

Preordering and Word Order Freedom in Machine Translation

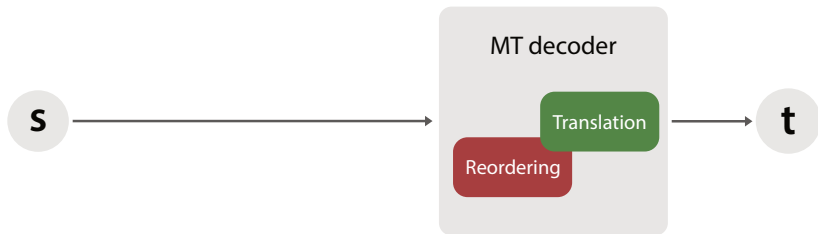


Joachim Daiber, Miloš Stanojević, Wilker Aziz and Khalil Sima'an

*Institute for Logic, Language and Computation
University of Amsterdam*

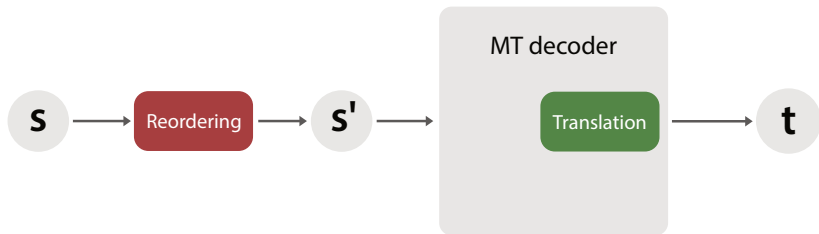


Reordering in PBMT



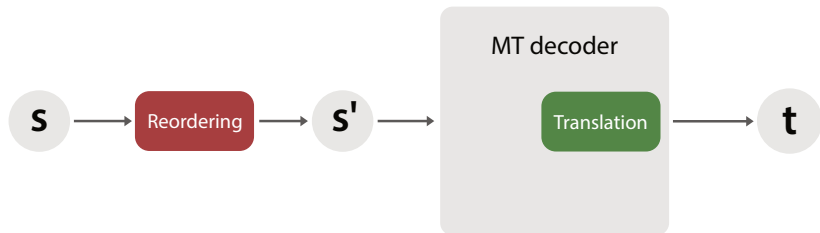


Reordering in PBMT





Reordering in PBMT

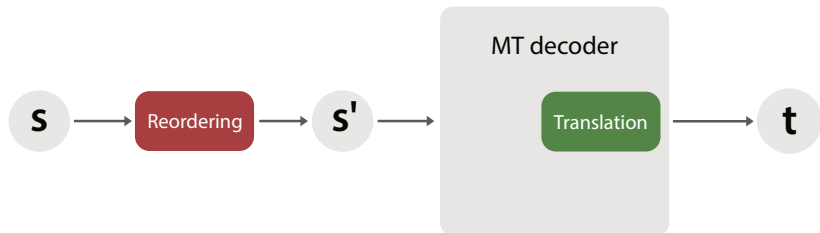


Preordering

+ Works well for English–Japanese, English–Korean, ...



Reordering in PBMT



Preordering

- + Works well for English–Japanese, English–Korean, ...
- Not that well for English–German, English–Czech, ...



Reordering in PBMT

Why the partial success?

- ▶ **Free word order** languages are difficult: many potential target word orders



Reordering in PBMT

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- ▶ **Source side often not enough** to predict a unique target word order



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Potential solution:

- ▶ Narrow down space of target word orders and let decoder decide



Reordering in PBMT

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- ▶ **Source side often not enough** to predict a unique target word order

Potential solution:

- ▶ Narrow down space of target word orders and let decoder decide
- ▶ Prepare MT system for this kind of input



Outline

1. Word Order Freedom

- What exactly do we mean by free word order?
- Can we measure it?



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2. Translation with Permutation Lattices

- Permutation Lattices
- Lattice Silver Training



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1. Word Order Freedom

- What exactly do we mean by free word order?
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2. Translation with Permutation Lattices

- Permutation Lattices
- Lattice Silver Training

3. Experiments with two language pairs

- English–Japanese
Strict word order language
- English–German
Free word order language



What Does Free Word Order Mean?

- ▶ Multiple word orders expressing the same meaning

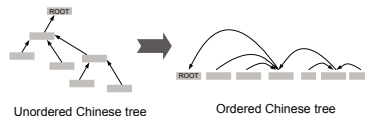
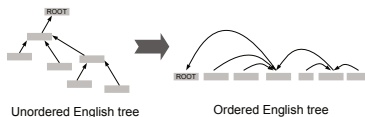
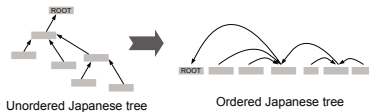


What Does Free Word Order Mean?

