

Supplementary Material: Content preserving text generation with attribute controls

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A Qualitative comparison

Restaurant Reviews	
negative → positive	
Query	<i>sorry but i do n't get the rave reviews for this place .</i>
Ctrl gen	i ordered the nachos , have perfect seasonal beans on amazing amazing .
Cross-align	sorry , i do n't be the best experience ever .
Ours	thanks but i love this place for lunch .
Query	<i>however my recent visit there made me change my mind entirely .</i>
Ctrl-gen	not like other target stores .
Cross-align	best little one time to go for in charlotte .
Ours	overall my experience here was great as well .
Query	<i>okay so this place has been a pain even after i already moved out .</i>
Ctrl gen	like i mentioned , i thought to this these fun .
Cross-align	food and this place has been a good place to be back .
Ours	overall this is a great place to go when i 'm in town .
Query	<i>personally i 'd rather spend my money at a business that appreciates my business .</i>
Ctrl gen	i became quite a gem at the beginning but we amazing fantastic amazing .
Cross align	then i will be back my time to get a regular time .
Ours	definitely i 'll definitely be back for a good haircut .
positive → negative	
Query	<i>best chinese food i 've had in a long time .</i>
Ctrl gen	very lousy texture and ruined .
Cross align	worst chinese food i 've had in a long in years .
Ours	worst food i 've had in a long time .
Query	<i>my appetizer was also very good and unique .</i>
Ctrl gen	both were ruined . ruined
Cross align	my wife was just very bland and no flavor .
Ours	my chicken was very dry and had no flavor .
Query	<i>everything tasted great and the service was excellent .</i>
Ctrl gen	but the real pleasure is the service department .
Cross align	everything tasted horrible and the service was very bad .
Ours	everything tasted bad and the service was horrible .
Query	<i>atmosphere is cozy and comfortable .</i>
Ctrl gen	atmosphere is not good .
Cross align	rude is dirty and way in .
Ours	restaurant is dirty and dirty .

Table 1: Query sentences from the restaurant reviews test dataset modified with opposite sentiment by Ctrl gen [1], Cross-align [2] and our model, respectively.

Movie Reviews	
negative → positive	
Query	<i>this is the most vapid movie i have ever seen .</i>
Ctrl gen	if this grabs your interest , you may want to give it a try
Cross-align	this is a great movie that is so good .
Ours	this is the most beautiful movie i have ever seen .
Query	<i>this 1944 film is too awful as it 's just incredible .</i>
Ctrl gen	<unk> the three dead world and <unk> 's <unk> is a cult in a life
Cross-align	this film is one of the best movies ever made .
Ours	this film is an excellent and it is definitely worth it .
Query	<i>1 out of 10 .</i>
Ctrl gen	he 's cold and hateful exactly what his part <unk>
Cross-align	my rating of the cast .
Ours	10 out of 10 .
Query	<i>i always thought she was a colorless , plain jane .</i>
Ctrl gen	a great comedy all wrapped up in a tiny package !
Cross-align	i think that is the best of the film .
Ours	i also thought she was a beautiful , talented actor .
Query	<i>her character is truly hateful and her acting , if you can call it that , is strictly wretched .</i>
Ctrl gen	a great ' proper ' summer movie
Cross-align	<unk> , is the <unk> , and you can be able to be more than it to be .
Ours	his character is very funny , and in fact , it 's just what he does n't disappoint .
positive → negative	
Query	<i>this is one of his best efforts .</i>
Ctrl gen	as david <unk> picked up the franchise , it has just <unk> to pieces
Cross-align	this is a complete waste of time .
Ours	this is one of the worst films .
Query	<i>if you love silent films , you 'll adore this one .</i>
Ctrl gen	nice photographic effects as jessica <unk> the process
Cross-align	if you 're no , but it is not bad .
Ours	if you love horror movies , do n't see this one .
Query	<i>and congratulations to kino for a superb video restoration .</i>
Ctrl gen	peter <unk> is not that she 's not gone bad movie
Cross-align	but then , it 's a waste of time .
Ours	and save your money on this piece of garbage .
Query	<i>the characters are portrayed vividly and realistically .</i>
Ctrl gen	problem is , not enough good work went into this
Cross-align	the characters are <unk> and <unk> .
Ours	the characters are completely unsympathetic and annoying .
Query	<i>there are some of the most stunning and grisly combat scenes ever filmed .</i>
Ctrl gen	unfortunately the only thing you see is <unk>
Cross-align	there is no a <unk> , and the <unk> , <unk> and <unk> .
Ours	there are some of the most boring and boring scenes ever made .

