

# CV\_siamese\_v5.nbconvert

July 7, 2023

## 1 run load\_data.ipynb BEFORE running this!

```
[1]: import numpy as np
import torch
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
s = {
    'problem'          : "regression",
    'approach'         : "few-shot learning",
    'method'           : "non-parametric",
    'algorithm'        : "siamese network",
    'goal'              : "learn a distribution using few samples from it",
    'input'             : "samples from a distribution",
    'input type'       : "vectors",
    'input meaning'    : "spectrum",
    'output'           : "samples from a distribution",
    'output type'      : "one number",
    'output meaning'   : "temperature or pressure, depending on distribution",
    'number of ways'   : 2,
    'number of shots'  : 1,
    'number of folds'  : 8,
    'support-query ratio': 0.8,
    'task size'        : 5,
    'learning rate'    : 5e-5,
    'feature dimension' : 16,
    'epoch'            : 1000,
    'epoch development' : 36,
    'cross validation round': 16,
    'cross validation round-development' : 3,
    'best model folder' : 'siamese_best_model/'
}
```

```
[2]: import data_accessor as acc
data_names_list = [
    'temperature_230509_discrete',
    'pressure_230516_discrete'
]
data_dictionary = acc.setup(data_names_list)
```

```
loading temperature_230509_discrete_-----
input shape (number, dimension): (6000, 10000)
label shape (number, dimension): (6000, 1)
there are 16 folds
4200 for training, 600 for validating, 1200 for testing
loading pressure_230516_discrete_-----
input shape (number, dimension): (5000, 10000)
label shape (number, dimension): (5000, 1)
there are 16 folds
3500 for training, 500 for validating, 1000 for testing
```

```
[3]: np.asarray(data_dictionary['temperature_230509_discrete']['train indices']).
      ↪shape
```

```
[3]: (16, 4200)
```

```
[4]: # task layout July 5, 2023

# siamese network extract feature space difference
# auxiliary network convert that difference into label difference
```

```
[5]: input_dimension = 10000
      output_dimension = 1
```

```
[6]: import torch.nn as nn
class SiameseNetwork(torch.nn.Module):
    def __init__(self, device, input_dimension, feature_dimension,
    ↪output_dimension):
        """ Input: input, anchor, anchor label
        Output: prediction for input"""
        super().__init__()
        self.input_dimension = input_dimension
        self.hidden_dimension = 32
        self.feature_dimension = feature_dimension
        self.output_dimension = output_dimension
        self.device = device
        self.feature_sequential = torch.nn.Sequential(
            torch.nn.Linear(self.input_dimension, self.hidden_dimension),
            nn.ReLU(),
            torch.nn.Linear(self.hidden_dimension, self.hidden_dimension),
            nn.ReLU(),
            torch.nn.Linear(self.hidden_dimension, self.feature_dimension)
        )
        self.auxiliary_sequential = torch.nn.Sequential(
            torch.nn.Linear(self.feature_dimension, self.feature_dimension),
            nn.ReLU(),
            torch.nn.Linear(self.feature_dimension, self.feature_dimension),
```

```

        nn.ReLU(),
        torch.nn.Linear(self.feature_dimension, self.output_dimension)
    )
    self.to(device)
    self.float()
    def forward(self, input, anchor, anchor_label):
        feature_input = self.feature_sequential(input)
        feature_anchor = self.feature_sequential(anchor)
        feature_space_difference_input_from_anchor = feature_input -
↪feature_anchor
        label_difference_input_from_anchor = self.
↪auxiliary_sequential(feature_space_difference_input_from_anchor)
        prediction = anchor_label + label_difference_input_from_anchor
        return prediction

```

```

[7]: import matplotlib.pyplot as plt
def plot_loss(train_loss, valid_loss):
    plt.subplot()
    plt.plot(train_loss)
    plt.plot(valid_loss)
    plt.legend(["train loss", "valid loss"], loc="upper right")
    plt.show()

```

```

[8]: from tools import SaveBestModel, PatienceEarlyStopping, Scheduler
class Manager:
    """ DOES: train & evaluate a Siamese network """
    def __init__(self, epoch, cross_validation_round):
        self._network = SiameseNetwork(device, 10000, 16, 1)
        self._network.apply(self.initializer)
        self._learning_rate = 1e-4
        self._optimizer = torch.optim.Adam(
            params=self._network.parameters(), lr=self._learning_rate,
            weight_decay=3e-3)
        self._energy = nn.MSELoss()
        self._train_loss = []
        self._valid_loss = []
        self._test_loss = []
        self._epoch = epoch
        self._stopper = PatienceEarlyStopping(patience=10, min_delta=1e-7)
        self._cross_validation_round = cross_validation_round
        self._saver = SaveBestModel(s['best model folder'])
        self._scheduler = Scheduler(optimizer=self._optimizer,
            minimum_learning_rate=1e-6,
            patience=5,
            factor=0.5)
    def initializer(self, layer):

```

```

        if type(layer) == nn.Linear:
            nn.init.kaiming_normal_(layer.weight) # normal version
def _step(self, job):
    input, input_label, anchor, anchor_label = job
    # print(f"input dtype is {input_1.dtype}")
    prediction = self._network(input, anchor, anchor_label)
    loss = self._energy(input_label, prediction)
    return loss
def train(self, train_dataloader, valid_dataloader):
    """ DOES: calculate loss from tasks
        NOTE: we have a BATCH of tasks here """
    for e in range(self._epoch):
        # print(f"train() epoch {e}")
        batch_train_loss = []
        for _, batch in enumerate(train_dataloader):
            loss = self._step(batch)
            loss.backward()
            self._optimizer.step()
            batch_train_loss.append(loss.item())
        self._train_loss.append(np.mean(batch_train_loss))
        batch_valid_loss = []
        with torch.no_grad():
            for _, batch in enumerate(valid_dataloader):
                loss = self._step(batch)
                batch_valid_loss.append(loss.item())
            self._valid_loss.append(np.mean(batch_valid_loss))
            # saving, early stopping, scheduler for EACH epoch!
            self._saver(current_loss=np.mean(batch_valid_loss),
                        model=self._network,
                        round=self._cross_validation_round
                        )
            self._scheduler(np.mean(batch_valid_loss))
            self._stopper(np.mean(batch_valid_loss))
            if self._stopper.early_stop == True:
                print(f"EARLY STOPPING @ epoch {e}")
                break
    # summary printout, after we're done with epochs
    print(f"min train loss: {np.min(self._train_loss)}")
    print(f"min valid loss: {np.min(self._valid_loss)}")
    plot_loss(self._train_loss, self._valid_loss)
    return np.min(self._valid_loss)
def test(self, test_dataloader):
    with torch.no_grad():
        batch_test_loss = []
        with torch.no_grad():
            for _, batch in enumerate(test_dataloader):
                loss = self._step(batch)

