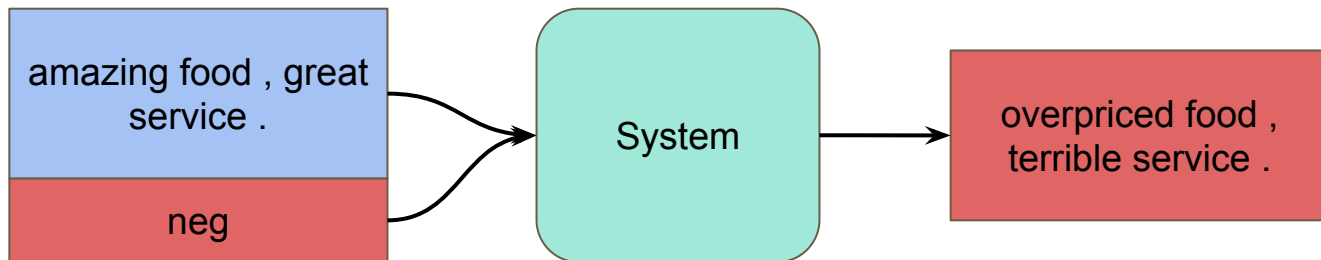

HW5-1 Text Style Transfer

TA 張致強 蔡翔陞 鍾起鳴

5.1.1 Style-Transformer

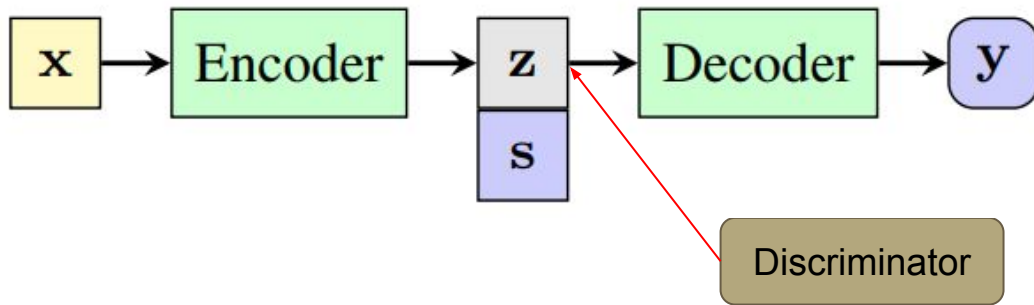
Text Style Transfer



Typical Unsupervised Approach

Disentangle **content** and **style** in the latent space.

A discriminator is used to make sure z is style independent.



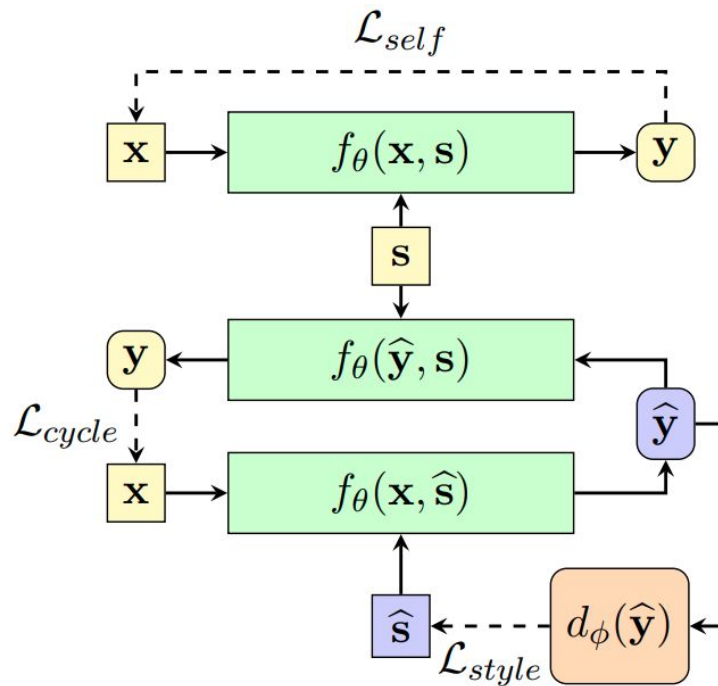
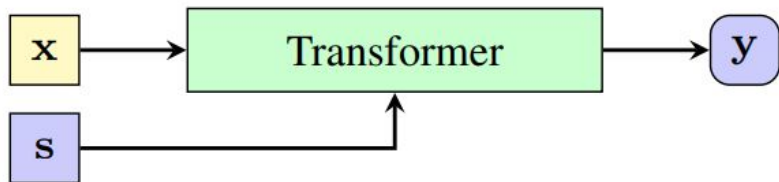
Style Transformer: Unpaired Text Style Transfer without Disentangled Latent Representation

