```
In [ ]: %load_ext autoreload
        %autoreload 2
```

The autoreload extension is already loaded. To reload it, use: %reload\_ext autoreload

### **Imports**

```
In [ ]: import pathlib
         import random
         import time
         import matplotlib.pyplot as plt
         import tqdm
         import torch
         from torch import nn, optim
         from torch.nn.utils import clip_grad_norm_
         from torch.nn.functional import interpolate
         from torchvision.utils import save_image
         from torch.utils.data import DataLoader
         \textbf{from} \  \, \textbf{torchvision.transforms.functional} \  \, \textbf{import} \  \, \textbf{adjust\_brightness}
         from super_resolution.src.sen2venus_dataset import (
             create_train_test_split,
         {\bf from} \ {\bf super\_resolution.src.visualization} \ {\bf import} \ {\bf plot\_gallery}
         from super_resolution.src.srgan import SRResNet, Discriminator
In [ ]: DATA_DIR = pathlib.Path("C:/Users/Mitch/stat3007_data")
         SITES_DIR = DATA_DIR / "sites"
         PREPROCESSING_DIR = DATA_DIR / "preprocessing"
         RESULTS_DIR = DATA_DIR / "results"
```

#### **Prepare Data**

```
In [ ]: sites = {
            "FR-BIL",
            "NARYN",
        train_patches, test_patches = create_train_test_split(
            str(SITES_DIR) + "\\", sites=sites
        print(f"Num train {len(train_patches)}\n" f"Num test {len(test_patches)}")
       Num train 8159
       Num test 3498
In [ ]: def image_transform(x, y):
            x = x[:3, :, :]
            y = y[:3, :, :]
            x = torch.clamp(x, 0, 1)
            y = torch.clamp(y, 0, 1)
            return x, y
In [ ]: train_patches.set_transform(image_transform)
        test_patches.set_transform(image_transform)
In [ ]: train_loader = DataLoader(train_patches, batch_size=1)
In [ ]: (low_res, high_res) = next(train_loader.__iter__())
In [ ]: index = random.randint(0, len(low_res) - 1)
        low_res_example = low_res[index]
        high_res_example = high_res[index]
        plot_gallery(
                adjust brightness(low res example, 2).permute(1, 2, 0),
                adjust_brightness(high_res_example, 2).permute(1, 2, 0),
            titles=["low res", "high res"],
            xscale=5,
            yscale=5,
```