```

```

        batch_test_loss.append(loss.item())
    self._test_loss.append(np.mean(batch_test_loss))
    return np.min(self._test_loss)

```

```

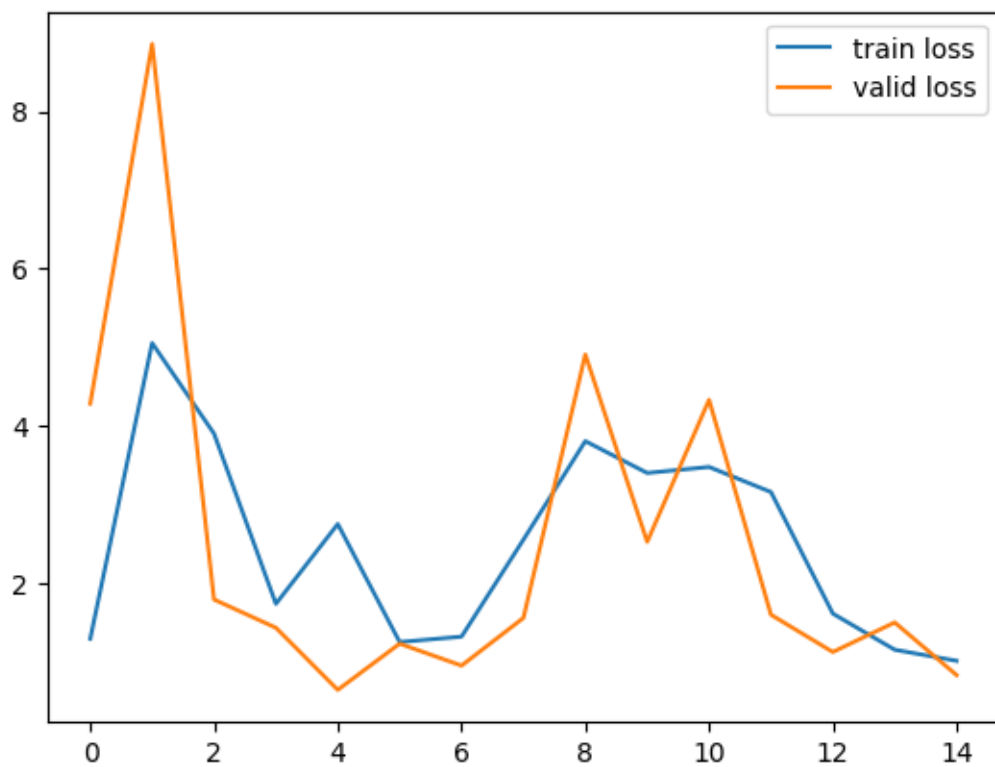
[9]: from torch.utils.data import DataLoader
from tools import SiameseDataset, SaveBestCrossValidationModel
CV_saver = SaveBestCrossValidationModel(s['best model folder'])
test_indices = data_dictionary['temperature_230509_discrete']['test indices']
epoch = s['epoch']
for cross_validation_round, (train, valid) in enumerate(zip(
    data_dictionary['temperature_230509_discrete']['train indices'],
    data_dictionary['temperature_230509_discrete']['valid indices'])):
    if cross_validation_round < s['cross validation round']:
        print(f"CV round {cross_validation_round}")
        network_object = Manager(epoch, cross_validation_round)
        valid_loss = network_object.train(
            DataLoader(SiameseDataset(
                data_dictionary['temperature_230509_discrete']['data'],
                data_dictionary['temperature_230509_discrete']['label'],
                data_dictionary['temperature_230509_discrete']['train_
↳indices'][cross_validation_round],
                device=device,), shuffle=False, batch_size=32),
            DataLoader(SiameseDataset(
                data_dictionary['temperature_230509_discrete']['data'],
                data_dictionary['temperature_230509_discrete']['label'],
                data_dictionary['temperature_230509_discrete']['valid_
↳indices'][cross_validation_round],
                device=device,), shuffle=False, batch_size=32))
        CV_saver(current_loss=valid_loss, round=cross_validation_round)
    print(f"\nbest model is: {CV_saver.best_model_name} with {CV_saver.
↳current_best_loss}")

