

The RWTH Machine Translation System for IWSLT 2008

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1 Introduction

- Combination of phrase-based and hierarchical SMT systems
- Chinese-to-English and Arabic-to-English
- Investigated the effect of
 - b different preprocessing techniques
 - reordering methods (including reordering of speech lattices)
 - syntax-based enhancements
- System ranked 6th in CE (all conditions) and 3rd in AE (all conditions)
- Combination of AE and CE outputs

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Outline

- **1** Introduction
- 2 Translation Models
- 3 Extensions
- **4** Experimental Results
- **5** Conclusions



2 Translation Models

2.1 Phrase-based Model

Well-known model

- Scores computed by relative frequencies
- Two different reordering models (depending on language pair)
 - IBM Reordering
 - Jump Reordering



2.2 Hierarchical Model

- Extension of the phrase-based model
- Allow for "gaps" in the phrases
- Formalized as a CF grammar (translation as parsing process)
- Example rules:

$$X \to \langle \oplus X^{\sim 0}$$
那个 $X^{\sim 1}$, It's the $X^{\sim 1}$ in the $X^{\sim 0} \rangle$
 $X \to \langle \oplus X^{\sim 0} - \oplus X^{\sim 1}$, like to $X^{\sim 0}$ some $X^{\sim 1}$ too \rangle

2.3 Common Models

- Word-based Lexicon Model
- Target Language Model (6-gram, Kneser-Ney discounting)
- Phrase Count Features
- Phrase Penalty
- Word Penalty



3 Extensions

3.1 Syntactical Features

- Extension for the hierarchical model
- Additional (soft) feature extracted at training time
- A rule is "syntactically consistent" if the "involved" original phrases correspond to yields of a syntax tree
- Done for source and target part independently
- Possibility of smoothing the "syntactic constraints"
- Detailed description in [Vilar et al. 2008]





3.2 Chunk-based Reordering for Chinese

- Reordering of the Chinese source sentence
- Syntactic chunk-level rules, automatically learned from the training data
- Reordered possibilities represented as n-best lists (with small n)
- Each reordering scored with the product of the probability of each of the rules
- Reordered training data added to the original data
- Detailed description in [Zhang et al. 2007]



3.3 Source Preprocessing

Chinese

- Unigram segmenter obtains better results than ictclass
- LDC-like segmentation without text normalization
- Maximize the joint probability of all the words in the sentence
- Splitting long sentence pairs
- Detailed description in [Xu et al. 2008]



Experiments with MADA and MorphTagger (POS-tagging tool)

- Three segmentation schemes
 - Splitting only the prefixes w+, l+, k+, b+, s+ (PRE)
 - Additionally splitting the determiner AI+ (PRE+DET)
 - Additionally splitting the pronominal suffixes (PREF+SUF)
- Tested normalizing Yaa and Alef
- Best results:
 - MorphTagger: PRE+SUF and no normalization
 - MADA: PRE and normalization



3.4 Translation of Speech Lattices

- Translation of word lattices including reordering
- Acoustic and source language model scores
- Cardinality synchronous search
 - Define cardinality in terms of "slots" (CN-like)
 - Allow for reordering without the over-generalization of CN
- Mapping from ASR vocabulary to MT vocabulary (segmentation)
- No improvements on this task (regretfully)
- Detailed description in [Matusov et al. 2008]

3.5 System Combination

- Approach used in last year's evaluation
- Build a confusion network for each sentence
 - Select one system as primary system
 - Align the single-best output of this system with the other hypotheses
 - Build a confusion network
 - Repeat with each system as primary
- The resulting confusion networks are joined into a word graph
- Weight with system specific factors and a trigram LM trained on the MT hypotheses
- Detailed description in [Matusov et al. 2006]

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4 Experimental Results

- Arabic-to-English and Chinese-to-English translation directions
- Provided training data + HIT Corpus for CE
 - Selected sentences with 60% of the words in the IWSLT data
- Preprocessing of English:
 - Tokenization
 - Expansion of contractions
- GIZA++ for alignments
 - Tested different variants of word classes, model sequences and combination heuristics
- Optimized for BLEU on IWSLT 2004 eval data
- IWSLT 2005 eval data for system combination



4.1 Chinese-to-English

BTEC Task

CRR					
System	BLEU	TER	WER	PER	
System Combination	46.1	37.7	43.9	39.4	
Phrase Based (PBT)	42.5	36.6	45.3	40.6	
PBT + Chunk Reordering	42.6	39.9	47.8	42.4	
PBT + New Segmentation	44.3	40.3	47.3	42.0	
Hierarchical	41.2	41.5	48.1	42.7	
Hierarchical + Syntax	41.4	40.6	47.3	42.8	
ASR					
System	BLEU	TER	WER	PER	
System Combination	39.7	42.5	49.6	44.5	
Phrase Based (PBT)	37.3	41.2	50.0	45.1	
PBT + Chunk Reordering	38.5	42.8	51.2	46.4	
Hierarchical	31.6	49.6	56.5	49.5	
Hierarchical + Syntax	36.6	44.1	51.4	47.0	
Lattices	32.2	48.6	57.1	51.5	



Challenge Task

CRR					
System	BLEU	TER	WER	PER	
System Combination	39.1	40.7	48.3	44.1	
Phrase Based (PBT)	32.1	42.7	51.9	47.8	
PBT + Chunk Reordering	32.6	43.6	52.5	48.5	
PBT + New Segmentation	37.2	41.8	49.3	44.5	
Hierarchical	30.7	47.1	54.6	48.9	
Hierarchical + Syntax	30.2	45.5	53.6	48.5	
ASR					
System	BLEU	TER	WER	PER	
System Combination	34.3	43.6	51.1	46.1	
Phrase Based (PBT)	27.8	46.0	55.4	51.1	
PBT + Chunk Reordering	29.4	45.7	55.0	50.5	
Hierarchical	26.4	51.0	59.2	51.9	
Hierarchical + Syntax	30.2	45.6	53.7	48.6	
Lattices	25.0	56.6	62.8	56.7	



4.2 Arabic-to-English

CRR					
System	BLEU	TER	WER	PER	
System Combination	53.5	33.0	37.6	33.9	
PBT + MADA	50.0	33.7	39.7	36.0	
PBT + MorphTagger	51.8	33.8	38.1	33.9	
Hierarchical + MADA	49.2	36.6	41.3	36.7	
Hierarchical + MorphTagger	49.3	35.9	41.3	38.0	
ASR					
System	BLEU	TER	WER	PER	
System Combination	44.5	37.6	43.4	39.9	
PBT + MADA	42.6	38.2	45.3	41.7	
PBT + MorphTagger	44.0	38.0	43.4	39.4	
Hierarchical + MADA	41.3	42.1	47.7	42.7	
Hierarchical + MorphTagger	41.3	40.7	47.2	43.9	

► Note: bug in the hierarchical system (Corrected score for CRR: 54.1%)





4.3 Arabic&Chinese-to-English

System combination of the best performing systems for both language pairs

CRR				
System	BLEU	TER	WER	PER
System Combination	56.2	31.7	36.0	32.6



5 Conclusions

- Presented RWTH system for the IWSLT 2008 evaluation
- Combination of different statistical machine translation approaches
 - Phrase-based and hierarchical systems + extensions
- Combination of Arabic-to-English and Chinese-to-English systems increases performance