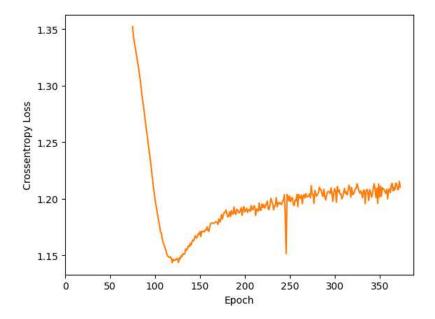
In [ ]: del train\_loader

## **Training**

```
In [ ]: # For clearing GPU memory
        # 1 / 0
In [ ]: # For clearing GPU memory
        import gc
        gc.collect()
        torch.cuda.empty_cache()
In [ ]: BATCH_SIZE = 16
        GEN_LEARNING_RATE = 1e-5
        DISCRIM_LEARNING_RATE = 1e-5
        GEN_WEIGHT_DECAY = 1e-8
        DISCRIM_WEIGHT_DECAY = 1e-8
        DISCRIM_WEIGHT = 1e-6
In [ ]: device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
        torch.cuda.empty_cache()
        should_pin_memory = {"cuda": True, "cpu": False}
        train_loader = DataLoader(
            train_patches,
             shuffle=True,
            batch_size=BATCH_SIZE,
            drop_last=True,
            pin_memory=should_pin_memory[device.type],
             num_workers=0,
In [ ]: device
Out[ ]: device(type='cuda')
In [ ]: generator = SRResNet(scaling_factor=2, n_blocks=8)
In [ ]: discriminator = Discriminator(n_blocks=2, fc_size=128)
In [ ]: generator = generator.to(device)
         generator = generator.train()
        discriminator = discriminator.to(device)
discriminator = discriminator.train()
In [ ]: gen_optimizer = optim.Adam(
             generator.parameters(), 1r=GEN_LEARNING_RATE, weight_decay=GEN_WEIGHT_DECAY
        discrim_optimizer = optim.Adam(
            discriminator.parameters(),
             1r=DISCRIM_LEARNING_RATE,
```

```
weight_decay=DISCRIM_WEIGHT_DECAY,
In [ ]: # save_file = RESULTS_DIR / "continuingbestgen_374epoch.pkl"
        # Loaded_experiment = torch.load(save_file, map_location=device)
        # # generator = SRResNet(scaling_factor=2, n_blocks=16).to(device)
        # # discriminator = Discriminator(n_blocks=3, fc_size=128)
# generator.load_state_dict(loaded_experiment["gen_state"])
        # discriminator.load_state_dict(loaded_experiment["discrim_state"])
        # gen_optimizer.load_state_dict(loaded_experiment["gen_optimizer_state"])
        # discrim_optimizer.load_state_dict(loaded_experiment["discrim_optimizer_state"])
        # gen_losses = loaded_experiment["gen_losses"]
        # discrim_losses = Loaded_experiment["discrim_losses"]
        # train_time = loaded_experiment["train_time"]
In [ ]: pixel_criterion = nn.MSELoss()
        discrim_criterion = nn.BCEWithLogitsLoss()
In [ ]: NUM_EPOCHS = 2000
In [ ]: train_time = 0.0
        gen_losses = []
        discrim_losses = []
In [ ]: generator = generator.to(device)
        generator = generator.train()
        discriminator = discriminator.to(device)
        discriminator = discriminator.train()
        for epoch in range(NUM_EPOCHS):
            progress_bar = tqdm.tqdm(train_loader, total=len(train_loader), ncols=100)
            gen_epoch_loss = 0.0
            discrim_epoch_loss = 0.0
            num_batches = 0
            for i, (low_res_batch, high_res_batch) in enumerate(progress_bar):
               num batches += 1
                start_time = time.time()
                # Push to GPU
               low_res_batch = low_res_batch.to(device)
               high_res_batch = high_res_batch.to(device)
                # Update generator
                super_resolved = generator(low_res_batch)
               natural_probs = discriminator(super_resolved)
                pixel_loss = pixel_criterion(super_resolved, high_res_batch)
                adversarial_loss = discrim_criterion(
                   natural_probs, torch.ones_like(natural_probs)
                gen_loss = pixel_loss
                gen_loss += DISCRIM_WEIGHT * adversarial_loss
                gen_optimizer.zero_grad()
                gen_loss.backward()
                gen_optimizer.step()
                # Update discriminator
                true_natural_probs = discriminator(high_res_batch)
                # Detach to skip generator computations
                fake_natural_probs = discriminator(super_resolved.detach())
                true_natural_loss = discrim_criterion(
                   {\tt true\_natural\_probs, torch.ones\_like(true\_natural\_probs)}
                fake_natural_loss = discrim_criterion(
                   fake_natural_probs, torch.zeros_like(fake_natural_probs)
                discrim_loss = true_natural_loss + fake_natural_loss
                discrim_optimizer.zero_grad()
                discrim_loss.backward()
                discrim_optimizer.step()
                # Collect data
                gen_epoch_loss += gen_loss.item()
                discrim_epoch_loss += discrim_loss.item()
                progress_bar.set_postfix(
                   epoch=epoch.
                   gen_loss=f"{gen_epoch_loss/num_batches:.8f}",
```

```
discrim_loss=f"{discrim_epoch_loss/num_batches:.8f}",
                )
                end_time = time.time()
                train_time += end_time - start_time
            gen_epoch_loss /= len(train_loader)
            discrim_epoch_loss /= len(train_loader)
            gen_losses.append(gen_epoch_loss)
            discrim_losses.append(discrim_epoch_loss)
            print(
                f"Epoch: {epoch} / gen_loss: {gen_epoch_loss:.8f} / discrim_loss: {discrim_epoch_loss:.8f}"
In [ ]: train_time / 60
Out[]: 1053.3067023237547
In [ ]: len(gen_losses)
Out[ ]: 374
In [ ]: gen_losses[-5:]
Out[ ]: [2.1545394862930416e-05,
         2.1526079207431426e-05,
         2.174403593542232e-05,
         2.1463911347496594e-05,
         2.1551264829290158e-05]
In [ ]: plt.plot(gen_losses[:75], label="Pre-training generator loss")
        plt.plot(range(75, len(gen_losses)), gen_losses[75:], label="Adversarial learning loss")
         # plt.title("Generator losses")
        plt.xlabel("Epoch")
        plt.ylabel("Loss")
        plt.legend()
        plt.show()
          0.0007
                                                            Pre-training generator loss
                                                            Adversarial learning loss
          0.0006
          0.0005
          0.0004
          0.0003
          0.0002
          0.0001
          0.0000
                     0
                                     100
                                             150
                                                     200
                                                              250
                                                                      300
                                                                               350
                                                  Epoch
In [ ]: plt.plot([])
        plt.plot(range(75, len(discrim_losses)), discrim_losses[75:])
        plt.xlim(left=0)
         # plt.title("Discriminator Losses")
        plt.xlabel("Epoch")
        plt.ylabel("Crossentropy Loss")
        plt.show()
```