```

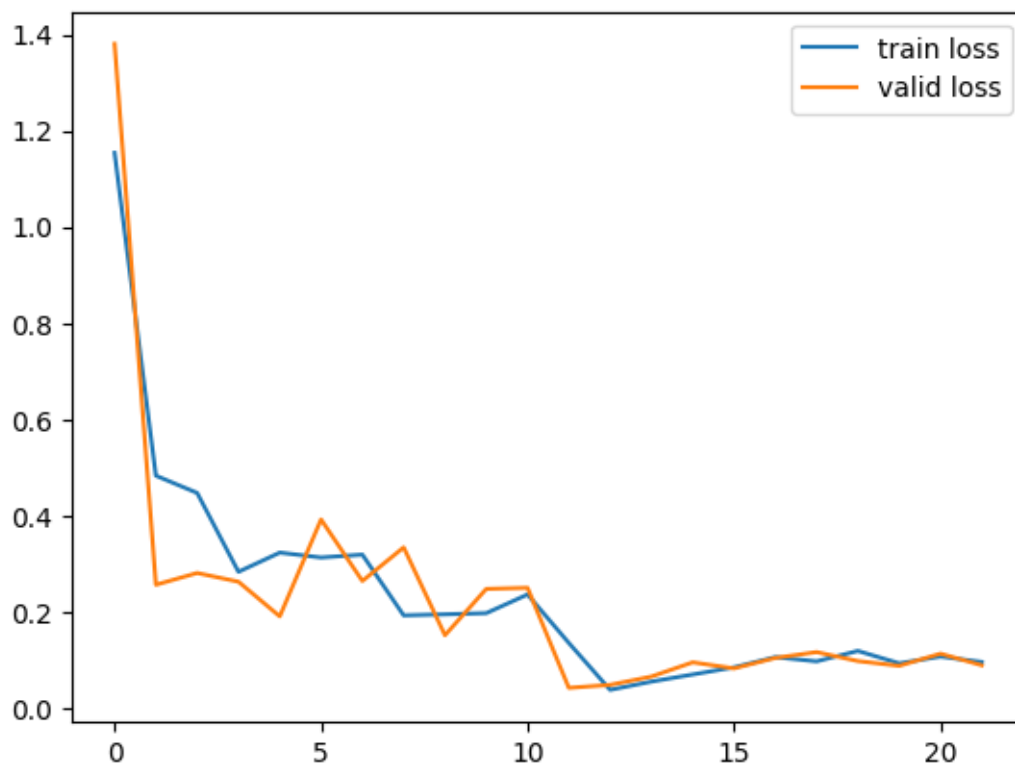
```

CV round 0
EARLY STOPPING @ epoch 14
min train loss: 1.0147766145792874
min valid loss: 0.645540072729713

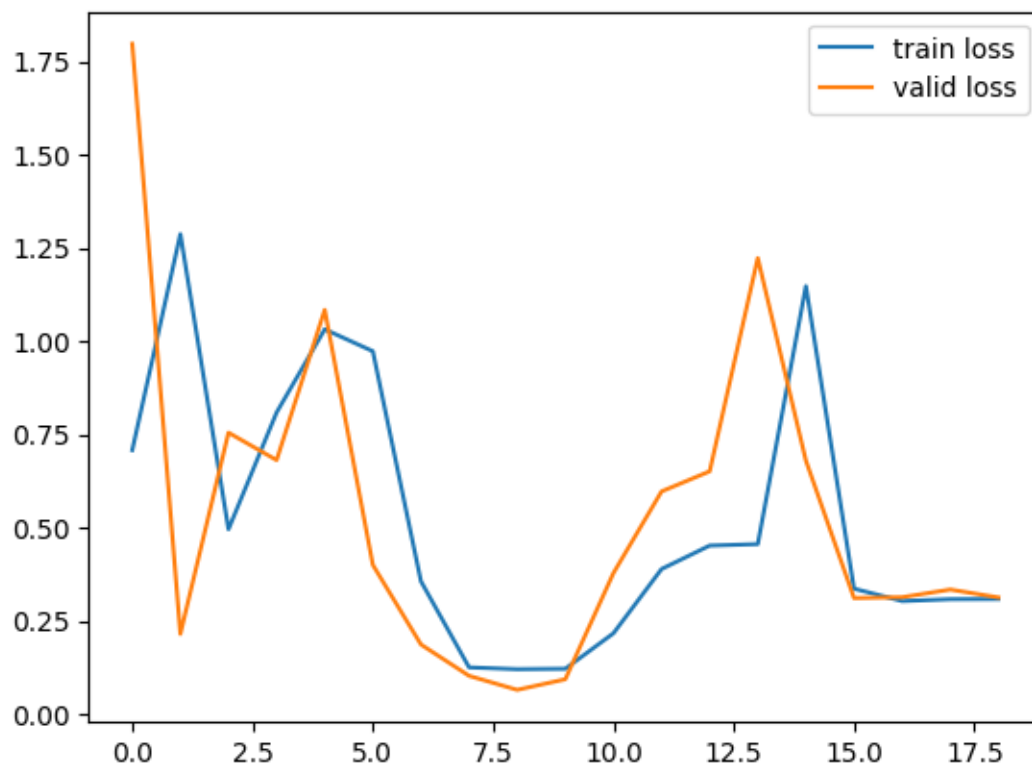
```



CV round 1  
EARLY STOPPING @ epoch 21  
min train loss: 0.038293136340199097  
min valid loss: 0.0424345610173125

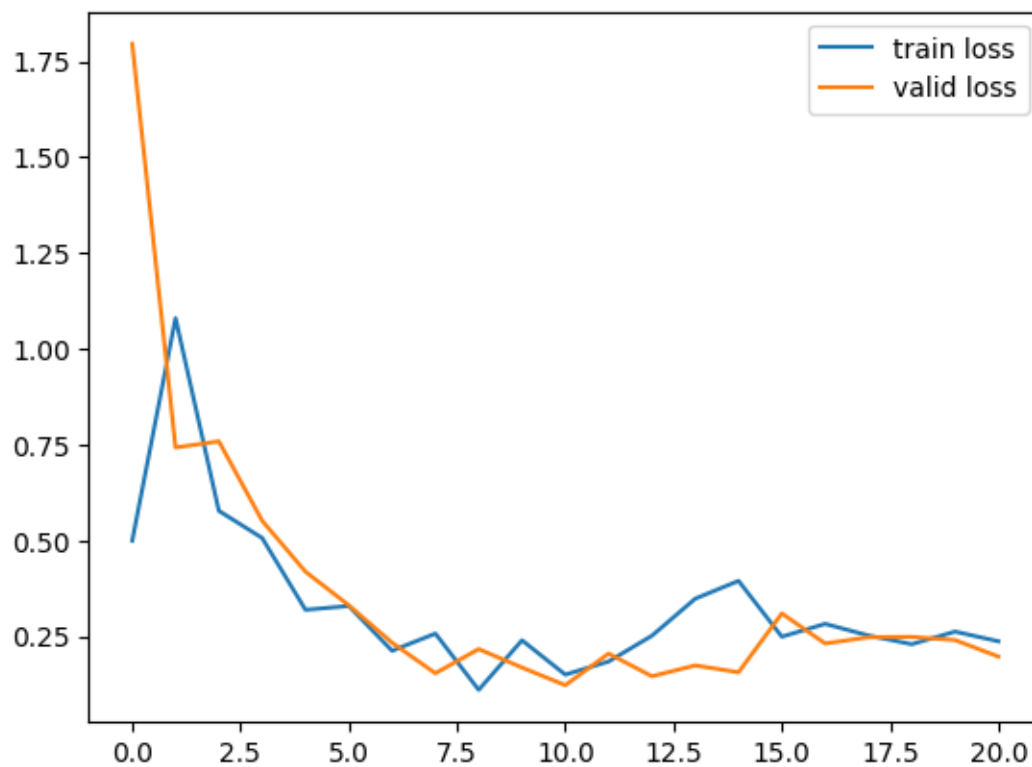


CV round 2  
EARLY STOPPING @ epoch 18  
min train loss: 0.12034989410842006  
min valid loss: 0.06495706521366772

