Recent Advances In Document-level Neural Machine Translation

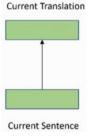
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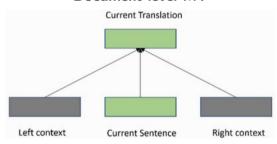
July 10, 2020

What is Document-level Machine Translation

Sentence-level MT

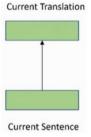


Document-level MT

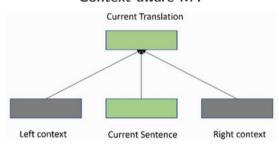


Document-level MT ↔ Context-aware MT

Context-agnostic MT



Context-aware MT



Why Document-level NMT?

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► Some recent results suggest that neural machine translation (NMT) "approaches the accuracy achieved by average bilingual human translators [on some test sets] [Wu et al., 2016]

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- Some recent results suggest that neural machine translation (NMT) "approaches the accuracy achieved by average bilingual human translators [on some test sets] [Wu et al., 2016]
- "In a pairwise ranking experiment, human raters assessing adequacy and fluency show a stronger preference for human over machine translation when evaluating documents as compared to isolated sentences." [Lubli et al., 2018]

B: How are you today?

B: How are you today?

SENTENCE-LEVEL TRANSLATION

B: Comment vas-tu aujourd'hui?

A: Good Morning, Mr. President.

B: How are you today?

SENTENCE-LEVEL TRANSLATION

B: Comment vas-tu aujourd'hui?

A: Good Morning, Mr. President.

B: How are you today?

SENTENCE-LEVEL TRANSLATION

B: Comment vas-tu aujourd'hui?

CONTEXT-AWARE TRANSLATION

B: Comment allez-vous aujourd'hui?

How frequent are inconsistencies?

[Voita et al., 2019] undertake a human study on context agnostic translation :

- ▶ 2000 pairs of consecutive English sentences (S1 + S2) from OpenSubtitles2018
- ► translate to Russian with Transformer model [Vaswani et al., 2017]

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all	one/both bad	both good	
		bad pair	good pair
2000	211	140	1649
100%	11%	7%	82%

Which kind of inconsistencies?

type of phenomena	frequency	
deixis	37%	
ellipsis	29%	
lexical cohesion	14%	
ambiguity	9%	
anaphora	6%	
other	5%	

Figure: Types of phenomena causing inconsistencies between English-Russian context-agnostic translations of consecutive sentences when placed in the context of each other.

Objectives

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 Design translation models and learning techniques that solve inconsistencies by taking context into account;

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- Evaluate such models in a proper way;

Plan

Thank you for your attention!

References I



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