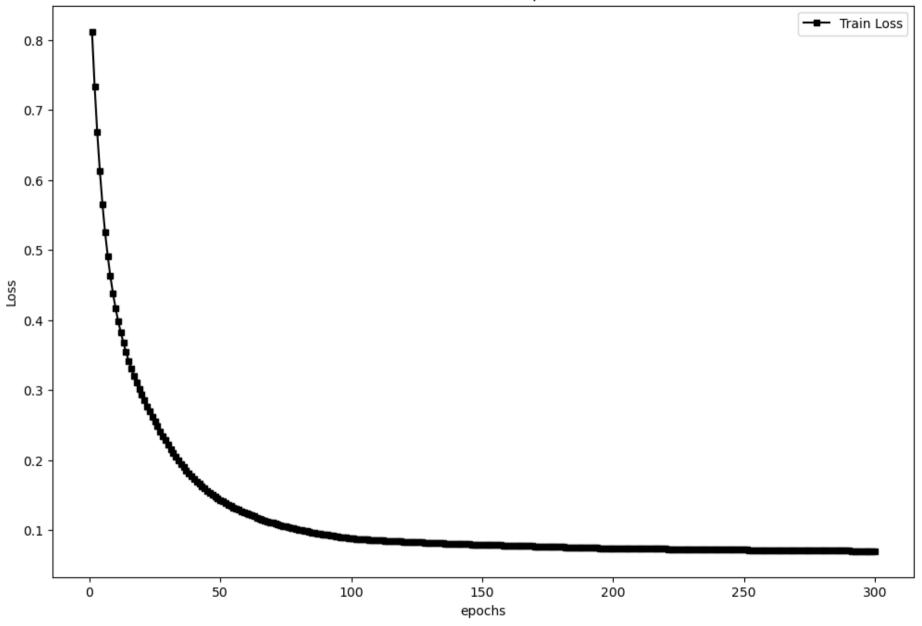
Homework 7

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```
In [133... import numpy as np
          import matplotlib.pvplot as plt
          import torch
          import torch.nn as nn
          import torch.optim as optim
          from PIL import Image
          from IPython.display import display
          import torchvision.transforms as visiontrans
In [134... def train nn(hidden layers):
             model = nn.Sequential(*hidden layers)
              trns img = visiontrans.Compose([
             visiontrans.Grayscale(),
              visiontrans.ToTensor()1)
             input_pixels = trns_img(Image.open("bird038.png"))
             flattened pxl = input pixels.flatten()
             mean pxl = input pixels.mean()
              std pxl = input pixels.std()
             std flattened pxl = (flattened pxl - mean pxl) / std pxl
             x coordinates = torch.arange(97).float()
             y coordinates = torch.arange(128).float()
             xx, yy = torch.meshgrid(x coordinates, y coordinates)
              coordinates = torch.stack((xx, yy), dim=-1).view(-1, 2)
              coord mean = coordinates.mean(dim=0)
              coord std = coordinates.std(dim=0)
              custom standardized axis = (coordinates - coord mean) / coord std
              custom standardized axis = custom standardized axis.numpy()
              std_flattened_pxl = std_flattened_pxl.numpy()
```

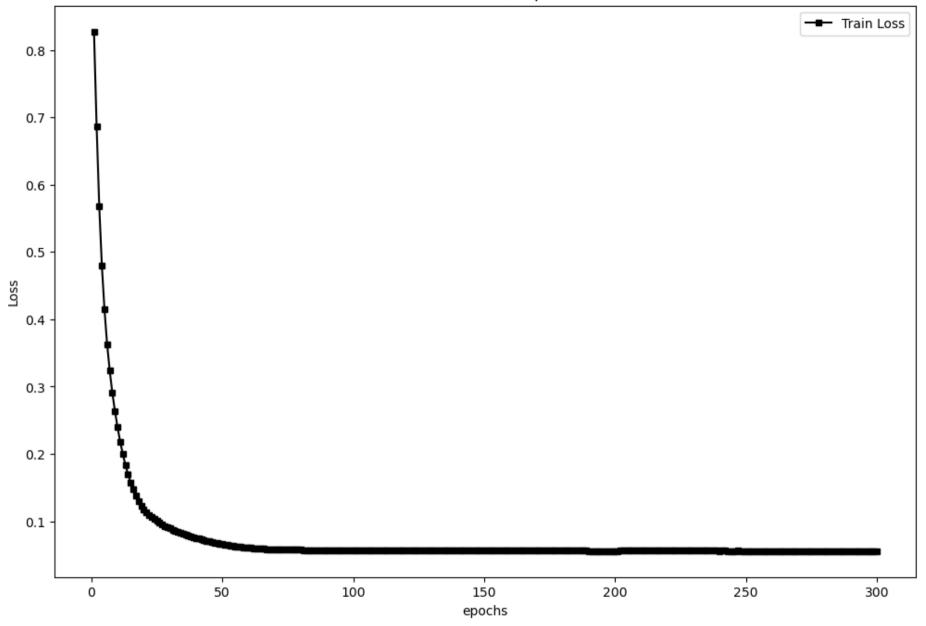
```
custom data = list(zip(custom standardized axis, std flattened pxl))
np.random.shuffle(custom data)
batch size = 64
num batches = len(custom data) // batch size
model.to(torch.device("cuda" if torch.cuda.is available() else "cpu"))
adam optim = optim.Adam(model.parameters(), lr=0.0001)
loss fn = nn.MSELoss()
step size = 100
