

Statistical NLP Spring 2010

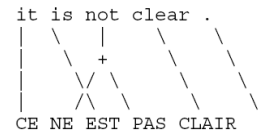


Lecture 18: Phrase / Syntactic MT

Dan Klein – UC Berkeley

Decoding

- First, consider word-to-word models
 - Finding best alignments is easy
 - Finding translations is hard (why?)



Bag “Generation” (Decoding)

Exact reconstruction (24 of 38)

Please give me your response as soon as possible.
⇒ Please give me your response as soon as possible.

Reconstruction preserving meaning (8 of 38)

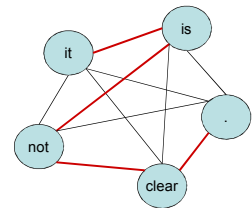
Now let me mention some of the disadvantages.
⇒ Let me mention some of the disadvantages now.

Garbage reconstruction (6 of 38)

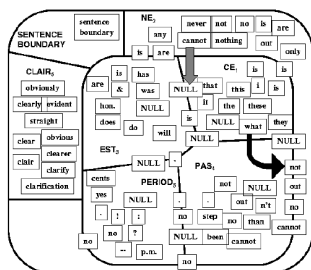
In our organization research has two missions.
⇒ In our missions research organization has two.

Bag Generation as a TSP

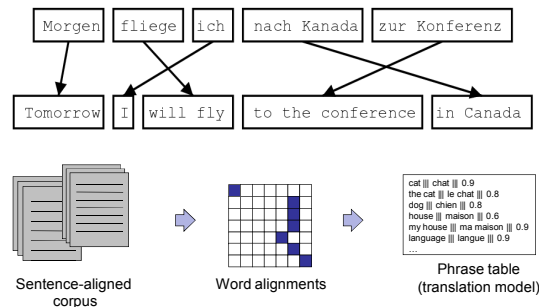
- Imagine bag generation with a bigram LM
 - Words are nodes
 - Edge weights are $P(w|w')$
 - Valid sentences are Hamiltonian paths
- Not the best news for word-based MT (and it doesn't get better with phrases)



IBM Decoding as a TSP



Phrase-Based Systems



The Pharaoh "Model"

[Koehn et al, 2003]

$P(e|g) = P(\{\bar{g}_i\}|g) \prod_i \phi(\bar{e}_i|\bar{g}_i)d(a_i - b_{i-1})$

Segmentation Translation Distortion

The Pharaoh "Model"

$$P(f|e) = P(\{\bar{e}_i\}|e) \prod_i \phi(\bar{f}_i|\bar{e}_i)d(a_i - b_{i-1})$$

$\frac{1}{K}$ $\frac{\text{count}(\bar{f}_i, \bar{e}_i)}{\text{count}(\bar{e}_i)}$ $\alpha^{|a_i - b_{i-1}|}$

Where do we get these counts?

Counting Phrase Pairs

Input:

Gracias , lo haré de muy buen grado .
Thank you , I shall do so gladly .

*First, we learn word alignments,
then we infer aligned phrases.*

Gloss

Gracias Thanks
, ,
lo haré , that
de do [first, future]
muy of
buen very
grado good
 degree
Thank you , I shall do so gladly .

Phrase-Based Decoding

这 7人 中包括 来自 法国 和 俄罗斯 的 宇航 员 .

the	7	people	including	by	some	and	the	the	astronauts													
it	7	people	including	by	france	and	the	the	astronauts	international	astronautical											
this	7	people	including	from	the	france	and	the	the	astronauts	international	astronautical										
those	7	among	including	from	the	france	and	of	the	astronauts	international	astronautical										
that	7	persons	including	from	the	france	and	of	the	astronauts	international	astronautical										
	7	persons	including	from	the	france	and	of	the	astronauts	international	astronautical										
	7	members	include	from	france	and	the	the	astronauts	international	astronautical											
	7	populations	include	from	france	and	the	the	astronauts	international	astronautical											
	7	populations	include	from	france	and	the	the	astronauts	international	astronautical											
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Decoder design is important: [Koehn et al. 03]

The Pharaoh Decoder

Maria	no	dio	una	bofetada	a	la	bruja	verde
Mary	not	gave	a	slap	to	the	witch	green
	did not		a	slap	by			green witch
	no		slap		to	the		
	did not	gave			to			
				slap		the	witch	

