

```
In [ ]: import itertools

from tqdm import tqdm
import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split

import torch
import torch.nn as nn
```

Question 1

(a)

```
In [ ]: import pickle
from torch.utils.data import Dataset, DataLoader

class MnistDataset():
    def __init__(self, X, y, transform_X=lambda x: x):
        assert X.shape[0] == y.shape[0]
        self.X = transform_X(X)
        self.y = y

    def __len__(self):
        return len(self.y)

    def __getitem__(self, idx):
        return self.X[idx], self.y[idx]

def load_mnist(path):
    with open(path, 'rb') as f:
        train_data, test_data = pickle.load(f)

    X_train = torch.tensor(train_data[0], dtype=torch.float).unsqueeze(1)
    y_train = torch.tensor(train_data[1], dtype=torch.long)
    X_test = torch.tensor(test_data[0], dtype=torch.float).unsqueeze(1)
    y_test = torch.tensor(test_data[1], dtype=torch.long)
    return X_train, y_train, X_test, y_test
```

```
In [ ]: # load data and normalize by dividing the maximum value
X_train, y_train, X_test, y_test = load_mnist('mnist.pkl')
X_train /= torch.max(X_train)
X_test /= torch.max(X_test)
```

```
In [ ]: # further split the train to train/validation with 80/20
X_further_train, X_val, y_further_train, y_val = train_test_split(X_train, y
# use MnistDataset class to handle the data
```

```
train_data = MnistDataset(X_further_train, y_further_train)
val_data = MnistDataset(X_val, y_val)
test_data = MnistDataset(X_test, y_test)
```

(b)

```
In [ ]: class VAE(nn.Module):
    def __init__(self, in_channels=1, z_dim=32):
        super().__init__()
        self.encoder = nn.Sequential(
            # conv1, input_channel -> 4
            nn.Conv2d(in_channels, 4, kernel_size=4, padding=1, stride=2),
            nn.ReLU(), # relu
            # conv2, channel 4 -> 8
            nn.Conv2d(4, 8, kernel_size=4, padding=1, stride=2),
            nn.ReLU(), # relu
            nn.Conv2d(8, 16, kernel_size=4, padding=1, stride=2), # conv3,
            nn.ReLU(), # relu
            nn.Conv2d(16, 32, kernel_size=4, padding=1, stride=2), # conv4,
            nn.ReLU(), # relu
            nn.Flatten(), # flatten
        )

        # manually calculate the dimension after all convolutions
        dim_after_conv = 2
        hidden_dim = 32 * dim_after_conv * dim_after_conv

        # Readout layer is mu
        self.readout_mu = nn.Linear(hidden_dim, z_dim)
        # Readout layer
        self.readout_sigma = nn.Linear(hidden_dim, z_dim)

        # You can use nn.ConvTranspose2d to decode
        self.decoder = nn.Sequential(
            nn.Linear(z_dim, hidden_dim),
            nn.Unflatten(1, (32, dim_after_conv, dim_after_conv)),
            nn.ConvTranspose2d(32, 16, kernel_size=4, stride=2, padding=1),
            nn.ReLU(), # relu
            nn.ConvTranspose2d(16, 8, kernel_size=4, stride=2, padding=1), #
            nn.ReLU(), # relu
            nn.ConvTranspose2d(8, 4, kernel_size=4, stride=2, padding=1), #
            nn.ReLU(), # relu
            nn.ConvTranspose2d(4, in_channels, kernel_size=4, stride=2, padding=1),
            nn.Sigmoid(), # use a sigmoid activation to squeeze the outputs
        )

    def reparameterize(self, mu, sigma):
        """
        Reparameterize, i.e. generate a  $z \sim N(\mu, \sigma)$ 
        """
        # generate epsilon ~ N(0, I)
        # hint: use torch.randn or torch.randn_like
        epsilon = torch.randn_like(sigma)
        #  $z = \mu + \sigma * \epsilon$ 
```

```

        z = mu + sigma * epsilon
        return z

    def encode(self, x):
        # call the encoder to map input to a hidden state vector
        h = self.encoder(x)
        # use the "readout" layer to get \mu and \sigma
        mu = self.readout_mu(h)
        sigma = self.readout_sigma(h)
        return mu, sigma

    def decode(self, z):
        # call the decoder to map z back to x
        return self.decoder(z)

    def forward(self, x):
        mu, sigma = self.encode(x)
        z = self.reparameterize(mu, sigma)
        x_recon = self.decode(z)
        return x_recon, mu, sigma

```

(c)

