MLDS 2017 Spring HW3 - Generative Adversarial Networks

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Outline

- Task Introduction
 - Text2image generation
 - Dataset collection
- Model
 - Conditional GAN
 - Tips for training
 - Discriminator loss function
 - Objective function
- Submission and grading

Task Introduction - text2image generation

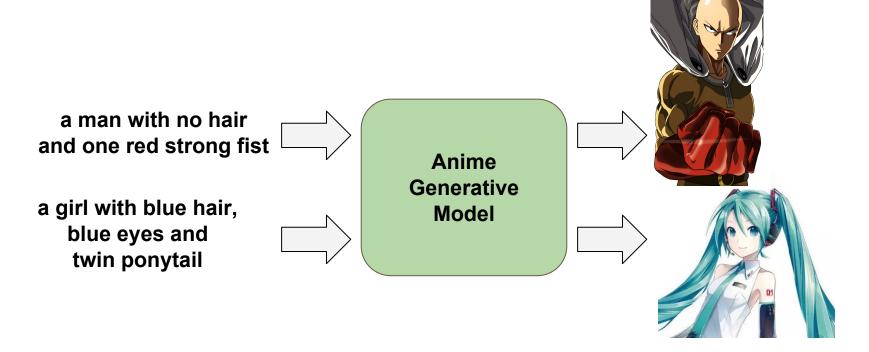
an all black bird

Bird
Generative Model

this flower is white and pink

Flower
Generative Model

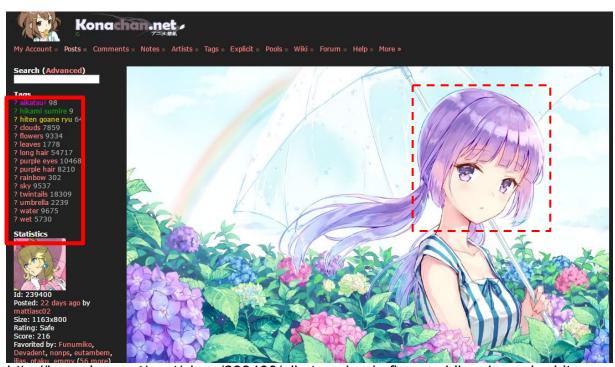
Task Introduction - text2image generation



Data Collection

Tags ? aikatsu! 98 ? hikami sumire 9 ? hiten goane ryu 64 ? clouds 7859 ? flowers 9334 ? leaves 1778 ? long hair 54717 ? purple eyes 10468 ? purple hair 8210 ? rainbow 302 ? sky 9537 ? twintails 18309 ? umbrella 2239 ? water 9675 ? wet 5730

Not all tags are useful

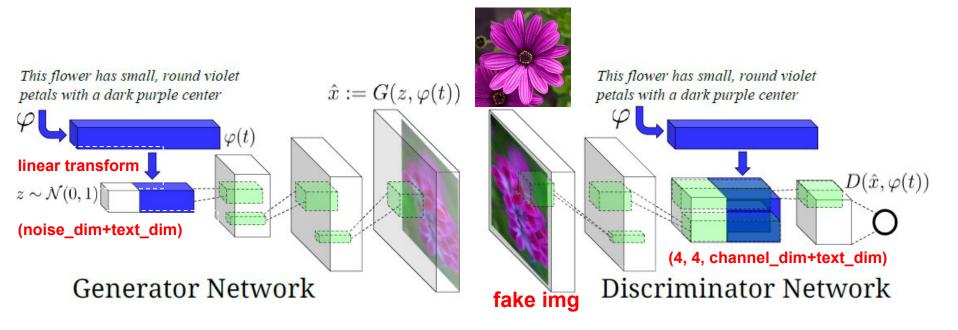


http://konachan.net/post/show/239400/aikatsu-clouds-flowers-hikami_sumire-hiten_goane_r

感謝樊恩宇助教蒐集data

Model and training tips

Conditional GAN for text2image generation real img



Paper: https://arxiv.org/pdf/1605.05396.pdf

Details for training

- Updates between Generator and Discriminator
 - o 1:1 or 2:1
- ADAM with Ir = 0.0002, momentum = 0.5
- gaussian or uniform noise dim = 100
- batch size = 64
- epoch = 600
 - Practically, epoch = 300 is enough

Text feature process tool - Skip-thought vector

```
In [2]: import skipthoughts
In [3]: model = skipthoughts.load_model()
Loading model parameters...
Compiling encoders...
Loading tables...
Packing up...
In [4]: vecs = skipthoughts.encode(model, ['i love machine learning', 'text to image generation yeah'])
4
5
In [5]: vecs.shape
Out[5]: (2, 4800)
```

No matter which tool you use to process text input, please make sure you include that pre-trained model in your repository to let us run your code successfully.

