

Diffusion project

Exercise 3

Advanced Deep Learning in Computer Vision

September 2023

In this exercise, you are asked to build a diffusion model that generates 16×16 sprites:



Figure 1: Example sprites

You tasks are as follows:

1. Calculate β_t , α_t , and $\bar{\alpha}_t$ in the `__init__` function of the Diffusion class.
(See files `ddpm.py` and `playground.py`)
2. Complete the implementation of forward process in the function `q_sample`.
(See files `ddpm.py` and `playground.py`)
3. Complete the implementation of the reverse process in the function `p_sample`.
(See files `ddpm.py` and `playground.py`)
4. Implement the training function.
(See file `ddpm_train.py`). Training takes around 1 hour on a CPU with a reduced dataset size of 40K images. You should be able to see reasonable image generations between epochs 20-30 (tested with SEED=1). Set DATASET_SIZE to None if you want to train on the full dataset.

5. (OPTIONAL) Implement a cosine schedule in the function *get_betas* and analyze all the differences. See Figures 3 and 5 in the paper "[Improved Denoising Diffusion Probabilistic Models](#)"

Notation: In the lecture, we follow the notation of the [ddpm paper](#), while in the code, we follow the notation from the OpenAI code repository. Here we provide a mapping between the two.

- T is the total number of diffusion steps
- x_t = image at timestep t
- $x_T \sim \mathcal{N}(0, \mathbf{I})$
- $\beta_t = \text{betas}[t]$
- $\alpha_t = \text{alphas}[t]$
- $\bar{\alpha}_t = \text{alphas_bar}[t]$
- $q(x_t|x_0) = \text{q_sample}$
- $p_\theta(x_{t-1}|x_t) = \text{p_sample}$