

train_LaengeNet_all

February 5, 2022

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[1]: from Neural_Nets.LaengeNet.Development.LaengeNetTorch import LaengeNet, \
      ↪LaengeNetLossFunc
from Neural_Nets.ThermoDataset.Development.ThermoDataset import ThermoDataset
from Neural_Nets.ThermoNetActFuncs.Development.ThermoNetActFuncs import \
      ↪ChenSundman, Softplus
from Utils.PlotHandler.Development.PlotHandler import PlotHandler
import torch
from torch.utils.data import DataLoader, Dataset
import torch.nn as nn
from torch.optim import Rprop
from Data_Handling.SGTEHandler.Development.SGTEHandler import SGTEHandler
import numpy as np
import matplotlib.pyplot as plt

[2]: def epoch(net: LaengeNet, dataloader, loss_func, optimizer):
      epoch_losses = np.zeros([len(dataloader), ])

      for i, (temp, g, s, h, c) in enumerate(dataloader):
          temp = temp.unsqueeze(-1)

          # Input scaling
          #temp /= temp.max()

          # Forward pass
          gibbs_energy, entropy, enthalpy, heat_cap = net(temp, temp, temp, temp, \
          ↪debug=False)

          # Output scaling
          #scale = 100000
          scale = 1
          gibbs_energy, entropy, enthalpy, heat_cap = gibbs_energy/scale, entropy/
          ↪scale, enthalpy/scale, heat_cap/scale
          g, s, h, c = g/scale, s/scale, h/scale, c/scale

          # Get the loss
          loss = loss_func(gibbs_energy.float(), g.float(), entropy.float(), s.
          ↪float(), enthalpy.float(), h.float(),
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        heat_cap.float(), c.float(), debug=False)

    # Backward pass
    net.zero_grad()
    loss.backward()
    #torch.nn.utils.clip_grad_norm_(net.parameters(), 100)
    optimizer.step()
    epoch_losses[i] = loss

mean_epoch_loss = epoch_losses.mean()
print('Mean epoch loss: ', mean_epoch_loss)
return mean_epoch_loss

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[17]: def train(net, dataset):
    # Hyperparameters
    n_epochs = 100
    lr = 0.01
    batch_size = 128
    std_thresh = 0.05
    loss_weights = [1, 0, 0, 0]

    # Data
    dataloader = DataLoader(dataset, batch_size=batch_size, shuffle=False)

    # Optimizer
    optimizer = Rprop(net.parameters(), lr=lr)
    loss_func = LaengeNetLossFunc(weights=None)

    losses = []

    # Keep track of epoch where learning rate was reduced last
    lr_reduced_last = 0

    for i in range(n_epochs):
        print('-----\nEpoch %i:\n' % i)
        loss = epoch(net, dataloader, loss_func, optimizer)
        losses.append(loss)

        # Adapt learning rate if standard deviation over the last 10 epochs is
        ↳ below a threshold
        if np.array(losses[-10:]).std() < std_thresh and (i - lr_reduced_last)
        ↳ >= 10:
            print('Learning rate halfed! \n')
            lr_reduced_last = i
            lr /= 2

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[18]: net = LaengeNet(init_args=(-0.2, -0.1))

      element = 'Fe'
      phase = ['BCC_A2']
      dataset = ThermoDataset(element, phase, scaling=False)