— Ning Dai, Jianze Liang, Xipeng Qiu* , —
Xuanjing Huang

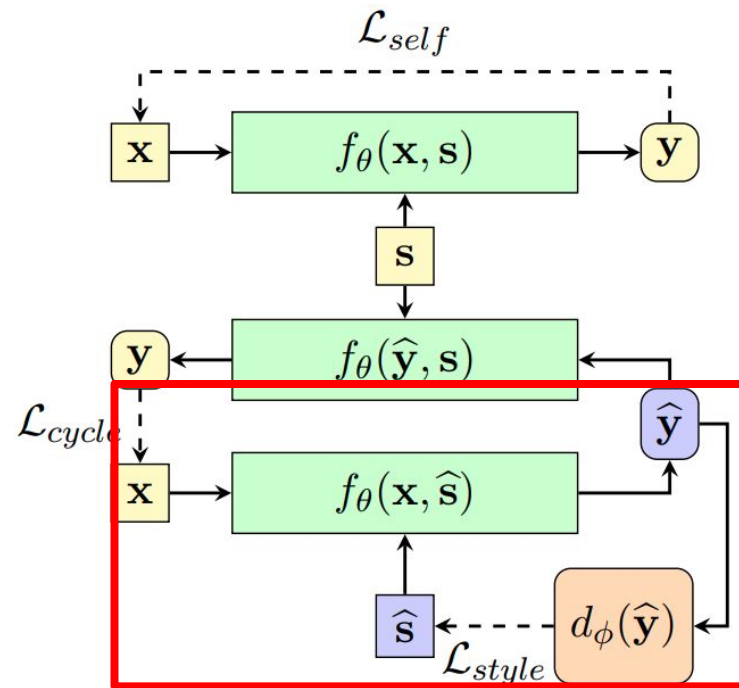
Fudan University

Approach

- No disentanglement
- No compressed representation



GAN!



Training Style-Transformer

- Self Reconstruction

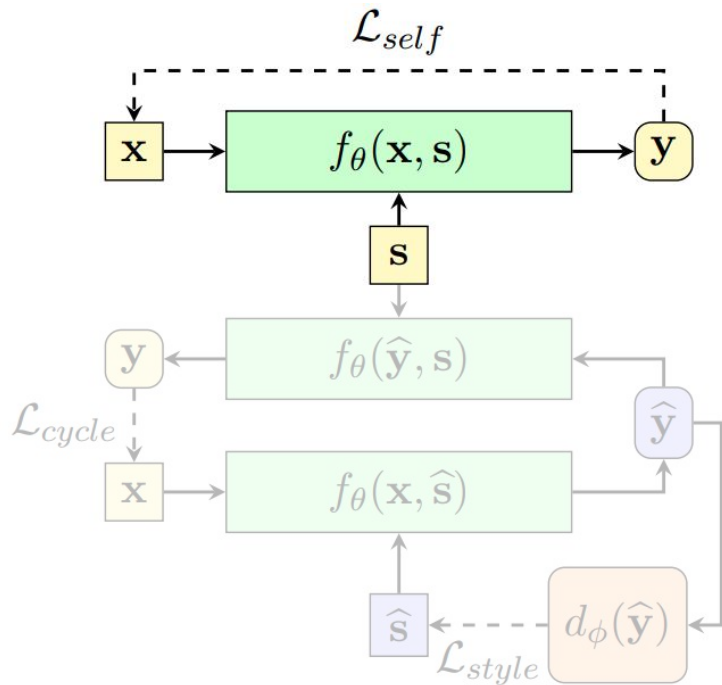
$$\mathcal{L}_{self}(\theta) = -p_{\theta}(\mathbf{y} = \mathbf{x} | \mathbf{x}, \mathbf{s})$$