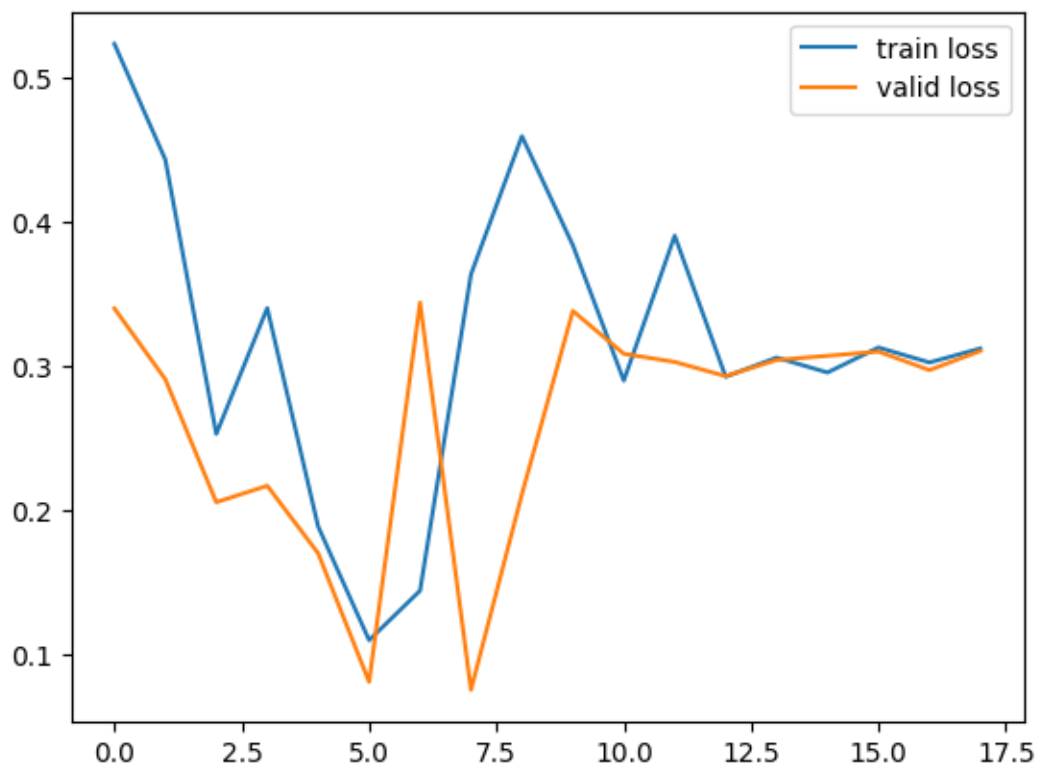


CV round 3  
EARLY STOPPING @ epoch 20  
min train loss: 0.1100511385946337  
min valid loss: 0.12221350207140572





CV round 4  
EARLY STOPPING @ epoch 17  
min train loss: 0.10957405377518047  
min valid loss: 0.075070479198506