      train(net, dataset)
```

Fe successfully selected!

Epoch 0:

Mean epoch loss: 446430902.85714287

Epoch 1:

Mean epoch loss: 151830509.7142857

Epoch 2:

Mean epoch loss: 224236730.85714287

Epoch 3:

Mean epoch loss: 248201387.42857143

Epoch 4:

Mean epoch loss: 157879505.7142857

Epoch 5:

Mean epoch loss: 193528710.2857143

Epoch 6:

Mean epoch loss: 253607486.2857143

Epoch 7:

Mean epoch loss: 143991017.14285713

Epoch 8:

Mean epoch loss: 151755648.57142857

Epoch 9:

Mean epoch loss: 202552853.14285713

Epoch 10:

Mean epoch loss: 222032878.2857143

Epoch 11:

Mean epoch loss: 145088085.7142857

Epoch 12:

Mean epoch loss: 131331097.71428572

Epoch 13:

Mean epoch loss: 97225667.14285715

Epoch 14:

Mean epoch loss: 72244658.28571428

Epoch 15:

Mean epoch loss: 34005273.64285714

Epoch 16:

Mean epoch loss: 25026904.39285714

Epoch 17:

Mean epoch loss: nan

Epoch 18:

Mean epoch loss: nan

Epoch 19:

Mean epoch loss: nan

Epoch 20:

Mean epoch loss: nan

Epoch 21:

Mean epoch loss: nan

Epoch 22:

Mean epoch loss: nan

Epoch 23:

Mean epoch loss: nan

Epoch 24:

Mean epoch loss: nan

Epoch 25:

Mean epoch loss: nan

Epoch 26:

Mean epoch loss: nan

Epoch 27:

Mean epoch loss: nan

Epoch 28:

Mean epoch loss: nan

Epoch 29:

Mean epoch loss: nan

Epoch 30:

Mean epoch loss: nan

Epoch 31:

Mean epoch loss: nan

Epoch 32:

Mean epoch loss: nan

Epoch 33:

Mean epoch loss: nan

Epoch 34:

Mean epoch loss: nan

Epoch 35:

Mean epoch loss: nan

Epoch 36:

Mean epoch loss: nan