Maria no dio una bofetada a la bruja verde

Mary did not slap the green witch

- Probabilities at each step include LM and TM

Hypothesis Lattices

Maria	no	dio	una	bofetada	a	la	bruja	verde
Mary	not	gave	a	slap	to	the	witch	green
	did not		a	slap	by			green witch
	no		slap		to	the		
	did not	gave			to			
				slap		the	witch	

Joe: [] → did not give (p=0.092)

Mary: [] → did not give (p=0.534) → did not (p=0.164) → give (p=0.092)

Pruning

Maria no dio una bofetada a la bruja verde

e: Mary did not
f: **-----
p: 0.154

better
partial
translation

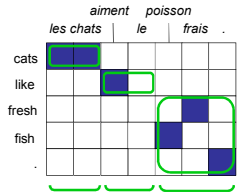
e: the
f: -----*--
p: 0.354

covers
easier part
--> lower cost

- Problem: easy partial analyses are cheaper
 - Solution 1: use beams per foreign subset
 - Solution 2: estimate forward costs (A*-like)

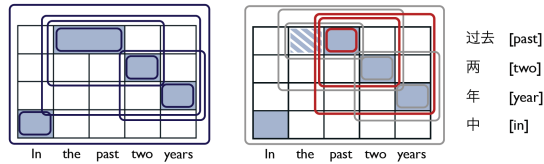
Phrase Scoring

$$\phi_{\text{score}}(\bar{e}_j | f_i) = \frac{c(f_i, \bar{e}_j)}{c(f_i)}$$



- Learning weights has been tried, several times:
 - [Marcu and Wong, 02]
 - [DeNero et al, 06]
 - ... and others
- Seems not to work well, for a variety of partially understood reasons
- Main issue: big chunks get all the weight, obvious priors don't help
 - Though, [DeNero et al 08]

Extraction Sets



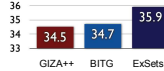
$$\phi(\mathcal{A}) = \sum_{(i,j) \in \mathcal{A}_s} \phi(i,j) + \sum_{[g,h] \Rightarrow [k,l] \in R_{\mathcal{A}}(\mathcal{A})} \phi(g,h,k,l)$$

$P(\text{in}|\text{中}) = 0.8$
InDictionary = 1.0

Count(the past two years, 过去两年) = 7
Size(4,3) = 1

Translation quality for Chinese-to-English

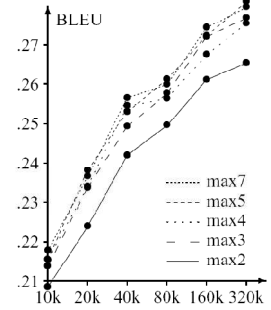
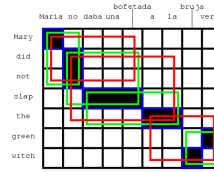
BLEU



[DeNero and Klein, in submission]

Phrase Size

- Phrases do help
 - But they don't need to be long
 - Why should this be?

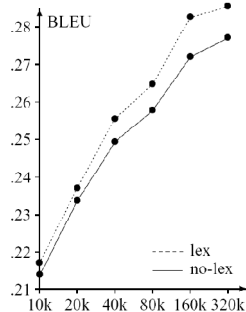


Lexical Weighting

$$\phi(\bar{f}_i | \bar{e}_i) = \frac{\text{count}(\bar{f}_i, \bar{e}_i)}{\text{count}(\bar{e}_i)} p_w(\bar{f}_i | \bar{e}_i)$$

f1 f2 f3
NULL -- -- ##
e1 ## -- --
e2 -- ## --
e3 -- ## --

$$p_w(\bar{f} | \bar{e}, a) = p_w(f_1 f_2 f_3 | e_1 e_2 e_3, a) \\ = w(f_1 | e_1) \\ \times \frac{1}{2} (w(f_2 | e_2) + w(f_2 | e_3)) \\ \times w(f_3 | \text{NULL})$$



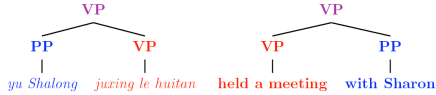
WSD?

- Remember when we discussed WSD?
 - Word-based MT systems rarely have a WSD step
 - Why not?

