Problem 5

- Please train UNet2DModel using MNIST (hand written digits) without conditioning.
- Report the loss curve.
- Test the trained model by sampling 2 times 4 new digits.

```
!pip install datasets
!pip install torch
!pip install nbdev
!pip install genaibook
!pip install transformers
!pip install torchvision
!pip install diffusers
!pip install accelerate

Requirement accelerate

Requirement accelerate
```

Requirement already satisfied: sympy in /usr/local/lib/python3.10/dist-packa Requirement already satisfied: networkx in /usr/local/lib/python3.10/dist-pa Requirement already satisfied: jinja2 in /usr/local/lib/python3.10/dist-pack Requirement already satisfied: fsspec in /usr/local/lib/python3.10/dist-pack Requirement already satisfied: triton==2.1.0 in /usr/local/lib/python3.10/di Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-pa Requirement already satisfied: llvmlite<0.42,>=0.41.0dev0 in /usr/local/lib/ Requirement already satisfied: platformdirs>=2.5.0 in /usr/local/lib/python3 Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-pa Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/py Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dis Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3. Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3. Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/pythor Requirement already satisfied: cffi>=1.0 in /usr/local/lib/python3.10/dist-r Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/ Requirement already satisfied: mpmath>=0.19 in /usr/local/lib/python3.10/dis Requirement already satisfied: pycparser in /usr/local/lib/python3.10/dist-r Installing collected packages: genaibook Successfully installed genaibook-0.0.3 Requirement already satisfied: transformers in /usr/local/lib/python3.10/dis Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-pa Requirement already satisfied: huggingface-hub<1.0,>=0.19.3 in /usr/local/li Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.10/dist Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/ Requirement already satisfied: pyyaml>=5.1 in /usr/local/lib/python3.10/dist Requirement already satisfied: regex!=2019.12.17 in /usr/local/lib/python3.1 Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-pa

```
Requirement already satisfied: tokenizers<0.19,>=0.14 in /usr/local/lib/pyth
    Requirement already satisfied: safetensors>=0.4.1 in /usr/local/lib/python3.
    Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.10/dist-
    Requirement already satisfied: fsspec>=2023.5.0 in /usr/local/lib/python3.10
    Requirement already satisfied: typing-extensions>=3.7.4.3 in /usr/local/lib/
    Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/py
    Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dis
    Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.
    Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.
    Requirement already satisfied: torchvision in /usr/local/lib/python3.10/dist
    Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packa
    Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-pa
    Requirement already satisfied: torch==2.1.0 in /usr/local/lib/python3.10/dis
    Requirement already satisfied: pillow!=8.3.*,>=5.3.0 in /usr/local/lib/pythc
    Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-pa
    Requirement already satisfied: typing-extensions in /usr/local/lib/python3.1
    Requirement already satisfied: sympy in /usr/local/lib/python3.10/dist-packa
    Requirement already satisfied: networkx in /usr/local/lib/python3.10/dist-pa
    Requirement already satisfied: jinja2 in /usr/local/lib/python3.10/dist-pack
    Requirement already satisfied: fsspec in /usr/local/lib/python3.10/dist-pack
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    Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/py
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    Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.
    Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/
    Requirement already satisfied: mpmath>=0.19 in /usr/local/lib/python3.10/dis
    Collecting diffusers
      Downloading diffusers-0.26.3-py3-none-any.whl (1.9 MB)
                                                  1.9/1.9 MB 10.0 MB/s eta 0:00:
import torch
print(torch.__version__)
print("GPU Available:", torch.cuda.is_available())
if torch.cuda.is available():
    device = torch.device("cuda:0")
else:
```

print(f"Using device type: {device}")
2.1.0+cu121
GPU Available: True

Using device type: cuda:0

device = "cpu"

```
from psutil import virtual_memory
ram gb = virtual memory().total / 1e9
print('Your runtime has {:.1f} gigabytes of available RAM\n'.format(ram_gb))
if ram qb < 20:
  print('Not using a high-RAM runtime')
else:
  print('You are using a high-RAM runtime!')
    Your runtime has 89.6 gigabytes of available RAM
    You are using a high-RAM runtime!
from datasets import load_dataset
dataset = load_dataset("mnist", split="train")
    /usr/local/lib/python3.10/dist-packages/huggingface hub/utils/ token.py:88: Us
    The secret `HF TOKEN` does not exist in your Colab secrets.
    To authenticate with the Hugging Face Hub, create a token in your settings tak
    You will be able to reuse this secret in all of your notebooks.
    Please note that authentication is recommended but still optional to access pu
      warnings.warn(
```