Epoch 37:

Mean epoch loss: nan

Epoch 38:

Mean epoch loss: nan

Epoch 39:

Mean epoch loss: nan

Epoch 40:

Mean epoch loss: nan

Epoch 41:

Mean epoch loss: nan

Epoch 42:

Mean epoch loss: nan

Epoch 43:

Mean epoch loss: nan

Epoch 44:

Mean epoch loss: nan

Epoch 45:

Mean epoch loss: nan

Epoch 46:

Mean epoch loss: nan

Epoch 47:

Mean epoch loss: nan

Epoch 48:

Mean epoch loss: nan

Epoch 49:

Mean epoch loss: nan

Epoch 50:

Mean epoch loss: nan

Epoch 51:

Mean epoch loss: nan

Epoch 52:

Mean epoch loss: nan

Epoch 53:

Mean epoch loss: nan

Epoch 54:

Mean epoch loss: nan

Epoch 55:

Mean epoch loss: nan

Epoch 56:

Mean epoch loss: nan

Epoch 57:

Mean epoch loss: nan

Epoch 58:

Mean epoch loss: nan

Epoch 59:

Mean epoch loss: nan

Epoch 60:

Mean epoch loss: nan

Epoch 61:

Mean epoch loss: nan

Epoch 62:

Mean epoch loss: nan

Epoch 63:

Mean epoch loss: nan

Epoch 64:

Mean epoch loss: nan

Epoch 65:

Mean epoch loss: nan

Epoch 66:

Mean epoch loss: nan

Epoch 67:

Mean epoch loss: nan

Epoch 68:

Mean epoch loss: nan

Epoch 69:

Mean epoch loss: nan

Epoch 70:

Mean epoch loss: nan

Epoch 71:

Mean epoch loss: nan

Epoch 72:

Mean epoch loss: nan

Epoch 73:

Mean epoch loss: nan

Epoch 74:

Mean epoch loss: nan

Epoch 75:

Mean epoch loss: nan

Epoch 76:

Mean epoch loss: nan

Epoch 77:

Mean epoch loss: nan

Epoch 78:

Mean epoch loss: nan

Epoch 79:

Mean epoch loss: nan

Epoch 80:

Mean epoch loss: nan

Epoch 81:

Mean epoch loss: nan

Epoch 82:

Mean epoch loss: nan

Epoch 83:

Mean epoch loss: nan

Epoch 84:

Mean epoch loss: nan

Epoch 85:

Mean epoch loss: nan

Epoch 86:

Mean epoch loss: nan

Epoch 87:

Mean epoch loss: nan

Epoch 88:

Mean epoch loss: nan

Epoch 89:

Mean epoch loss: nan

Epoch 90:

Mean epoch loss: nan

Epoch 91:

Mean epoch loss: nan

Epoch 92:

Mean epoch loss: nan

Epoch 93:

Mean epoch loss: nan

Epoch 94:

Mean epoch loss: nan

Epoch 95:

Mean epoch loss: nan

Epoch 96:

Mean epoch loss: nan

Epoch 97:

Mean epoch loss: nan

Epoch 98:

Mean epoch loss: nan

Epoch 99:

Mean epoch loss: nan

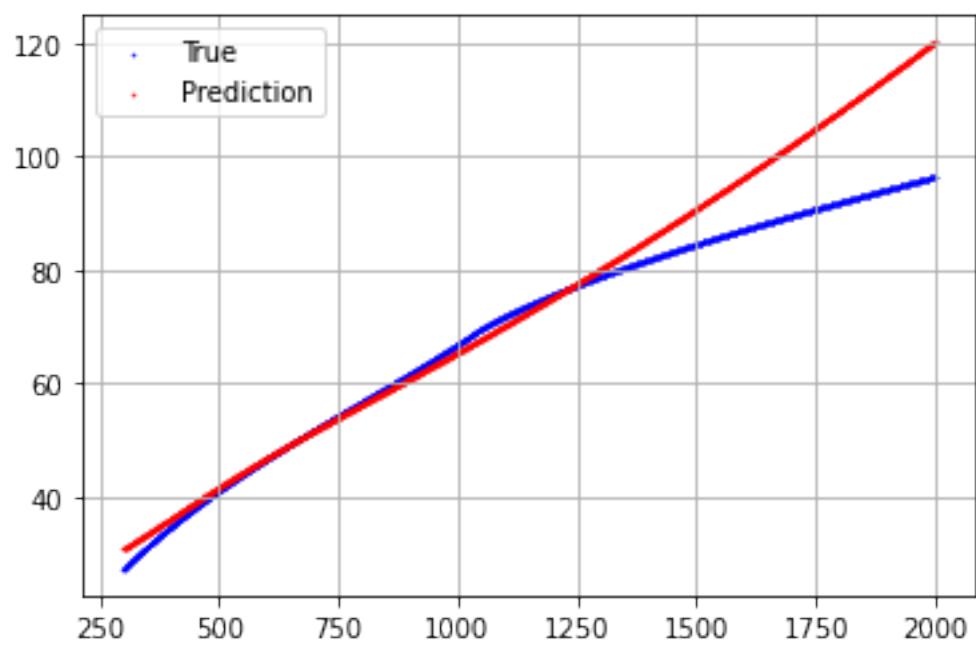
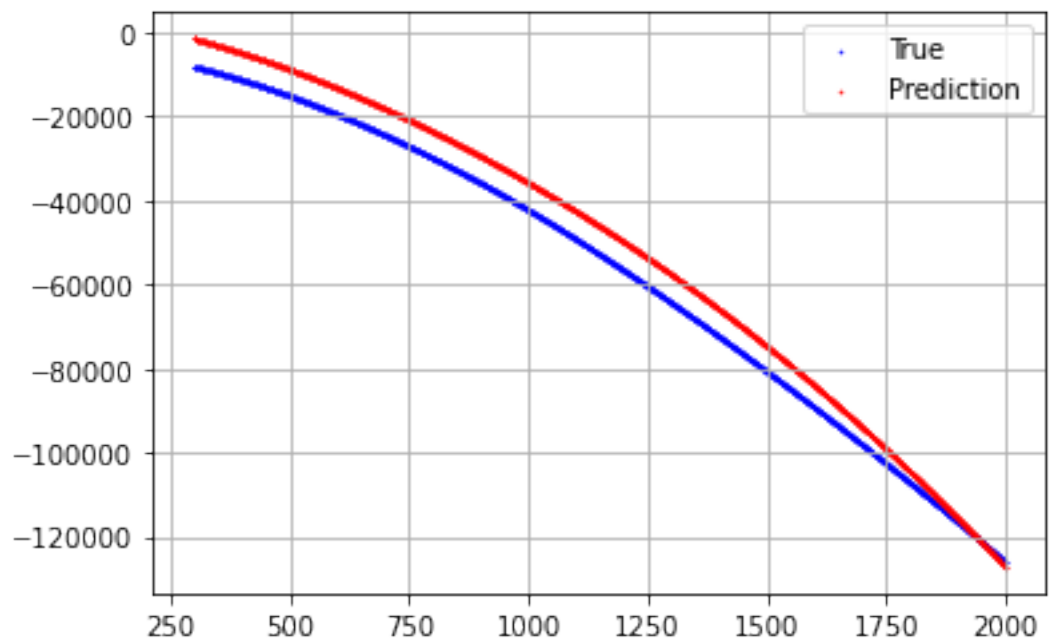
```
[19]: print('theta_E: ', net.sub_net_1.act_1.theta_E)
      print('E0: ', net.sub_net_1.act_1.E0)
      print('a: ', net.sub_net_1.act_1.a)
      print('b: ', net.sub_net_1.act_1.b)
```

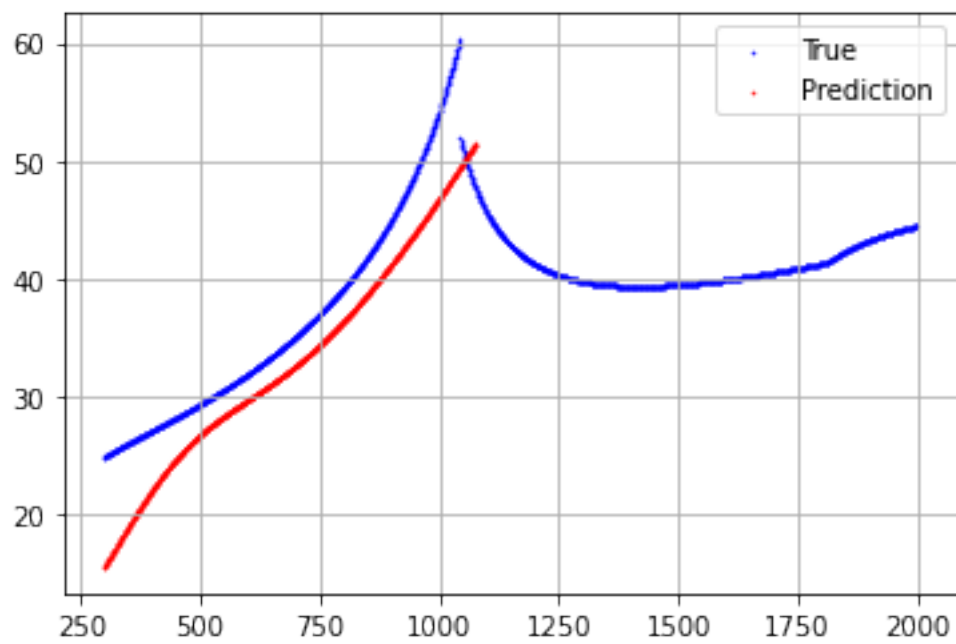
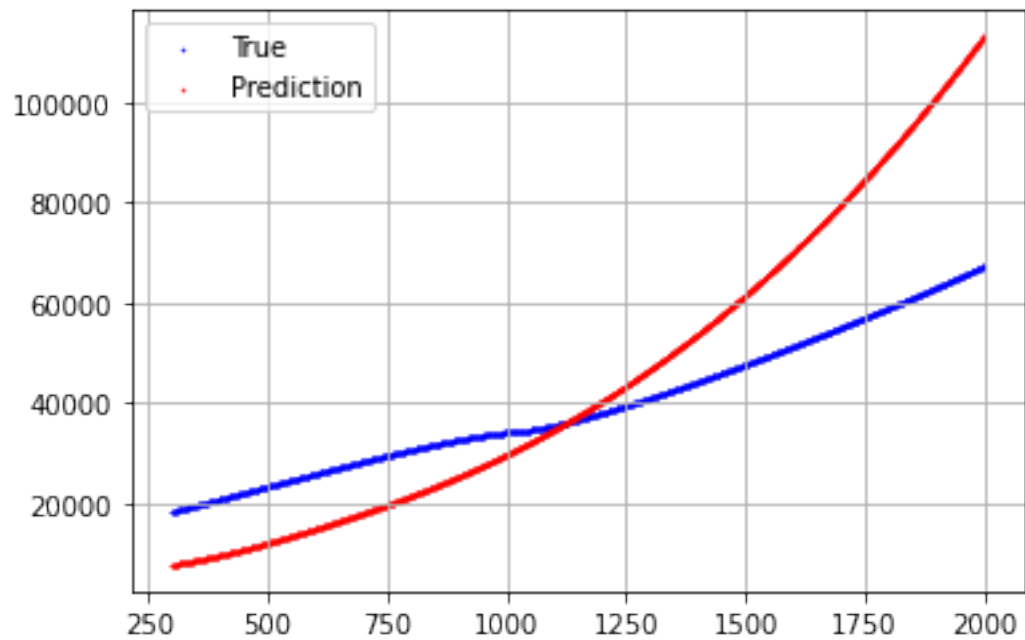
```
theta_E: Parameter containing:
tensor(-1., requires_grad=True)
E0: Parameter containing:
tensor(77104.9062, requires_grad=True)
a: Parameter containing:
tensor(-37.4346, requires_grad=True)
b: Parameter containing:
tensor(1.8301, requires_grad=True)
```

```
[20]: ph = PlotHandler('Laenge')

      ph.properties_temp(net, element, phase, scaling=False)
```

Fe successfully selected!





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[7]: #torch.save(net, 'LaengeNet/Models/model_12_01_22_1535')
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