For debugging: The `test_kld_loss_func` should output 1.3863

```

In [ ]: def kld_loss_func(mu, sigma):
        """
        KL-Divergence:  $KLD = 0.5 * \sum(\mu^2 + \sigma^2 - \ln(\sigma^2) - 1)$ 

        Parameters
        -----
        mu: torch.Tensor
            Mean vector in the VAE bottleneck region
        sigma: torch.Tensor
            Standard Deviation vector in the VAE bottleneck region

        Return
        -----
        kld: torch.Tensor
            KL-Divergence loss (a scalar)
        """
        return 0.5 * torch.sum(mu.pow(2) + sigma.pow(2) - torch.log(sigma.pow(2)))

    def vae_loss_func(recon_x, x, mu, sigma):

        bce_loss = nn.BCELoss(reduction='sum')(recon_x, x)
        kld_loss = kld_loss_func(mu, sigma)

        return bce_loss + kld_loss

    def test_kld_loss_func():
        mu = torch.tensor([0.5, 0.5, 1.0])

```

```
sigma = torch.tensor([1.0, 0.5, 0.5])
print(kld_loss_func(mu, sigma))

test_kld_loss_func()
```

tensor(1.3863)

```
In [ ]: import torch.nn.functional as F

class VAETrainer:

    def __init__(self, model, learning_rate, batch_size, epoch, l2):
        self.model = model
        num_params = sum(item.numel() for item in model.parameters())
        print(f"{model.__class__.__name__} - Number of parameters: {num_params}")

        self.optimizer = torch.optim.Adam(model.parameters(), learning_rate,

        self.epoch = epoch
        self.batch_size = batch_size

    def train(self, train_data, val_data, early_stop=True, verbose=True, device='cpu',
              train_loader = DataLoader(train_data, batch_size=self.batch_size, sh

        train_loss_list = []
        val_loss_list = []

        weights = self.model.state_dict()
        lowest_val_loss = np.inf

        for n in tqdm(range(self.epoch), leave=False):
            self.model.train()
            epoch_loss = 0.0
            for X_batch, y_batch in train_loader:
                batch_importance = 1 / len(train_data)
                # call the model
                X_batch_recon, mu, sigma = self.model(X_batch)
                batch_loss = vae_loss_func(X_batch_recon, X_batch, mu, sigma)

                self.optimizer.zero_grad()
                batch_loss.backward()
                self.optimizer.step()

                epoch_loss += batch_loss.detach().cpu().item() * batch_importance

            train_loss_list.append(epoch_loss)

            val_loss = self.evaluate(val_data, print_loss=False)
            val_loss_list.append(val_loss)

            if early_stop:
                if val_loss < lowest_val_loss:
                    lowest_val_loss = val_loss
                    weights = self.model.state_dict()
```

```

    if draw_curve:
        x_axis = np.arange(self.epoch)
        fig, ax = plt.subplots(1, 1, figsize=(5, 4))
        ax.plot(x_axis, train_loss_list, label="Train")
        ax.plot(x_axis, val_loss_list, label="Validation")
        ax.set_title("Total Loss")
        ax.set_xlabel("# Epoch")

    if early_stop:
        self.model.load_state_dict(weights)

    return {
        "train_loss_list": train_loss_list,
        "val_loss_list": val_loss_list,
    }

def evaluate(self, data, print_loss=True):
    self.model.eval()
    loader = DataLoader(data, batch_size=self.batch_size)
    total_loss = 0.0
    for X_batch, y_batch in loader:
        with torch.no_grad():
            batch_importance = 1 / len(data)
            X_batch_recon, mu, sigma = self.model(X_batch)
            batch_loss = vae_loss_func(X_batch_recon, X_batch, mu, sigma)
            total_loss += batch_loss.detach().cpu().item() * batch_importance
    if print_loss:
        print(f"Total Loss: {total_loss}")
    return total_loss

```

```

In [ ]: train_loader = DataLoader(train_data, batch_size=128, shuffle=True)
        for X_batch, y_batch in train_loader:
            print(X_batch.shape)