- ▶ Multiple word orders expressing the same meaning
→ Fuzzy definition!
- ▶ Can we quantify and compare languages' word order freedom?

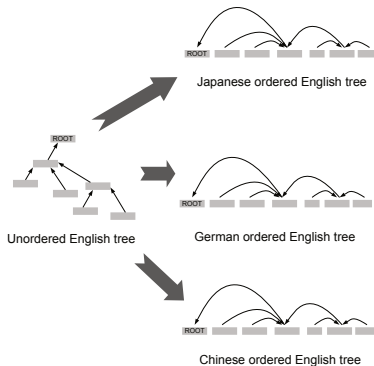


Estimating Word Order Freedom: Futrell et al. (2015)



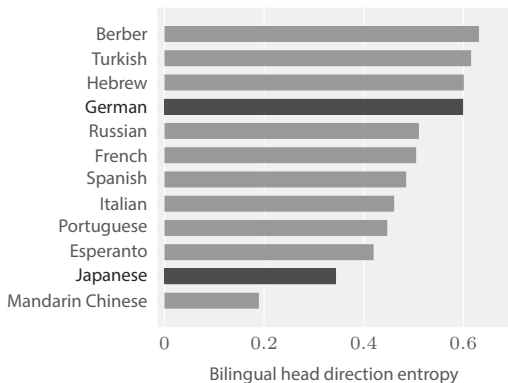


Estimating Word Order Freedom: Our Bilingual Approach





Word Order Freedom

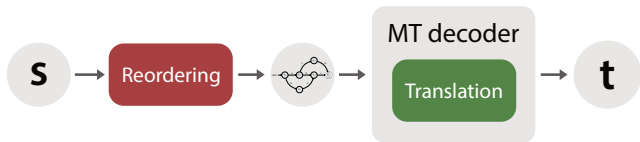




Free Word Order

What to do about Free Word Order?

- ▶ Instead of choosing 1-best word order, work with n -best

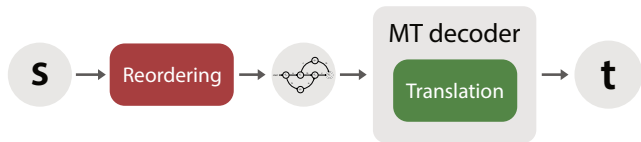




Free Word Order

What to do about Free Word Order?

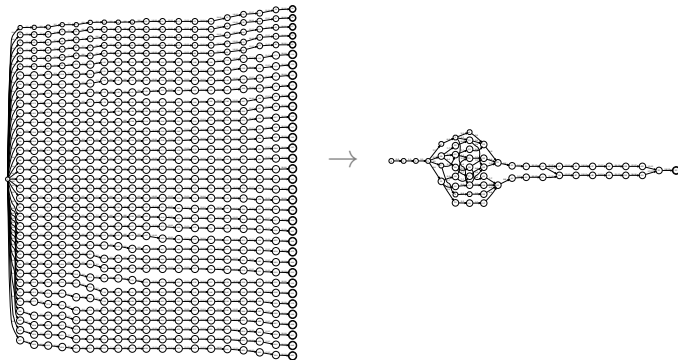
- ▶ Instead of choosing 1-best word order, work with n -best



- ▶ Decoding with n -best permutations:
 - 1) **Inference:** Compactly encode n -best WOs into lattice
 - 2) **Training:** Estimate parameters for translating lattices



Permutation Lattices





Lattice Training

Q: How to extract the phrase table?

Gold Training

(target-order s' , t)

+ compact

+ less noisy

- may not cover

preordered input



Lattice Training

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Gold Training

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Predicted Training

(predicted s' , t)

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- more noisy

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Lattice Training

Q: How to extract the phrase table?

Gold Training

(target-order s' , t)

- + compact
- + less noisy
- may not cover preordered input

Predicted Training

(predicted s' , t)

- ? less compact
- more noisy
- + covers preordered input

Lattice Silver Training

(s' from n -best most similar to gold, t)

- + compact
- + less noisy
- + covers preordered input



Preordering Models

Reordering Grammar

Stanojević and Sima'an (2015)

- ▶ Hierarchical, probabilistic model
- ▶ Unsupervised (no syntax)
- ▶ Beyond ITG (based on PETs)

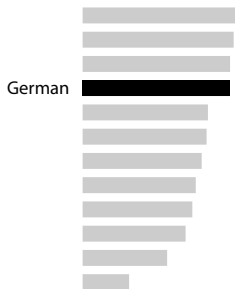
Neural Preordering

de Gispert et al. (2015); Jehl et al. (2014)

- ▶ Preorders dep. tree recursively
- ▶ NN to decide pairwise ordering of subtrees
- ▶ We extend search to n -best branch-and-bound



Translation Experiment: English→German



On WMT newstest 2015.