- Cycle Reconstruction

$$\mathcal{L}_{cycle}(\theta) = -p_{\theta}(\mathbf{y} = \mathbf{x} | f_{\theta}(\mathbf{x}, \hat{\mathbf{s}}), \mathbf{s})$$

- Style Controlling

$$\mathcal{L}_{style}(\theta) = -p_{\phi}(\mathbf{c} = \hat{\mathbf{s}} | f_{\theta}(\mathbf{x}, \hat{\mathbf{s}}))$$



Training Style-Transformer

- Self Reconstruction

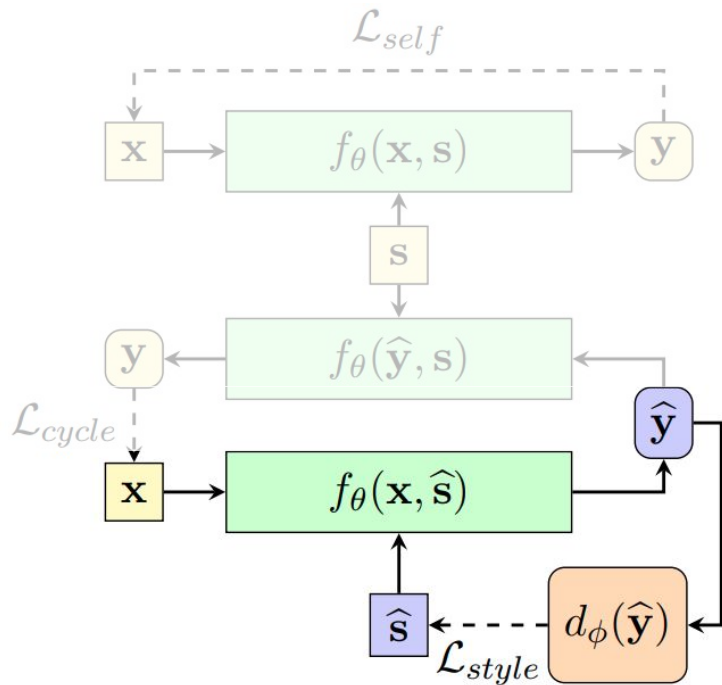
$$\mathcal{L}_{self}(\theta) = -p_{\theta}(\mathbf{y} = \mathbf{x} | \mathbf{x}, \mathbf{s})$$

- Cycle Reconstruction

$$\mathcal{L}_{cycle}(\theta) = -p_{\theta}(\mathbf{y} = \mathbf{x} | f_{\theta}(\mathbf{x}, \hat{\mathbf{s}}), \mathbf{s})$$

- Style Controlling

$$\mathcal{L}_{style}(\theta) = -p_{\phi}(\mathbf{c} = \hat{\mathbf{s}} | f_{\theta}(\mathbf{x}, \hat{\mathbf{s}}))$$