```

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

```
In [ ]: vae(x)
```

```

Out[ ]: (tensor([[[[0.5465, 0.5466, 0.5461, ..., 0.5467, 0.5459, 0.5462],
                    [0.5431, 0.5441, 0.5418, ..., 0.5430, 0.5411, 0.5471],
                    [0.5462, 0.5461, 0.5440, ..., 0.5463, 0.5431, 0.5472],
                    ...,
                    [0.5431, 0.5445, 0.5424, ..., 0.5439, 0.5422, 0.5468],
                    [0.5462, 0.5462, 0.5440, ..., 0.5469, 0.5421, 0.5468],
                    [0.5446, 0.5432, 0.5440, ..., 0.5433, 0.5442, 0.5450]]],

               [[ [0.5465, 0.5465, 0.5461, ..., 0.5467, 0.5459, 0.5462],
                   [0.5431, 0.5441, 0.5418, ..., 0.5430, 0.5411, 0.5471],
                   [0.5462, 0.5462, 0.5440, ..., 0.5463, 0.5431, 0.5472],
                   ...,
                   [0.5431, 0.5445, 0.5423, ..., 0.5439, 0.5422, 0.5468],
                   [0.5462, 0.5462, 0.5440, ..., 0.5469, 0.5421, 0.5468],
                   [0.5446, 0.5432, 0.5440, ..., 0.5434, 0.5442, 0.5450]]],

               [[ [0.5465, 0.5465, 0.5461, ..., 0.5467, 0.5459, 0.5462],
                   [0.5430, 0.5442, 0.5418, ..., 0.5430, 0.5411, 0.5471],
                   [0.5462, 0.5462, 0.5440, ..., 0.5463, 0.5431, 0.5472],
                   ...,
                   [0.5431, 0.5446, 0.5424, ..., 0.5439, 0.5422, 0.5468],
                   [0.5462, 0.5462, 0.5440, ..., 0.5469, 0.5421, 0.5468],
                   [0.5446, 0.5432, 0.5440, ..., 0.5433, 0.5442, 0.5450]]],

               ...,

               [[ [0.5465, 0.5465, 0.5461, ..., 0.5467, 0.5459, 0.5462],
                   [0.5430, 0.5442, 0.5418, ..., 0.5430, 0.5411, 0.5471],
                   [0.5462, 0.5462, 0.5440, ..., 0.5463, 0.5431, 0.5472],
                   ...,
                   [0.5431, 0.5445, 0.5424, ..., 0.5439, 0.5422, 0.5468],
                   [0.5462, 0.5462, 0.5439, ..., 0.5469, 0.5421, 0.5468],
                   [0.5446, 0.5432, 0.5440, ..., 0.5433, 0.5442, 0.5450]]],

               [[ [0.5465, 0.5466, 0.5461, ..., 0.5467, 0.5459, 0.5462],
                   [0.5430, 0.5442, 0.5418, ..., 0.5430, 0.5411, 0.5471],
                   [0.5462, 0.5462, 0.5440, ..., 0.5463, 0.5430, 0.5472],
                   ...,
                   [0.5431, 0.5445, 0.5423, ..., 0.5438, 0.5422, 0.5468],
                   [0.5462, 0.5462, 0.5439, ..., 0.5469, 0.5420, 0.5468],
                   [0.5446, 0.5432, 0.5440, ..., 0.5434, 0.5442, 0.5450]]],

               [[ [0.5465, 0.5465, 0.5461, ..., 0.5467, 0.5459, 0.5462],
                   [0.5430, 0.5442, 0.5418, ..., 0.5430, 0.5411, 0.5471],
                   [0.5462, 0.5462, 0.5440, ..., 0.5463, 0.5431, 0.5472],
                   ...,
                   [0.5431, 0.5445, 0.5424, ..., 0.5439, 0.5422, 0.5468],
                   [0.5462, 0.5462, 0.5440, ..., 0.5469, 0.5421, 0.5468],
                   [0.5446, 0.5432, 0.5440, ..., 0.5434, 0.5442, 0.5450]]]]],
        grad_fn=<SigmoidBackward0>),

```

```

tensor([[-0.0063,  0.0882,  0.0155, ..., -0.0328,  0.0110, -0.0878],
        [-0.0067,  0.0876,  0.0164, ..., -0.0327,  0.0111, -0.0903],
        [-0.0060,  0.0866,  0.0179, ..., -0.0334,  0.0124, -0.0885],
        ...,
        [-0.0091,  0.0878,  0.0183, ..., -0.0336,  0.0114, -0.0815],
        [-0.0089,  0.0875,  0.0205, ..., -0.0341,  0.0110, -0.0853],
        [-0.0068,  0.0873,  0.0178, ..., -0.0346,  0.0101, -0.0884]],
       grad_fn=<AddmmBackward0>),
tensor([ [ 0.0068,  0.0253,  0.0311, ...,  0.0323,  0.0757, -0.0555],
        [ 0.0057,  0.0251,  0.0289, ...,  0.0334,  0.0748, -0.0554],
        [ 0.0035,  0.0267,  0.0292, ...,  0.0345,  0.0760, -0.0548],
        ...,
        [ 0.0020,  0.0261,  0.0306, ...,  0.0363,  0.0740, -0.0537],
        [ 0.0047,  0.0252,  0.0276, ...,  0.0314,  0.0759, -0.0552],
        [ 0.0063,  0.0269,  0.0267, ...,  0.0357,  0.0750, -0.0543]],
       grad_fn=<AddmmBackward0>))