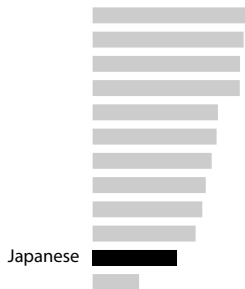
- ▶ Big oracle WO gain:
+4.92 over baseline (21.76 BLEU)
- ▶ 1-best with DL 6 performs poorly
- ▶ Lattices with Silver Training and DL 0 on par with baseline +0.12

Software: Moses (with extended lattice support). Optimizer: Batch MIRA (15 iterations, BLEU averaged over 3 runs).

Training data: 4.5m sentence pairs (WMT16). LM (5-gram): 189m sentences (WMT16). Dev/Test: newstest2014/newstest2015.



Translation Experiment: English→Japanese



On NTCIR Patent translation task.

- ▶ Big oracle WO gain:
+4.57 over baseline (29.65 BLEU)
- ▶ 1-best with DL 6 performs well: +2.49
- ▶ Lattices with DL 0 enables additional improvement: +2.85

Software: Moses (with extended lattice support). Optimizer: Batch MIRA (15 iterations, BLEU averaged over 3 runs).

Training data: NTCIR-8 Patent Translation Task. Target tokenization: KyTea 5 Neubig et al. (2011). LM (5-gram): NTCIR-8. Dev/Test: NTCIR-7/NTCIR-9.



Summary of our Contributions

Bilingual Head Direction Entropy:

- ▶ Information-theoretic measure of word order freedom

Permutation Lattices produced from preordering models:

- ▶ Valuable tool to address uncertainty in word order
- ▶ Strict word order (En→Ja)
 - Additional improvements over 1-best preordering with DL 0
- ▶ Free word order (En→De)
 - Lattice Silver Training important for model estimation
 - Enable automatic preordering with DL 0



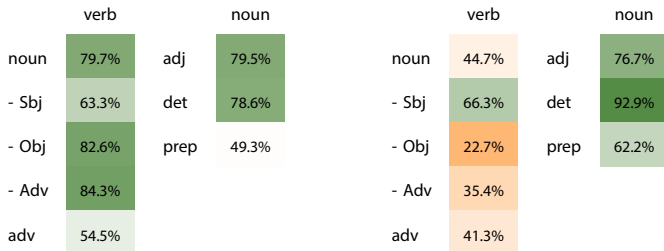
Thank You!
Questions?



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Word Order Freedom: Micro Level



(a) English–Japanese

(b) English–German



Number of Permutations vs. Lattice Size

Permutations	Kendall τ	Lattice	
		States	Transitions
Monotone	83.78	23	22
5	84.69	24	52
10	85.23	33	69
100	86.20	72	138
1000	86.75	123	233

Table: Permutations and lattice size (En-De).



Experiments: Neural Reordering on English-German

	DL	Translation	Word order
		BLEU	Kendall τ
Baseline	6	21.76	54.75
Oracle order	6	26.68	58.05
	0	26.41	57.92
First-best	6	21.21 ^A	53.44
Lattice (silver)	0	21.88 ^B	54.51

Software: Moses (with extended lattice support).

Optimizer: Batch MIRA (15 iterations, BLEU averaged over 3 runs).

Training data: 4.5m sentence pairs (WMT16). LM (5-gram): 189m sentences (WMT16).

Dev/Test: newstest2014/newstest2015.

^AStat. significant against baseline. ^BStat. significant against first-best.



Experiments: Reordering Grammar on English-Japanese

	DL	Translation	Word order
		BLEU	Kendall τ
Baseline	6	29.65	44.87
Oracle order	6	34.22	56.23
	0	30.55	53.98
First-best	6	32.14 ^A	49.68
Lattice	0	32.50 ^{AB}	50.79

Software: Moses (with extended lattice support). Optimiser: Batch MIRA (15 iterations).
Training data: NTCIR-8 Patent Translation Task. Target tokenization: KyTea 5 Neubig et al. (2011). LM
(5-gram): NTCIR-8. Dev/Test: NTCIR-7/NTCIR-9.

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