Simultaneous Neural Machine Translation with Constituent Label Prediction

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Simultaneous translation

• Consecutive translation

Input: I am a student . Output: 私は学生です

• Simultaneous translation

Input: I am a student . Output: 私は学生です。

Related work: Meaningful Unit [Zhang+, 2020]



Language pairs with different word orders

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• SVO(Subject-Verb-Object) \rightarrow SOV

En) I <mark>bought a pen</mark>. Ja) Watashi wa <mark>pen wo katta</mark> .

 \rightarrow Use syntax information

Proposed: Constituent Label Prediction



Simple rules based on predicted labels

• Segment the input coming just before constituents labeled S and VP.

I / saved time by / doing this. <u>VP</u> NP PP <u>S</u> NP

• If the previous label is S or VP, do not segment the input.

I / can (/) save time . <u>VP</u><u>VP</u>NP

 If the chunk is shorter than the minimum length, do not segment the input. [Change minimum length to adjust latency]

> I (/) bought a pen . VP NP NN

Minimum length = 2

Experiment of ICLP

- Data
 - Train: Penn Treebank 3 [Marcus+, 1993]
 - dev: 1% of training data
 - test: NAIST-NTT TED Talk Treebank [Neubig+, 2014]
- Model
 - BERT [Devlin+,2019]
- Result (VP)

Model	Precision	Recall
0 future words	0.75	0.80
1 future word	0.89	0.97
1 future word (LSTM)	0.91	0.94

Experiment of simultaneous translation

- Data [En-Ja]
 - Pretrain: 20M (WMT2020
 - Fin-tune: 200K (IWSLT2021
 - Dev: 5.3K (IWSLT dev2010, tst2011, tst2012, and tst2013
 - Test: 1.5K (IWSLT2021 dev
- Subwords
 - Joint vocabulary size 16k (BPE)
- NMT Model
 - Transformer [Vaswani+, 2017]
- Evaluation metrics
 - Quality: BLEU
 - Latency: AL (Average Lagging) [Ma+, 2019]

Result [En \rightarrow Ja]



Proposed method outperformed baselines for wide range of AL.

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Length Ratio [En \rightarrow Ja]



Translation of short segments tends to be longer than expected

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Segment Length Distribution (test $[En \rightarrow Ja]$)

Meaningful Unit (1 future word) AL: 7.26 BLEU: 16.53 Proposed: ICLP (1 future word) AL: 7.23 BLEU:17.22



Controllability of latency $[En \rightarrow Ja]$

Meaningful Unit

Proposed: ICLP



Result [En \rightarrow De]



Look-ahead approaches did not improve the performance.

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Conclusion

- Novel segmentation method: simple rules and label predictor
 - Higher BLEU than baselines in En-Ja simultaneous translation
 - Easy to control latency
 - Not dependent on trained NMT model
- Future work
 - Extract rules automatically
 - Improvement for other language pairs

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