In []: !pip install torch-fidelity

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vjitlink-cu12-12.4.127 torch-fidelity-0.3.0
```

```
In [ ]: import os
        import numpy as np
        import torch
        import torch.nn as nn
        import torch.nn.functional as F
        from torch.utils.data import DataLoader, TensorDataset
        from torch.optim import Adam
        from torch.cuda.amp import GradScaler, autocast
        from tgdm import tgdm
        import matplotlib.pyplot as plt
        from torch fidelity import calculate metrics
In [ ]: !pip install gdown
        import gdown
        import zipfile
        file id = "1cJyPQzV0zsCZQctNBuHCqxHn0Y7v7UiA"
        output = "dataset.zip"
        gdown.download(f"https://drive.google.com/uc?id={file id}", output, quiet =
        with zipfile.ZipFile("dataset.zip", "r") as zip ref:
          zip ref.extractall("/content/dataset")
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       t-packages (from requests[socks]->gdown) (3.10)
       Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.
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       v7UiA
       From (redirected): https://drive.google.com/uc?id=1cJyPQzVOzsCZQctNBuHCqxHnO
       Y7v7UiA&confirm=t&uuid=c3609420-c9d5-48b7-8078-ac84e41c92cb
       To: /content/dataset.zip
       100%
                      | 304M/304M [00:08<00:00, 34.5MB/s]
```

```
In [ ]: data dir = "/content/dataset/Samples"
        npy files = [f for f in os.listdir(data dir) if f.endswith('.npy')]
In []: images = []
        for file in npy files:
            img = np.load(os.path.join(data dir, file))
            images.append(torch.from numpy(img).float())
In [ ]: images tensor = torch.stack(images)
        print(f"Dataset loaded. Shape: {images tensor.shape}, Min: {images tensor.mi
       Dataset loaded. Shape: torch.Size([10000, 1, 150, 150]), Min: 0.0, Max: 1.0
In [ ]: dataset = TensorDataset(images tensor)
        dataloader = DataLoader(dataset, batch size=32, shuffle=True)
In [ ]: device = torch.device('cuda' if torch.cuda.is available() else 'cpu')
In [ ]: class Diffusion(nn.Module):
          def init (self, T = 1000, beta starts = 1e-4, beta end = 0.02, device =
            super().__init ()
            self.T = T
            self.device = device
            betas = torch.linspace(beta_starts, beta_end, T).to(device)
            alphas = 1 - betas
            alpha bars = torch.cumprod(alphas, dim = 0)
            self.register buffer('sqrt alpha bars', torch.sqrt(alpha bars))
            self.register buffer('sqrt one minus alpha bars', torch.sqrt(1 - alpha bars')
          def q sample(self, x0, t, noise = None):
            if noise is None:
              noise = torch.randn like(x0)
            batch\_size = x0.shape[0]
            sqrt alpha bar = self.sqrt alpha bars[t].view(batch size, 1, 1, 1)
            sqrt one minus alpha bar = self.sqrt one minus alpha bars[t].view(batch
            return sqrt alpha bar * x0 + sqrt one minus alpha bar * noise
          def ddim sample(self, model, n samples, image size, channels, steps=50):
            model.eval()
            with torch.no grad():
              total steps = self.T
              step indices = torch.linspace(0, total steps-1, steps, dtype=torch.lor
              timesteps = step indices
              x = torch.randn(n_samples, channels, image_size, image_size, device=de
              for i in range(len(timesteps) - 1, -1, -1):
                t = timesteps[i].repeat(n samples)
                t next = timesteps[i-1] if i > 0 else -1
                predicted noise = model(x, t)
                alpha bar t = self.sqrt alpha bars[t]**2
                alpha bar t next = self.sqrt alpha bars[t next]**2 if t next >= 0 el
                sigma = 0
                x0_pred = (x - torch.sqrt(1 - alpha_bar_t) * predicted noise) / torce
                if t next >= 0:
```

