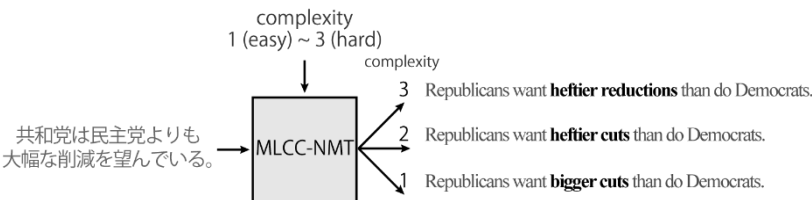


Introduction

- Style-controllable NMT has recently received much attention.
- Multi-Level Complexity-Controllable MT (MLCC-MT)
 - Controls the complexity of a target language sentence at three or more levels
 - To allow translation tailored to the user's reading level

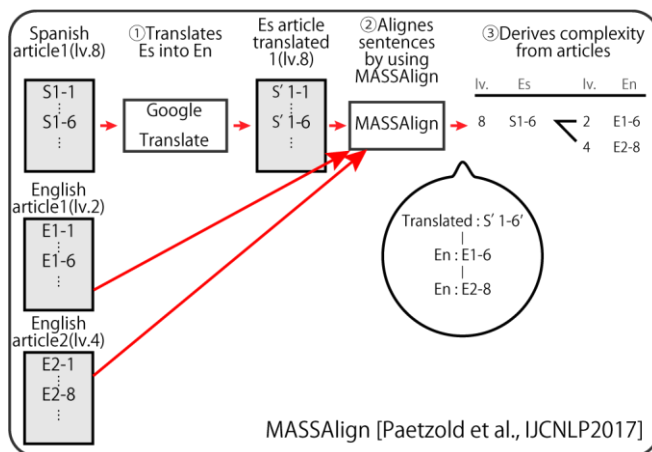


Problem: Existing test dataset cannot precisely measure model performance

- **Objectives:**
 - Construct a test dataset to properly evaluate MLCC-MT models
 - Provide benchmark performance by evaluating two MLCC-MT models (i.e., pipeline and multi-task models)

Problems of Existing Test Dataset [Agrawal et al., EMNLP 2019]

- Es⇔En dataset automatically generated from Newsela corpus
 - Newsela corpus:
 - a document-level comparable corpus with document-level complexity (i.e., grade level), composed of English and Spanish news articles.
 - Creation procedure:



□ Issues:

1. **Incorrect translation pairs**
2. **Difference of granularity of information among target language sentences with different complexity levels**

Examples of Issue 2 (proper noun insertion)

level	English
8	However, she says that there are times when "you just need to get away."
5	Yet she says that there are times when "you just need to get away."
3	Still, Bopp says that there are times when "you just need to get away."

3. **Incorrect sentence-level complexity**

Examples of Issue 3

level	English
12	So few Indians drink brewed coffee that virtually all its best crop is exported to countries such as Italy, where the beans are used in name-brand espresso blends and sold at a huge markup.
9	There the beans are used in name-brand espresso blends and sold at a huge price increase.
7	There_ the beans are used in name-brand espresso blends and sold for a huge price increase.

Benchmark Test Dataset

- A new benchmark test dataset for Ja-to-En MLCC-MT
- Proposed creation procedure:
 - Step1:** Generates the sets of English sentences with the same content written at multiple complexity levels
 - 1-1: Extracts English aligned sentences from Newsela-auto [Jiang et al, ACL 2020]
 - 1-2: Automatically removes exactly the same sentences or grade levels diff ≤ 1 (\rightarrow solves **Issue 3**)
 - 1-3: Manually removes the sets where new content appears (\rightarrow solves **Issue 2**)
 - Step2:** Manually translates English sentences into Japanese (\rightarrow solves **Issue 1**)
- Result: **1,014 sets** (1 Japanese \Leftrightarrow 3~5 English with multi-level complexity)

□ Release: <https://github.com/K-T4N1/A-BenchmarkDataset-for-ComplexityControllableNMT>

Samples in our test dataset

Japanese	level	English
公衆衛生について	12	A primer about public health is in order.
の入門書が必要だ。	9	A short explanation about public health is in order.
	6	A short explanation about public health is needed.

Benchmark Experiments

- Implement two Transformer-based MLCC-MT models and evaluate them on our test dataset to serve as benchmark performance for future research

□ Benchmark models:

- Pipeline model: Ja-to-En NMT \rightarrow En multi-level simplification
 - Ja-to-En NMT: Transformer NMT [Kiyono et al., wmt2020]
 - En multi-level simplification model: Incorporation of special tokens representing target complexity [Scarton et al., ACL 2018]
- Multi-task model: based on the following three losses

$$loss = L_{MT} + L_{Simplify} + L_{CMT}$$

- L_{MT} : the loss for conventional MT
- $L_{Simplify}$: the loss for text simplification
- L_{CMT} : the loss for complexity-controllable MT

□ Evaluation Metrics:

- BLEU [Papineni et al., ACL 2002]: MT performance
- SARI [Xu et al., TACL 2016]: text simplification performance
- MAE_{fkg} (Mean absolute error of FKGL) [Nishihara et al., ACLSRW 2019]: complexity controlling performance

□ Results:

model	BLEU (%) \uparrow	SARI (%) \uparrow	MAE_{fkg} \downarrow
Pipeline	15.12	23.89	5.10
Multi-task	20.17	26.78	4.83

- Multi-task model outperforms the pipeline model in term of BLEU, SARI and MAE_{fkg} .

Conclusion

- Create a new benchmark **test dataset** for Ja-En MLCC-MT
 - The proposed creation procedure includes **automatic filtering**, **manual check**, and **manual translation** to make our test dataset more appropriate than existing test datasets.
- Implement two Transformer-based MLCC-NMT (pipeline and multi-task) and evaluate them as **benchmark performance**
- Future work:
 - Increase the size of our dataset and create a multi-lingual dataset for MLCC-MT