```

```

In [ ]: vae = VAE()
        trainer = VAETrainer(vae, 1e-3, 128, 10, 1e-5)
        # train
        trainer.train(train_data, val_data)

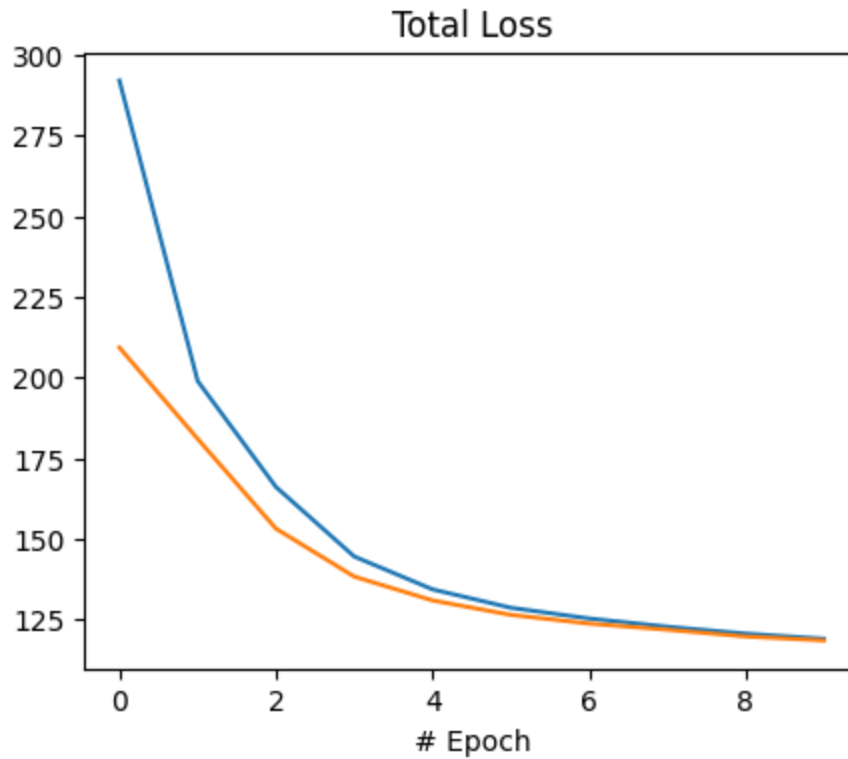
```

VAE – Number of parameters: 34201

```

Out[ ]: {'train_loss_list': [292.1656652832029,
                             198.88523193359373,
                             166.01494095865885,
                             144.44174377441396,
                             134.2405961710612,
                             128.53765653483072,
                             125.20061944580074,
                             122.6587432657878,
                             120.50614562988284,
                             118.90336954752603],
         'val_loss_list': [209.27612076822913,
                            180.93946402994783,
                            153.07922412109372,
                            138.20236848958334,
                            130.80785856119792,
                            126.32972094726566,
                            123.6293977050781,
                            121.72244881184905,
                            119.62093229166669,
                            118.39482324218758]}