Syntax-Based MT

- synchronous context-free grammars (SCFGs)
- context-free grammar in two dimensions
- generating pairs of strings/trees simultaneously
- co-indexed nonterminal further rewritten as a unit

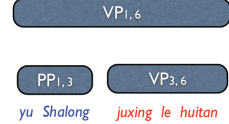
$VP \rightarrow PP^{(1)} VP^{(2)}, \quad VP^{(2)} PP^{(1)}$
 $VP \rightarrow \textit{juxing le huitan}, \textit{held a meeting}$
 $PP \rightarrow \textit{yu Shalong}, \textit{with Sharon}$



Translation by Parsing

- translation with SCFGs => monolingual parsing
- parse the source input with the source projection
- build the corresponding target sub-strings in parallel

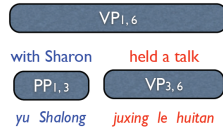
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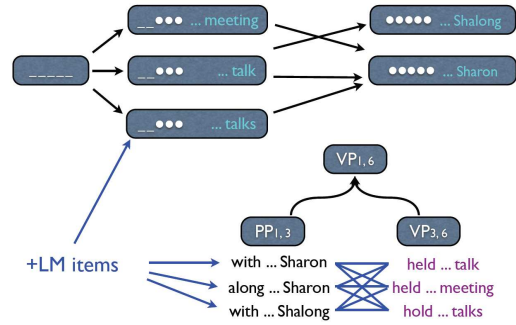
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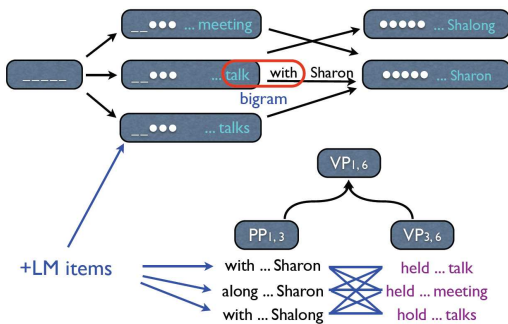
$VP \rightarrow pp^{(1)} vp^{(2)}, \quad vp^{(2)} pp^{(1)}$
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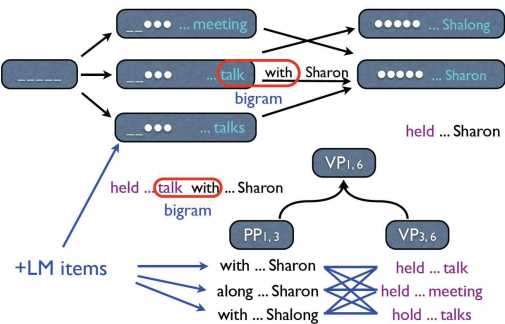
Compact Forests