Downloading data: 100% 2.60M/2.60M [00:00<00:00, 30.1MB/s]

Generating train split: 100% 60000/60000 [00:00<00:00, 552229.67 examples/s]

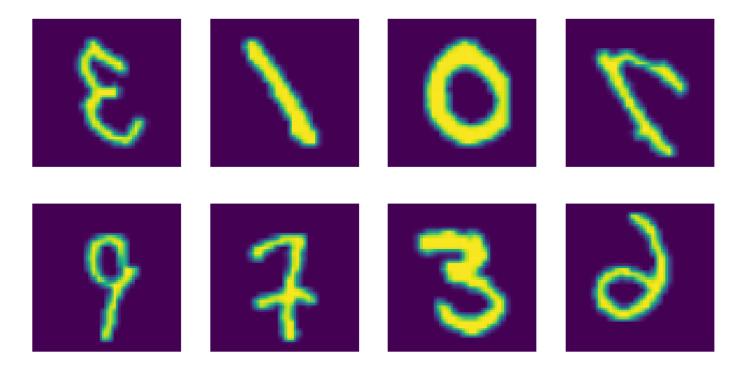
Generating test split: 100% 10000/10000 [00:00<00:00, 326445.63 examples/s]

```
from torchvision import transforms
image_size = 64
# Define data augmentations
preprocess = transforms.Compose(
    [
        transforms.Resize((image_size, image_size)), # Resize
        transforms.RandomHorizontalFlip(), # Randomly flip (data augmentation)
        transforms.ToTensor(), # Convert to tensor (0, 1)
        transforms.Normalize([0.5], [0.5]), # Map to (-1, 1)
    ]
)
import torch
batch_size = 16
def transform(examples):
   examples = [preprocess(image) for image in examples["image"]]
    return {"images": examples}
dataset.set_transform(transform)
train_dataloader = torch.utils.data.DataLoader(
   dataset, batch size=batch size, shuffle=True
)
```

from genaibook.core import show_images

batch = next(iter(train_dataloader))

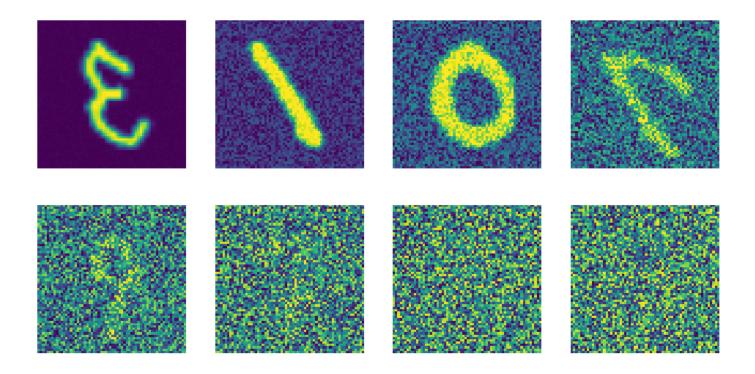
We map back to (0, 1) for display
show_images(batch["images"][:8] * 0.5 + 0.5)



```
from diffusers import DDPMScheduler
```

```
# We'll learn about beta_start and beta_end in the next sections
scheduler = DDPMScheduler(
    num_train_timesteps=1000, beta_start=0.001, beta_end=0.02
)
timesteps = torch.linspace(0, 999, 8).long()

# We load 8 images from the dataset and
# add increasing amounts of noise to them
x = batch["images"][:8]
noise = torch.rand_like(x)
noised_x = scheduler.add_noise(x, noise, timesteps)
show_images((noised_x * 0.5 + 0.5).clip(0, 1))
```



```
from diffusers import UNet2DModel
# Create a UNet2DModel
model = UNet2DModel(
    in_channels=1, # 1 channel for MNIST images
    sample_size=64, # Specify our input size
    # The number of channels per block affects the model size
    block_out_channels=(64, 128, 256, 512),
    down block types=(
        "DownBlock2D",
        "DownBlock2D",
        "AttnDownBlock2D",
        "AttnDownBlock2D",
        ),
    out channels=1.
    up_block_types=("AttnUpBlock2D", "AttnUpBlock2D", "UpBlock2D", "UpBlock2D"),
).to(device)
# Pass a batch of data through to see it works
with torch.no grad():
    out = model(noised_x.to(device), timestep=timesteps.to(device)).sample
out.shape
    torch.Size([8, 1, 64, 64])
from torch.nn import functional as F
num_epochs = 4 # Decreasing epochs to overcome colab runtime disconnect issue
lr = 1e-4 # What learning rate should we use
optimizer = torch.optim.AdamW(model.parameters(), lr=lr)
losses = [] # Somewhere to store the loss values for later plotting
# Train the model (this takes a while!)
for epoch in range(num_epochs):
    for batch in train dataloader:
        # Load the input images
        clean_images = batch["images"].to(device)
        # Sample noise to add to the images
        noise = torch.randn(clean_images.shape).to(device)
        # Sample a random timestep for each image
        timesteps = torch.randint(
```

