Neural Machine Translation of Text from Non-Native Speakers

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NAACL 2019
The sad truth

Non-native English speakers outnumber native ones by:

3:1

[Crystal, 2003]

…and we make "mistakes"

Yet, can our models handle non-standard input?

Do our datasets cover non-standard English?
Machine Translation and Noise

Yonatan2: "Synthetic and Natural Noise both break NMT"

If you can raed tihs, you msut be ralely samrt

…but the MT system is not so it struggles to translate it!
This is not such a big deal due to the restart system, but is enough to spoil the fun.
This is not such a big deal due to the restart system, but is enough to spoil the fun.

No se trata de mucho debido as sistema de reinicio, sino que basta con echar por tierra el fun.
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This is not such a big deal due to the restart system, but is enough to spoil of the fun.
Machine Translation and Noise

**eng:** This is not such a big deal due to the restart system, but is enough to spoil the fun.

**spa:** No se trata de mucho debido al sistema de reinicio, sino que basta con echar por tierra el fun.

**eng:** This is not such a big deal due to the restart system, but is enough to spoil of the fun.

**spa:** No se trata de Esto no es mucho debido al sistema de reinicio, sino que basta con echar echarnos por tierra el fun.
Working with Realistic Noise

Grammar Error Correction Corpora

English from non-native speakers with corrections.

Create confusion sets for some error types:

\[ p(\text{error} | \text{correct}) \]

<table>
<thead>
<tr>
<th></th>
<th>a</th>
<th>an</th>
<th>the</th>
<th>Ø</th>
</tr>
</thead>
<tbody>
<tr>
<td>a</td>
<td>0.01</td>
<td>0.18</td>
<td>0.8</td>
<td></td>
</tr>
<tr>
<td>an</td>
<td>0.06</td>
<td>0.14</td>
<td>0.78</td>
<td></td>
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<tr>
<td>the</td>
<td>0.02</td>
<td>0.0</td>
<td>0.96</td>
<td></td>
</tr>
<tr>
<td>Ø</td>
<td>0.08</td>
<td>0.01</td>
<td>0.9</td>
<td></td>
</tr>
</tbody>
</table>
Working with Realistic Noise

We added a **single** error on a WMT English test set

- **clean:** Yet the debit market is breaking records today.
- **typo:** Yet the debit **maret** is breaking records today.
- **article:** Yet the debit market is breaking records the today.
- **prep:** Yet the debit market is breaking records in today.
- **noun num:** Yet the debit market is breaking record today.
- **sva:** Yet the debit market are breaking records today.
Testing on Grammatical Noise

WMT Test Set

<table>
<thead>
<tr>
<th>Clean</th>
<th>Typo</th>
<th>Article</th>
<th>Preposition</th>
<th>Noun Number</th>
<th>Subject-Verb Agreement</th>
</tr>
</thead>
<tbody>
<tr>
<td>33.0</td>
<td>-3.4</td>
<td>-1.7</td>
<td>-1.0</td>
<td>-3.7</td>
<td>-0.9</td>
</tr>
</tbody>
</table>

Train on Clean
Solution

Add noise to the training data…

…and train with the synthesized data.

First, add a single type of error.
1. Improvements only on the respective test set
2. Worse performance on clean data
Solution

Add noise to the training data…

…and train with the synthesized data.

…and train on clean and synthesized data (concatenated).

First, add a single type of error.
Testing on Grammatical Noise

<table>
<thead>
<tr>
<th>Clean</th>
<th>Typo</th>
<th>Article</th>
<th>Preposition</th>
<th>Noun Number</th>
<th>Subject-Verb Agreement</th>
</tr>
</thead>
<tbody>
<tr>
<td>Train on Clean</td>
<td>Train on Typo</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>33.0</td>
<td>33.0</td>
<td>-0.1</td>
<td>-2</td>
<td>-2.8</td>
<td>-3.4</td>
</tr>
<tr>
<td>-1.25</td>
<td>-1.25</td>
<td>-1.6</td>
<td>-1.7</td>
<td>-1.6</td>
<td>-3.4</td>
</tr>
<tr>
<td>-2.5</td>
<td>-2.5</td>
<td>-3</td>
<td>-3</td>
<td>-3</td>
<td>-3.4</td>
</tr>
<tr>
<td>-3.75</td>
<td>-3.75</td>
<td>-3.7</td>
<td>-3.5</td>
<td>-3.7</td>
<td>-4.7</td>
</tr>
<tr>
<td>-5</td>
<td>-5</td>
<td>-0.9</td>
<td>-2.4</td>
<td>-3.5</td>
<td>-4.7</td>
</tr>
</tbody>
</table>
Add noise to the training data…

…and train with the synthesized data.

…and train on both clean and synthesized data.

First, add a single type of error.

… adding all types of errors.
Testing on Grammatical Noise

WMT Test Set

Clean | Typo | Article | Preposition | Noun Number | Subject-Verb Agreement
--- | --- | --- | --- | --- | ---
33.0 | -0.3 | -1.1 | -1.0 | -1.0 | -0.9
-1.0 | -1.7 | -1.6 | -1.0 | -1.0 | -1.0
-2.4 | -3.4 | -3.7 | -2.4 | -2.4 | -2.4

Train on Clean

Average BLEU: 31.2 ± 1.5
Working with Real Noise

JHU Fluency-Extended GUG Corpus (JFLEG)

Prompt:

*Please translate the following sentences. Note that some sentences will have grammatical errors or typos in English. Don’t try to translate the sentences word for word (e.g. replicate the error in Spanish).*
Testing on Real Grammatical Noise

JFLEG-ES Test Set

- 2.4 BLEU loss due to noise
- What if we used a GEC system as a pre-processing step?
- Small loss on clean, but improvement on noisy
Summary

We need to account for non-native speakers and language learners.

We need to create appropriate datasets. We provide:

1. train, dev, and test WMT datasets with realistic noise on the English side on 8 language pairs: {cs,de,et,fi,ru,tr,zh}-en

2. JFLEG-ES: an evaluation set (dev+test) with professional translations in Spanish
   Available at: https://bitbucket.org/antonis/nmt-grammar-noise/

Towards robustness:

   combine synthesized noisy data with clean ones