```
noise_dir = torch.sqrt(1 - alpha_bar_t_next - sigma**2) * predicte
                  x = torch.sqrt(alpha bar t next) * x0 pred + noise dir
                else:
                  x = x0 pred
            return x
In [ ]: class SinusoidalEmbedding(nn.Module):
            def init (self, dim):
                super().__init__()
                self.dim = dim
            def forward(self, t):
                device = t.device
                half dim = self.dim // 2
                emb = torch.arange(half dim, device=device).float()
                emb = torch.exp(-emb * (np.log(10000) / half dim))
                emb = t[:, None] * emb[None, :]
                return torch.cat([torch.sin(emb), torch.cos(emb)], dim=1)
In [ ]: class UNetBlock(nn.Module):
            def init (self, in channels, out channels, time dim):
                super(). init ()
                self.time mlp = nn.Linear(time dim, out channels)
                self.conv1 = nn.Conv2d(in channels, out channels, 3, padding=1)
                self.conv2 = nn.Conv2d(out channels, out channels, 3, padding=1)
                self.norm1 = nn.GroupNorm(8, out channels)
                self.norm2 = nn.GroupNorm(8, out channels)
                self.relu = nn.ReLU()
            def forward(self, x, t):
                h = self.conv1(x)
                time emb = self.relu(self.time mlp(t))
                h = h + time emb[:, :, None, None]
                h = self.norm1(h)
                h = self.relu(h)
                h = self.conv2(h)
                h = self.norm2(h)
                return self.relu(h)
In [ ]: class UNet(nn.Module):
            def init (self, time dim=128):
                super(). init ()
                self.time emb = SinusoidalEmbedding(time dim)
                self.time mlp = nn.Sequential(nn.Linear(time dim, time dim), nn.ReLU
                # Downsampling
                self.down1 = UNetBlock(1, 64, time dim)
                self.pool1 = nn.MaxPool2d(2) # 150 -> 75
                self.down2 = UNetBlock(64, 128, time dim)
                self.pool2 = nn.MaxPool2d(2) # 75 -> 37
                # Bottleneck
                self.bottleneck = UNetBlock(128, 256, time dim)
                # Upsampling
```

```
self.upconv1 = nn.ConvTranspose2d(256, 128, 2, stride=2) # 37 -> 74
                                     self.up1 = UNetBlock(256, 128, time dim)
                                     self.upconv2 = nn.ConvTranspose2d(128, 64, 2, stride=2) # 74 -> 148
                                     self.up2 = UNetBlock(128, 64, time dim)
                                     # Output
                                     self.out = nn.Conv2d(64, 1, 1)
                            def forward(self, x, t):
                                     t = self.time = 
                                     t emb = self.time mlp(t emb)
                                     d1 = self.down1(x, t emb)
                                     d2 = self.pool1(d1)
                                     d3 = self.down2(d2, t emb)
                                     d4 = self.pool2(d3)
                                     b = self.bottleneck(d4, t emb)
                                     u1 = self.upconv1(b)
                                     u1 = torch.cat([F.interpolate(u1, size=d3.shape[2:], mode='bilinear'
                                     u2 = self.up1(u1, t emb)
                                     u3 = self.upconv2(u2)
                                     u3 = torch.cat([F.interpolate(u3, size=d1.shape[2:], mode='bilinear'
                                     u4 = self.up2(u3, t emb)
                                     return self.out(u4)
In [ ]: diffusion = Diffusion(device=device).to(device)
                   model = UNet().to(device)
                   optimizer = Adam(model.parameters(), lr=1e-4)
                   scaler = GradScaler()
                   num epochs = 50
                <ipython-input-13-105fb84a9e88>:4: FutureWarning: `torch.cuda.amp.GradScaler
                (args...)` is deprecated. Please use `torch.amp.GradScaler('cuda', args...)`
                instead.
                   scaler = GradScaler()
In [ ]: for epoch in range(num epochs):
                            model.train()
                            total loss = 0
                            for batch in tqdm(dataloader, desc=f"Epoch {epoch+1}/{num epochs}"):
                                     x0 = batch[0].to(device)
                                     t = torch.randint(0, diffusion.T, (x0.shape[0],), device=device)
                                     noise = torch.randn like(x0)
                                     xt = diffusion.q sample(x0, t, noise)
                                     with autocast():
                                              predicted noise = model(xt, t)
                                              loss = F.mse loss(predicted noise, noise)
                                     optimizer.zero grad()
                                     scaler.scale(loss).backward()
                                     scaler.step(optimizer)
                                     scaler.update()
                                     total loss += loss.item()
                            avg loss = total loss / len(dataloader)
```