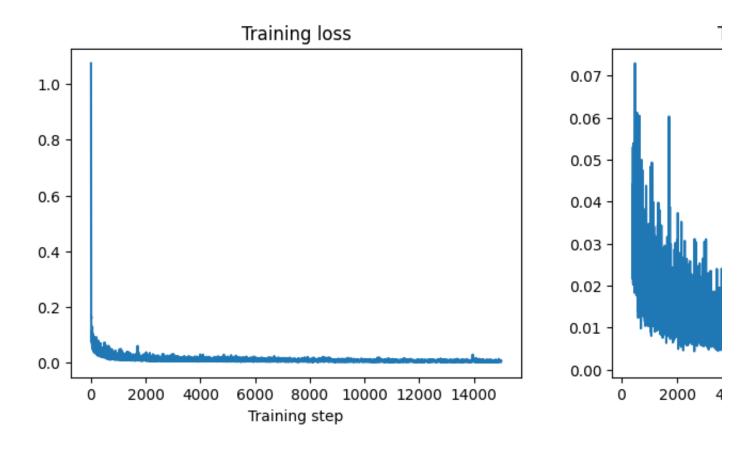
```
0.
        scheduler.config.num_train_timesteps,
        (clean images.shape[0],),
        device=device,
    ).long()
   # Add noise to the clean images according
   # to the noise magnitude at each timestep
   noisy_images = scheduler.add_noise(clean_images, noise, timesteps)
   # Get the model prediction for the noise
   # The model also uses the timestep as an input
   # for additional conditioning
    noise_pred = model(noisy_images, timesteps, return_dict=False)[0]
   # Compare the prediction with the actual noise
    loss = F.mse_loss(noise_pred, noise)
   # Store the loss for later plotting
    losses.append(loss.item())
   # Update the model parameters with the optimizer based on this loss
    loss.backward(loss)
    optimizer.step()
   optimizer.zero_grad()
print("epoch", epoch)
print("loss", loss)
epoch 0
loss tensor(0.0077, device='cuda:0', grad_fn=<MseLossBackward0>)
epoch 1
loss tensor(0.0056, device='cuda:0', grad fn=<MseLossBackward0>)
loss tensor(0.0069, device='cuda:0', grad_fn=<MseLossBackward0>)
loss tensor(0.0048, device='cuda:0', grad fn=<MseLossBackward0>)
```

```
from matplotlib import pyplot as plt

plt.subplots(1, 2, figsize=(12, 4))

plt.subplot(1, 2, 1)
plt.plot(losses)
plt.title("Training loss")
plt.xlabel("Training step")

plt.subplot(1, 2, 2)
plt.plot(range(400, len(losses)), losses[400:])
plt.title("Training loss from step 400")
plt.xlabel("Training step");
```



from diffusers import DDPMPipeline

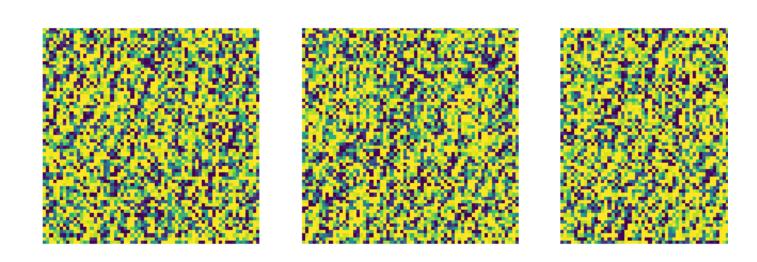
- At this point I again was disconnected from colab. As I had already used the majority of my available GPU credits on this
- assignment attempting to run the full training for the Unet model,
 I am executing the code below on output from an untrained Unet model.

```
# Random starting point (4 random images):
sample = torch.randn(4, 1, 64, 64).to(device)

for t in scheduler.timesteps:
    # Get the model prediction
    with torch.no_grad():
        noise_pred = model(sample, t)["sample"]

# Update sample with step
sample = scheduler.step(noise_pred, t, sample).prev_sample

show_images(sample.clip(-1, 1) * 0.5 + 0.5, nrows=1)
```



Start coding or generate with AI.