

Statistical NLP Spring 2010

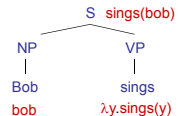


Lecture 20: Compositional Semantics

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Truth-Conditional Semantics

- Linguistic expressions:
 - "Bob sings"



- Logical translations:
 - sings(bob)
 - Could be $p_{1218}(e_{397})$

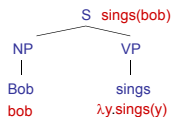
- Denotation:
 - $[[\text{bob}]]$ = some specific person (in some context)
 - $[[\text{sings}(\text{bob})]]$ = ???

- Types on translations:
 - bob : e (for entity)
 - sings(bob) : t (for truth-value)

Truth-Conditional Semantics

- Proper names:
 - Refer directly to some entity in the world
 - Bob : bob $[[\text{bob}]]^w \rightarrow ???$

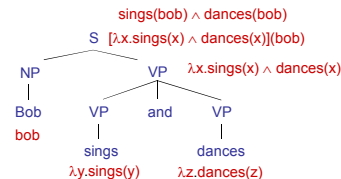
- Sentences:
 - Are either true or false (given how the world actually is)
 - Bob sings : sings(bob)



- So what about verbs (and verb phrases)?
 - sings must combine with bob to produce sings(bob)
 - The λ -calculus is a notation for functions whose arguments are not yet filled.
 - sings : $\lambda x.sings(x)$
 - This is *predicate* – a function which takes an entity (type e) and produces a truth value (type t). We can write its type as $e \rightarrow t$.
 - Adjectives?

Compositional Semantics

- So now we have meanings for the words
- How do we know how to combine words?
- Associate a combination rule with each grammar rule:
 - S : $\beta(\alpha) \rightarrow NP : \alpha \quad VP : \beta$ (function application)
 - VP : $\lambda x . \alpha(x) \wedge \beta(x) \rightarrow VP : \alpha \quad \text{and} : \emptyset \quad VP : \beta$ (intersection)
- Example:



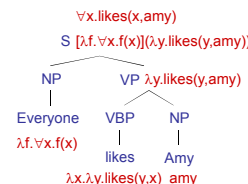
Denotation

- What do we do with logical translations?
 - Translation language (logical form) has fewer ambiguities
 - Can check truth value against a database
 - Denotation ("evaluation") calculated using the database
 - More usefully: assert truth and modify a database
 - Questions: check whether a statement in a corpus entails the (question, answer) pair:
 - "Bob sings and dances" \rightarrow "Who sings?" + "Bob"
 - Chain together facts and use them for comprehension

Other Cases

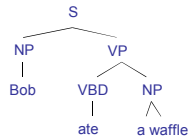
- Transitive verbs:
 - likes : $\lambda x.\lambda y.likes(y,x)$
 - Two-place predicates of type $e \rightarrow (e \rightarrow t)$.
 - likes Amy : $\lambda y.likes(y,Amy)$ is just like a one-place predicate.

- Quantifiers:
 - What does "Everyone" mean here?
 - Everyone : $\lambda f.\forall x.f(x)$
 - Mostly works, but some problems
 - Have to change our NP/VP rule.
 - Won't work for "Amy likes everyone."
 - "Everyone likes someone."
 - This gets tricky quickly!



Indefinites

- First try
 - "Bob ate a waffle" : $\text{ate}(\text{bob}, \text{waffle})$
 - "Amy ate a waffle" : $\text{ate}(\text{amy}, \text{waffle})$
- Can't be right!
 - $\exists x : \text{waffle}(x) \wedge \text{ate}(\text{bob}, x)$
 - What does the translation of "a" have to be?
 - What about "the"?
 - What about "every"?



Grounding

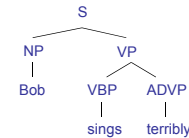
- Grounding
 - So why does the translation $\text{likes} : \lambda x. \lambda y. \text{likes}(y, x)$ have anything to do with actual liking?
 - It doesn't (unless the denotation model says so)
 - Sometimes that's enough: wire up **bought** to the appropriate entry in a database
- Meaning postulates
 - Insist, e.g. $\forall x, y. \text{likes}(y, x) \rightarrow \text{knows}(y, x)$
 - This gets into lexical semantics issues
- Statistical version?

