Imports & private packages

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

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

- Run

Network definition

```
1 # Hyperparameters
2 seq_len = 5
3 measurement = 'C'
4 batch_size = 256
5 nr_epochs = 250
6 lr = 0.001
7 hidden size linear = 128
```

```
8 \text{ hidden layers} = 2
 9 ctan = 0 05
 1 def create_data_loaders(elements, splits, seq_len, validation=False, measurement='G', s
       dc = DatasetCreator(elements=elements, splits=splits, validation=validation, seq_l
       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
 9
          val_loader = DataLoader(val_dataset)
10
      else:
          val_loader = None
11
12
       return train_loader, test_loader, val_loader
13
 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 = ['Ag', 'Au', 'Fe', 'Cu', 'Pt', 'Mg', 'C', 'Ca', 'Co', 'S',
               'Rh', 'Ru', 'Ti', 'W', 'Yb', 'Zn', 'Eu', 'Ga', 'Dy', 'Cd',
 2
               'Er', 'B', 'Hg', 'Ho', 'In', 'Na', 'Nb', 'Nd', 'Sb', 'Sc',
 3
               'As', 'Ba', 'Be', 'Bi', 'Ce', 'Cr', 'Cs', 'Ge', 'Hf', 'O',
 4
               'Ni', 'Np', 'Pa', 'Pb', 'Pr', 'Pd', 'Rb', 'Al', 'Y',
 5
               'Gd', 'Ir', 'K', 'La', 'Li', 'Lu', 'Mn', 'Mo', 'N',
 6
               'Pu', 'Re', 'Se', 'Sm', 'Sn', 'Sr', 'Ta', 'Tb', 'Tc', 'Te',
 7
               'Th', 'Tl', 'Tm', 'U', 'V', 'Am', 'Zr', 'Si', 'P']
 8
 9
10 print(len(elements))
11
12 elements = None
14 train loader, test loader, val loader = create data loaders(elements, (0.8, 0.2), seq 1
     (411319, 5, 3)
 1 best net = train(net, optimizer, train loader, test loader, batch size, seq len, measur
 2 test(best_net, test_loader)
     tensor(77) tensor(|77|)
     tensor(23) tensor([23])
```

```
tensor(67) tensor([67])
tensor(67) tensor([67])
tensor(63) tensor([63])
tensor(3) tensor([3])
tensor(56) tensor([56])
tensor(14) tensor([14])
tensor(15) tensor([15])
tensor(15) tensor([15])
tensor(38) tensor([38])
tensor(21) tensor([21])
tensor(5) tensor([5])
tensor(59) tensor([59])
tensor(76) tensor([76])
tensor(39) tensor([39])
tensor(53) tensor([53])
tensor(61) tensor([61])
tensor(24) tensor([24])
tensor(56) tensor([56])
tensor(58) tensor([58])
tensor(39) tensor([39])
tensor(41) tensor([41])
tensor(48) tensor([48])
tensor(8) tensor([8])
tensor(31) tensor([31])
tensor(41) tensor([41])
tensor(32) tensor([32])
tensor(39) tensor([39])
tensor(2) tensor([2])
tensor(72) tensor([72])
tensor(38) tensor([38])
tensor(72) tensor([72])
tensor(67) tensor([67])
tensor(29) tensor([29])
tensor(54) tensor([54])
tensor(15) tensor([15])
tensor(32) tensor([32])
tensor(36) tensor([36])
tensor(49) tensor([49])
tensor(39) tensor([39])
tensor(44) tensor([44])
tensor(26) tensor([26])
tensor(4) tensor([4])
tensor(6) tensor([6])
tensor(20) tensor([20])
tensor(49) tensor([49])
tensor(18) tensor([18])
tensor(20) tensor([20])
tensor(26) tensor([26])
tensor(61) tensor([61])
tensor(24) tensor([24])
tensor(31) tensor([31])
tensor(47) tensor([47])
tensor(12) tensor([12])
tensor(1) tensor([1])
Test accuracy: 0.9782315040670159
0.9782315040670159
```

1 test(best net, test loader)

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

1

✓ 1h 32m 6s completed at 5:16 PM