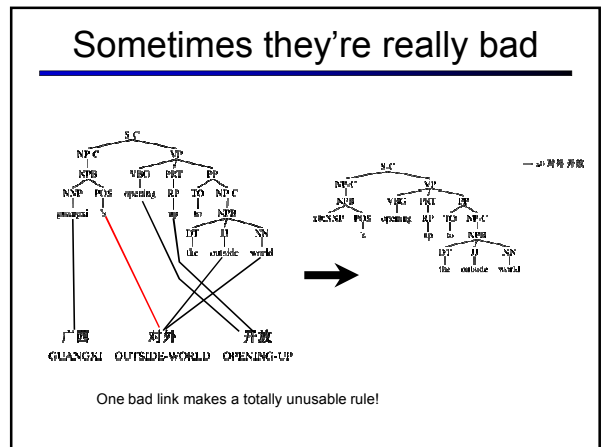
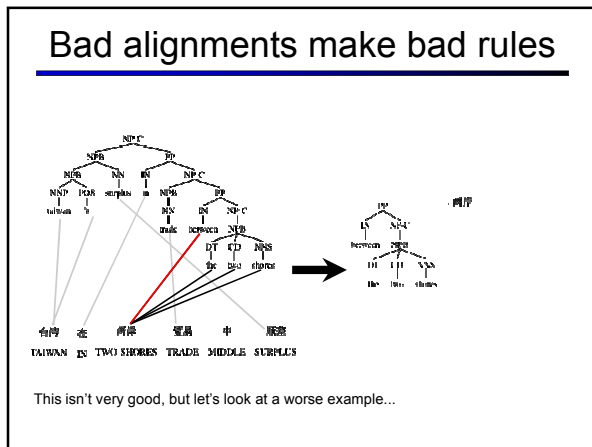
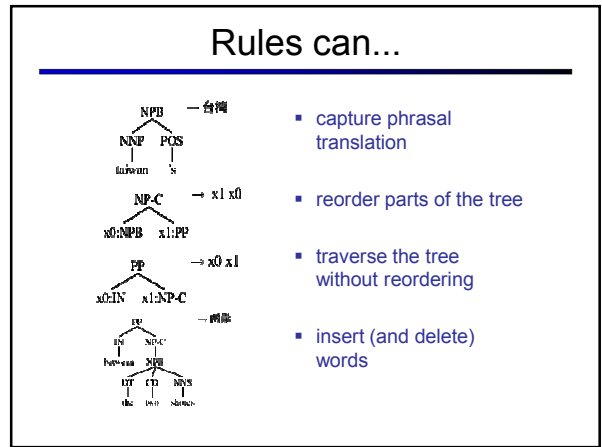
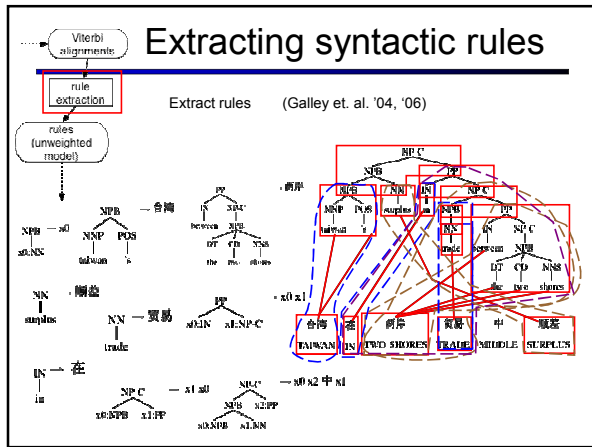
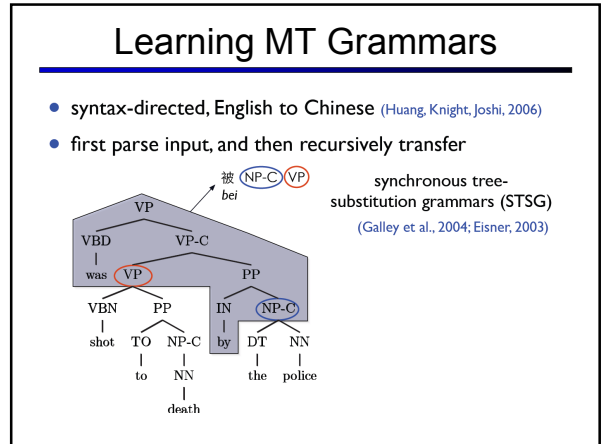
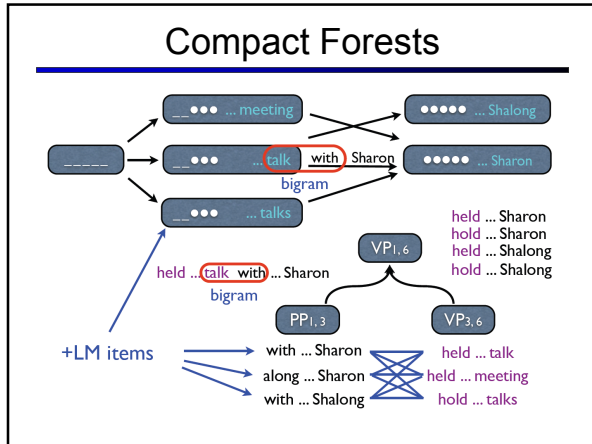


Compact Forests

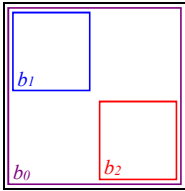


Compact Forests





Discriminative Block ITG



Features

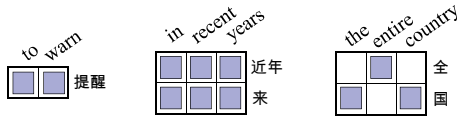
$$\phi(b_0, s, s')$$

$$\phi(b_1, s, s')$$

$$\phi(b_2, s, s')$$

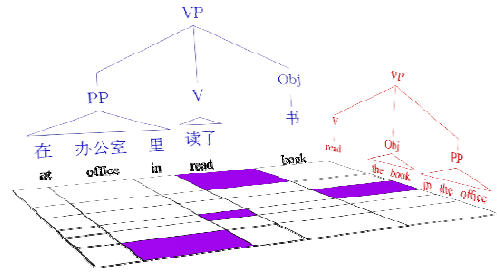
$$\phi(A) = \sum_{b \in A} \phi(b, s, s')$$

