

# TALP: Xgram-based Spoken Language Translation System



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# Outline

- Overview
- Translation generation
- Training
- IWSLT'04 Chinese-English supplied task results
- Conclusion and further work

# Overview

- TALP Statistical Machine Translation (SMT)
- Integrated speech-text approach
- Finite-State Transducer (FST) implementation
- Automatically learnt from parallel corpus
- Bilingual units called tuples

# Translation generation

- Maximising joint probability
- Variable-length N-gram of bilingual units (tuples)

$$\hat{f} = \underset{f}{\operatorname{argmax}} p(e, f)$$

$$p(e, f) = \prod_{n=1}^N p((e, f)_n | (e, f)_{n-1} \dots (e, f)_{n-X+1})$$

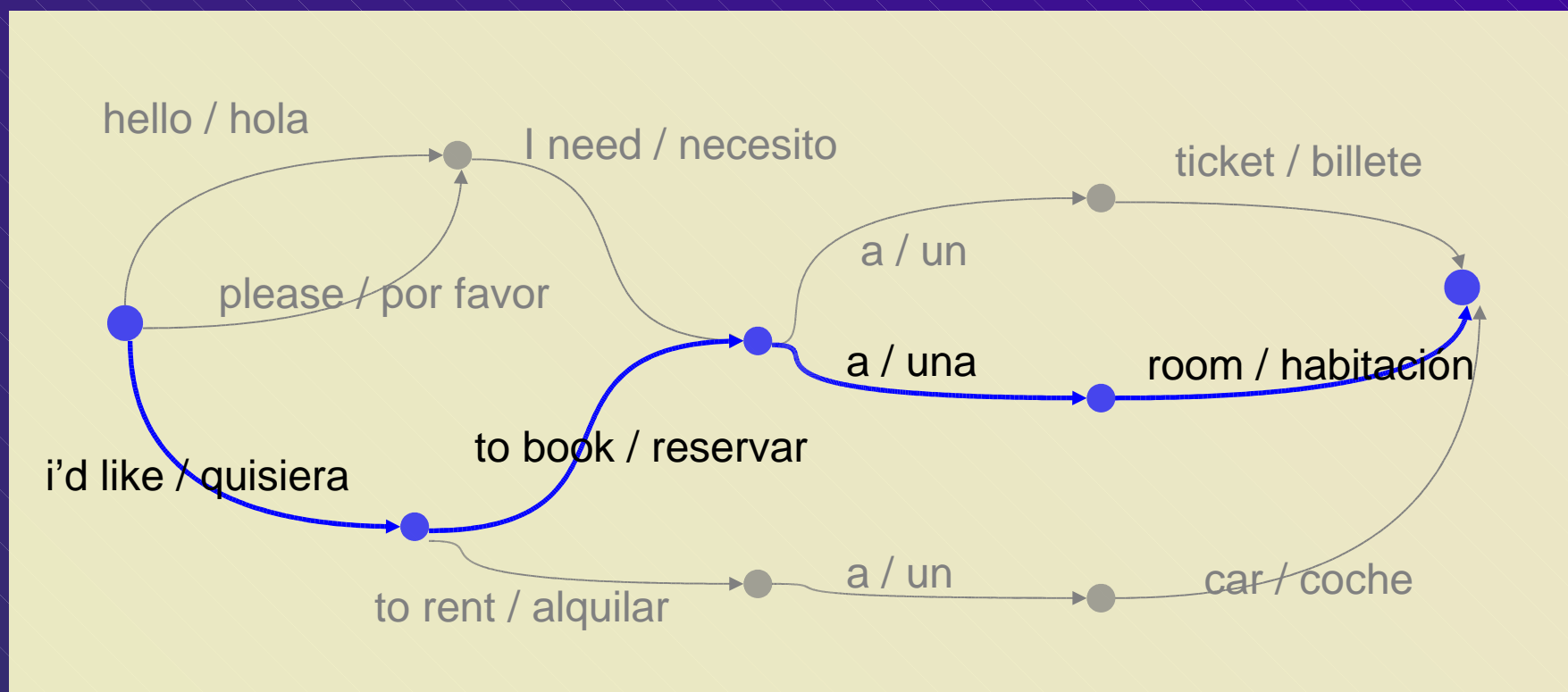
*X-gram*

*tuple*

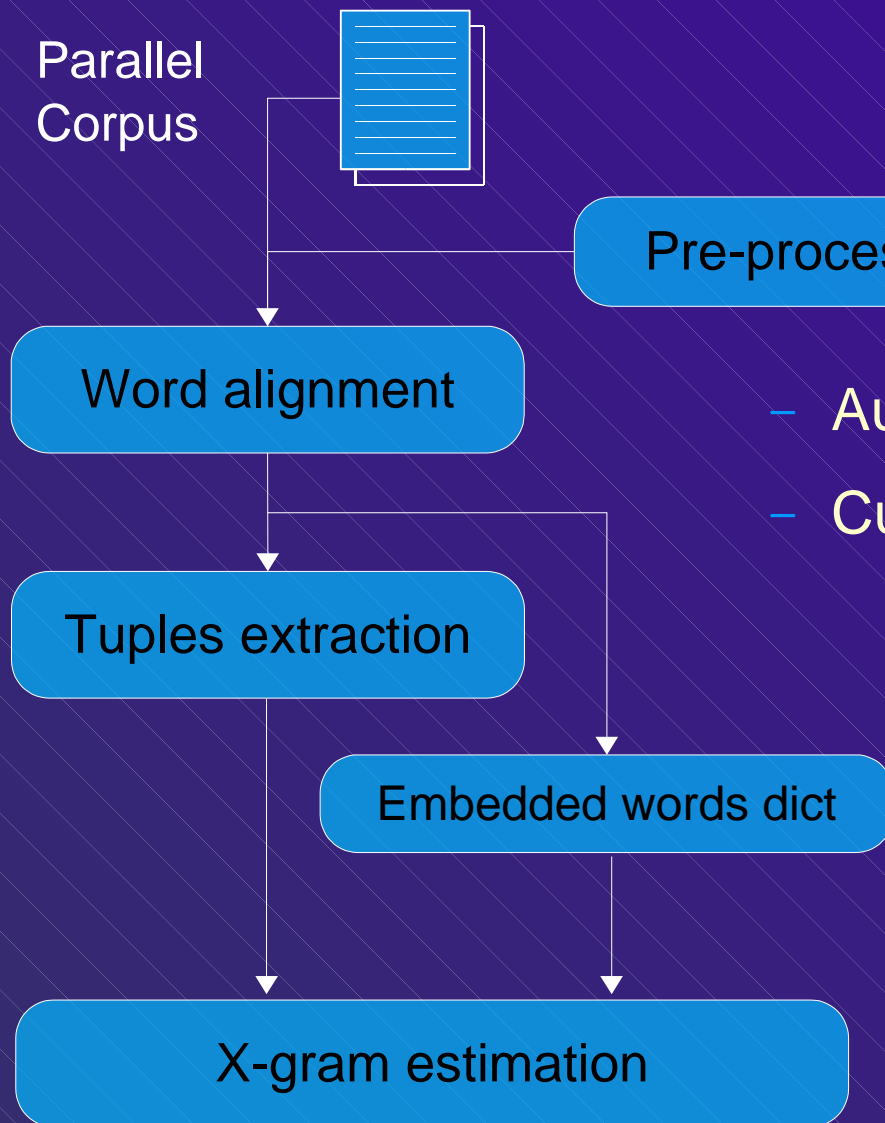
$$(e, f)_n = (e_{i(n)} \dots e_{i(n)+I(n)}, f_{j(n)} \dots f_{j(n)+J(n)})$$

# FST implementation

- Search for best-scoring path
- Speech translation: include acoustic models



# Training



- Automatic from parallel corpus
- Currently based on Giza++ alignments

# Preprocessing

- Particular for each pair of languages
- Categorisation *personal names, dates, times, numbers, ...*
- **Chinese-English** IWSLT'04 supplied track:
  - Clearing out punctuation *no gain*
  - Segmentation of longer sentences