```
print(f"Epoch {epoch+1}/{num epochs}, Avg Loss: {avg loss:.6f}")
    # Save checkpoint every 10 epochs
    if (epoch + 1) % 10 == 0:
        torch.save(model.state dict(), f'diffusion model epoch {epoch+1}.pth
                        | 0/313 [00:00<?, ?it/s]<ipython-input-14-9e358da
Epoch 1/50:
0f5fd>:9: FutureWarning: `torch.cuda.amp.autocast(args...)` is deprecated. P
lease use `torch.amp.autocast('cuda', args...)` instead.
 with autocast():
Epoch 1/50: 100% | 313/313 [01:24<00:00, 3.70it/s]
Epoch 1/50, Avg Loss: 0.061173
Epoch 2/50: 100%
                       313/313 [01:22<00:00, 3.79it/s]
Epoch 2/50, Avg Loss: 0.008365
Epoch 3/50: 100% | 313/313 [01:22<00:00, 3.79it/s]
Epoch 3/50, Avg Loss: 0.006109
Epoch 4/50: 100% 313/313 [01:22<00:00, 3.79it/s]
Epoch 4/50, Avg Loss: 0.004949
Epoch 5/50: 100% | 313/313 [01:22<00:00, 3.78it/s]
Epoch 5/50, Avg Loss: 0.004373
Epoch 6/50: 100% | 313/313 [01:22<00:00, 3.79it/s]
Epoch 6/50, Avg Loss: 0.004251
Epoch 7/50: 100%
                       | 313/313 [01:22<00:00, 3.79it/s]
Epoch 7/50, Avg Loss: 0.003748
Epoch 8/50: 100% 313/313 [01:22<00:00, 3.79it/s]
Epoch 8/50, Avg Loss: 0.003838
Epoch 9/50: 100% | 313/313 [01:22<00:00, 3.80it/s]
Epoch 9/50, Avg Loss: 0.003494
Epoch 10/50: 100%| 313/313 [01:22<00:00, 3.80it/s]
Epoch 10/50, Avg Loss: 0.003297
Epoch 11/50: 100%| 313/313 [01:22<00:00, 3.80it/s]
Epoch 11/50, Avg Loss: 0.003168
Epoch 12/50: 100% | 313/313 [01:22<00:00, 3.79it/s]
Epoch 12/50, Avg Loss: 0.003616
Epoch 13/50: 100% | 313/313 [01:22<00:00, 3.80it/s]
Epoch 13/50, Avg Loss: 0.003116
Epoch 14/50: 100% | 313/313 [01:22<00:00, 3.79it/s]
Epoch 14/50, Avg Loss: 0.002822
Epoch 15/50: 100%
                        | 313/313 [01:22<00:00, 3.79it/s]
Epoch 15/50, Avg Loss: 0.002839
Epoch 16/50: 100% | 313/313 [01:22<00:00, 3.79it/s]
Epoch 16/50, Avg Loss: 0.002829
Epoch 17/50: 100%| 313/313 [01:22<00:00, 3.79it/s]
Epoch 17/50, Avg Loss: 0.002805
Epoch 18/50: 100% | 313/313 [01:22<00:00, 3.79it/s]
Epoch 18/50, Avg Loss: 0.002809
Epoch 19/50: 100% | 313/313 [01:22<00:00, 3.80it/s]
Epoch 19/50, Avg Loss: 0.002782
Epoch 20/50: 100%
                        | 313/313 [01:22<00:00, 3.79it/s]
Epoch 20/50, Avg Loss: 0.002889
Epoch 21/50: 100% | 313/313 [01:22<00:00, 3.79it/s]
```