$$P(A) \propto \exp(\theta, \phi(A))$$



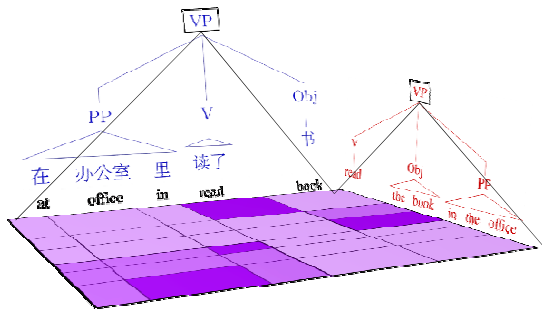
[Haghighi, Blitzer, Denero, and Klein, ACL 09]

Syntactic Correspondence

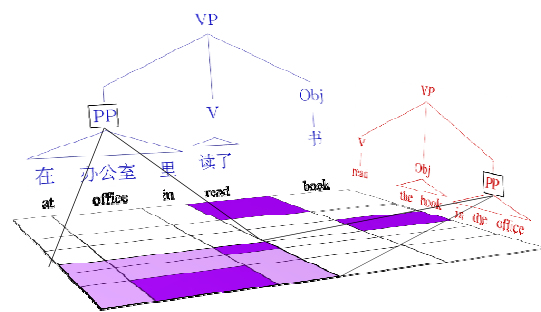


Build a model $p_{\theta}(\Delta, \triangle, \text{读}, \text{中文}, \text{EN})$

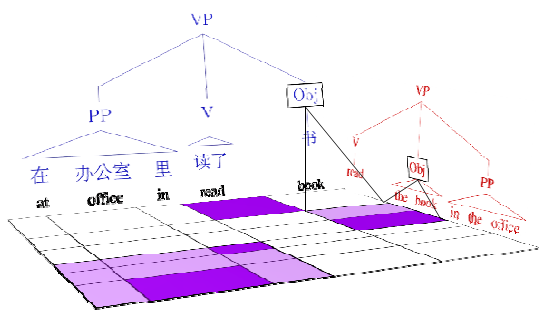
Synchronous Grammars?



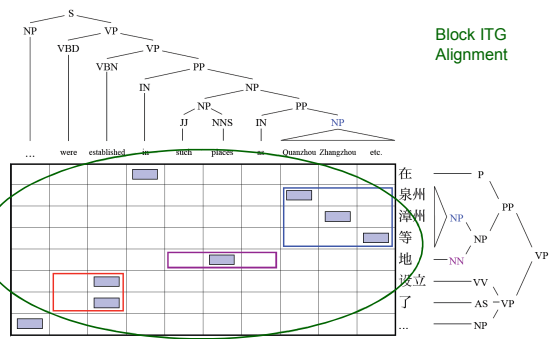
Synchronous Grammars?



Synchronous Grammars?



Adding Syntax: Weak Synchronization



Adding Syntax: Weak Synchronization

Separate PCFGs

Adding Syntax: Weak Synchronization

Get points for synchronization; not required

Weakly Synchronous Features

Parsing	Alignment
$\phi_{\mathcal{P}}(IP, s)$	$\phi_{\mathcal{A}}(b_0, s, s')$
$\phi_{\mathcal{P}}(NP, s)$	$\phi_{\mathcal{A}}(b_1, s, s')$
$\phi_{\mathcal{P}}(VP, s)$	$\phi_{\mathcal{A}}(b_2, s, s')$
Agreement	
$\phi_{\mathcal{E}}(S, s')$	$\phi_{\mathcal{D}}(IP, b_0)$
$\phi_{\mathcal{E}}(NP, s')$	$\phi_{\mathcal{D}}(b_0, S)$
$\phi_{\mathcal{E}}(AP, s')$	$\phi_{\mathcal{D}}(b_1, NP)$
$\phi_{\mathcal{E}}(VP, s')$	$\phi_{\mathcal{D}}(IP, b_0, S)$