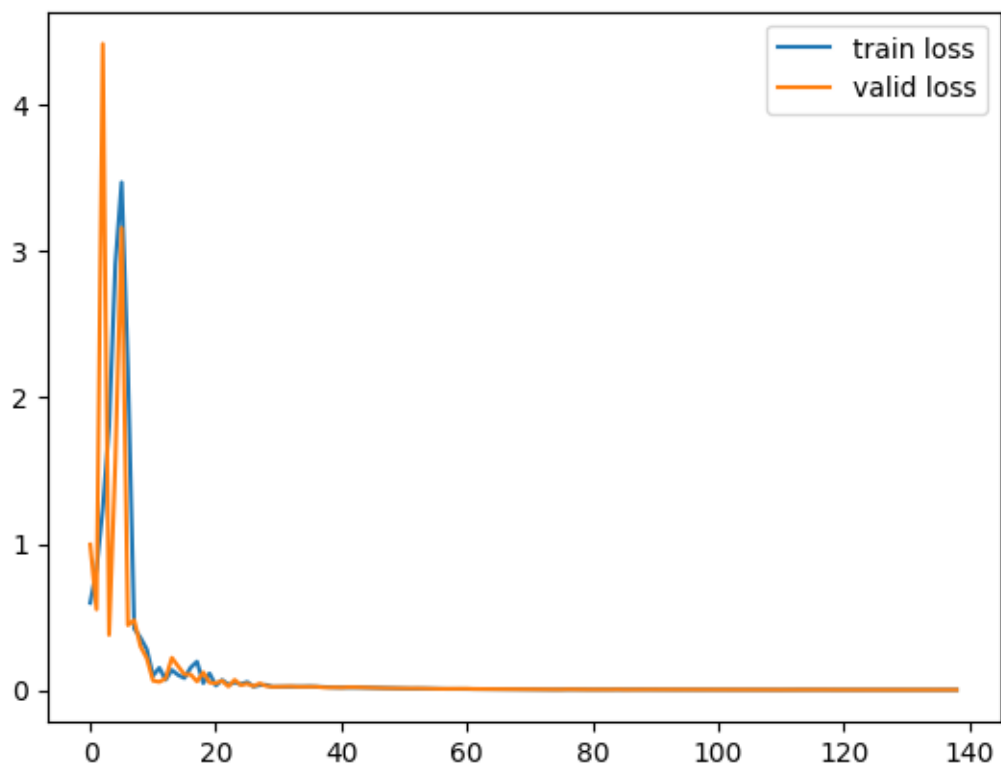


CV round 5

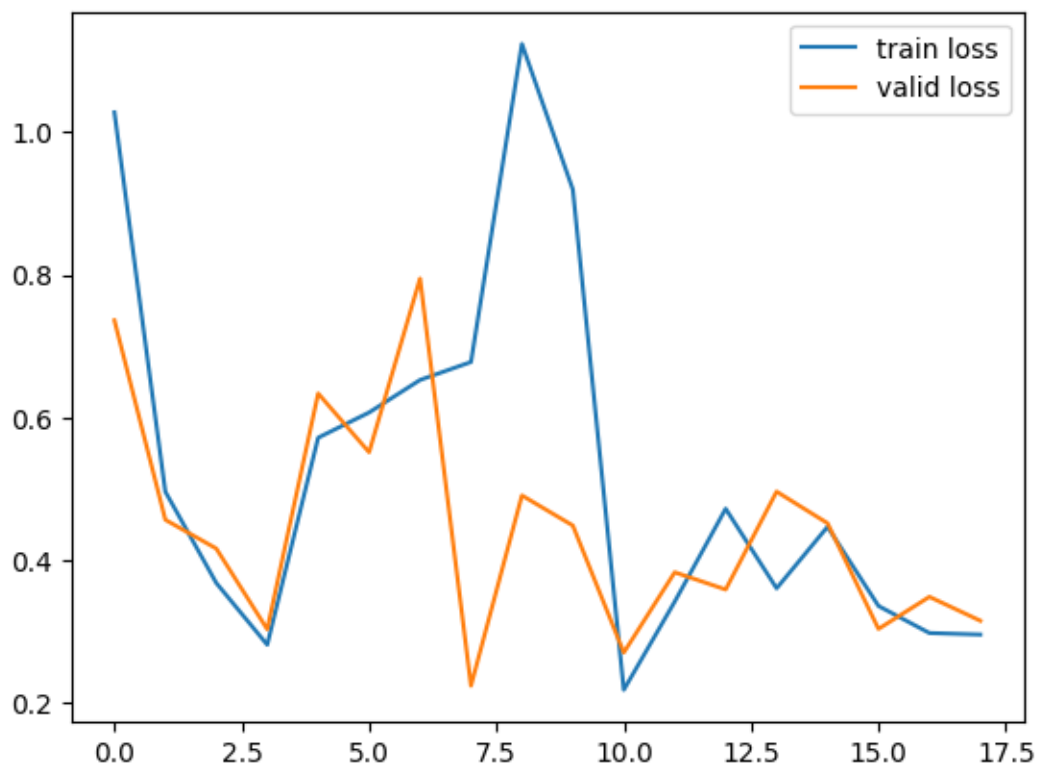
EARLY STOPPING @ epoch 138

min train loss: 0.0017095510578876588

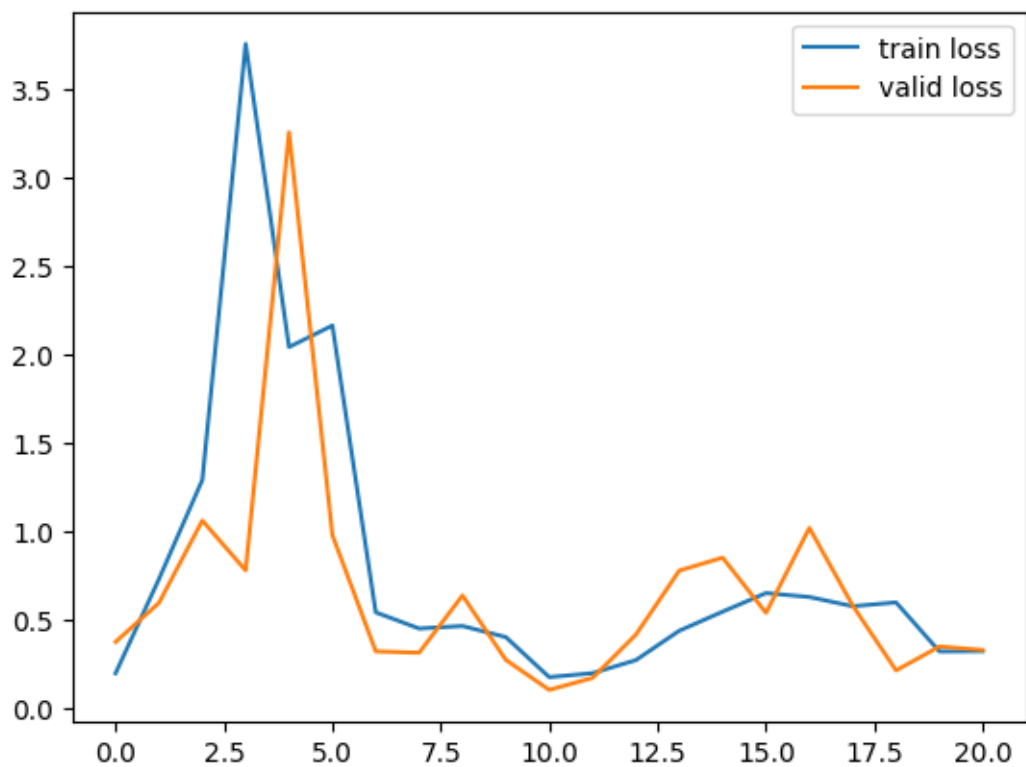
min valid loss: 0.0015852679529129283



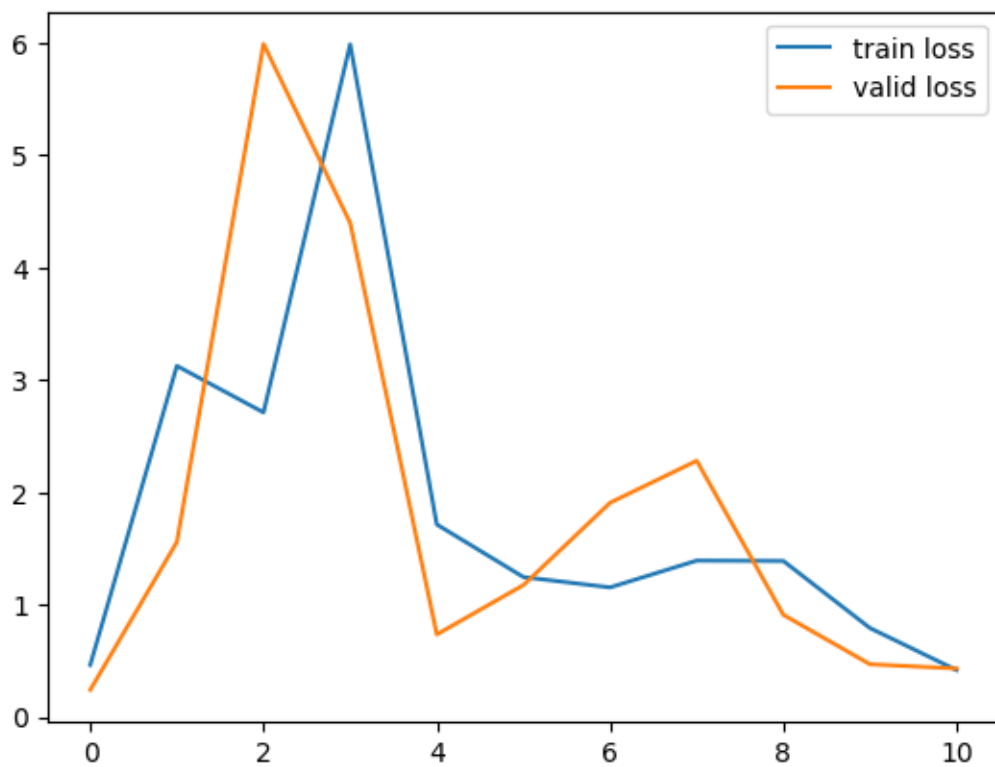
CV round 6  
EARLY STOPPING @ epoch 17  
min train loss: 0.2177621406700575  
min valid loss: 0.2234966284350345