```

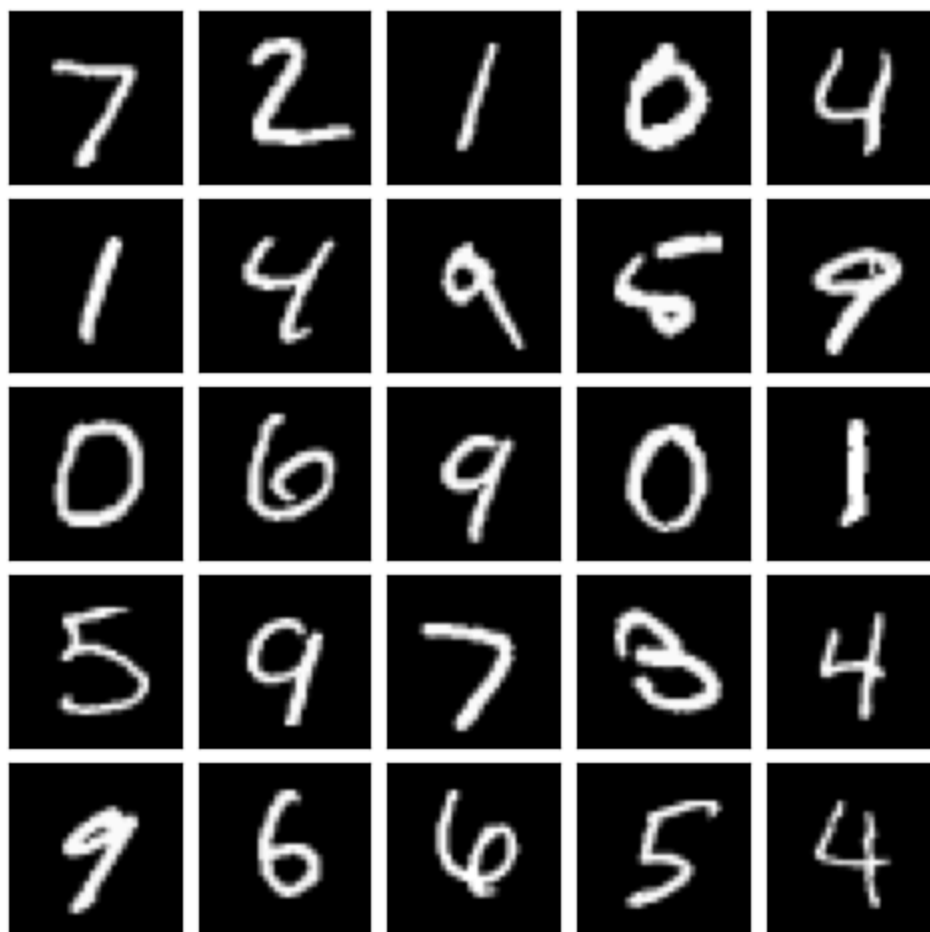


```
In [ ]: # evaluate the quality of reconstruction
def plot_digits(data, title):
    fig, axes = plt.subplots(5, 5, figsize=(6, 6))
    fig.subplots_adjust(hspace=0.1, wspace=0.1)
    fig.suptitle(title)
    for i, ax in enumerate(axes.flatten()):
        im = ax.imshow(data[i].reshape(32, 32), cmap='gray')
        ax.set_xticks([])
        ax.set_yticks([])

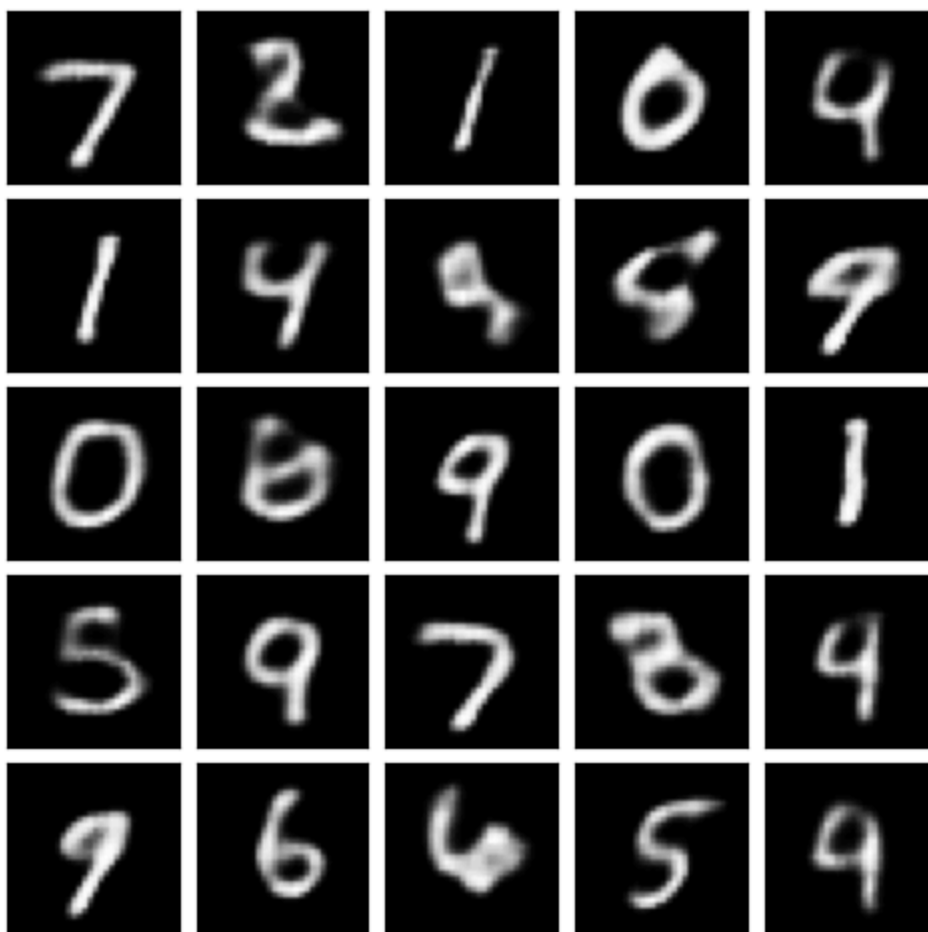
def compare_reconstruct(model, X):
    plot_digits(X, "Original Data")
    with torch.no_grad():
        X_recon, _, _ = model(X)
    plot_digits(X_recon, "Reconstructed Data")

compare_reconstruct(trainer.model, X_test[:100])
```

Original Data



Reconstructed Data



Reconstruction worked well! If I had more time to train on additional epochs, then reconstruction could potentially look even better!

Question 2

```
In [ ]: from torch_geometric.datasets import QM9
        from torch_geometric.loader import DataLoader as GraphDataLoader
        from torch_geometric.utils import scatter
```

(a)

```
In [ ]: def load_qm9(path="./QM9"):
        def transform(data):
            edge_index = torch.tensor(
                list(itertools.permutations(range(data.x.shape[0]), 2)),
                dtype=torch.long
            ).T
            edge_feature = 1 / torch.sqrt(
                torch.sum(
```

```

        (data.pos[edge_index[0]] - data.pos[edge_index[1]]) ** 2,
        axis=1, keepdim=True
    )
    data.edge_index = edge_index
    data.edge_attr = edge_feature
    data.y = data.y[:, [-7]]
    return data

qm9 = QM9(path, transform=transform)
return qm9

qm9 = load_qm9("./QM9")
train_data, test_data = train_test_split(qm9, train_size=0.8, test_size=0.2)

