

Toshiba MT System Description for the WAT2015 Workshop

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Motivations

Rule-Based Machine Translation (RBMT)

- We have been developed RBMT for more than 30 years.
- Japanese⇔English, Japanese⇔Chinese, Japanese⇔Korean
- Large technical dictionaries and translation rules

Pre-ordering SMT and Tree/Forest to String

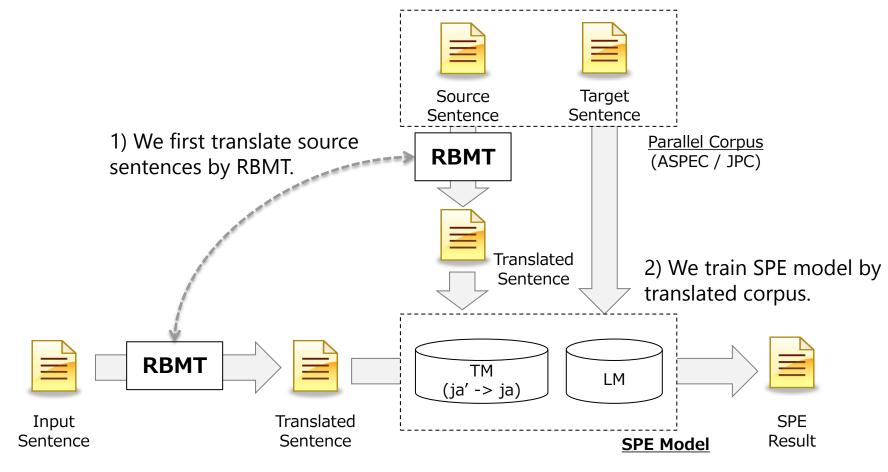
- Effective solutions for Asian language translation (WAT2014)
- But, pre-ordering rules and parsers are needed.

Our approach:

- Statistical Post Editing (SPE) (same as WAT2014)
 - Verify effectiveness in all tasks
- System combination between SPE and SMT (new in WAT2015)

Statistical Post Editing (SPE)

Translating RBMT results to post-edited results.



本发明具有以下效果。

本発明は以下効果を持っている。

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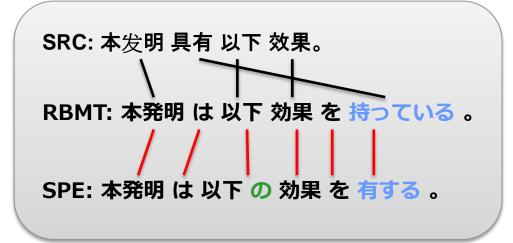
Features of SPE

From RBMT's standpoint

- Correct mistranslations / Translate unknown words
 - Phrase-level correction (domain adaptation)
- Improve fluency
 - Use of more fluent expressions
 - Insertion of particles
- Recover translation failure

From SMT's standpoint

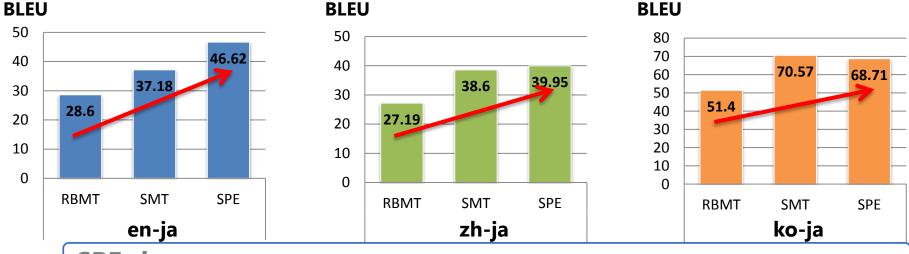
- Pre-ordering by RBMT
- Reduction of NULL alignment (subject/particle)
- Use of syntax information (polarity/aspect)
- Enhancement of lexicon



SPE for Patent Translation

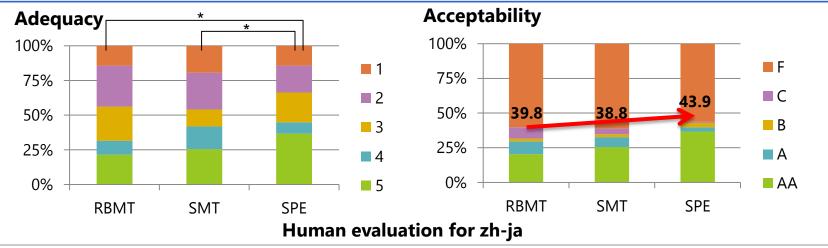
Corpus: JPO-NICT patent corpus # of training data: 2M(en-ja), 1M(zh-ja/ko-ja)

of automatic evaluation: 2,000 # of human evaluation: 200



SPE shows:

- Better scores than PB-SMT in automatic evaluation
- Improvements of understandable level (>=C in acceptability)



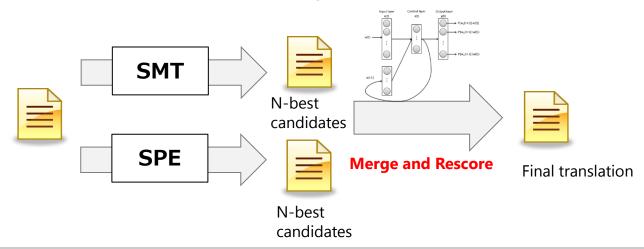
System Combination

How combine systems?