skip-thought source code: https://github.com/ryankiros/skip-thoughts

Image process tool - skimage and scipy.misc

```
In [35]: # convert img to tensor
In [36]: import skimage
In [37]: import skimage.io
In [38]: img = skimage.io.imread('sample.jpg')
In [39]: # resize img
In [40]: import skimage.transform
In [41]: img_resized = skimage.transform.resize(img, (64, 64))
In [42]: img.shape
Out[42]: (96, 96, 3)
In [43]: img_resized.shape
Out[43]: (64, 64, 3)
```

Install:

- sudo apt-get install python-skimage
- sudo pip install --user numpy scipy

Little Demo

input text: black hair blue eyes



input text: pink hair green eyes



input text: green hair green eyes



input text: blue hair red eyes



Tips for training

- Discriminator output:
 - (real img, right text): 1
 - (fake img, right text): 0
 - o (real img, wrong text): 0
 - (wrong img, right text): 0
- Different objective function
 - Wasserstein GAN (WGAN)
 - Least Squares GAN (LSGAN)
 - Boundary-Seeking GAN (BGAN)

Wasserstein GAN

Earth-Mover (Wasserstain-1) distance

$$\begin{split} W(P_r, P_g) &= \inf_{\substack{\gamma \in \Pi(P_r, P_g) \\ |f|_L \leq 1}} E_{(x,y) \sim \gamma[\|x - y\|]} \\ &= \sup_{\substack{\{E \text{M distance is Continuous JS-divergence is not } \\ |f|_L \leq 1}} \underbrace{E_{x \sim P_r[f(x)] - E_{x \sim P_{\theta}}[f(x)]}_{\text{EM distance is Continuous JS-divergence is not }} \\ &= \sup_{\substack{\{E_x \sim P_r[f(x)] - E_{x \sim P_{\theta}}[f(x)]\}}} \underbrace{E_{x \sim P_r[f(x)] - E_{x \sim P_{\theta}}[f(x)]}_{\text{JS-divergence is not }} \end{split}$$

The optimization problem is thus:

$$\max_{w \in W} \{ E_{x \sim P_r}[f_w(x)] - E_{x \sim P_{\theta}}[f_w(x)] \}$$

$$= \max_{w \in W} \{ E_{x \sim P_r}[f_w(x)] - E_{z \sim P_z(z)}[f_w(g_{\theta}(z))] \}$$

Reference: Arjovsky, Martin, Soumith Chintala, and Léon Bottou. "Wasserstein gan." arXiv preprint arXiv:1701.07875 (2017).

Wasserstein GAN

For Discriminator

The output of D is thus not probability anymore!

The D loss turn to be a measure of distance

$$\max_{D} \{ E_{x \sim P_r}[D(x)] - E_{z \sim P_z(z)}[D(G(z))] \}$$

For Generator

$$\max_{G} \{ E_z \sim P_z(z) [D(G(z))] \}$$

Wasserstein GAN

Implementation Notes:

- Do not apply sigmoid at the output of D
- Clip the weight of D
- Use RMSProp instead of Adam
- Train more iteration of D (the paper use 5)

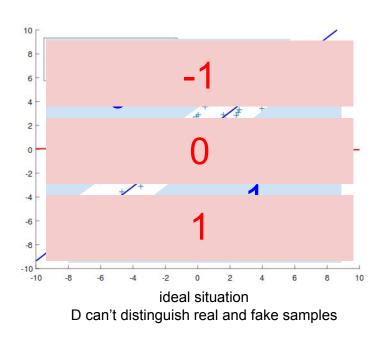
For Discriminator

$$\min_{G} \Big\{ \frac{1}{2} E_{x} \sim p_{data}(x) [(D(x) - b)^{2}] + \frac{1}{2} E_{z} \sim p_{z}(z) [(D(G(z)) - a)^{2}] \Big\}$$

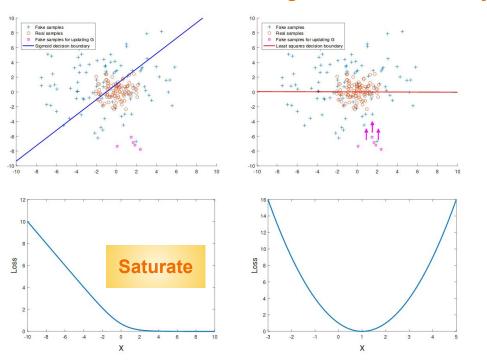
For Generator

$$\min_{G} \frac{1}{2} E_z \sim p_z(z) [(D(G(z)) - c)^2]$$

Reference: Mao, Xudong, et al. "Least squares generative adversarial networks." arXiv preprint ArXiv:1611.04076 (2016).