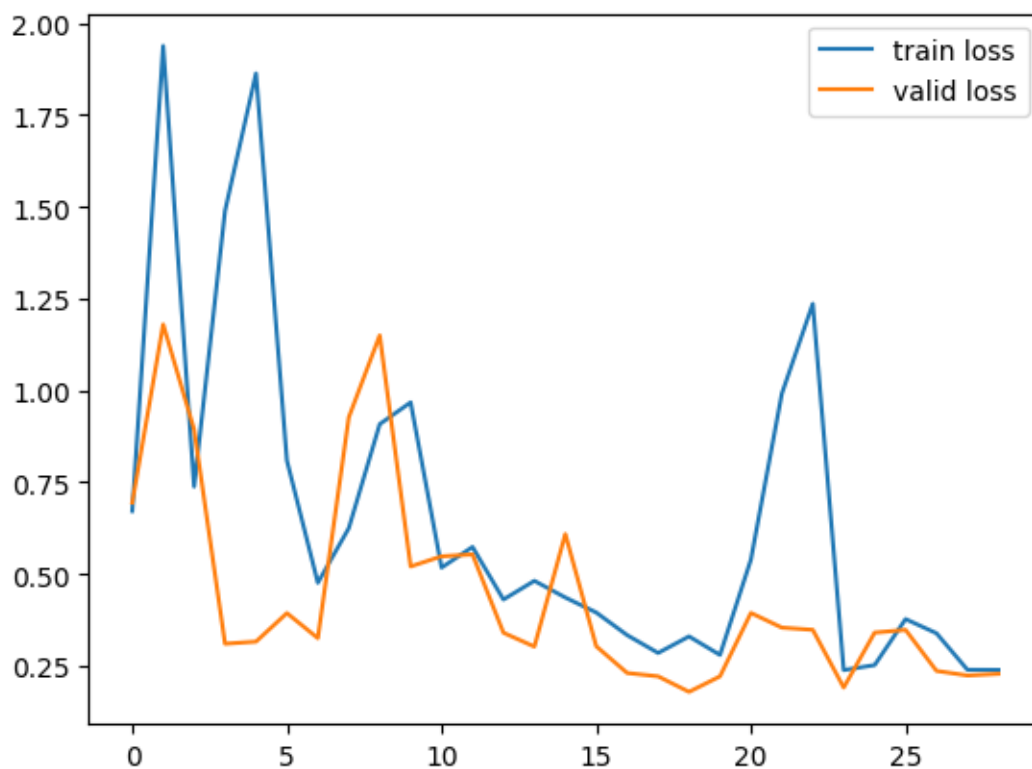
CV round 7  
EARLY STOPPING @ epoch 20  
min train loss: 0.1737676223344875  
min valid loss: 0.10159156745985935



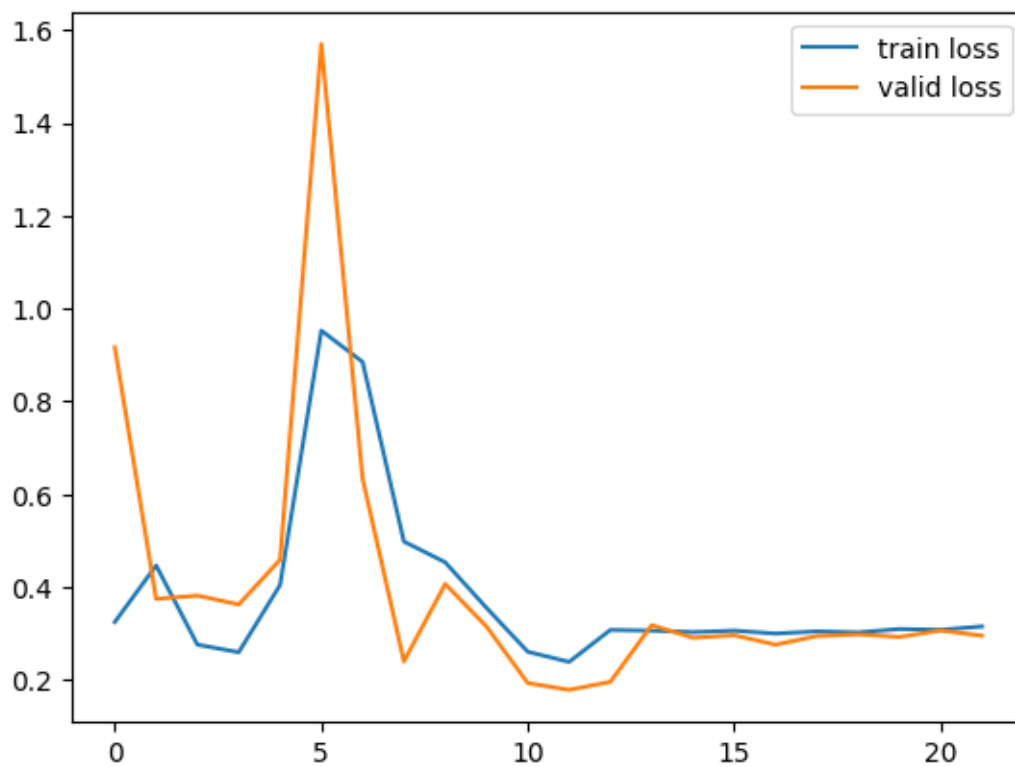
CV round 8  
EARLY STOPPING @ epoch 10  
min train loss: 0.41822355443781073  
min valid loss: 0.2427428756889544



CV round 9  
EARLY STOPPING @ epoch 28  
min train loss: 0.2389822488820011  
min valid loss: 0.17969518311713872