Training Style-Transformer

- Self Reconstruction

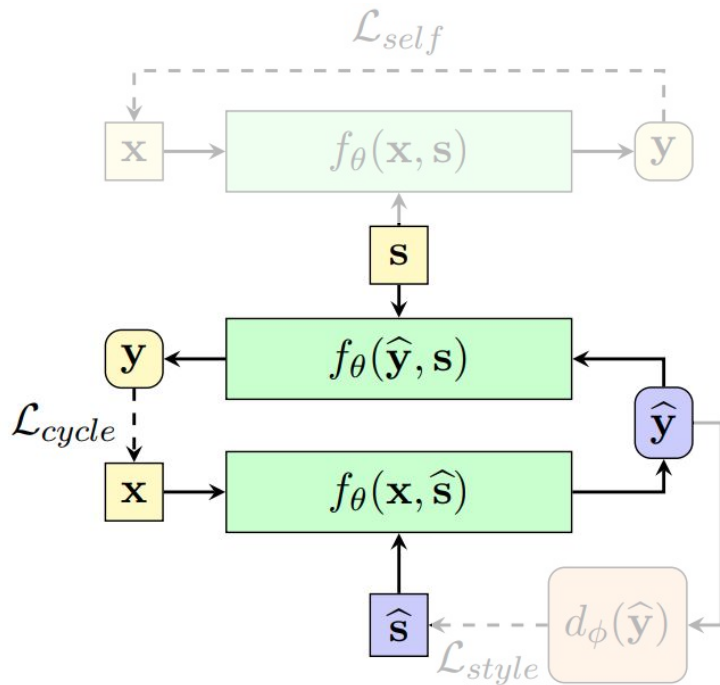
$$\mathcal{L}_{self}(\theta) = -p_{\theta}(\mathbf{y} = \mathbf{x} | \mathbf{x}, \mathbf{s})$$

- Cycle Reconstruction

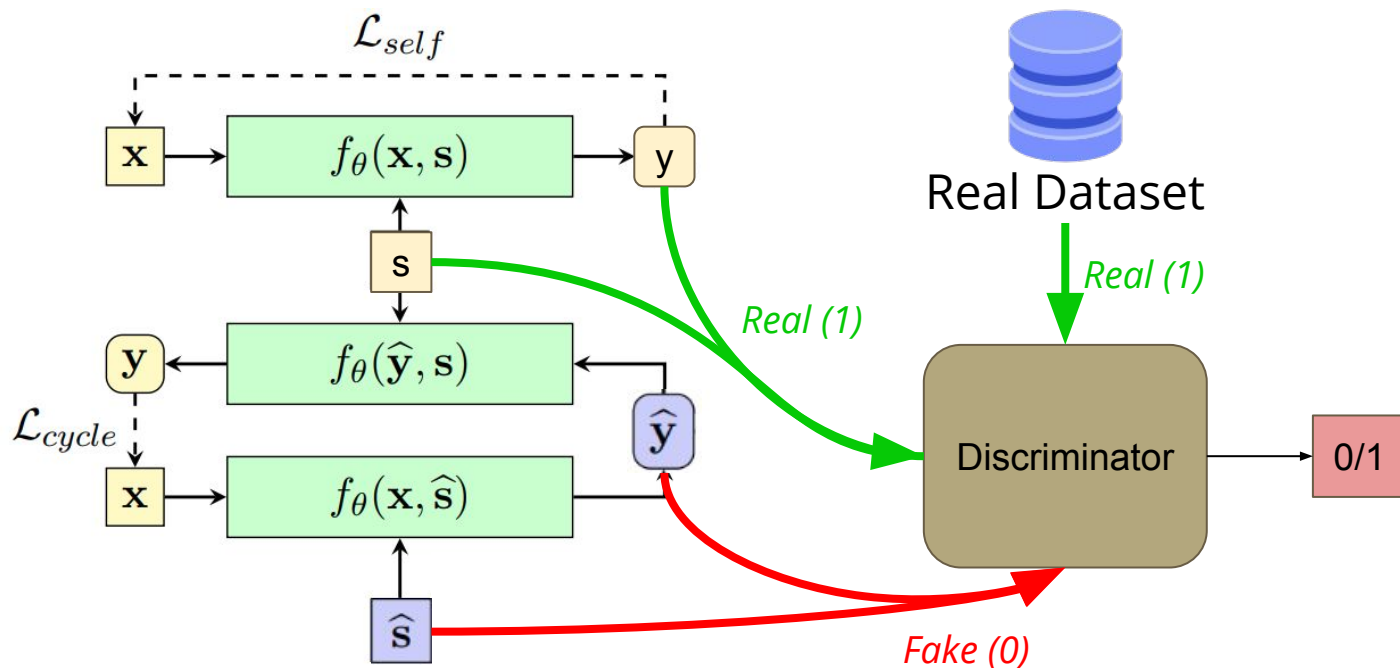
$$\mathcal{L}_{cycle}(\theta) = -p_{\theta}(\mathbf{y} = \mathbf{x} | f_{\theta}(\mathbf{x}, \hat{\mathbf{s}}), \mathbf{s})$$