Tense and Events

- In general, you don't get far with verbs as predicates
- Better to have event variables e
 - "Alice danced" : $\text{danced}(\text{alice})$
 - $\exists e : \text{dance}(e) \wedge \text{agent}(e, \text{alice}) \wedge (\text{time}(e) < \text{now})$
- Event variables let you talk about non-trivial tense / aspect structures
 - "Alice had been dancing when Bob sneezed"
 - $\exists e, e' : \text{dance}(e) \wedge \text{agent}(e, \text{alice}) \wedge \text{sneeze}(e') \wedge \text{agent}(e', \text{bob}) \wedge (\text{start}(e) < \text{start}(e') \wedge \text{end}(e) = \text{end}(e')) \wedge (\text{time}(e') < \text{now})$

Adverbs

- What about adverbs?
 - "Bob sings terribly"
 - $\text{terribly}(\text{sings}(\text{bob}))?$
 - $(\text{terribly}(\text{sings}))(\text{bob})?$
 - $\exists e \text{ present}(e) \wedge \text{type}(e, \text{singing}) \wedge \text{agent}(e, \text{bob}) \wedge \text{manner}(e, \text{terrible})?$
 - It's really not this simple..



Propositional Attitudes

- "Bob thinks that I am a gummi bear"
 - $\text{thinks}(\text{bob}, \text{gummi}(\text{me}))?$
 - $\text{thinks}(\text{bob}, \text{"I am a gummi bear"})?$
 - $\text{thinks}(\text{bob}, \wedge \text{gummi}(\text{me}))?$
- Usual solution involves intensions ($\wedge X$) which are, roughly, the set of possible worlds (or conditions) in which X is true
- Hard to deal with computationally
 - Modeling other agents models, etc
 - Can come up in simple dialog scenarios, e.g., if you want to talk about what your bill claims you bought vs. what you actually bought

Trickier Stuff

- Non-Intersective Adjectives
 - $\text{green ball} : \lambda x. [\text{green}(x) \wedge \text{ball}(x)]$
 - $\text{fake diamond} : \lambda x. [\text{fake}(x) \wedge \text{diamond}(x)]? \rightarrow \lambda x. [\text{fake}(\text{diamond}(x))]$
- Generalized Quantifiers
 - $\text{the} : \lambda f. [\text{unique-member}(f)]$
 - $\text{all} : \lambda f. \lambda g. [\forall x. f(x) \rightarrow g(x)]$
 - $\text{most}?$
 - Could do with more general second order predicates, too (why worse?)
 - $\text{the}(\text{cat}, \text{meows}), \text{all}(\text{cat}, \text{meows})$
- Generics
 - "Cats like naps"
 - "The players scored a goal"
- Pronouns (and bound anaphora)
 - "If you have a dime, put it in the meter."
- ... the list goes on and on!

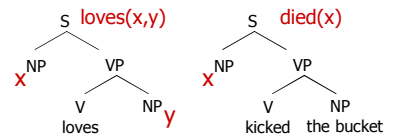
Multiple Quantifiers

- Quantifier scope
 - Groucho Marx celebrates quantifier order ambiguity:
 - "In this country a woman gives birth every 15 min.
 - Our job is to find that woman and stop her."
- Deciding between readings
 - "Bob bought a pumpkin every Halloween"
 - "Bob put a warning in every window"
 - Multiple ways to work this out
 - Make it syntactic (movement)
 - Make it lexical (type-shifting)

Implementation, TAG, Idioms

- Add a "sem" feature to each context-free rule
 - $S \rightarrow NP \text{ loves } NP$
 - $S[\text{sem}=\text{loves}(x,y)] \rightarrow NP[\text{sem}=x] \text{ loves } NP[\text{sem}=y]$
 - Meaning of S depends on meaning of NPs

- TAG version:



- Template filling: $S[\text{sem}=\text{showflights}(x,y)] \rightarrow$
I want a flight from NP[sem=x] to NP[sem=y]

Modeling Uncertainty

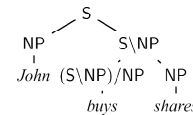
- Gaping hole warning!
- Big difference between statistical disambiguation and statistical reasoning.
 - The scout saw the enemy soldiers with night goggles.*
 - With probabilistic parsers, can say things like "72% belief that the PP attaches to the NP."
 - That means that *probably* the enemy has night vision goggles.
 - However, you can't throw a logical assertion into a theorem prover with 72% confidence.
 - Not clear humans really extract and process logical statements symbolically anyway.
 - Use this to decide the expected utility of calling reinforcements?
- In short, we need probabilistic reasoning, not just probabilistic disambiguation followed by symbolic reasoning!

CCG Parsing

- Combinatory
Categorial
Grammar

- Fully (mono-)lexicalized grammar
- Categories encode argument sequences
- Very closely related to the lambda calculus
- Can have spurious ambiguities (why?)