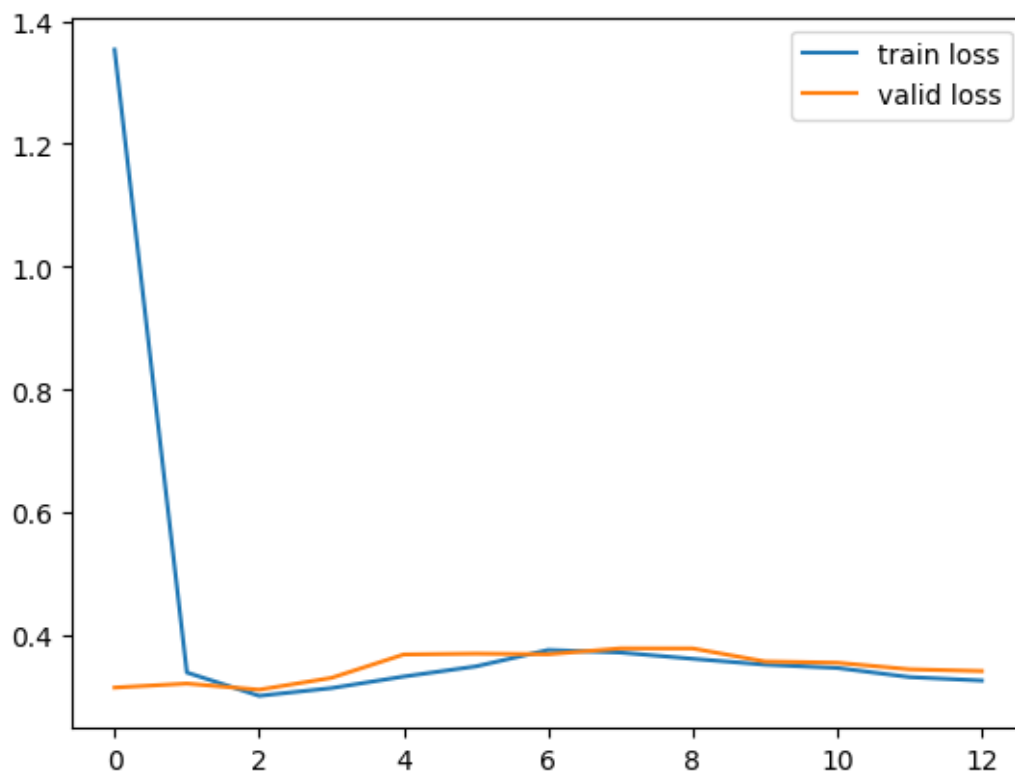


CV round 10  
EARLY STOPPING @ epoch 21  
min train loss: 0.2392647930731376  
min valid loss: 0.17908812706407748

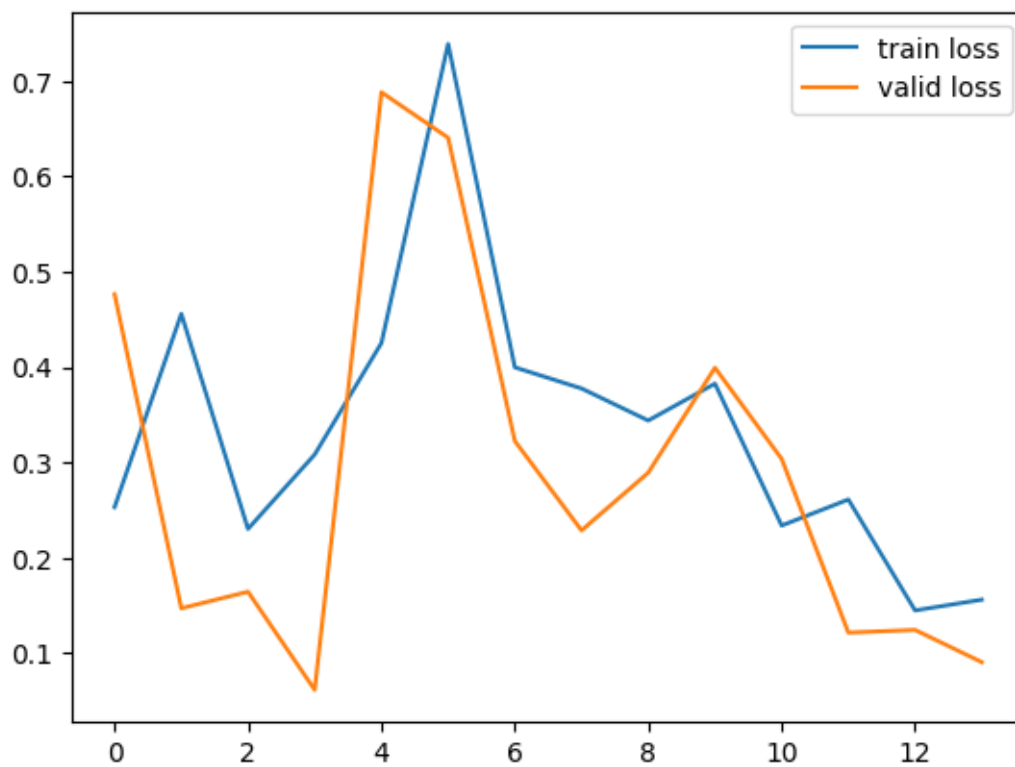


CV round 11  
EARLY STOPPING @ epoch 12  
min train loss: 0.30046337210770807  
min valid loss: 0.31045905931999807