```

In []: `qm9[0].x.shape`

Out[]: `torch.Size([5, 11])`

In []: `qm9[0].edge_attr.shape`

Out[]: `torch.Size([20, 1])`

In []: *# find out the dimension of node input features*
`node_input_dim = 11`
`edge_input_dim = 1`

(b)

```

In [ ]: class Layer(nn.Module):
    """
    Basic layer, a linear layer with a ReLU activation
    """
    def __init__(self, in_dim, out_dim):
        super().__init__()
        self.layers = nn.Sequential(
            nn.Linear(in_dim, out_dim), # linear layer
            nn.ReLU() # relu
        )

    def forward(self, x):
        return self.layers(x)

class MessagePassingLayer(nn.Module):
    """
    A message passing layer that updates nodes/edge features
    """
    def __init__(self, node_hidden_dim, edge_hidden_dim):
        super().__init__()
        # figure out the input/output dimension
        self.edge_net = Layer(2*node_hidden_dim + edge_hidden_dim, edge_hidden_dim)
        # figure out the input/output dimension

```

```

        self.node_net = Layer(node_hidden_dim + edge_hidden_dim, node_hidden_dim)

    def forward(self, node_features, edge_features, edge_index):
        """
        Update node and edge features

        Parameters
        -----
        node_features: torch.Tensor
            Node features from the previous layer
        edge_features: torch.Tensor
            Edge features from the previous layer
        edge_index: torch.Tensor
            A sparse matrix (n_edge, 2) in which each column denotes node id
        """
        # concatenate previous edge features with node features forming the edge features
        # hint: use edge_features[edge_index[0](or 1)] to get node features
        concatenate_edge_features = torch.cat([
            node_features[edge_index[0]], # features of one node
            node_features[edge_index[1]], # features of the other node
            edge_features # previous edge features
        ], dim=1)

        # pass through the "edge_net" to map it back to the original dimension
        updated_edge_features = self.edge_net(concatenate_edge_features)

        # use scatter to aggregate the edge features to nodes
        aggr_edge_features = scatter(updated_edge_features, edge_index[0])
        # concatenate it with previous node features
        concatenate_node_features = torch.cat([aggr_edge_features, node_features])
        # pass through the "node_net" to map it back to the original dimension
        updated_node_features = self.node_net(concatenate_node_features)

        return updated_node_features, updated_edge_features

class GraphNet(nn.Module):
    def __init__(self, node_input_dim, edge_input_dim, node_hidden_dim, edge_hidden_dim):
        super().__init__()
        # embed the input node features
        self.node_embed = Layer(node_input_dim, node_hidden_dim)
        # embed the input edge features
        self.edge_embed = Layer(edge_input_dim, edge_hidden_dim)
        # message passing layer
        self.message_passing = MessagePassingLayer(node_hidden_dim, edge_hidden_dim)
        # use a linear layer as readout to get the "atomic" energy contribution
        self.readout = Layer(node_hidden_dim, 1)

    def forward(self, node_features, edge_features, edge_index, batch):
        """
        Update node and edge features

        Parameters
        -----

```

```

node_features: torch.Tensor
    Node features from the previous layer
edge_features: torch.Tensor
    Edge features from the previous layer
edge_index: torch.Tensor
    A sparse matrix (n_edges, 2) in which each column denotes node i and j
batch: torch.Tensor
    A 1-D tensor (n_nodes,) that tells you each node belongs to which batch
"""

node_hidden = self.node_embed(node_features) # call the node embedding layer
edge_hidden = self.edge_embed(edge_features) # call the edge embedding layer
updated_node_hidden, updated_edge_hidden = self.message_passing(node_hidden, edge_hidden, edge_index)
readout = self.readout(updated_node_hidden) # use the readout layer
out = scatter(readout, batch) # use the scatter function to aggregate
return out

```

```

In [ ]: class GNNTrainer:
    def __init__(self, model, batch_size, learning_rate, epoch, l2):
        self.model = model

        num_params = sum(item.numel() for item in model.parameters())
        print(f"{model.__class__.__name__} - Number of parameters: {num_params}")

        self.batch_size = batch_size
        self.optimizer = torch.optim.Adam(model.parameters(), learning_rate, l2)
        self.epoch = epoch

    def train(self, dataset, draw_curve=True):
        self.model.train()
        loader = GraphDataLoader(dataset, batch_size=self.batch_size, shuffle=True)

        loss_func = nn.MSELoss()
        batch_loss_list = []
        for i in range(self.epoch):
            print(f"Epoch: {i}")
            for batch_data in tqdm(loader, leave=False):
                batch_size = batch_data.y.shape[0]
                batch_pred = self.model(batch_data.x, batch_data.edge_attr, batch_data.y)
                batch_loss = loss_func(batch_pred, batch_data.y)

                self.optimizer.zero_grad()
                batch_loss.backward()
                self.optimizer.step()

            batch_loss_list.append(batch_loss.detach().numpy())

        if draw_curve:
            fig, ax = plt.subplots(1, 1, figsize=(5, 4), constrained_layout=True)
            ax.set_yscale("log")
            ax.plot(np.arange(len(batch_loss_list)), batch_loss_list)
            ax.set_xlabel("# Batch")
            ax.set_ylabel("Loss")

        return batch_loss_list

    def evaluate(self, dataset, draw_curve=True):