$John \vdash NP : john'$
 $shares \vdash NP : shares'$
 $buys \vdash (S \setminus NP) / NP : \lambda x. \lambda y. buys'xy$
 $sleeps \vdash S \setminus NP : \lambda x. sleeps'x$
 $well \vdash (S \setminus NP) \setminus (S \setminus NP) : \lambda f. \lambda x. well''(f(x))$



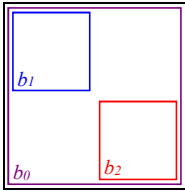
Syntax-Based MT

- 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

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



Discriminative Block ITG



Features

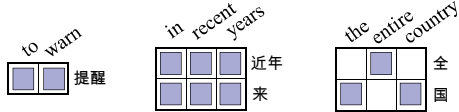
$$\phi(b_0, s, s')$$

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

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

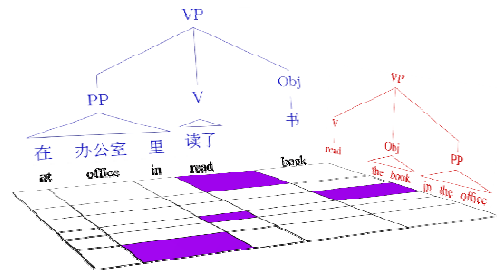
$$\phi(A) = \sum_{b \in A} \phi(b, s, s')$$

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



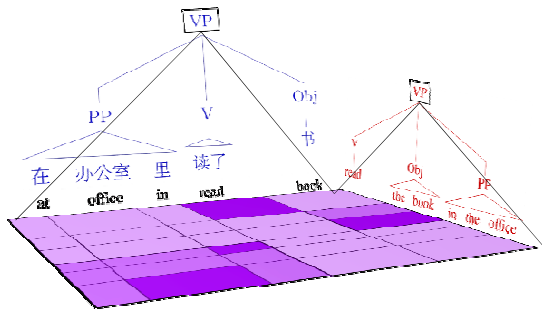
[Haghighi, Blitzer, Denero, and Klein, ACL 09]

Syntactic Correspondence

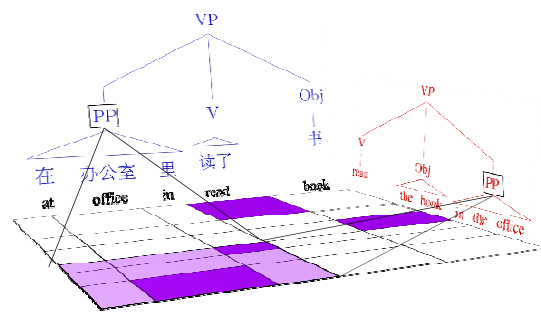


Build a model $p_{\theta}(\triangle, \blacktriangle, \square | \text{中文, EN})$

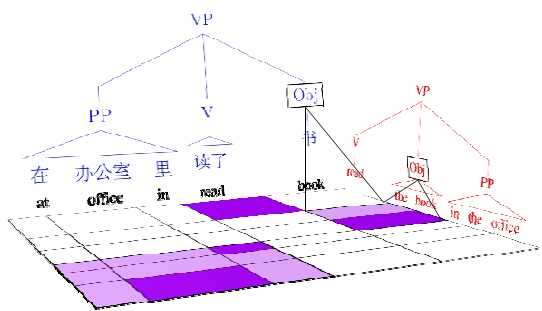
Synchronous Grammars?



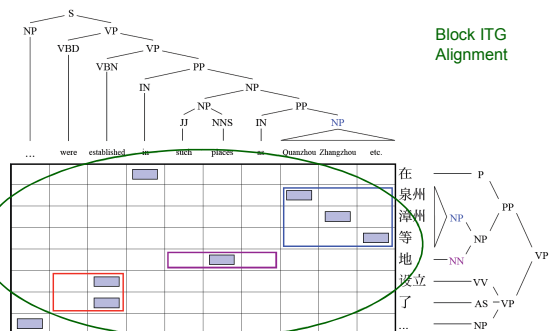
Synchronous Grammars?



Synchronous Grammars?



