Hybrid Rule-Based – Example-Based MT: Feeding Apertium with Sub-sentential Translation Units

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Motivation

Predictability of rule-based MT (RBMT) systems:

- Lexical and structural selection is consistent
- Errors can be attributed to a particular module
- Eases postedition for dissemination

Usually RBMT systems do not benefit from the postedition effort of professional translators

- Incorporating postedition work is not trivial
- Some RBMT may benefit from the translation units found in translation memories (usually whole sentences)

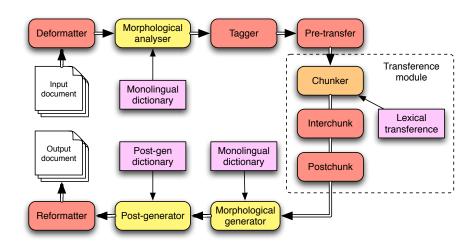
Goal

Integrate sub-sentential translation units into the Apertium free/open-source MT platform

 Sub-sentential translation units are more likely to be re-used than whole sentences

Test the bilingual chunks automatically obtained using the maker-based chunkers and chunk aligners of Matrex

Apertium rule-based MT engine



Example of translation /1

Source text:

Francis' car is broken

De-formatter:

Francis'[]car[]is broken

Morphological analyser:

```
^Francis'/Francis<np><ant><m><sg>+'s<gen>$[ <strong>]
^car/car<n><sg>$[</strong>]
```

Part-of-speech tagger:

```
^Francis<np><ant><m><sg>$ ^'s<gen>$[ <strong>]
```

```
^car<n><sg>$[</strong>]^be<vbser><pri><p3><sg>$
```

[^]is/be<vbser><pri><p3><sq>\$

[^]broken/break<vblex><pp>\$

[^]break<vblex><pp>\$

Apertium: Example of translation /2

Structural transfer (prechunk) + Lexical transfer:

```
^nom<SN><UNDET><m><sg>{^Francis<np><ant><3><4>$}$
^pr<GEN>{}$[ <strong>]
^nom<SN><UNDET><m><sg>{^coche<n><3><4>$}$
[</strong>] ^be_pp<Vcop><vblex><pri><p3><sg><GD>{
^estar<vblex><3><4><5>$
^romper<vblex><pp><6><5>$}$
```

Structural transfer (interchunk):

```
[<strong>]^nom<SN><PDET><m><sg>{^coche<n><3><4>$}$
[</strong>]^pr<PREP>{ ^de<pr>$}$
^nom<SN><PDET><m><sg>{ ^Francis<np><ant><3><4>$}$
^be_pp<Vcop><vblex><pri><p3><sg><m>{
^estar<vblex><3><4><5>$
^romper<vblex><pp><6><5>$}$
```

Apertium: Example of translation /3

Structural transfer (postchunk):

```
[<strong>]^el<det><def><m><sg>$ ^coche<n><m><sg>$
[</strong>] ^de<pr>$ ^Francis<np><ant><m><sg>$
^estar<vblex><pri><p3><sg>$
^romper<vblex><pp><m><sg>$
```

Morphological generator and post-generator:

[]el coche[]de Francis está roto

De-formatter:

el coche de Francis está roto

Target text:

el coche de Francis está roto

Considerations

Requirements

- Not break the application of structural transfer rules
- Use the longest possible chunks

How

- Introducing chunks delimiters as format information
 ... is [BCH_12_0] the chunk [ECH_12_0] that ...
- Chunks can be then recognised after the translation
 ... es [BCH_12_0]el segmento[ECH_12_0] que ...

Side effect

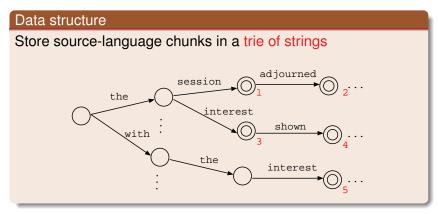
- Format information may be moved around
- Or deleted by some rules (bug)

Translation approach

Algorithm

- apply a dynamic-programming algorithm to compute the best coverage of the input sentence
 - Introduce chunk delimiters as format information
- translate the input sentence as usual by Apertium
 - Detected chunks are also translated
- use a language model to choose one of the possible translations for each of the bilingual chunks detected
 - One source-language chunk may have different target-language translations
 - Also consider Apertium translation