Table 2: Query sentences from the movie reviews test dataset modified with opposite sentiment by Ctrl gen [1], Cross-align [2] and our model, respectively.

B Human Evaluation

Sections B.1, B.2, B.3 describe the setup for human evaluations done using Amazon Mechanical Turk (AMT) to obtain annotations for generated sentences.

B.1 Content compatibility

Given a reference sentence and a set of candidate sentences, pick the candidates that have the **same semantic content as the reference sentence but have the opposite sentiment** (i.e., mean the opposite). Select all that apply. If you think neither of the given sentences have this property, choose *No preference* (This can happen when all the candidate sentences are either semantically irrelevant to the reference sentence or have the incorrect sentiment).

Example:

Reference sentence: *This is a great movie !*

You would pick sentences such as

✓ *This is not a good movie.*

✓ *This is a bad movie.*

The following sentences do not fit the criteria because they are either semantically irrelevant to the reference sentence or have the incorrect sentiment.

✗ *I did not like the salad.*

✗ *This is a wonderful movie.*

B.2 Attribute compatibility

Pick the best sentiment based on the following criterion.

<i>Sentiment</i>	<i>Guidance</i>
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Positive	Sentence conveys positive sentiment. Eg: "I really liked the food."
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Negative	Sentence conveys negative sentiment. Eg: "This was the worst experience ever."
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Neutral	Sentence does not carry any sentiment information.
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B.3 Fluency

Rate the grammaticality/fluency of the sentence based on the following criterion.

<i>Fluency</i>	<i>Guidance</i>
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5	The sentence is grammatical and does not have any grammar errors.
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4	Sentence is mostly grammatical except for one/two mistakes.
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3	Parts of the sentence are grammatical and sentence is somewhat coherent, but there are glaring errors.
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2	Too many grammatical errors and sentence is incoherent.
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1	Sentence is completely ungrammatical.
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C Sampling strategy

In this section we compare soft and hard sampling during training. For the soft-sampling model, we use an exponential decay temperature annealing schedule with an initial temperature of 1. The temperature decays until it reaches 0.01 and remains constant afterwards. Other parameters of the model are identical to section 4.1 of the paper. We use the Yelp dataset for this experiment. Table 3 compares the models with respect to the metrics in section 4.2.

Models learned with soft-sampling produce sentences judged to be highly attribute compatible. However, the content compatibility is considerably poor and generated sentences have lower fluency. This supports our claim that the training and inference behavior are mismatched when soft-sampled sequences are used for training.

Model	Attribute ↑ compatibility	Content ↑ compatibility		Fluency ↓ Perplexity
		BLEU-1	BLEU-4	
Hard-sampling	90.50%	53.0	7.5	133
Soft-sampling	92.33%	43.6	3.1	137

Table 3: Comparison between soft and hard sampling. Higher numbers are better for accuracy and content compatibility and lower numbers are better for perplexity.

References

- [1] Zhiting Hu, Zichao Yang, Xiaodan Liang, Ruslan Salakhutdinov, and Eric P Xing. Controllable text generation. *arXiv preprint arXiv:1703.00955*, 2017.
- [2] Tianxiao Shen, Tao Lei, Regina Barzilay, and Tommi Jaakkola. Style transfer from non-parallel text by cross-alignment. *arXiv preprint arXiv:1705.09655*, 2017.