Adding Syntax: Weak Synchronization



Block ITG Alignment

Adding Syntax: Weak Synchronization

Separate PCFGs

Adding Syntax: Weak Synchronization

Get points for synchronization; not required

Weakly Synchronous Features

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

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
 $\phi(\blacktriangle, \text{EN})$

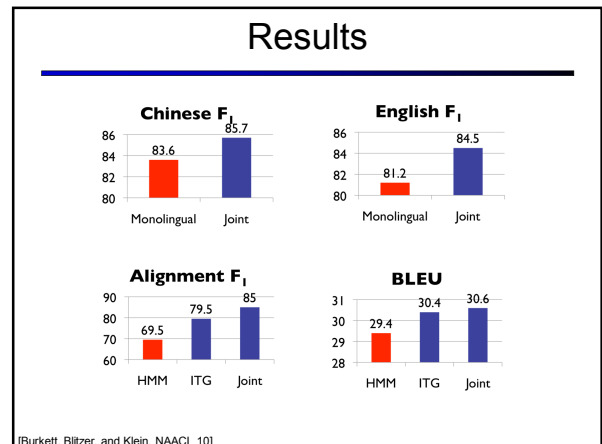
Feature Type 3: Agreement
 $\phi(\triangle, \blacktriangle, \text{办公室})$

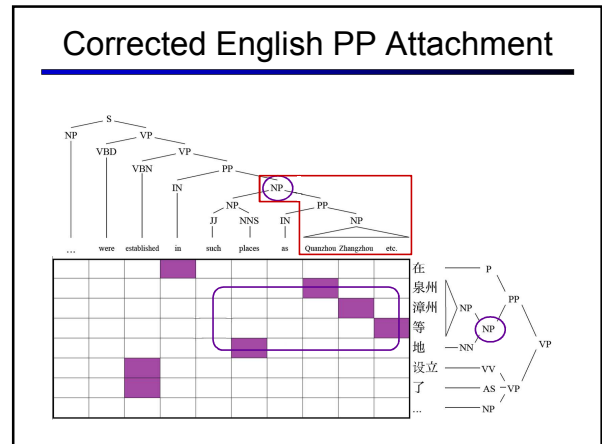
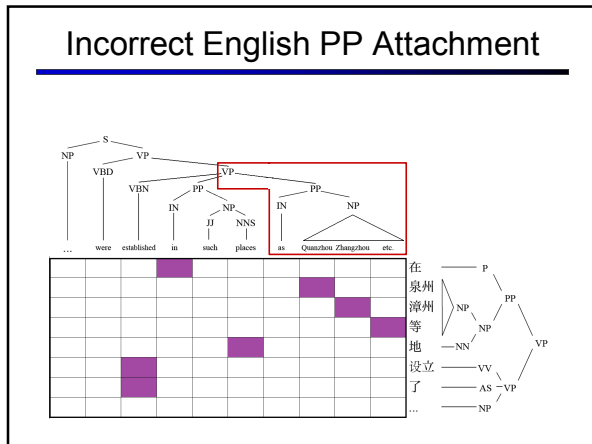
Inference: Structured Mean Field

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





Improved Translations

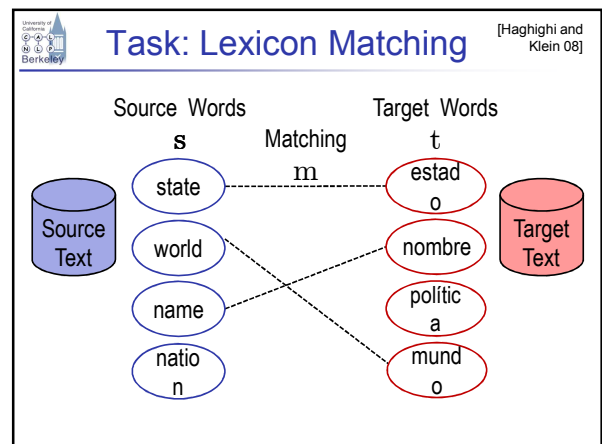
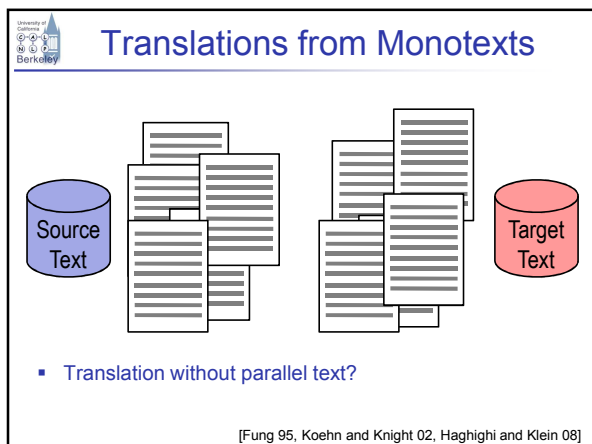
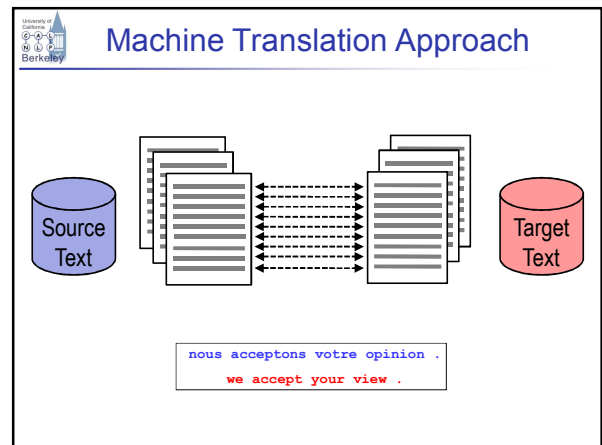
目前导致飞机相撞的原因尚不清楚，当地民航部门将对此展开调查