- Style Controlling

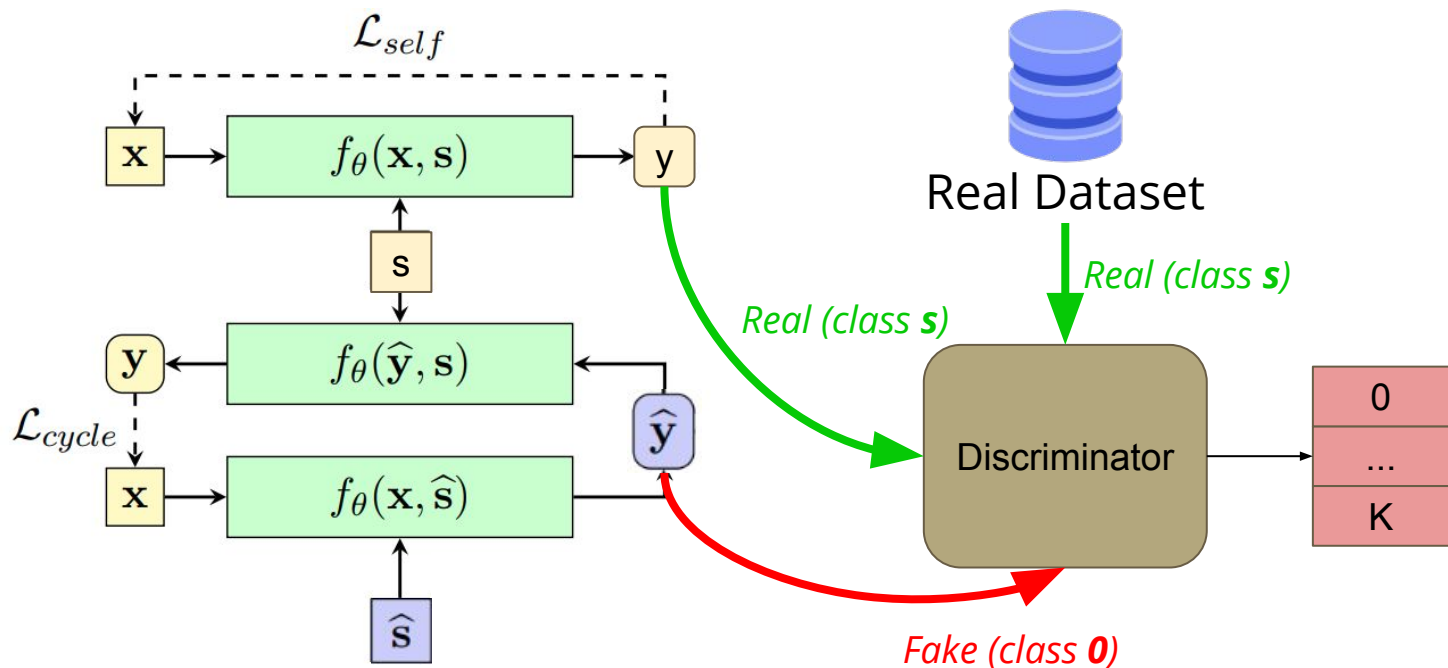
$$\mathcal{L}_{style}(\theta) = -p_{\phi}(\mathbf{c} = \hat{\mathbf{s}} | f_{\theta}(\mathbf{x}, \hat{\mathbf{s}}))$$



Training the Discriminator (Conditional)



Training the Discriminator (Multiclass)



Evaluating Text Style Transfer System

- Accuracy: Classification accuracy of transferred sentence.
- BLEU: Content preservation.
 - self-BLEU: score by source sentence
 - ref-BLEU: score by reference (target) sentence
- Perplexity: Sentence fluency.
- Human Evaluation

5.1.2 Tasks

Code

Github - <https://github.com/MarvinChung/HW5-TextStyleTransfer>

See README.md for instructions to run.