```
Epoch 21/50, Avg Loss: 0.002983
Epoch 22/50: 100%
                       | 313/313 [01:22<00:00, 3.79it/s]
Epoch 22/50, Avg Loss: 0.002647
Epoch 23/50: 100%| 313/313 [01:22<00:00, 3.80it/s]
Epoch 23/50, Avg Loss: 0.002530
Epoch 24/50: 100%| 313/313 [01:22<00:00, 3.79it/s]
Epoch 24/50, Avg Loss: 0.002551
Epoch 25/50: 100% | 313/313 [01:22<00:00,
                                              3.79it/s
Epoch 25/50, Avg Loss: 0.002328
Epoch 26/50: 100% | 313/313 [01:22<00:00,
                                              3.79it/s
Epoch 26/50, Avg Loss: 0.002647
Epoch 27/50: 100%
                       | 313/313 [01:22<00:00,
                                              3.79it/s
Epoch 27/50, Avg Loss: 0.002692
Epoch 28/50: 100%| 313/313 [01:22<00:00, 3.79it/s]
Epoch 28/50, Avg Loss: 0.002683
Epoch 29/50: 100% | 313/313 [01:22<00:00,
                                              3.79it/s
Epoch 29/50, Avg Loss: 0.002508
Epoch 30/50: 100%| 313/313 [01:22<00:00, 3.78it/s]
Epoch 30/50, Avg Loss: 0.002139
Epoch 31/50: 100%| 313/313 [01:22<00:00,
                                               3.79it/s
Epoch 31/50, Avg Loss: 0.002213
Epoch 32/50: 100%| 313/313 [01:22<00:00, 3.79it/s]
Epoch 32/50, Avg Loss: 0.002151
Epoch 33/50: 100% | 313/313 [01:22<00:00, 3.80it/s]
Epoch 33/50, Avg Loss: 0.002427
Epoch 34/50: 100% | 313/313 [01:22<00:00,
                                              3.79it/s
Epoch 34/50, Avg Loss: 0.002314
Epoch 35/50: 100%
                       | 313/313 [01:22<00:00, 3.79it/s]
Epoch 35/50, Avg Loss: 0.002134
Epoch 36/50: 100%| 313/313 [01:22<00:00, 3.79it/s]
Epoch 36/50, Avg Loss: 0.002307
Epoch 37/50: 100% | 313/313 [01:22<00:00, 3.79it/s]
Epoch 37/50, Avg Loss: 0.001954
Epoch 38/50: 100% | 313/313 [01:22<00:00,
                                              3.79it/s
Epoch 38/50, Avg Loss: 0.002340
Epoch 39/50: 100% | 313/313 [01:22<00:00,
                                              3.78it/s
Epoch 39/50, Avg Loss: 0.002319
Epoch 40/50: 100%
                       | 313/313 [01:22<00:00,
                                              3.78it/s
Epoch 40/50, Avg Loss: 0.001903
Epoch 41/50: 100% | 313/313 [01:22<00:00,
                                              3.78it/s
Epoch 41/50, Avg Loss: 0.002443
Epoch 42/50: 100% | 313/313 [01:22<00:00,
                                              3.79it/s
Epoch 42/50, Avg Loss: 0.001773
Epoch 43/50: 100% 3.79it/s]
Epoch 43/50, Avg Loss: 0.002127
Epoch 44/50: 100%| 313/313 [01:22<00:00, 3.78it/s]
Epoch 44/50, Avg Loss: 0.002165
Epoch 45/50: 100% | 313/313 [01:22<00:00, 3.78it/s]
Epoch 45/50, Avg Loss: 0.002119
```

```
OutOfMemorvError
                                          Traceback (most recent call last)
<ipython-input-17-c4d3fef7a10e> in <cell line: 0>()
----> 1 generated images = diffusion.ddim sample(model, n samples=1000, imag
e size=150, channels=1, steps=50)
      2 generated images = torch.clamp(generated images, 0, 1)
<ipython-input-9-77580367b9d0> in ddim sample(self, model, n samples, image
size, channels, steps)
     30
                t = timesteps[i].repeat(n samples)
     31
                t next = timesteps[i-1] if i > 0 else -1
---> 32
                predicted noise = model(x, t)
     33
                alpha bar t = self.sqrt alpha bars[t]**2
     34
                alpha bar t next = self.sqrt alpha bars[t next]**2 if t next
>= 0 else 1.0
/usr/local/lib/python3.11/dist-packages/torch/nn/modules/module.py in wrapp
ed call impl(self, *args, **kwargs)
   1734
                    return self. compiled call impl(*args, **kwargs) # typ
e: ignore[misc]
   1735
                else:
-> 1736
                    return self. call impl(*args, **kwargs)
   1737
   1738
            # torchrec tests the code consistency with the following code
/usr/local/lib/python3.11/dist-packages/torch/nn/modules/module.py in call
impl(self, *args, **kwargs)
                        or global backward pre hooks or global backward ho
   1745
oks
   1746
                        or global forward hooks or global forward pre hook
s):
-> 1747
                    return forward call(*args, **kwargs)
   1748
   1749
                result = None
<ipython-input-12-b7592765fa37> in forward(self, x, t)
     27
                t emb = self.time mlp(t emb)
     28
---> 29
                d1 = self.down1(x, t emb)
     30
                d2 = self.pool1(d1)
     31
                d3 = self.down2(d2, t emb)
/usr/local/lib/python3.11/dist-packages/torch/nn/modules/module.py in wrapp
ed call impl(self, *args, **kwargs)
   1734
                    return self. compiled call impl(*args, **kwargs) # typ
e: ignore[misc]
   1735
                else:
-> 1736
                    return self. call_impl(*args, **kwargs)
   1737
   1738
            # torchrec tests the code consistency with the following code
/usr/local/lib/python3.11/dist-packages/torch/nn/modules/module.py in call
impl(self, *args, **kwargs)
   1745
                        or global backward pre hooks or global backward ho
oks
   1746
                        or global forward hooks or global forward pre hook
```