## Saving

```
In [ ]: experiment = {
             "<mark>gen_losses"</mark>: gen_losses,
             "discrim_losses": discrim_losses,
             "gen_state": generator.state_dict(),
             "discrim_state": discriminator.state_dict(),
             "gen_optimizer_state": gen_optimizer.state_dict(),
             "discrim_optimizer_state": discrim_optimizer.state_dict(),
             "gen_learning_rate": GEN_LEARNING_RATE,
             "discrim_learning_rate": DISCRIM_LEARNING_RATE,
             "gen_weight_decay": GEN_WEIGHT_DECAY,
             "discrim_weight_decay": DISCRIM_WEIGHT_DECAY,
             "discrim_weight": DISCRIM_WEIGHT,
             "train_time": train_time,
In [ ]: save file = RESULTS DIR / "continuingbestgen 378epoch.pkl"
In [ ]: if not save_file.exists():
             torch.save(experiment, save_file)
             print(f"Saved to {save_file}")
```

Saved to C:\Users\Mitch\stat3007\_data\results\continuingbestgen\_378epoch.pkl

# **Generated Examples**

```
In [ ]: BRIGHT_FACTOR = 3.5
In [ ]: generator = generator.to("cpu")
In [ ]: loader = DataLoader(train_patches, batch_size=100)
         (low_res, high_res) = next(loader.__iter__())
In [ ]: import random
        index = random.randint(0, len(low_res) - 1)
        low_res_example = low_res[index]
        high_res_example = high_res[index]
        out = generator(low_res_example.unsqueeze(0)).detach().clamp(0, 1)
        bicubic_out = interpolate(
            low res example.unsqueeze(0),
            size=(256, 256),
            mode="bicubic",
        ).float()
        plot_gallery(
                 adjust_brightness(low_res_example, BRIGHT_FACTOR).permute(1, 2, 0),
                 adjust\_brightness(high\_res\_example, \ BRIGHT\_FACTOR).permute(1,\ 2,\ 0),
                 adjust_brightness(out[0], BRIGHT_FACTOR).permute(1, 2, 0),
                 adjust_brightness(bicubic_out[0], BRIGHT_FACTOR).permute(1, 2, 0),
            titles=["low res", "high res", "SRGAN", "Bicubic"],
```

```
xscale=5,
            yscale=5,
In [ ]: loader = DataLoader(test_patches, batch_size=300)
        (low_res, high_res) = next(loader.__iter__())
In [ ]: import random
        index = random.randint(0, len(low_res) - 1)
        low_res_example = low_res[index]
        high_res_example = high_res[index]
        out = generator(low_res_example.unsqueeze(0)).detach().clamp(0, 1)[0]
        out = adjust_brightness(out, BRIGHT_FACTOR)
        bicubic\_out = (
            interpolate(low_res_example.unsqueeze(0), size=(256, 256), mode="bicubic")
             .float()
            .clamp(0, 1)
        bicubic_out = adjust_brightness(bicubic_out, BRIGHT_FACTOR)
        low_res_example = adjust_brightness(low_res_example, BRIGHT_FACTOR)
        high_res_example = adjust_brightness(high_res_example, BRIGHT_FACTOR)
        plot_gallery(
            [
                low_res_example.permute(1, 2, 0),
                high_res_example.permute(1, 2, 0),
                out.permute(1, 2, 0),
                bicubic_out.permute(1, 2, 0),
            titles=["low res", "high res", "SRGAN", "Bicubic"],
            xscale=5,
            yscale=5,
```

# Save images

```
In []: IMAGE_DIR = RESULTS_DIR / "images/final3_5bright"

In []: index

In []: srgan_file = IMAGE_DIR / f"{index}srgan.png"
    bicubic_file = IMAGE_DIR / f"{index}bicubic.png"
    low_res_file = IMAGE_DIR / f"{index}lowres.png"
    high_res_file = IMAGE_DIR / f"{index}highres.png"

In []: save_image(low_res_example, low_res_file)
    save_image(high_res_example, high_res_file)
    save_image(out, srgan_file)
    save_image(bicubic_out, bicubic_file)
```

#### **Metrics**

```
In [ ]: del loader
In [ ]: from super_resolution.src.testing import compute_metrics
In [ ]: # 256 is largest I can handle on cpu
        metric_loader = DataLoader(test_patches, batch_size=min(len(test_patches), 256))
In [ ]: metrics = compute_metrics(lambda x: generator(x).clamp(0, 1), metric_loader)
      100%|
                                                                       | 14/14 [14:58<00:00, 64.16s/it]
In [ ]: metrics
Out[]: Metrics(mse=2.52257554425991e-05, psnr=45.9937013898577, ssim=0.9888156950473785, fid=0.21431732177734375)
In [ ]: bicubic_metrics = compute_metrics(
            lambda x: interpolate(x, size=(256, 256), mode="bicubic"), metric_loader
        )
                                                                  | 14/14 [09:37<00:00, 41.28s/it]
      100%|
In [ ]: bicubic_metrics
Out[]: Metrics(mse=4.024813668885534e-05, psnr=43.96280016217913, ssim=0.9836511441639492, fid=1.3964385986328125)
```