         if i % 10 == 0:
27             print(train_losses[i])
28
29     return best_net, train_losses, train_accuracies, test_losses, test_accuracies
30
31 def test(net, test_loader, loss_func=None):
32     correct = 0
33     incorrect = 0
34
35     test_losses = np.zeros([len(test_loader), ])
36
37     for i, d in enumerate(test_loader):
38         inp = d[:, :, :-1]
39         #inp = d[:, :, [0, 2]]
40         inp[:, :, 0] /= 1000
41         predictions = net(inp.float())
42         targets = d[:, :, -1][:, 0].long()
43
44         correct += (predictions.argmax(dim=-1).flatten() == targets).sum().item()
45         incorrect += len(targets) - (predictions.argmax(dim=-1).flatten() == targets).s
46
47         # Get the testing loss
48         if loss_func is not None:
49             loss = loss_func(predictions.squeeze(), targets)
50             test_losses[i] = loss
51
52     accuracy = correct/(correct + incorrect)
53     print('Test accuracy: ', accuracy)
54
55     return test_losses.mean(), accuracy

```

## ▼ Network definition

```

1 # Hyperparameters
2 seq_len = 1
3 batch_size = 256
4
5 # Create the dataset
6 measurement = 'C'
7 dc = DatasetCreator(elements=None, splits=(0.8, 0.2), validation=False, seq_len=seq_len)
8 train_dataset, test_dataset, val_dataset = dc.get_datasets()
9
10 # Create the DataLoaders
11 train_loader = DataLoader(train_dataset, batch_size=batch_size, shuffle=True)
12 test_loader = DataLoader(test_dataset, batch_size=batch_size, shuffle=True)
13 if val_dataset:
14     val_loader = DataLoader(val_dataset)

    Dataset shape: (205592, 1, 4)

1 # Create the network
2 t = True
3 net = PhaseClassifier(train=t, measurement=measurement, hidden_layers=2, hidden_size=12

```

## ▼ Training

```

1 # Train the network
2 best_net, train_losses, train_accuracies, test_losses, test_accuracies = train(net, tra
3
4 # Test the trained network
5 test(best_net, test_loader)

    Test accuracy:  0.9877167068335536
    Training accuracy:  0.9899023308299933
    Test accuracy:  0.8980214568918604
    Training accuracy:  0.9859089847854002
    Test accuracy:  0.9861424240068414
    Training accuracy:  0.9931563484960504
    Test accuracy:  0.9812252196221721
    Training accuracy:  0.9892797385112261
    Test accuracy:  0.9915260825623883
    Training accuracy:  0.9864148410443986
    Test accuracy:  0.9876584000621939
    Training accuracy:  0.993730300789914
    Test accuracy:  0.9901461556402084
    Training accuracy:  0.9884431300828826
    Test accuracy:  0.984159993780611
    Training accuracy:  0.9904422351064244
    0.030177996339089935
    Test accuracy:  0.9856565342455104

```

```
Training accuracy: 0.9893478345460913
Test accuracy: 0.9873668662053953
Training accuracy: 0.9883458500330752
Test accuracy: 0.9024916426961052
Training accuracy: 0.9932974045682711
Test accuracy: 0.9835574904765607
Training accuracy: 0.9914198996069886
Test accuracy: 0.9958796548239135
Training accuracy: 0.9867164091988015
Test accuracy: 0.9924784264945969
Training accuracy: 0.99486361337017
Test accuracy: 0.9924201197232372
Training accuracy: 0.9894353865909179
Test accuracy: 0.9962100598616186
Training accuracy: 0.9879761858438072
Test accuracy: 0.9919731011428127
Training accuracy: 0.9921786839954863
Test accuracy: 0.995510378605302
Training accuracy: 0.9879859138487879
0.039907462411011056
Test accuracy: 0.9957047345098344
Training accuracy: 0.9933168605782327
Test accuracy: 0.9940721449117624
Training accuracy: 0.985169656406864
Test accuracy: 0.9924784264945969
Training accuracy: 0.9918917078485544
Test accuracy: 0.9894076032029853
Training accuracy: 0.9890024903692751
Test accuracy: 0.9950050532535178
Training accuracy: 0.9906854352309429
Test accuracy: 0.9866477493586255
Training accuracy: 0.9922273240203899
Test accuracy: 0.9920314079141724
Training accuracy: 0.9876113856570294
Test accuracy: 0.9964627225375107
Training accuracy: 0.9901309389470407
Test accuracy: 0.9562893570706678
Training accuracy: 0.9908751313280673
Test accuracy: 0.9562893570706678
(0.0, 0.9562893570706678)
```

```
1 torch.save(best_net, 'PhaseClassifier_9563.pth')
```

```
1 fig, ax = plt.subplots(1, 2, figsize=(14,7))
2 ax[0].plot(train_losses, label='Training loss')
3 ax[0].plot(test_losses, label='Test loss')
4 ax[0].set_xlabel('Epochs'), ax[0].set_ylabel('Loss')
5 ax[0].set_title('Losses over time')
6 ax[0].grid()
7 ax[0].legend()
8 ax[1].plot(train_accuracies, label='Training accuracy')
9 ax[1].plot(test_accuracies, label='Test accuracy')
10 ax[1].set_xlabel('Epochs'), ax[1].set_ylabel('Accuracy')
11 ax[1].set_title('Training and test accuracy over time')
12 ax[1].grid()
13 ax[0].legend()
14 plt.show()
```

[illegible]



☒ ☐