```
s):
-> 1747
                    return forward call(*args, **kwargs)
   1748
   1749
                result = None
<ipython-input-11-05f8304e98a8> in forward(self, x, t)
     11
            def forward(self, x, t):
---> 12
                h = self.conv1(x)
     13
                time emb = self.relu(self.time mlp(t))
     14
                h = h + time emb[:, :, None, None]
/usr/local/lib/python3.11/dist-packages/torch/nn/modules/module.py in wrapp
ed call impl(self, *args, **kwargs)
   1734
                   return self. compiled call impl(*args, **kwargs) # typ
e: ignore[misc]
   1735
                else:
-> 1736
                    return self. call impl(*args, **kwargs)
  1737
   1738
           # torchrec tests the code consistency with the following code
/usr/local/lib/python3.11/dist-packages/torch/nn/modules/module.py in call
impl(self, *args, **kwargs)
   1745
                        or global backward pre hooks or global backward ho
oks
                        or global forward hooks or global forward pre hook
   1746
s):
-> 1747
                    return forward call(*args, **kwargs)
   1748
   1749
                result = None
/usr/local/lib/python3.11/dist-packages/torch/nn/modules/conv.py in forward
(self, input)
    552
    553
            def forward(self, input: Tensor) -> Tensor:
--> 554
                return self. conv forward(input, self.weight, self.bias)
    555
    556
/usr/local/lib/python3.11/dist-packages/torch/nn/modules/conv.py in conv fo
rward(self, input, weight, bias)
    547
                        self.groups,
    548
--> 549
                return F.conv2d(
    550
                    input, weight, bias, self.stride, self.padding, self.dil
ation, self.groups
    551
OutOfMemoryError: CUDA out of memory. Tried to allocate 5.37 GiB. GPU 0 has
a total capacity of 14.74 GiB of which 916.12 MiB is free. Process 5293 has
13.84 GiB memory in use. Of the allocated memory 11.57 GiB is allocated by P
yTorch, and 2.14 GiB is reserved by PyTorch but unallocated. If reserved but
unallocated memory is large try setting PYTORCH CUDA ALLOC CONF=expandable s
egments:True to avoid fragmentation. See documentation for Memory Managemen
t (https://pytorch.org/docs/stable/notes/cuda.html#environment-variables)
```

```
In [ ]: generated images rgb = generated images.repeat(1, 3, 1, 1).cpu()
        real_images_rgb = images_tensor[:1000].repeat(1, 3, 1, 1)
        metrics = calculate metrics(
            input1=generated images rgb,
            input2=real images rgb,
            fid=True,
            cuda=True,
            isc=False,
            kid=False
        fid = metrics['frechet inception distance']
        print(f"FID Score: {fid}")
In [ ]: fig, axes = plt.subplots(1, 5, figsize=(15, 3))
        for i in range(5):
            axes[i].imshow(generated_images[i].squeeze().cpu().numpy(), cmap='gray')
            axes[i].axis('off')
        plt.suptitle("Generated Strong Lensing Images")
        plt.show()
```

This notebook was converted with convert.ploomber.io