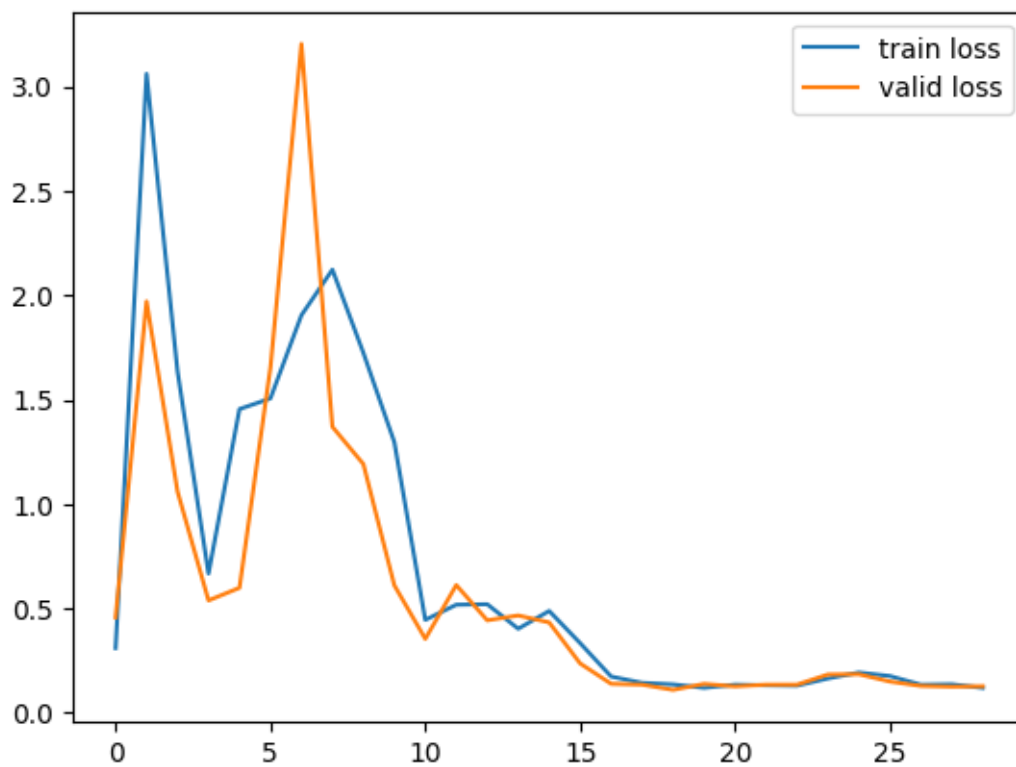




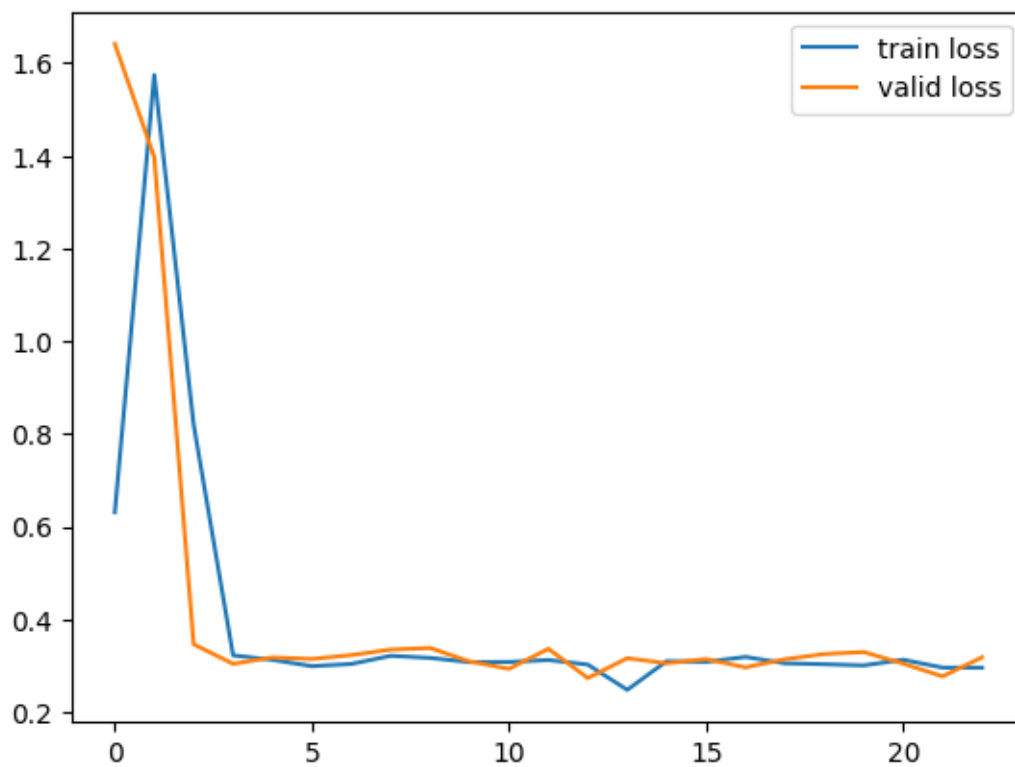
CV round 12  
EARLY STOPPING @ epoch 13  
min train loss: 0.14465605919108246  
min valid loss: 0.06134305659093355



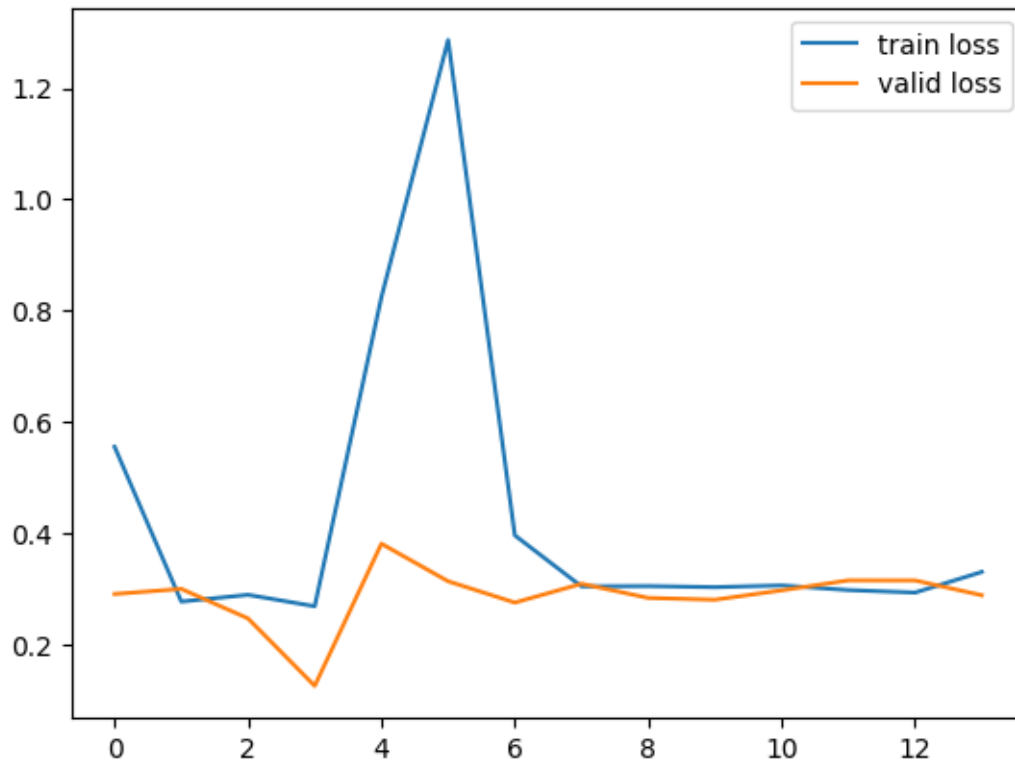
CV round 13  
EARLY STOPPING @ epoch 28  
min train loss: 0.12012304269680471  
min valid loss: 0.11064708977937698



CV round 14  
EARLY STOPPING @ epoch 22  
min train loss: 0.2480857027976802  
min valid loss: 0.2735394984483719



CV round 15  
EARLY STOPPING @ epoch 13  
min train loss: 0.26818944937126205  
min valid loss: 0.12498640582749718



best model is: CV=5.pth with 0.0015852679529129283

```
[10]: network_object._network.load_state_dict(torch.load(s['best model folder'] +
    ↪CV_saver.best_model_name))
test_loss = network_object.test(
    DataLoader(SiameseDataset(
        data_dictionary['temperature_230509_discrete']['data'],
        data_dictionary['temperature_230509_discrete']['label'],
        data_dictionary['temperature_230509_discrete']['test indices'],
        device=device,), shuffle=False, batch_size=32))
print(f"testing loss: {test_loss}")
```

testing loss: 0.0018470123744153074

```
[11]: # test_model = SiameseNetwork(device, 10000, 16, 1)
# test_model.load_state_dict(torch.load(s['best model folder'] + CV_saver.
    ↪best_model_name))
# test_network_object = Manager(1, 1)
# test_network_object._network = test_model
# test_loss = test_network_object.test(
#     DataLoader(SiameseDataset(
```

```
#         data_dictionary['temperature_230509_discrete']['data'],
#         data_dictionary['temperature_230509_discrete']['label'],
#         data_dictionary['temperature_230509_discrete']['test_indices'],
#         device=device,), shuffle=False, batch_size=32))
# print(f"testing loss: {test_loss}")
```