	# sentences	Lavg
Chinese	20 K (22.2 K)	9.1 (8.2)
English		9.4 (8.5)

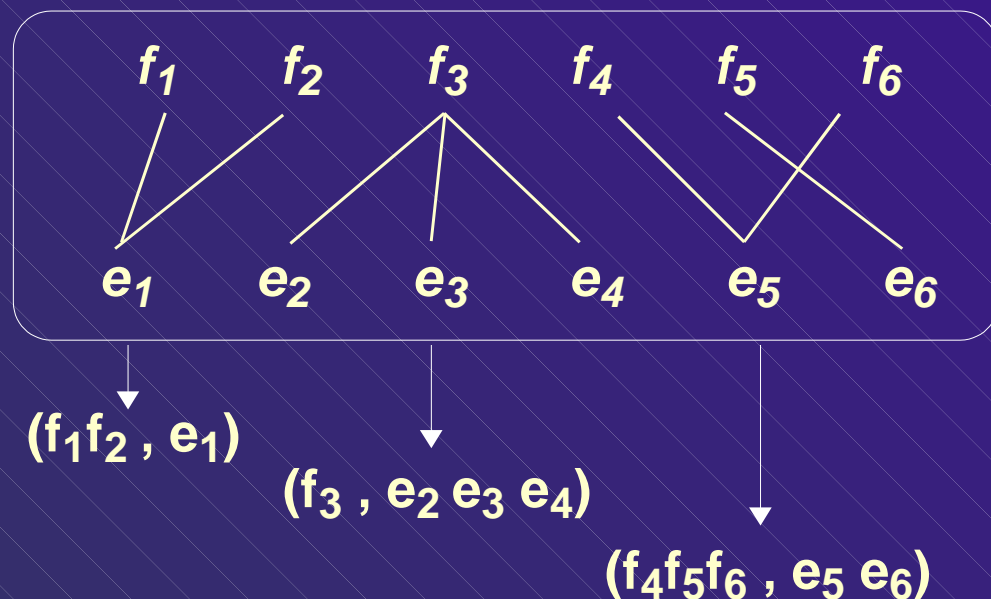
# Word alignment

- Standard GIZA++ alignments  $1^5H^53^34^3$ 
  - Source-to-Target (s2t)
  - Target-to-Source (t2s)
  - Union
  - Intersection
- Tuples can be extracted from any alignment
- Usually **union** and **s2t** are used



# Tuples extraction

- Tuples are bilingual units containing
  - one or more source words
  - zero, one or more target words
- Subset of phrases, unique under following conditions
- Example