Computation of the best coverage /1



It allows to compute the best coverage in a efficient way

Computation of the best coverage /2

Algorithm

- A set of alive states in the trie is maintained to compute all the possible ways to cover the input sentence
- At each position the best coverage until that position is stored
 - Only the best coverage up to the last / word
- A new search is started at every word
- Is applied to text segments shorter than sentences
 - The best coverage can be retrieved when there are no more alive states



Computation of the best coverage /3

The best coverage

- is the one that uses the least possible number of chunks
 - longest possible chunks
- each not covered word counts like one chunk
- if two coverages use the same number of chunks, the one that uses the most frequent chunks is used

Experimental setup /1

Corpora

- Corpora distributed for the WMT 09 Workshop for MT
- Language pairs: Spanish→English (es-en), English→Spanish (en-es)
- Linguistic data: apertium-en-es; SVN revision 9284

Tools

- Apertium
- Giza++ and Moses to calculate word alignments and lexical probabilities
- SRILM to train 5-gram language models
- Matrex to segment training corpora and to align chunks

Experimental setup /2

Training corpus preprocessing

Max. sentence length: 45 words

Max. word ratio: 1.5 words (mean ration + std. dev.)

Corpus	Sentences	English words	Spanish words
Training	1,187,905	26,983,025	27,951,388
Development	2,050	49,884	52,719
Test	3,027	77,438	80,580

Experimental setup /3

Marker-based bilingual chunks

- Based on the 'Marker Hypothesis'
- Marker words: prepositions, pronouns, determiners, etc.
- Chunks start with a maker word
- Chunks contain at least one non-marker word

Chunks filtering

- There must be at least one word aligned in each side
- \blacksquare Chunks not seen at least θ times are discarded
 - Tested values: $\theta \in [5, 80]$
- Chunks containing punctuation marks and numbers are discarded

Results: BLEU scores

Translation	Apertium		Apertium+chunks		
ITATISIALIOTI	dev	test	θ	dev	test
English→Spanish	17.10	18.51	11	17.41	18.94
Spanish→English	17.71	18.81	28	17.91	19.14

- Small improvement, not statistical significant
 - Statistical significance test: bootstrap resampling
- Improvement is larger in the test corpus

Translation	dev	test
English→Spanish		+0.43
Spanish→English	+0.20	+0.33

Results: Analysis

	Number of chunks (% words covered)			
Translation	Detected	Finally used		
	Detected	All	Apertium	
English→Spanish	6,812 (18%)	5,546 (15%)	2,662 (7%)	
Spanish→English	6,321 (17%)	5,488 (14%)	2,929 (8%)	

- Around half of the chunks finally used are translate the same way as Apertium
- Chunks detected and not used due to chunk delimiters placed in the wrong position

Spanish→English translation

Example

- S: desde hace muchos años un fenómeno misterioso ...
- R: for years, a mysterious phenomenon...
- A: from does a lot of years a mysterious phenomenon ...
- A+C: for many years a mysterious phenomenon ...
 - S: olmert devolvería ... las zonas ocupadas a cambio de la paz
 - R: olmert would return ... territories in exchange for peace
 - A: olmert it would give back ... zones to change of the peace
- A+C: olmert it would give back ... zones in exchange for peace
 - S: pero hay una cosa que nos une :
 - R: but there is one thing that connects us:
 - A: but there is a thing that joins us:
- A+C: but there is one thing that joins us:

Discussion /1

Novel approach to integrate sub-sentential translation units in Apertium

■ Uses of the longest possible chunks and a language model

Approach tested using marker-based chunks

- Small improvement
- Most of the chunks are translated the same way as Apertium

Noise introduced due to how Apertium manages format information

Some chunks are not applied because chunk delimiters are lost or moved to a wrong position

Discussion /2

Improvement of the computation of the best coverage when two coverages use the same number of chunks:

- Use the bilingual chunk that would produce the most-likely TL translation instead of the most frequent one
- How? Using a language model with gaps
 - \bigcirc in the session \bigcirc \bigcirc with the interest .

Improvement of the bilingual chunks filtering

- Current approach only based on chunks frequency
- Longer chunks are penalised in favour of shorter ones

Thanks for your attention!

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- An open-source implementation is available at http://sf.net/projects/apertium/files/ package name: apertium-chunks-mixer
- More information on the Apertium web page: http://www.apertium.org