Deep Learning: Homework 3

Instructed by Yi Wu Due on Apr 21, 2021

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1 Energy-Based Model

Almost all default hyper-parameters are modified according to [2]. To train, run:

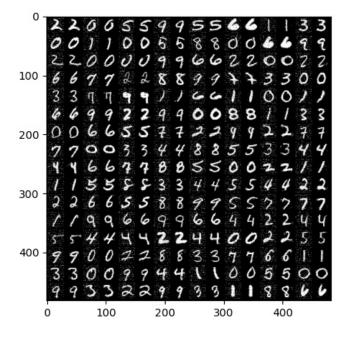
python main.py

To validate, run:

python main.py --play --load_dir=ebm_checkpoint_99.pth

Result:

For the two adjacent grids, the right one is the denoised picture.



2 Normalizing Flows (As Bonus)

Implemented according to [1].

To train, run:

python train.py

To validate, run:

python inpainting.py

Line 48-49, 58-63 are modified to load data and checkpoint correctly.

Result:



3 Variational Autoencoder

Loss function derived from [3].

To train variational autoencoder, run:

python vae.py

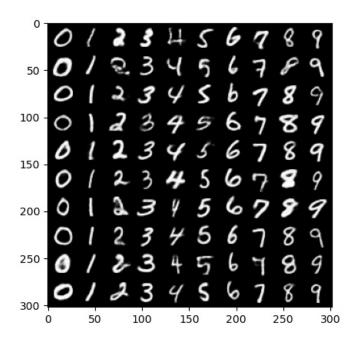
Before this, you need to train an MNIST classifier:

 $\verb"python mnist_classifier.py"$

To validate, run:

python vae.py --eval --load_path=best.pth

Result:



4 Generative Adversarial Network

Loss function derived from [3].

To train variational autoencoder, run:

```
python gan.py
```

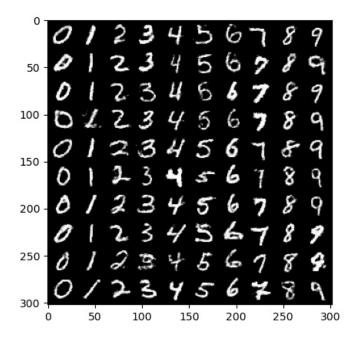
Before this, you need to train an MNIST classifier:

```
python mnist_classifier.py
```

To validate, run:

```
python gan.py --eval --load_path=best.pth
```

Result:



References

- [1] DINH, L., SOHL-DICKSTEIN, J., AND BENGIO, S. Density estimation using real nvp, 2017.
- [2] Du, Y., and Mordatch, I. Implicit generation and generalization in energy-based models, 2020.
- [3] KINGMA, D. P., AND WELLING, M. Auto-encoding variational bayes, 2014.