

The TÜBİTAK-UEKAE Statistical Machine Translation System for IWSLT 2007

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- System Description
- Training
 - Phrase table augmentation
- Decoding
 - Out of Vocabulary Words (OOV)
- Results
- Conclusion and Future Work



- Participated in translation tasks
 - Arabic-to-English
 - Japanese-to-English
- Built on phrase-based SMT software Moses
- Used only supplied data and Buckwalter Arabic Morphological Analyzer (BAMA)

System Description







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- Devset1-3 are included in the training with all 16 reference segments
- Train and Devset1-3 are given equal weight
- Language models
 - 3-gram for AR-EN
 - 4-gram for JP-EN
 - Trained with modified Kneser-Ney discounting and interpolation

JÜBİTAK	Training						
UEKAE	 Multi-sentence segments are split 						
	Before splitting After splitting						
	AR-EN	44,164 *	49,318				
	JP-EN	64,145 *	71,435				
		* train segments + 16 * dev1-3 segme	nts				



- Parameter tuning
 - Manually tested different set of parameters
 - Different data favored different parameters
 - Instead of selecting argmax, selected mode in a desirable interval to select a robust set of parameters



Phrase Table Augmentation

- Translation model is represented in a *phrase table*
- Bi-directional alignment and phrase extraction
 with grow-diag-final-and heuristics
- Source-language words without a one-word entry in phrase table are listed
- The words, which are in the list and have a lexical translation probability above a threshold in *GIZA++* word alignment, are added to phrase list



Phrase Table Augmentation

Corpus	AR-EN	JP-EN
Source vocabulary size	18,751	12,699
Number of entries in the original phrase table	408,052	606,432
Number of source vocabulary words without a one-word entry in the original phrase table	8,035	6,302
Number of one-word bi-phrases added to the phrase table	21,439	23,396
Number of entries in the augmented phrase-table	429,491	629,828



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Decoding



- Decoding is done on tokenized and punctuated data
 - Source-side punctuation insertion (for ASR data)
 - Target-side case restoration
- SRILM tools used for punctuation restoration

Decoding



 Merged 10 sentences to train punctuation restorer with more internal sentence boundaries

	Ν	Devset4	Devset5
	1	24.32	20.23
AR-EN	10	24.95	20.66
	1	15.59	14.26
JP-EIN	10	17.82	16.12



Out of Vocabulary Words

- Lexical Approximation
 - Find a set of candidate approximations
 - Select the candidate with least edit distance
 - In case of a tie, more frequently used candidate is chosen



- Arabic lexical approximation (2 pass)
 - Morphological root(s) of the word found by feature function using BAMA
 - If not, skeletonized version of the word is found by feature function
- Japanese lexical approximation (1 pass)
 Right-truncations of the word is found by feature function



Run-time Lexical Approximation

	Devset4		Devset5	
	# of OOVs	BLEU	# of OOVS	BLEU
Original	661	24.91	795	20.59
After LA#1	185	25.33	221	21.22
After LA#2	149	25.56	172	21.51
JP-EN	Devset4		Devset5	
	# of OOVs	BLEU	# of OOVS	BLEU
Original	119	23.68	169	20.44
After LA	10	23.84	17	20.69



mjAny~ is OOV



Out of Vocabulary Words

- Lexical approximation finds candidates
 mjAnyP, mjAnY, mjAnA, kjm, mjAny, mjAnAF
- mjAny has an edit distance of 1, so it's selected



Out of Vocabulary Words

After lexical approximation

hl hw mjAny?



Is it free ?



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UEKAE	AE Results					
		Clean Transcript	ASR Output			
	AR-EN	49.23	36.79			
	JP-EN	48.41	42.69			

Clean vs. ASR



- Possible causes of performance drop in ASR condition
 - Recognition errors of ASR
 - Punctuation restorer performance
 - Parameter tuning for clean transcript but not for ASR output

AR-EN vs. JP-EN



- Possible causes of higher performance drop in AR-EN than JP-EN
 - Lower accuracy of Arabic ASR data than Japanese data
 - Higher difficulty of punctuation insertion due to higher number of punctuation types
 - Less reliable punctuation insertion caused by higher recognition error rate

AR-EN vs. JP-EN



Lexical approximation is sensitive to recognition errors

	Clean transcript	ASR output	Clean-to-ASR degradation
Original source	38.48	31.82	17.3%
After LA	49.23	36.79	25.3%



Devset4-5 vs. Evaluation Set

 There is a dramatic variation in the improvement obtained with the lexical approximation technique on the evaluation and development sets

Devset4-5 vs. Evaluation Set

TÜBİTAK____

	Devset4	Devset5
Original source	24.91	20.59
After LA#1	25.33	21.22
After LA#2	25.56	21.51
Improvment	2.6%	4.5%
	Evaluation set clean transcript	Evaluation set ASR output
Original source	38.48	31.82
After LA	49.23	36.79
Improvment	27.9%	15.6%

Devset4-5 vs. Evaluation Set

- 167 of 489 evaluation set segments have at least one reference which is a perfect match with a training segment
- Only 19 of 167 have the source segment exactly the same as in the training set
- Remaining 148 segments represents a potential to obtain a perfect match

TÜBİTAK_____ U E K A E

Number of segments	Devset4	Devset5	Evaluation set
Exact match of at least one reference with a segment in the training set	12	4	167
Exact math of the source with a segment in the training set	1	0	19
Total	489	500	489



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Conclusion and Future Work

- Make the system more robust to ASR output. For this goal:
 - Using n-best/lattice ASR output
 - Tuning system for ASR output
 - Better punctuation performance



Thank you for your attention!