Configurations

You can (should) tune some of the following args to obtain better results.

Model related

argument	description
discriminator_method	Two <u>types</u> of discriminator described in paper.
embed_size, d_model, head, num_layers	Transformer architecture.
learned_pos_embed	Use trainable position encoding or sinusoidal.

Training related

argument	description
batch_size, lr_F, lr_D	Typical training hyperparameters.
L2, dropout, inp_drop_prob	Various regularizations.
iter_D, iter_F, F_pretrain_iter	Affects relative strengths of style-transformer and discriminator while training.
slf_factor, cyc_factor, adv_factor	Weight factors of three loss described in paper.

What you should do

1. Understand and train a Text Style Transfer model
2. Show your training configuration
 - a. show what's different from default
3. Show your training curves
 - a. Plot the three loss in style-transformer
 - b. Plot the discriminator loss
4. Evaluate your model
 - a. Report your best model's [accuracy, ref-BLEU and perplexity](#)
 - b. Submit your model's output on the 1000 yelp testing data
 - c. Show some results on sentences inside or outside of the yelp dataset

5.1.3 Sample Report

HW5-1

— B05902064 張致強 —

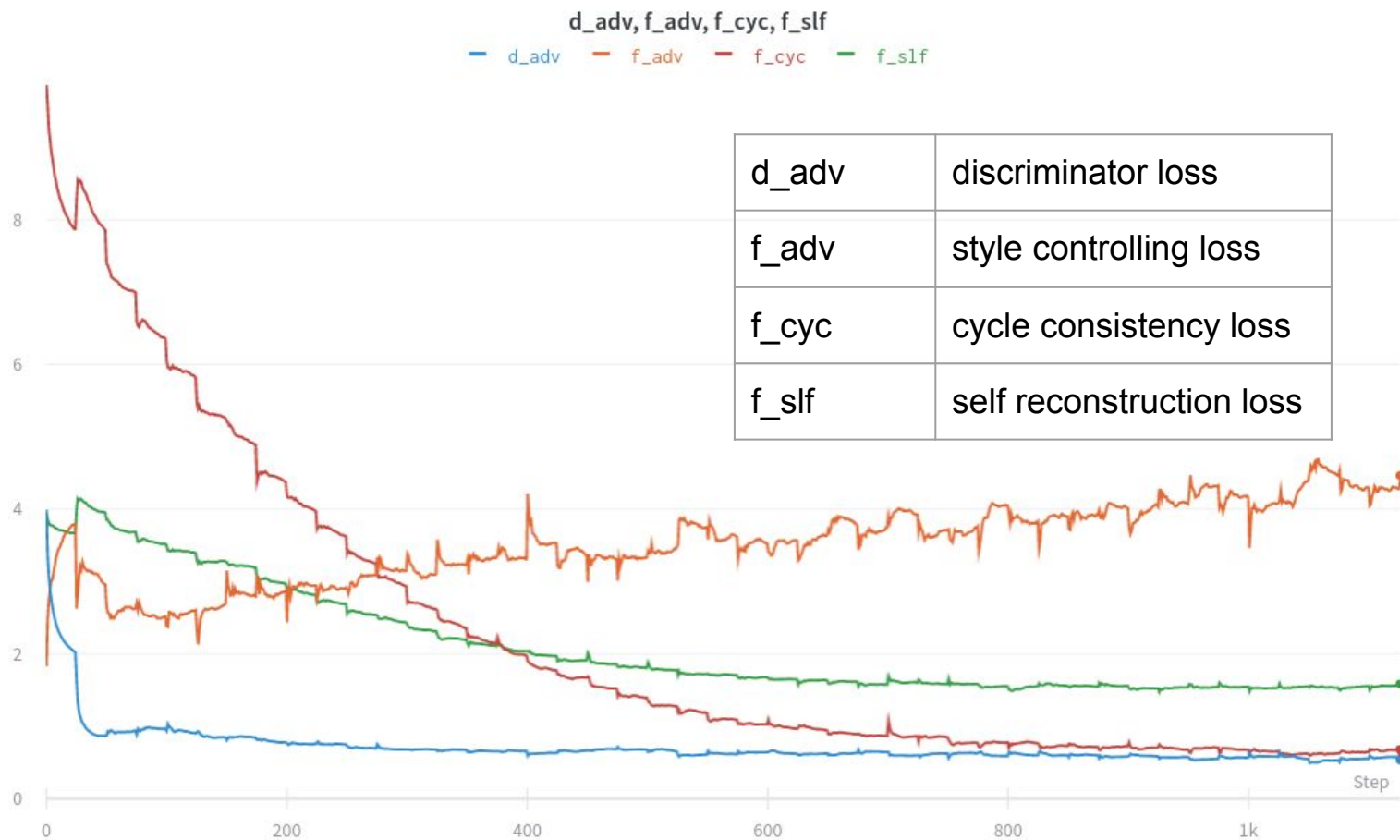
Configuration

We use the sinusoidal positional embedding. Also changed batch size to 128. Other settings follows the default.

Command

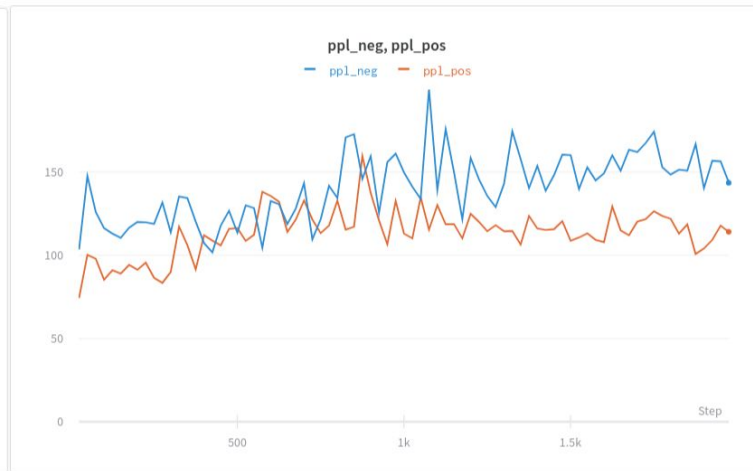
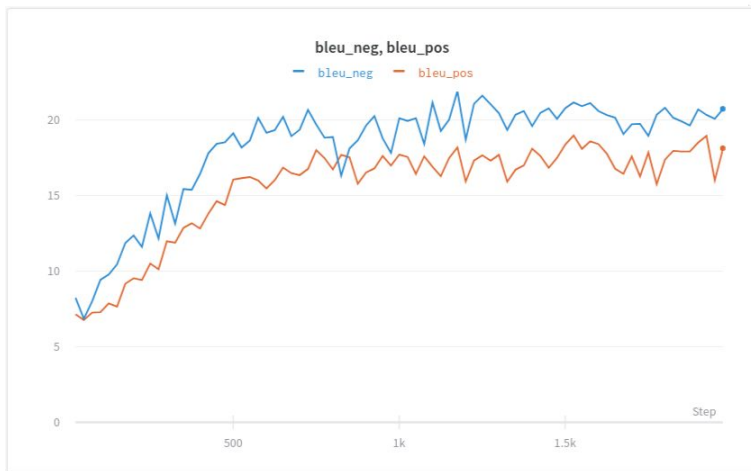
```
main.py --do_train -learned_pos_embed False -batch_size 128 --use_wandb
```

Loss



Evaluation - Metrics

	pos	neg	avg
accuracy	0.902	0.784	0.843
ref-bleu	18.132	20.734	19.433
perplexity	114.031	143.505	128.768