Drag to decision boundary



For Discriminator

$$\min_{G} \left\{ \frac{1}{2} E_{x} \sim p_{data}(x) [(D(x) - b)^{2}] + \frac{1}{2} E_{z} \sim p_{z}(z) [(D(G(z)) - a)^{2}] \right\}$$

For Generator

$$\min_{G} \frac{1}{2} E_z \sim p_z(z) [(D(G(z)) - c)^2]$$

Implementation Notes:

Do not apply sigmoid at the output of D

Boundary-Seeking GAN

Optimum Discriminator : $D^*(x) = \frac{P_{data}(x)}{P_{data}(x) + P_{a}(x)}$

$$P_{data}(x) = P_g(x) \frac{D^*(x)}{1 - D^*(x)}$$

For imperfect discriminator :
$$\widetilde{P}(x) = \frac{1}{Z} P_g(x) \frac{D(x)}{1 - D(x)}$$

Z: normalize term

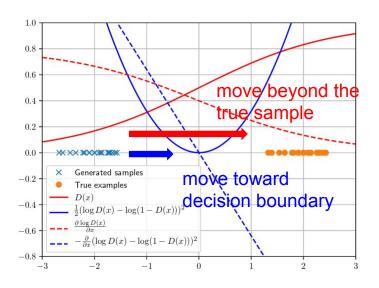
Boundary-Seeking GAN

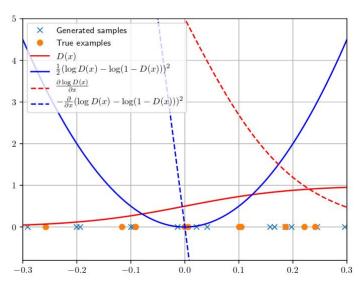
For Generator

$$\min_{G} E_{z} \sim P_{z(z)}[(log(D(G(z)) - log(1 - D(G(z)))^{2}]$$

Only change the loss function of G

Boundary-Seeking GAN





Submission and Grading

Homework 3 package

- Anime Dataset:
 - o training data: 33.4k (image, tags) pair
 - faces/, tags.csv, sample_testing.txt
- training tags file format
 - o img_id <comma> tag1 <colon> #_post <tab> tag2 <colon> ...

- testing text file format
 - testing_text_id <comma> testing_text

```
1, blue hair blue eyes
2, blue hair green eyes
3, blue hair red eyes
4, green hair blue eyes
testing_text.txt
```

blue eyes

short hair

red hair

tags.csv

- testing text only includes 'color hair' and 'color eyes', only alphabetic char involved.
- Data download link:
 - https://drive.google.com/open?id=0BwJmB7alR-AvMHEtczZZN0EtdzQ
 - If you want to do something cool beyond generating faces, mail us. We will give you original images

Submission

- Only Python with Tensorflow r1.0 for GAN
 - You are allowed to use any tool to process text
 - You are allowed to use any data you collect
- Deadline: 5/25(Thu.) 23:59:59 (UTC+8)
- MLDS2017/hw3 should contain following files
 - run.sh train.py, (pre-)trained_model, generate.py, samples/, report.pdf
 - If some files are too big, upload to your cloud and download them when running your run.sh
- TAs will run your run.sh to generate images given a text
 - bash run.sh [testing_text.txt]
 - run.sh must output in 10 minutes.

Submission on Github

- Only one branch master is needed
- master stores the model by using GAN structure
- Remember to put your pre-trained models or download scripts so that we can run your code successfully

Output Format Requirement

- The generated images should be in Directory samples/
 - o make sure it's **empty** before we run your code
- Each generated image must be 64 x 64 in size
 - o please resize all training images to 64 x 64 before training the model
- For each input text, you must generate 5 images
- Generated img should be named as
 - "sample_(testing_text_id)_(sample_id).jpg"
- Example:

```
andy@andy-All-Series|x86_64:samples:4$ ls
sample_1_1.jpg sample_1_3.jpg sample_1_5.jpg sample_2_2.jpg sample_2_4.jpg
sample_1_2.jpg sample_1_4.jpg sample_2_1.jpg sample_2_3.jpg sample_2_5.jpg
```

組別互評

- We will put your generated images in the grading platform
- Link will be sent to your mail after HW deadline
- Answer 2 scores for each image
 - How the image fits the text
 - How the image looks real
- Scores should be integer from 1 to 5
 - 1 to 5 corressponding to (super bad, bad, average, good, super good)
- You may score your results, so be fair when your are scoring :)

組別互評

- Separate scores with a comma (score for matching text, score for reality)
- Example:
 - 3 → Gray hair green eyes



4, 5



What report should cover?

- Environment (1%)
 - Ex. OS, CPU, GPU, Memory, libraries you used and version, etc.
- Model description(3%)
 - Must include model strucuture, objective function for G and D
- How do you improve your performance (5%)
- Experiment settings and observation (5%)
- Team division (1%)
- No more than 5 pages
- Please written in Chinese (unless you don't know how to type Chinese)

Grading Policy (20%)

- Wrong output format will not be graded
- Report (10%)
- Code (5%)
 - You will be scored only if you use GAN and output results in 10 minutes
- Baseline
- 組別互評 (5%)
- Bonus
 - 組別互評 top 3 (5%, 3%, 2%)
 - Generate images beyond faces and show results in report (10%)

Other Policy

- Late policy: 30% off per day late afterwards. [Delay form will be announced afterwards]
- No plagiarism is allowed.