- Selection based on SMT scores and/or other features.
- Selection based on estimated score (Adequacy? Fluency? ...)
 - Need data to learn the relationship...

Our approach in WAT2015:

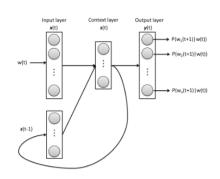
- Merge n-best candidates and rescore them.
- We used RNNLM for reranking.



RNNLM reranking and Tuning

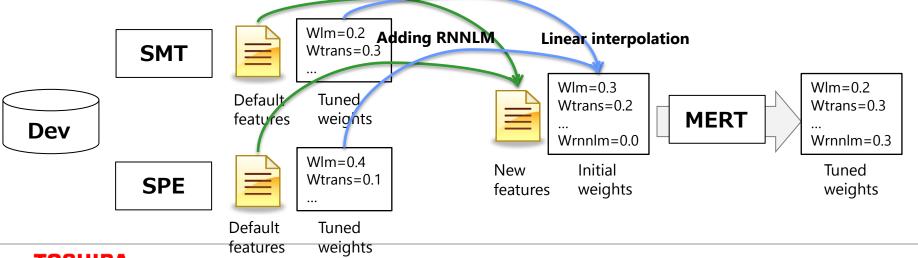
Reranking on the log-linear model

- Adding RNNLM score to default features of Moses.
- RNNLM trained by rnnlm toolkit (Mikolov '12).
 - 500,000 sentences for each language
 - # of hidden layer=500, # of class=50

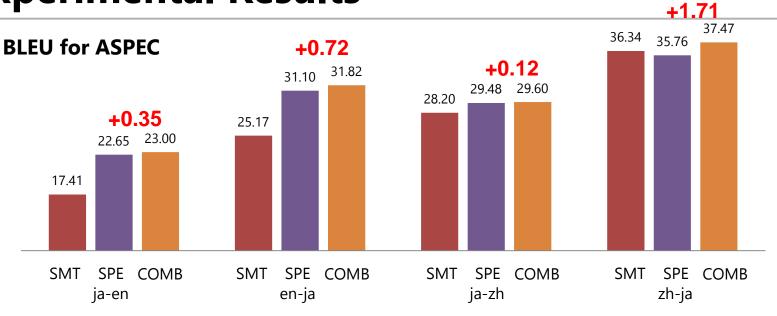


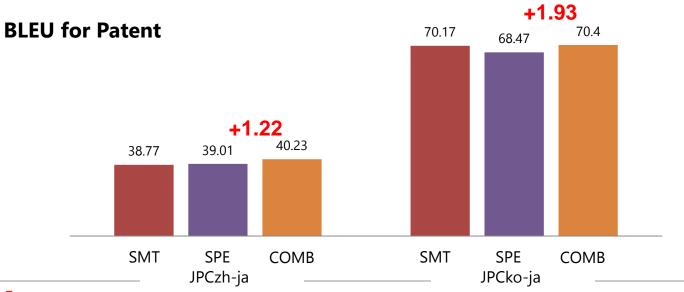
Tuning

 Using tuned weights without RNNLM, we ran only 1 iteration. (to reduce tuning time)



Experimental Results







Experimental Results

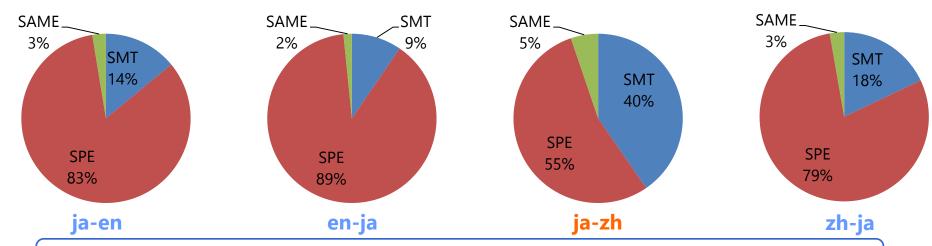
Systems	Rerank	ja-en		en-ja		ja-zh		zh-ja	
		BLEU	RIBES	BLEU	RIBES	BLEU	RIBES	BLEU	RIBES
RBMT	No	15.31	0.677	14.78	0.685	19.51	0.767	15.39	0.767
CNAT	No	17.41	0.620	25.17	0.642	28.20	0.810	36.34	0.810
SMT	Yes	17.85	0.619	25.37	0.643	28.46	0.809	36.69	0.809
CDE	No	22.65	0.717	31.10	0.767	29.48	0.809	35.76	0.809
SPE	Yes	22.92	0.718	31.73	0.770	29.49	0.809	36.06	0.809
СОМВ	Yes	23.00	0.716	31.82	0.770	29.60	0.810	37.47	0.810

System Combination (COMB) achieved improvements of BLEU and RIBES score than SPE.

