

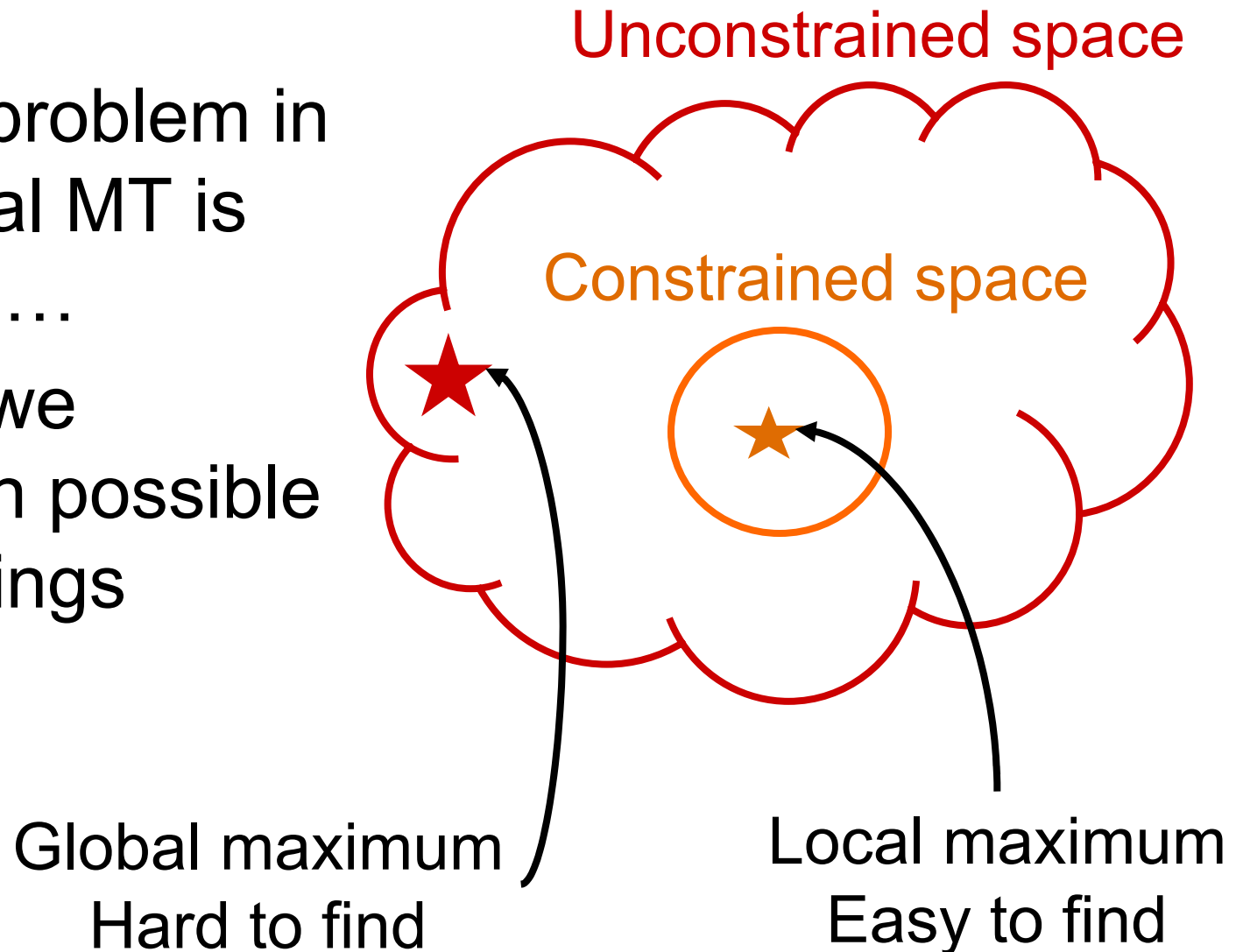
Reordering Constraints for Phrase-based MT

Richard Zens, Hermann Ney, Taro
Watanabe, and Eiichiro Sumita

Karim Filali Oct 27, 2004

Problem

- Search problem in Statistical MT is NP-hard...
- unless we constrain possible re-orderings



Review: Statistical MT

- Source channel approach

– Source: e_1^I Target: f_1^J

$$\hat{e}_1^I = \operatorname{argmax}_{e_1^I} \{ \Pr(e_1^I) \Pr(f_1^J | e_1^I) \}$$

- Alternative: model $\Pr(E|F)$ directly

$$\hat{e}_1^I = \operatorname{argmax}_{e_1^I} \{ \sum_{m=1}^M \lambda_m h_m(e^I, f^J) \}$$

Phrase-based MT /Translation Templates

- Train translation model in the two directions $f \rightarrow e$ and $e \rightarrow f$
- For each direction, calculate Viterbi alignment
- Create intersection alignment matrix from the above alignments
- Determine word classes
- Estimate word translation parameters and translation template probabilities
- For search, first find best segmentation of source sentence into alignment templates

Log-linear Model: Feature Functions Used

- Phrase translation model
- Word translation model
- Tri-gram LM
- Class-based five-gram LM
- Word penalty
- Alignment template penalty

