

KyotoEBMT System Description for the 1st Workshop on Asian Translation

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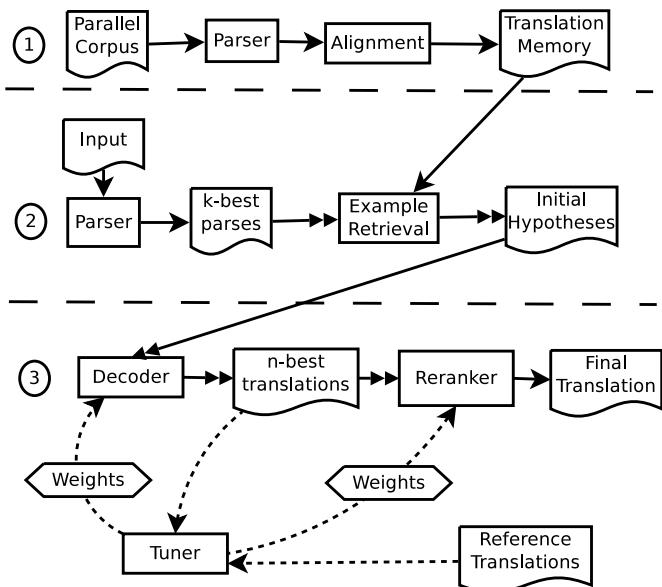
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KyotoEBMT System Pipeline



Web Interface of Translation

The web interface displays the following components:

- Flags:** Japan, China, UK.
- Text:** 本稿では 依存構造に基づく用例 ベース 機械翻訳 システムを 紹介する。
- Dependency Trees:**
 - Input and Output Dependency Trees: Shows Japanese input and its corresponding English output with dependency relations (e.g., r[1] 本稿, r[1] で, r[1] は, r[1] 依存, r[1] 基づく, r[1] 用例, r[1] ベース, r[1] 機械, r[1] 翻訳, r[1] を, r[1] 紹介, r[1] する).
 - List of Used Translation Examples: Shows examples from NICT_JE_SP-train-G-0654753 and NICT_JE_SP-train-R-0064303, such as r[1] 本稿, r[1] で, r[1] は, r[1] を, r[1] 紹介, r[1] する, r[1] における, r[1] CAD, etc.

WAT2014 Official Results

Dependency Parsers

Ja: KNP [Kawahara and Kurohashi, 2006]

En: NLParser [Charniak and Johnson, 2005] with rules

Zh: SKP [Shen et al., 2012]

Ranking Features

20-best parses with its parsing scores

7-gram language model with Modified Kneser-Ney smoothing

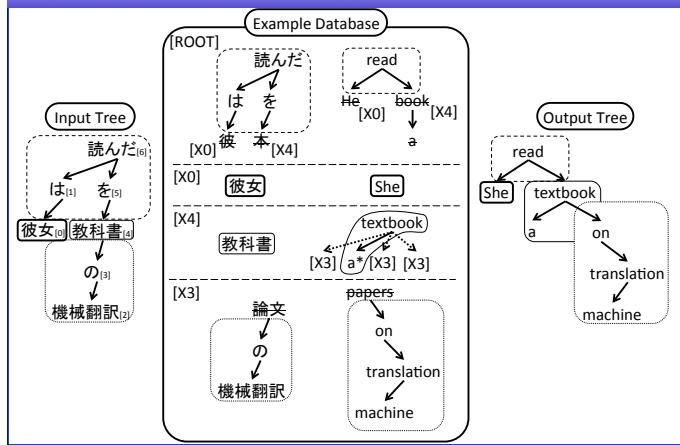
Recurrent Neural Network Language Model with hidden layer size 200

	Reranking	BLEU	RIBES	HUMAN
JE	NO	20.60	70.12	21.50
	YES	21.07	69.90	25.00
EJ	NO	29.76	75.21	33.75
	YES	31.09	75.96	38.00
JC	NO	27.21	79.13	-0.75
	YES	27.67	78.83	-8.75
CJ	NO	33.57	80.10	6.00
	YES	34.75	80.26	7.50

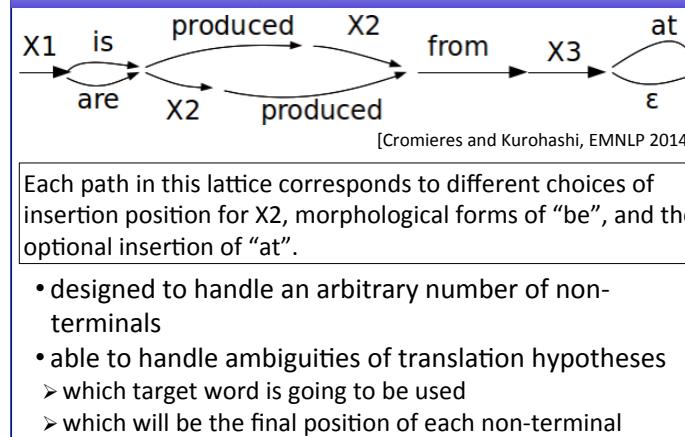
Findings

Team-based rankings of HUMAN score were 2nd, 3rd, 5th and 4th
Reranking works well for all the directions other than J->C
need more investigations

Illustration of Translation Process



Translation with Lattice Rules



Conclusion and Future Perspective

KyotoEBMT system

- source code available under a GPL license at <http://nlp.ist.i.kyoto-u.ac.jp/kyotoebmt/>
- uses both source and target dependency analysis
- online example retrieving
- availability of full translation examples at run time

Future perspectives

- use of the input parse forest instead of k-best parses
- parsing the example parallel corpus as forests
- use a target-side tree language model as a decoding/reranking feature
- online tuning of weights
- target-side structural features