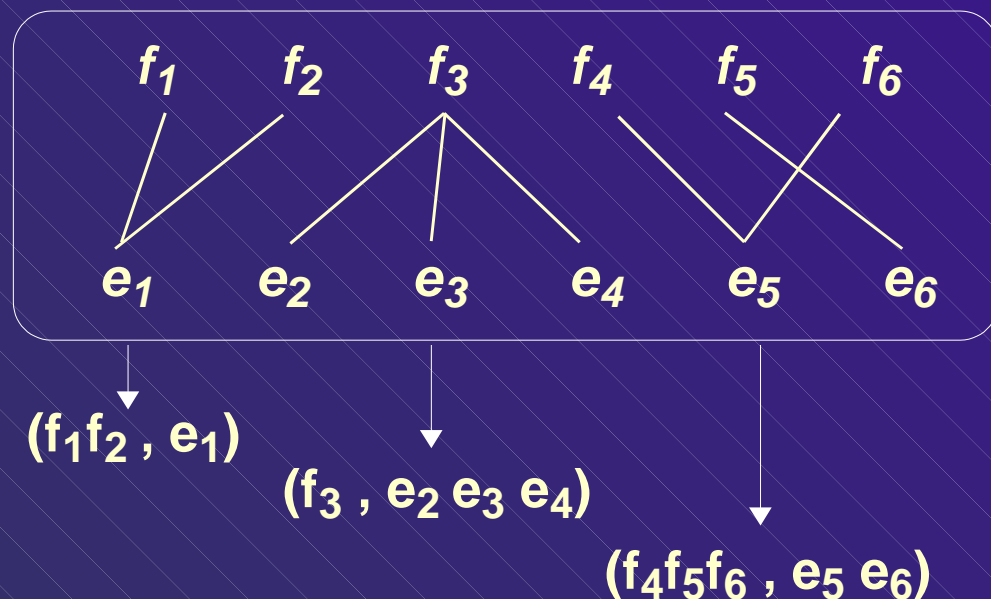


1. Monotonous segmentation of the pair
2. Words are consecutive along both source and target
3. No word in the tuple is aligned to a word outside the tuple
4. The tuple cannot be decomposed without violating 1-3

# Embedded words dictionary

- Embedded words
  - translation appears always inside a tuple (never isolated)
  - an 'accurate' dictionary is built *intersection*
- New unigrams (history independent)

## DICTIONARY ENTRIES:



Given a source word, look for the most freq. aligned words

1. Target words are consecutive
2. Target words are aligned *only* to the source word

# X-gram estimation

\$ (f<sub>1</sub>f<sub>2</sub> , e<sub>1</sub>) (f<sub>3</sub> , e<sub>2</sub> e<sub>3</sub> e<sub>4</sub>) (f<sub>4</sub>f<sub>5</sub>f<sub>6</sub> , e<sub>5</sub> e<sub>6</sub>) \$

$$p(e, f) = \prod_{n=1}^N p(T_n | T_{n-1} \dots T_{n-X+1})$$

- Usually, maximum memory is 3
- Pruning strategies
  - Min. number of times a certain history must occur  $k$
  - threshold of divergence between output prob. distributions for two nodes sharing recent history  $f$

*k is untouched (= 1)*

*f used for slight pruning (f = 0.2)*

# Chinese-English supplied track

- Results on development set
  - Union (aU) vs. s2t alignment (a2)
  - Normal vs. segmented corpus (seg)
  - Normal vs. FST pruning (f)

run	BLEU	NIST	WER	PER	GTM	E
aU	0.244	5.169	0.615	0.529	0.591	7
aU,seg	0.251	5.187	0.607	0.521	<b>0.595</b>	<b>7</b>
aU,seg,f	0.255	<b>5.210</b>	<b>0.603</b>	<b>0.518</b>	0.594	7
a2	<b>0.319</b>	3.789	0.614	0.552	0.573	16
a2,seg	0.318	3.871	0.606	0.546	0.573	18
a2,seg,f	0.314	3.678	0.607	0.548	0.570	19

→ *run A*

→ *run B*

*no embed.  
dictionary*

aU,seg,-D	0.264	4.741	0.606	0.524	0.592	7
a2,seg,-D	0.315	3.706	0.607	0.547	0.571	19

# Automatic evaluation results

## – Statistics of submitted runs

run	tuples	vcb	length	embed
aU,seg,f	97 K	27 K	3.9	4.7 K
a2,seg,f	140 K	29 K	2.9	1.5 K

- Longer tuples with Union (more embedded)
- Many tuples to NULL with s2t (28% over total, 7.5% union)

## – Results

run	BLEU	NIST	WER	PER	GTM	E
aU,seg,f	0.279	<b>6.778</b>	<b>0.556</b>	<b>0.465</b>	<b>0.647</b>	<b>5</b>
a2,seg,f	<b>0.331</b>	5.391	<b>0.550</b>	0.490	0.620	11

- Contrast BLEU vs. NIST (related to length)

# Manual evaluation results

## – Results

run	fluency	adequacy
aU,seg,f	2.792	3.022

- Expected fluency deficiency *no explicit long reordering*
- 'Much of the information' is transmitted

## • Examples:

*Translation: that what time start*

*Reference1: what time does it start*

*Translation: stomach very hurts*

*Reference1: i have a severe pain in my stomach*

# Conclusion and further research

- Tuple-based FST translation system presented
- Adequate for pairs of languages *similar in word-order*
- Further research
  - Embedded N-grams
  - Generalization of tuples
  - Explicit reordering techniques

# Thanks for attention



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