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# Face Generation

## REVIEW

## HISTORY

### Meets Specifications

Congratulations! Your project meets all the requirements and it is able to sample decent faces. Keep up the good work!

PD: Do not worry about the reference, your code speaks for itself.

### Section 1: Code quality

Scripts have an intuitive, easy-to-follow structure with code separated into logical functions. Naming for variables and functions follows the PEP8 style guidelines.

Good work. Code follows good programming practices, it is clear and uses type hints.

### Section 2: Generator and discriminator design

The generator should take a batched 1d latent vector as input and output a batch of RGB images (3 channels).

- ✓ The generator takes 1D latent vectors and outputs them to 3-channel RGB images.
- ✓ The generator uses transpose convolutions.
- ✓ The generator uses an activation function to map the output to  $(-1, 1)$ .
- ✓ The generator is not too complex nor too simple.

The discriminator should take as input a batch of images and output a score for each image in the batch.

- ✓ The discriminator takes images and outputs a score for each one.
- ✓ The discriminator downsamples the images using convolutions. You can try pooling methods too, as an alternative.
- ✓ The discriminator is not too complex nor too simple.

Generator and Discriminator are inheriting from the torch `Module` class. The layers are defined in the `__init__` method and called in the `forward` method.

Both discriminator and generator inherit from the Module class and are implemented correctly.

## Section 3: Data pipeline

(Provide specific details on how the student will fulfill the criteria.)

The `get_transform` function should output a `Compose` of different torchvision (or non torchvision) transforms.

Correct transform data pipeline. No data augmentation was used and Normalize is used correctly.

The custom dataset should have the `__len__` and the `__getitem__` methods implemented and working. The dataset should return a tensor image in the -1 / 1 range.

Nicely done!

## Section 4: Loss implementation and training

Both loss functions are accomplishing their roles: the discriminator should be trained to separate fake and real images and the generator should be trained to fool the discriminator. No specific functions are required.

Correct implementation. Gradient penalty was used.

Two optimizers are created, one for the generator and one for the discriminator. They are both using low learning rates.

Both optimizers correctly use the Adam optimizer with proper parameters.

The `discriminator_step` and `generator_step` functions are correctly implemented. The model is training for enough epochs.

Proper and easy-to-read implementation. Five epochs are more than enough for this exercise.

The student makes reasonable decisions based on initial results.

Nice analysis and sensible suggestions to try in the future.

The generated samples should have face attributes (eyes, nose, mouth, hair) and a rough resemblance to a face.

The model clearly generates recognizable faces, with the problems described in the questions section.

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