

RWTH Aachen Machine Translation System: {Arabic, Chinese, German}-English MT Track

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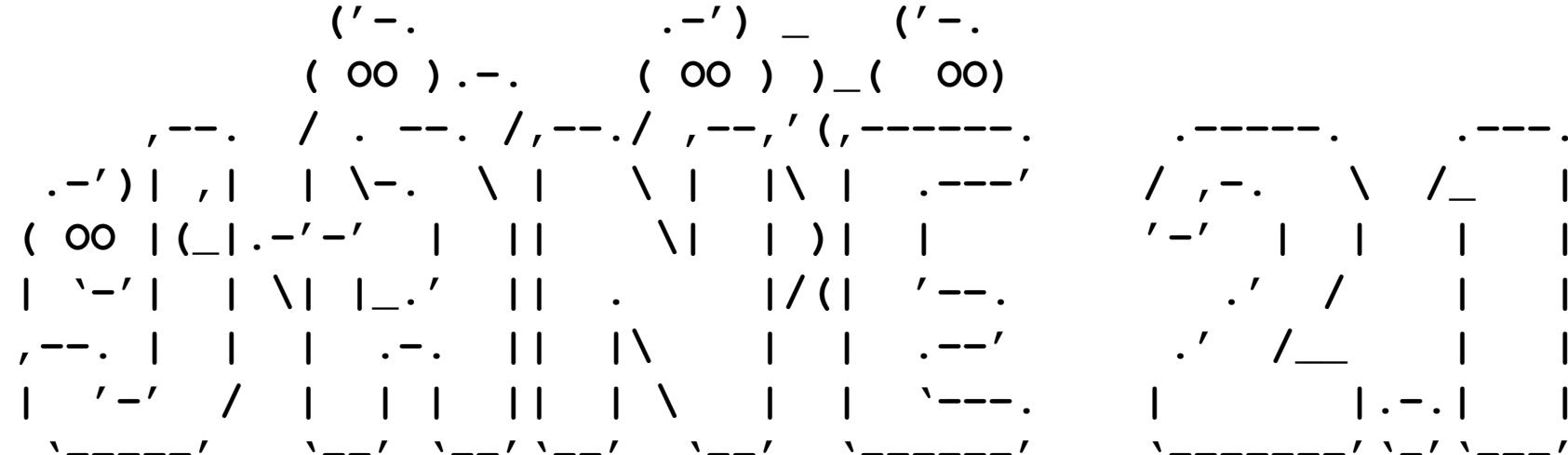
**Human Language Technology and Pattern Recognition
Lehrstuhl für Informatik 6
Computer Science Department
RWTH Aachen University, Germany**

Overview

- ▶ RWTH participated in 6 tracks this year:
 - ▷ English ASR
 - ▷ Arabic-English MT
 - ▷ English-French MT
 - ▷ Chinese-English MT
 - ▷ German-English MT
 - ▷ English-French SLT
- ▶ full results will be presented later today at the poster session:

The RWTH Aachen Speech Recognition and Machine Translation System for IWSLT 2012

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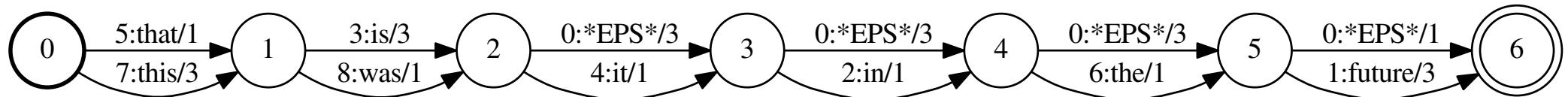
- ▶ RWTH's open-source translation toolkit
- ▶ new version Jane 2.1
- ▶ hierarchical phrase-based decoder [Huck & Peter⁺ 12]
- ▶ phrase-based decoder [Wuebker & Huck⁺ 12]
- ▶ applied in all MT and SLT tasks
- ▶ <http://www.hltpr.rwth-aachen.de/jane>

System Combination

- ▶ applied in following MT tasks:
 - ▷ Arabic-English
 - ▷ Chinese-English
 - ▷ English-French
- ▶ goal: produce consensus translation from multiple systems
- ▶ based on [Matusov & Leusch⁺ 08]
- ▶ in this work:
 - ▷ create word alignment with METEOR [Banerjee & Lavie 05]
 - ▷ feature weights optimization with MERT [Och 03]
 - ▷ implementation based on OpenFst [Allauzen & Riley⁺ 07]

System Combination

- ▶ select each hypothesis h in a set of hypotheses as primary system
 1. align all other hypotheses to h using METEOR
 2. construct confusion network
- ▶ unify all confusion networks
- ▶ add features to the arcs of the confusion networks
- ▶ find path with the best score (= consensus translation)



System Combination

- ▶ used features in system combination
 - ▷ word counts of the single systems
 - ▷ language model
 - ▷ word penalty
 - ▷ binary feature to mark primary system
- ▶ features are combined in a log-linear model
- ▶ feature weights are optimized with MERT
- ▶ in this work:
 - ▷ improvements of up to 0.9 points in BLEU over best single systems