```

```

self.model.eval()
loader = GraphDataLoader(dataset, batch_size=self.batch_size)
y_true, y_pred = [], []
with torch.no_grad():
    for batch_data in tqdm(loader, leave=False):
        batch_pred = self.model(batch_data.x, batch_data.edge_attr)
        y_pred.append(batch_pred.detach().numpy().flatten())
        y_true.append(batch_data.y.detach().numpy().flatten())

y_true = np.concatenate(y_true)
y_pred = np.concatenate(y_pred)
mse = np.mean((y_true - y_pred) ** 2)

if draw_curve:
    fig, ax = plt.subplots(1, 1, figsize=(5, 4), constrained_layout=True)
    ax.scatter(y_true, y_pred, label=f"MSE: {mse:.2f}", s=2)
    ax.set_xlabel("Ground Truth")
    ax.set_ylabel("Predicted")
    xmin, xmax = ax.get_xlim()
    ymin, ymax = ax.get_ylim()
    vmin, vmax = min(xmin, ymin), max(xmax, ymax)
    ax.set_xlim(vmin, vmax)
    ax.set_ylim(vmin, vmax)
    ax.plot([vmin, vmax], [vmin, vmax], color='red')
    ax.legend()

return mse

```

(c)

```

In [ ]: node_hidden_dim = 64
        edge_hidden_dim = 64

        net = GraphNet(node_input_dim, edge_input_dim, node_hidden_dim, edge_hidden_

In [ ]: loader = GraphDataLoader(test_data, batch_size=512)
        loader_iter = iter(loader)

In [ ]: batch_data = next(loader_iter)
        batch_data.y

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batch_pred)
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```

```
In [ ]: loss_func = nn.MSELoss()  
        loss_func(batch_pred, batch_data.y)
```