COMB is the best system except JPCko-ja task.

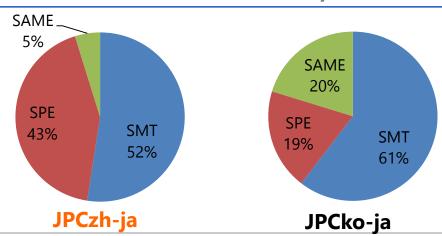
Cyctomo	Donords	JPCz	zh-ja	JPCko-ja		
Systems	Rerank	BLEU	RIBES	BLEU	RIBES	
RBMT	No	25.81	0.764	51.28	0.902	
SMT	No	38.77	0.802	70.17	0.943	
	Yes	39.18	0.805	70.89	0.944	
SPE	No	39.01	0.813	68.47	0.940	
	Yes	39.30	0.811	68.76	0.940	
СОМВ	Yes	40.23	0.813	70.40	0.942	

Which systems did the combination selected?



ja-en/en-ja/zh-ja: about 80% translations come from SPE.

ja-zh and JPCzh-ja: COMB selected SPE and SMT, equivalently. (Because RBMT couldn't translate well, % of SMT increased.)





Toshiba MT system of WAT2015

We additionally applied some pre/post processing.

Technical Term Dictionaries

Selecting RBMT dictionaries by devset.

+ JPO patent dictionary (2.2M words for JPCzh-ja)

English Word Correction

Edited-distance based correction

continous -> continuous
 behvior -> behavior
resolutin -> resolution

KATAKANA Normalization

Normalize to highlyfrequent notations for "—"

スクリュ -> スクリューサーバー -> サーバ

Post-translation

Translate remaining unknown words by RBMT.

アルキメデス数 ->阿基米德数 流入中하수 -> 流入マッハ数

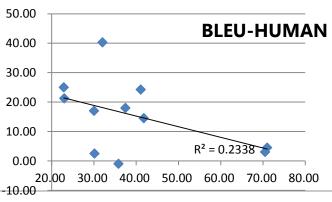
Official Results

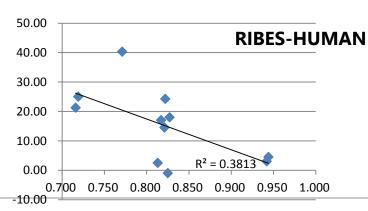
SPE and SMT ranked in the top 3 HUMAN in ja-en/ja-zh/JPCzh-ja.

Custom	ja-en		en-ja		ja-zh			zh-ja				
System	BLEU	RIBES	HUMAN									
SPE	22.89	0.719	25.00	32.06	0.771	40.25	30.17	0.813	2.50	35.85	0.825	-1.00
COMB	23.00	0.716	21.25	31.82	0.770	-	30.07	0.817	17.00	37.47	0.827	18.00

C		JPCzh-ja		JPCko-ja			
System	BLEU	RIBES	HUMAN	BLEU	RIBES	HUMAN	
SMT	-	-	-	71.01	0.944	4.50	
SPE	41.12	0.822	24.25	-	_	-	
СОМВ	41.82	0.821	14.50	70.51	0.942	3.00	

 The correlation between BLEU/RIEBES and HUMAN is not clear in our system.





Crowdsourcing Evaluation

- Analysis of JPCko-ja result (COMB vs Online A)
 - In in-house evaluation, COMB is better than Online A.

	DIELL	DIDEC	HUMAN				
	BLEU	RIBES	Baseline	СОМВ	Online A		
COMB	70.51	0.94	3.00	-	10.75		
Online A	55.05	0.9	38.75	-10.75	-		
Official In-house evalu (Crowdsourcing) results							

– Effected by differences in number expressions !?

SRC : 시스템(100) ⇒ Online A: システム(100)

COMB(SMT): システム100

- ⇒ Equally evaluated in-house evaluation.
- Crowd-workers should be provided an evaluation guideline by which such a difference is considered.

Summary

- Toshiba MT system achieved a combination method between SMT and SPE by RNNLM reranking.
- Our system ranked the top 3 HUMAN score in ja-en/ja-zh/JPCzh-ja.
- We will aim for practical MT system by more effective combination systems (SMT, SPE, RBMT and more...)

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