Evaluation - Good Examples

```
***** pos sample *****  
[gold] great quality , great service .  
[raw ] great quality , great service .  
[rev ] terrible quality , terrible service .
```

```
***** neg sample *****  
[gold] there chips are ok , but their salsa is really bland .  
[raw ] there chips are ok , but their salsa is really bland .  
[rev ] there chips are ok , but their salsa is really excellent .
```

```
***** pos sample *****  
[gold] the menudo here is perfect .  
[raw ] the menudo here is perfect .  
[rev ] the menudo here is crap .
```

Evaluation - Problematic Examples

```
***** pos sample *****  
[gold] the crispy mango fish was very tasty .  
[raw ] the crispy mango fish tastv very tastv .  
[rev ] the crispy mango fish was very disgusted .
```

food would be disgusting, and person disgusted.

```
***** neg sample *****  
[gold] i left hungry and unsatisfied , never again .  
[raw ] i left hungry and unsatisfied , never again .  
[rev ] i beyond hungry and honest , will again .
```

beyond hungry means very+++ hungry

Observation (optional)

- Despite being a text GAN, the training process is surprisingly stable using this set of hyperparameters.
- Our model seems to perform better on pos->neg transfer.

5.1.4 Human Evaluation

Procedure

- Each team will submit their outputs on testset (submission.txt)
- After the due date June 24, TA will announce the 10 sentences to be evaluated.
- Each team will review 2 other teams' output, totaling 20 reviews
- Peer review form will due on June 30 (23:59), no late submission
 - You can get 0.5% just by reviewing others !

submission.txt

Generate style transferred sentences on the provided yelp test set (1000 sentences), in the same order as the test set.

Do not change the **order** of the output sentences! TA will not order for you, sorry QQ.

```
ever since joes has grand hands it 's just gotten margaritas and margaritas .  
there is definitely not enough room in that part of the venue .  
so basically comfortable moist teaching .  
she works she 'd be back and lauren for a few minutes .  
i ca n't while how efficient this pharmacy is .  
just beyond and bright it high the bill .  
it is n't terrible , but it is n't very good either .  
definitely recommended that i could not use my birthday gift !  
new owner , i heard - but i bring to know the details .  
but it probably heaven too !  
we sit down and we got some really fast and always service .  
the live top include miso soup and a small salad .  
there was great i 'm honestly or how did everything come out .  
works we nails to sit at the table if we are to ordering dinner .  
the cash register area tasty offer and great place was watching the store front .  
there chips are ok , but their salsa is really excellent .  
the wine was very fantastic and the food was even world .  
staffed primarily by teenagers that bring their understand customer service .  
the burgers are usual cooked to the point the meat tasty crunchy .  
blue cheese selections tasty their the best by any means .  
my pad thai comfortable amazing thai rice noodles with barbeque sauce .  
she said `` yes , honestly .  
the store is brandon looking and staff family to change .  
there was actually meat and bread .  
when i handled in a polite cuisine , i was decent out the door .  
she was good happy being there .  
moving past the shape , they were dry and truly nicely .  
the associates program is great lunch an option .
```

Plain Text ▼ Tab Width: 8 ▼

5.1.5 Submission & Grading

Submission

- Github (DLHLP2020-SPRING/hw5/)
 - report.pdf
 - submission.txt

Grading

- Report (4%)
 - show configuration (1%)
 - show loss plot (1%)
 - show metrics (1%)
 - show examples & comment on them (1%)
- Results (2% + 1% bonus)
 - Your auto-metrics are as good or better than baseline (1%)
 - Accuracy and BLEU: the higher the better.
 - Perplexity: the lower the better.
 - within $\pm 5\%$ is considered “as good”.
 - You submitted *submission.txt* **that is intact** (1%)
 - Human evaluation
 - you reviewed others' results (0.5%)
 - your quality is in top 10% of class (0.5%)

baseline	(pos+neg)/2
accuracy	0.7
ref-bleu	17
perplexity	150

Q&A