```
Out [ ]: tensor(9634.5879, grad_fn=<MseLossBackward0>)
```

```
In [ ]: # 1e-3, 128, 10, 1e-5  
  
# train  
learning_rate = 1e-3  
n_epoch = 3  
batch_size = 512  
l2 = 1e-5  
  
trainer = GNNTrainer(net, batch_size, learning_rate, n_epoch, l2)  
trainer.train(train_data)
```

GraphNet – Number of parameters: 21569

Epoch: 0

Epoch: 1

Epoch: 2

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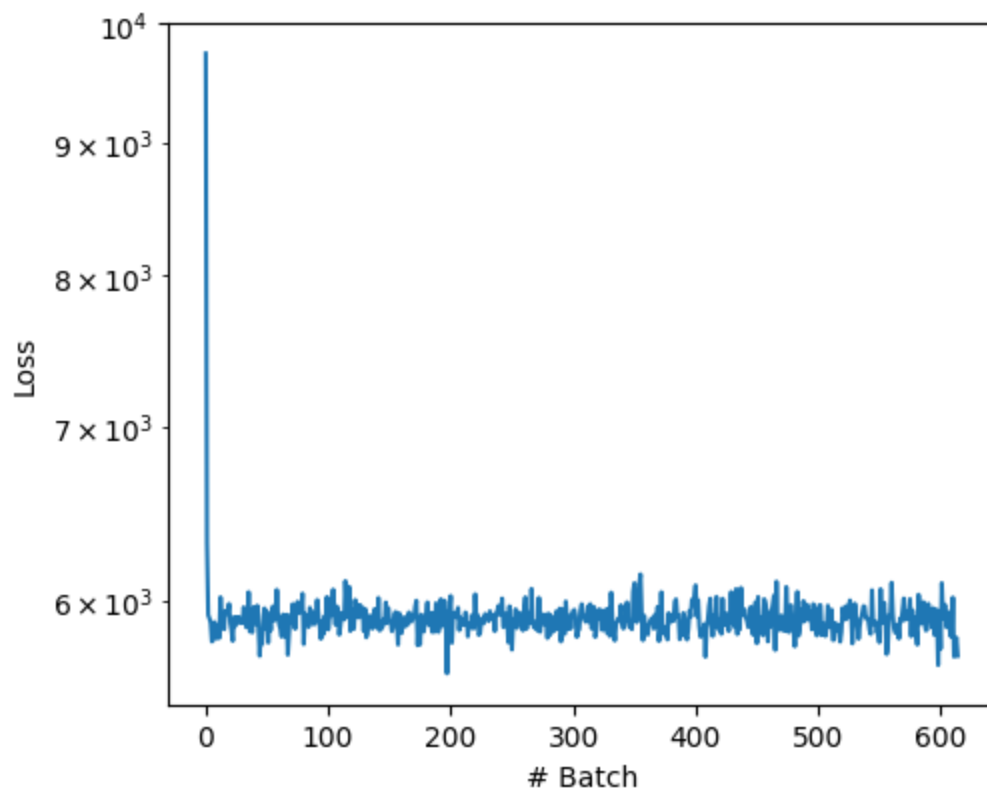
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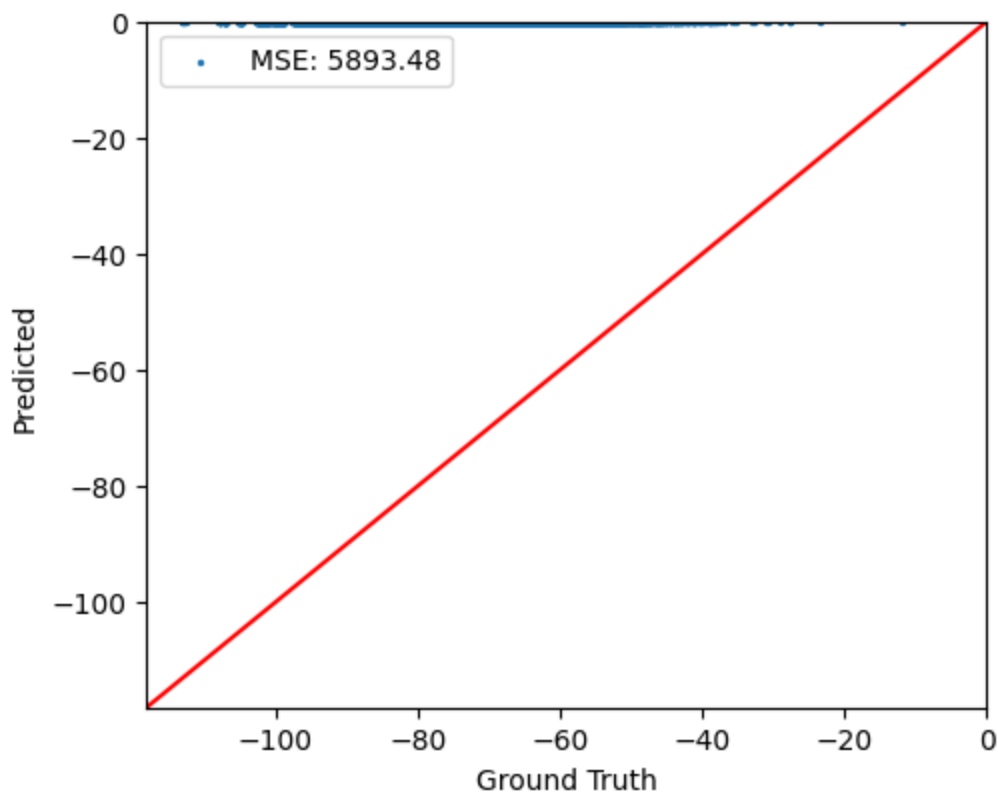
```
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array(5815.5244, dtype=float32),
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array(5908.2197, dtype=float32),
array(5952.8193, dtype=float32),
array(5795.3584, dtype=float32),
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array(5883.5635, dtype=float32),
array(5862.265, dtype=float32),
array(5844.225, dtype=float32),
array(5873.0645, dtype=float32),
array(5786.312, dtype=float32),
array(5836.0254, dtype=float32),
array(5961.745, dtype=float32),
array(5934.4927, dtype=float32),
array(5900.5527, dtype=float32),
array(5971.455, dtype=float32),
array(5909.792, dtype=float32),
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array(5777.41, dtype=float32),
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array(5900.6777, dtype=float32),
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array(5868.77, dtype=float32),
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array(5971.515, dtype=float32),
array(5968.745, dtype=float32),
array(5908.792, dtype=float32),
array(5909.5884, dtype=float32),
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array(5826.178, dtype=float32),
array(6055.9873, dtype=float32),
array(5878.679, dtype=float32),
array(5883.8677, dtype=float32),
array(5889.755, dtype=float32),
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array(5723.286, dtype=float32),
array(5776.2144, dtype=float32),
array(5924.982, dtype=float32),
array(5913.584, dtype=float32),
```

```
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array(5855.2666, dtype=float32),
array(5832.375, dtype=float32),
array(5837.949, dtype=float32),
array(5875.1904, dtype=float32),
array(5802.514, dtype=float32),
array(5907.1357, dtype=float32),
array(5890.121, dtype=float32),
array(5956.7847, dtype=float32),
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array(5949.674, dtype=float32),
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array(5771.5215, dtype=float32),
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array(5810.804, dtype=float32),
array(6017.117, dtype=float32),
array(5710.3623, dtype=float32),
array(5783.8438, dtype=float32),
array(5806.9355, dtype=float32),
array(5714.5273, dtype=float32)]
```



```
In [ ]: # evaluate
trainer.evaluate(test_data)
```

Out[]: 5893.482



It appears that the MSE is quite large, as we are always predicting a slightly positive enthalpy of formation, versus the true enthalpy of formations being negative.