train LaengeNetMod

February 8, 2022

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[26]: from Neural_Nets.LaengeNetModified.Development.LaengeNetModified import
       →LaengeNetModified, LaengeNetModifiedLossFunc
      from Neural_Nets.ThermoDatasetModified.Development.ThermoDatasetModified import_
      →ThermoDatasetModified
      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 *
      from Data_Handling.SGTEHandler.Development.SGTEHandler import SGTEHandler
      import numpy as np
      import matplotlib.pyplot as plt
 [2]: def epoch(net: LaengeNetModified, dataloader, loss func, optimizer):
          epoch_losses = np.zeros([len(dataloader), ])
          for i, (temp, target) in enumerate(dataloader):
              temp = temp.unsqueeze(-1)
              # Forward pass
              prediction = net(temp.float())
              # Get the loss
              loss = loss_func(prediction, target)
              # Backward pass
              net.zero grad()
              loss.backward()
              optimizer.step()
              epoch_losses[i] = loss
          mean_epoch_loss = epoch_losses.mean(axis=0)
          #print('Mean epoch loss: ', mean_epoch_loss)
          return mean_epoch_loss
[99]: def train(net, dataset):
```

Hyperparameters

```
n_{epochs} = 10000
lr = 0.001
batch\_size = 64
# Data
dataloader = DataLoader(dataset, batch_size=batch_size, shuffle=False)
# Optimizer
#optimizer = RMSprop(net.parameters(), lr=lr)
optimizer = RMSprop(net.parameters(), lr=lr)
loss_func = LaengeNetModifiedLossFunc()
losses = []
best_loss = epoch(net, dataloader, loss_func, optimizer)
best_net = net
for i in range(n_epochs):
    \#print('---- \land nEpoch \%i: \land n' \% i)
    loss = epoch(net, dataloader, loss_func, optimizer)
    losses.append(loss)
    if loss < best_loss:</pre>
        best_net = net
    if i % 10 == 0:
        print(loss)
return losses, net
```

```
[100]: net = LaengeNetModified(hidden_dim=4)

element = 'Fe'
phase = ['BCC_A2']
start_temp, end_temp = 200, 2000

dataset = ThermoDatasetModified(element, phase, step=2, start_temp=start_temp, userd_temp=end_temp)

losses, best_net = train(net, dataset)
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Fe successfully selected!

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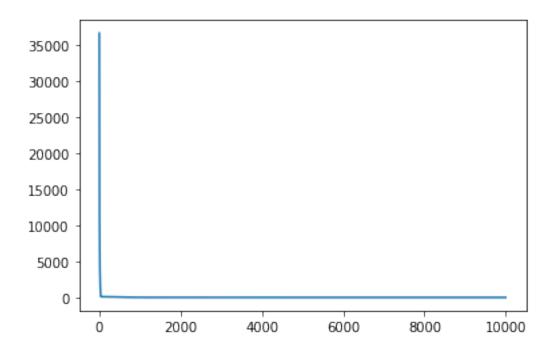
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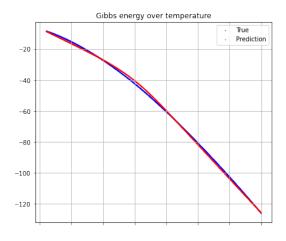
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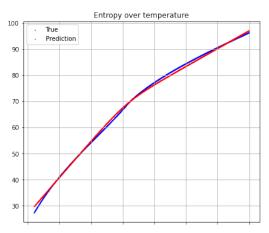
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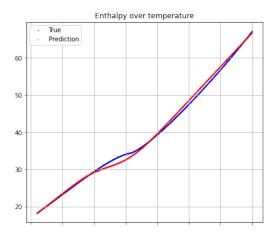
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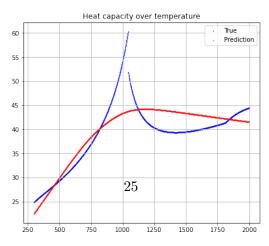


Fe successfully selected!









[77]:	#torch.save(net,	'Neural_Nets/LaengeNetModified/Models/model_07_02_22_2122')
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