Cur- cause plane crash DE reason yet not clear, local civil bureau will toward open investi-
rently acro- nautics 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 planes 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 .



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Data Representation

Orthographic Features

#st	1.0
tat	1.0
te#	1.0

Context Features

world	20.0
politics	5.0
society	10.0

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Data Representation

Orthographic Features

#st	1.0
tat	1.0
te#	1.0

Context Features

world	20.0
politics	5.0
society	10.0

Orthographic Features

#es	1.0
sta	1.0
do#	1.0

Context Features

mundo	17.0
politica	10.0
sociedad	6.0

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Generative Model (CCA)

Canonical Space \mathbb{R}^d

$z \sim \mathcal{N}(0, I_d)$

$W_s z + \text{noise}$

$W_t z + \text{noise}$

Source Space \mathbb{R}^{d_s}

Target Space \mathbb{R}^{d_t}

state

estado

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Generative Model (Matching)

Source Words s

Target Words t

Matching m

state

estado

world

nombre

name

politica

nation

mundo

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Inference: Hard EM

E-Step: Find best matching

$$w_{ij} = \log p(s_i, t_j | \mathbf{m}; W_s, W_t) - \log \text{NULL}_S(s_i) - \log \text{NULL}_T(t_j)$$

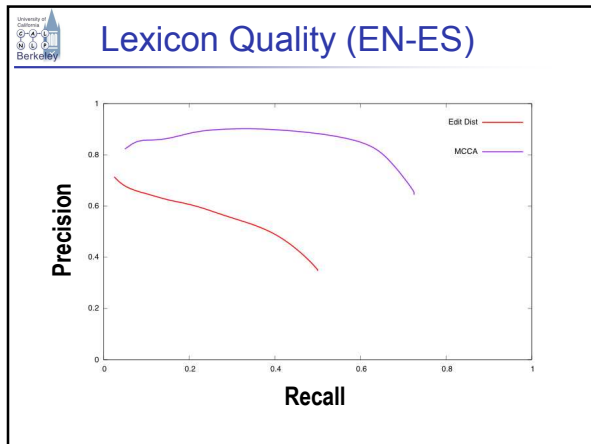
M-Step: Solve a CCA problem

$$\max_{(W_s, W_t)} \left[\sum_{(i,j) \in \mathbf{m}} \log p(s_i, t_j | \mathbf{m}; W_s, W_t) \right]$$

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Experimental Setup

- Data: 2K most frequent nouns, texts from Wikipedia
- Seed: 100 translation pairs
- Evaluation: Precision and Recall against lexicon obtained from Wiktionary
 - Report $p_{0.33}$, precision at recall 0.33



Analysis

English-Spanish		
Source	Target	Correct
education	educación	Y
pacto	pact	Y
stability	estabilidad	Y
corruption	corrupción	Y
tourism	turismo	Y
organisation	organización	Y
convenience	conveniencia	Y
syria	siria	Y
cooperation	cooperación	Y
culture	cultura	Y
protocol	protocolo	Y
north	norte	Y
health	salud	Y
action	reacción	N

Analysis

Interesting Matches		Interesting Mistakes	
health	salud	liberal	partido
traceability	rastreabilidad	Kirkhope	Gorsel
youth	juventud	action	reacción
report	informe	Albanians	Bosnia
advantages	ventajas	a.m.	horas

Language Variation

English-Chinese		
Source	Target	Correct
prices	价格	Y
network	网络	Y
population	人口	Y
reporter	孙	N
oil	石油	Y