DP Algorithm using IBM Constraints

$$Q(1, \emptyset, \$) = 1$$

$$Q(j, S, e) = \max\{$$

$$\max_{e', \tilde{e}}$$

$$\max_{j-M \leq j' < j} Q(j', S, e') p(f_{j'}^{j-1} | \tilde{e}) p(\tilde{e} | e'),$$

$$\max_{(j', l) \in S', S = S' \setminus (j', l)} Q(j, S', e') p(f_{j'}^{j'+l-1} | \tilde{e}) p(\tilde{e} | e') \},$$

$$\max_{j-M \leq j' < j, S': S = S \cup (j', j-j') \wedge |S'| < k} Q(j', S', e) \}$$

$$Q(J+2, \emptyset, \$) = \max_e Q(J+1, \emptyset, e) p(\$ | e)$$

ITG constraints for Phrase-based MT

- Inversion Transduction Grammars
 - $A \rightarrow [AA] \mid \langle AA \rangle \mid f/e \mid f/\varepsilon \mid \varepsilon/e$
 - $[]$: concatenation
 - $\langle \rangle$: inverted concatenation
- Example: [Leusch'02]
 - “We will meet at noon in the lobby”
 - “We will meet in the lobby at twelve o'clock”
 - $[we/we \text{ will/will } meet/meet \langle [at/at \text{ noon/twelve } \varepsilon/o'clock] [in/in \text{ the/the } lobby/lobby] \rangle$

DP algorithm using ITG Constraints

$$Q_{j_l, j_r, e_b, e_t} = \max_{j_l \leq k < j_r, e', e''} \{ Q_{j_l, j_r, e_b, e_t}^0, \\ Q_{j_l, k, e_b, e'} Q_{k+1, j_r, e'', e_t} p(e'' | e') p_m, \\ Q_{k+1, j_r, e_b, e'} Q_{j_l, k, e'', e_t} p(e'' | e') (1 - p_m) \}$$

Forbidden Reordering Patterns

4				
3				
2				
1				
	a	b	c	d

(1)

4				
3				
2				
1				
	a	b	c	d

(2)

Beam Search Algorithm

- Start with beam search decoder for unconstrained algorithm
- Let
 - jc : current source sentence position
 - jn : candidate source sentence position
 - $cov[j]$: j^{th} position has been translated
- Add constraints
 - If $jn < jc$ for all $jn < j < jc$: $cov[j] \rightarrow cov[j+1]$
 - If $jc < jn$ for all $jc < j < jn$: $cov[j] \rightarrow cov[j-1]$ (cannot move from uncovered position to covered one)
- We can prove
 - above constraints equivalent to ITG constraints, and
 - we can generate all reorderings that don't violate ITG constraints

Experiments

Setup

- Japanese-English tasks
 - Basic Travel Expression Corpus (BTEC): travel phrasebook entries
 - Spoken Language Database (SLDB): hotel reservation transcriptions
- Evaluation Criteria:
 - WER
 - PER
 - BLEU
 - NIST

BTEC Corpus Statistics

	Japanese	English
train sentences	152 K	
words	1044 K	893K
vocab	17047	12020
dev sentences	500	
words	3361	2858
test sentences	510	
words	3498	-

SLDB Corpus Statistics

	Japanese	English
train sentences	15 K	
words	201 K	190 K
vocab	4757	3663
test sentences	330	
words	3940	-

BTEC WER Results: Short vs. Long Sentences

	% WER			
	sentence length			time (ms)
reorder	short	long	all	
mon	11.4	16.6	12.7	73
skip 1	10.8	13.5	11.4	134
2	10.8	13.4	11.4	169
free	10.8	13.8	11.5	194
ITG	10.6	12.2	11.0	164

SLDB WER Results: Short vs. Long Sentences

	% WER			
	sentence length			time (ms)
reorder	short	long	all	
mon	32.0	52.6	48.1	911
skip 1	31.9	51.1	46.9	3175
2	32.0	51.4	47.2	4549
free	32.0	51.4	47.2	4993
ITG	31.8	50.9	46.7	4472

BTEC Results (510 sents): All Metrics

	error rates (%)		accuracy measures	
reorder	WER	PER	BLEU (%)	NIST
mon	12.7	10.6	86.8	14.14
skip 1	11.4	10.1	88.0	14.19
2	11.4	10.1	88.1	14.20
free	11.5	10.0	88.0	14.19
ITG	11.0	9.9	88.2	14.25

SLDB Results (330 sents): All Metrics

	error rates (%)		accuracy measures	
reorder	WER	PER	BLEU (%)	NIST
mon	48.1	35.5	54.4	9.45
skip 1	46.9	35.0	56.8	9.71
2	47.2	35.1	57.1	9.74
free	47.2	34.9	57.1	9.75
ITG	46.7	34.6	57.1	9.76

Conclusions

- ITG constraints best compared to
 - Unconstrained search (*)
 - Best in theory but 1) search space too large and 2) re-ordering probabilities not reliably estimated
 - (*) Difference doesn't seem significant for BLEU/NIST scores
 - IBM constraints
 - Performance equivalent to unconstrained case for skips > 2
 - Monotone search
 - Works reasonably well for short sentences (< 10 words)

Related Work

- Zens et al. “A comparative study on reordering constraints in SMT” ‘03
 - Word-based translation
 - Better BLEU % score using IBM constraints (42.5 vs 37.1)
 - Slightly better WER using ITG constraints (45.6 vs 46.2)
 - Extended ITG constraints: better alignment coverage
- Leusch et al. “A novel string-to-string distance measure with applications to MT evaluation” ‘02
- Doi and Sumita “Splitting Input Sentence for MT using LM with sentence similarity”