Arabic-English

- ▶ **phrase-based decoder**
- ▶ **preprocessing: different Arabic segmentations**
- ▶ **applied techniques:**
 - ▷ **data selection for LM and TM training [Moore & Lewis 10]**
 - ▷ **phrase table interpolation of in-domain (*in*) and out-of-domain (*ood*)**
 - ▷ **system combination**

Phrase Table Interpolation

► linear interpolation

- ▷ $p(\tilde{f}|\tilde{e}) = \lambda p_{in}(\tilde{f}|\tilde{e}) + (1 - \lambda)p_{ood}(\tilde{f}|\tilde{e})$
- ▷ interpolation weight λ was adjust on the development set

► log-linear interpolation

- ▷ fits directly into the SMT log-linear framework
- ▷ weights optimized using MERT
- ▷ no improvement

► ifelse method [Haddow & Koehn 12]

```
if ( $\tilde{f}, \tilde{e}$ ) exists in in-domain phrase table  
    assign  $p_{in}(\tilde{f}|\tilde{e})$   
else  
    assign  $p_{ood}(\tilde{f}|\tilde{e})$ 
```

Phrase Table Interpolation Results

system	dev2010		tst2010	
	BLEU	TER	BLEU	TER
TED	27.9	51.8	26.1	54.9
TED+UN	28.2	52.8	25.7	57.0
TED-linear-UN	29.0	51.0	26.8	54.6
TED-ifelse-UN	29.5	50.8	26.7	55.0

- ▶ TED: in-domain, UN: out-of-domain
- ▶ TED+UN: concatenation of in-domain and out-of-domain data

Arabic-English Results

system	tst2010	
	BLEU	TER
FST	26.5 +1.4	55.8 -1.2
SVM	26.6 +1.2	54.4 -3.0
HMM	26.9 +1.2	55.1 -1.8
CRF	26.9 +1.2	54.5 -2.2
MADA-D1	26.3 +1.6	55.4 -2.4
MADA-D2	26.9 +1.7	54.7 -2.4
MADA-D3	27.0 +1.6	54.0 -3.1
MADA-TB ALL	27.1 +1.0	54.4 -2.2
system combination	28.0 +1.0	53.4 -1.3

- ▶ a comparison between 2011 and 2012 systems, over tst2010
- ▶ for all segmentation methods: linear interpolation and same LM
- ▶ improvements of > 1% BLEU on all setups, including final system

Chinese-English

► decoders:

- ▷ **in-house phrase-based decoder (PBT)**
- ▷ **hierarchical decoder (HPBT)**

► applied techniques:

- ▷ **reverse translation**
- ▷ **system combination**

Reverse Translation

- ▶ reverse direction decoding (right-to-left) [Finch & Sumita 09]
- ▶ same data as the standard direction system
- ▶ reverse the word order of the corpora and test sets
 - ▷ retrain the word alignment
 - ▷ recompute the language model
- ▶ employ on PBT and HPBT
- ▶ obtain four different translations
- ▶ apply system combination to gain benefits from two-direction decoding

Chinese-English Results

system	dev2010		tst2010	
	BLEU	TER	BLEU	TER
PBT	12.2	80.0	14.2	73.7
PBT-reverse	11.9	79.6	13.7	74.3
HPBT	12.7	80.0	14.7	74.5
HPBT-reverse	12.8	81.0	14.5	76.2
HPBT-withUN-a	12.1	81.4	14.1	76.0
HPBT-withUN-b	12.5	80.4	14.0	75.5
system combination	13.7	78.9	15.4	74.1

► HPBT-withUN-*

- ▷ additional 800K bilingual sentences from UN data
- ▷ differently optimized feature weights

German-English

- ▶ **phrase-based decoder**
- ▶ **preprocessing:**
 - ▷ compound splitting [Koehn & Knight 03]
 - ▷ POS-based long-range verb reordering [Popović & Ney 06]
- ▶ **applied techniques:**
 - ▷ forced alignment [Wuebker & Mauser⁺ 10]
 - ▷ word class language model
 - ▷ two phrase tables (in-domain and out-of-domain)

German-English Results

system	dev2010		tst2010	
	BLEU	TER	BLEU	TER
allData	29.0	49.5	27.5	51.6
TED	29.9 +0.9	48.4 -0.9	28.4 +0.9	50.3 -1.3
+ForcedAlignment	30.3 +0.4	47.7 -0.7	28.5 +0.1	49.9 -0.4
+ShuffledNews	31.1 +0.8	47.9 +0.2	29.2 +0.7	50.2 +0.3
+WordClassLM	31.2 +0.1	47.8 -0.1	29.8 +0.6	49.7 -0.5
+oodDataTM	31.9 +0.7	47.4 -0.4	30.3 +0.5	49.3 -0.4
+Gigaword	32.6 +0.7	46.4 -1.0	30.8 +0.5	48.6 -0.7

- ▶ allData: all available bilingual data vs. TED: in-domain data
- ▶ oodDataTM: additional out-of-domain translation model
- ▶ incremental improvement of translation quality

Thank you for your attention

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