Weakly Synchronous Model

$p_{\theta}(\triangle, \blacktriangle, \text{办公室} | \text{EN, 中文})$

Feature Type 1: Word Alignment
 $\phi(\text{办公室}, \text{EN}, \text{中文})$

Feature Type 2: Monolingual Parser
 $\phi(\blacktriangle, \text{EN})$

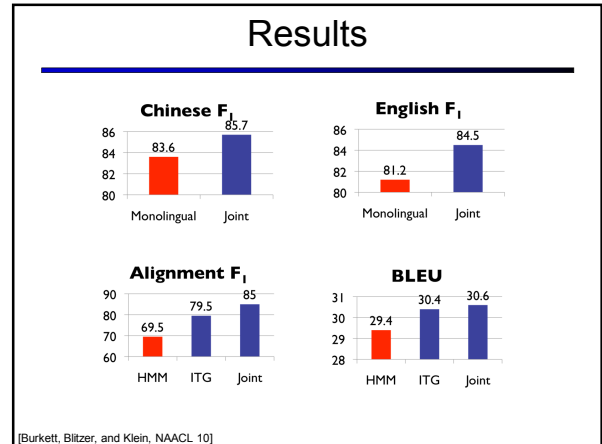
Feature Type 3: Agreement
 $\phi(\triangle, \blacktriangle, \text{办公室})$

Inference: Structured Mean Field

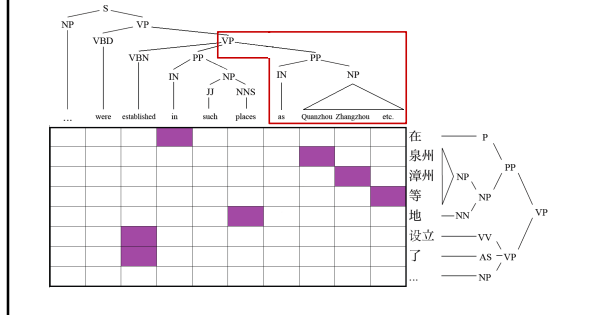
- Problem: Summing over weakly aligned hypotheses is intractable
- Factored approximation: $p_{\theta}(\triangle, \blacktriangle, \text{办公室} | \text{EN, 中文}) \approx q(\triangle)q(\blacktriangle)q(\text{办公室})$
- Set q to minimize $KL(q(\triangle)q(\blacktriangle)q(\text{办公室}), p_{\theta}(\triangle, \blacktriangle, \text{办公室} | \text{EN, 中文}))$

Algorithm

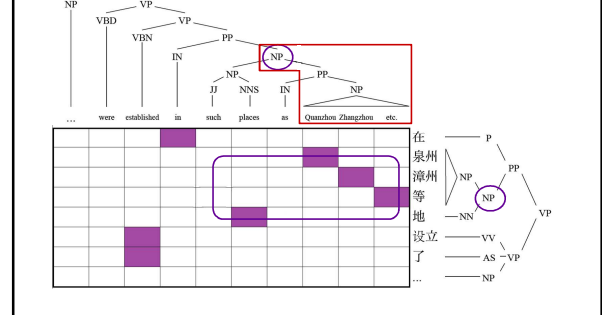
- Initialize: $q(\triangle)q(\blacktriangle)q(\text{办公室})$
- Iterate:
 - $q(\triangle) \propto \exp\{\langle \theta, \phi(\triangle, E_q(\blacktriangle), E_q(\text{办公室})) \rangle\}$
 - $q(\blacktriangle) \propto \exp\{\langle \theta, \phi(E_q(\triangle), \blacktriangle, E_q(\text{办公室})) \rangle\}$
 - $q(\text{办公室}) \propto \exp\{\langle \theta, \phi(E_q(\triangle), E_q(\blacktriangle), \text{办公室}) \rangle\}$



Incorrect English PP Attachment



Corrected English PP Attachment



Improved Translations

目前导致飞机相撞的原因尚不清楚，当地民航部门将对此展开调查
 Cur- cause plane crash DE reason yet not clear, local civil bureau will toward open investi-
 rently gations

Reference

At this point the cause of the plane collision is still unclear. The local caa will launch an investigation into this .

Baseline (GIZA++)

The cause of plane collision is still not clear yet, local civil aviation department will investigate this .

Bilingual Adaptation Model

The cause of plane collision remained unclear, local civil aviation departments will launch an investigation .