def reduce learning rate(optimizer, epoch):
    learning rate = 0.0001 * (0.5 ** (epoch // step size))
    for param group in optimizer.param groups:
        param group['lr'] = learning rate
epoch losses = []
epochs = 300
for epoch in range(epochs):
    for i in range(num batches):
        start idx = i * batch size
        end idx = (i + 1) * batch size
        batch data = custom data[start idx:end idx]
        X batch, y batch = zip(*batch data)
        X batch = torch.tensor(X batch).float()
        y batch = torch.tensor(y batch).float()
        adam optim.zero grad()
        pred = model(X batch).squeeze()
        loss = loss fn(pred, y batch)
        loss.backward()
        adam optim.step()
    epoch losses.append(loss.item())
    reduce learning rate(adam optim, epoch)
with torch.set grad enabled(False):
    model output = model(torch.tensor(custom standardized axis).float()).cpu()
    processed image = ((model output * std pxl + mean pxl).clamp(0., 1.) * 255).numpy().astype(np.uint8)
    final image = Image.fromarray(processed image.reshape(97, 128))
plt.figure(figsize=(12, 8))
```

```
plt.plot(range(1, 301), epoch losses, color='black', marker='s', markersize=5, linestyle='-', label='Train Loss')
            plt.xlabel('epochs')
            plt.ylabel('Loss')
            plt.title('Loss function v/s Epochs')
            plt.legend()
            plt.show()
            display(final_image)
In [135... hidden_layers = [
            torch.nn.Linear(2,128),
            torch.nn.ReLU(),
           torch.nn.Linear(128,1)
        print()
        print("-----")
        print()
        train_nn(hidden_layers)
        -----1A Solution-----
```



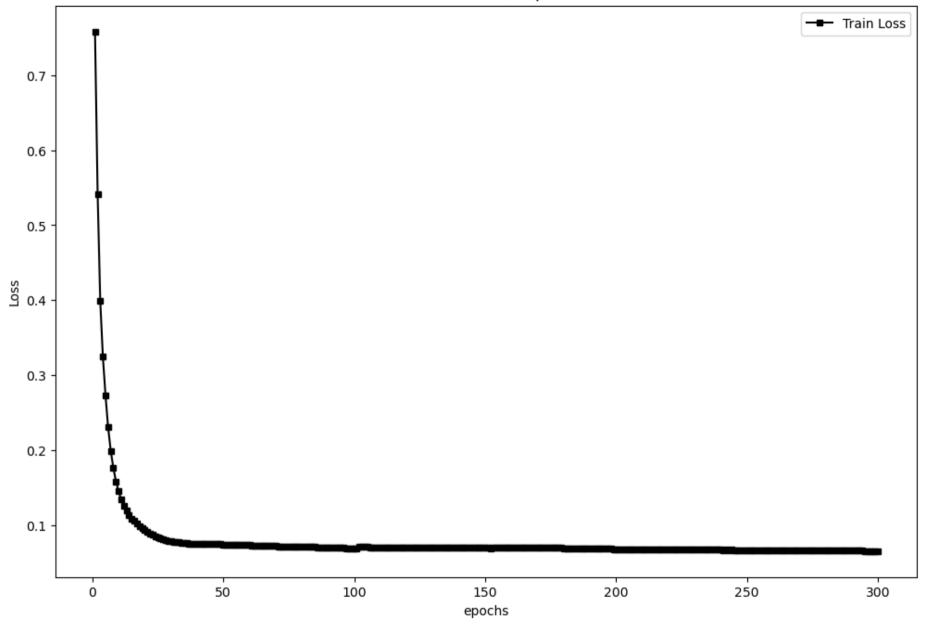


-----1B Solution-----



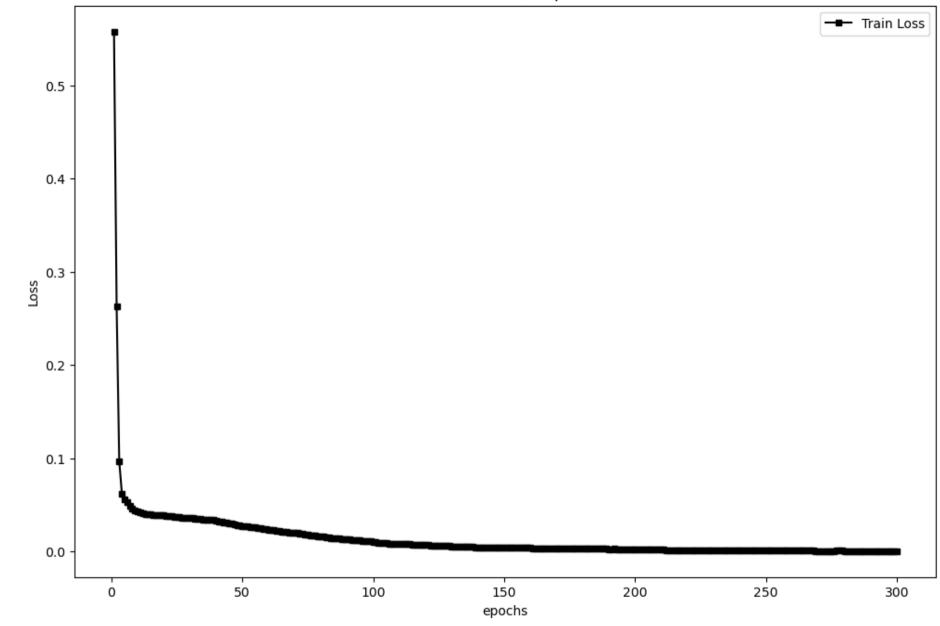


-----1C Solution-----





-----1D Solution-----





In []: