String Similarity Measures for Template Extraction Natalia Elita

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Outline

- Motivation
- □ Similarity Matrix
 - String Similarity Measures
 - Indexing
- □ Template Extraction
- Conclusion
- □ Further work

Motivation -1-

- (En) 1. The prosecution had charged Priebke with multiple and particularly ferocious homicide.
- (De) 1. Die Staatsanwaltschaft hatte Priebke des mehrfachen, besonders grausamen Mordes beschuldigt
- (En) 2. In the course of the trial, lasting three months, Priebke had admitted to have shot to death two people himself.
- (De) 2. Priebke hatte in dem 3 Monate dauernden Prozess zugegeben, 2 Menschen eigenhaendig erschossen zu haben.

Motivation -2-

- (En) 1. The prosecution had charged Priebke with multiple and particularly ferocious homicide.
- (En) 2. In **the** course of the trial, lasting three months, **Priebke** had admitted to have shot to death two people himself.
- (De) 1. Die Staatsanwaltschaft *hatte Priebke* des mehrfachen , besonders grausamen Mordes beschuldigt
- (De) 2. *Priebke hatte* in dem 3 Monate dauernden Prozess zugegeben, 2 Menschen eigenhaendig erschossen zu haben.

Motivation -3-

- (De) 1. Weitere Informationen finden Sie unter Sicherheitseinstellungen auf Seite NUM.
- (De) 2. Weitere Informationen hierzu finden Sie unter Sicherheitseinstellungen auf Seite NUM.
- (En) 1. For further information, see Security settings on page NUM
- (En) 2. For further information, see Security settings on page NUM.

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Similarity Matrix

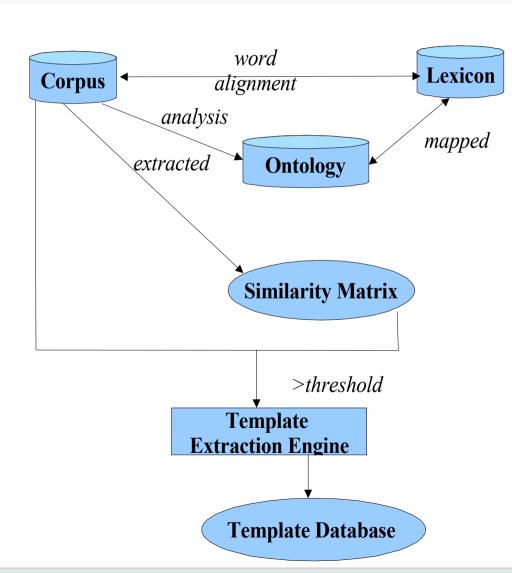
- □ Similarity Matrix:
 - □ For a monolingual corpus with *N* sentences, the Similarity Matrix *s* is formally defined:

$$s(i,j)=0$$
, for j

$$s(i,i)=1$$
, for $1 <= i=N$;

- s(i,j)=BSM(sentence_i, sentence_j), for j > i, 1 <= i, j <= N, where BSM = Best Similarity Measure
- to reduce the search space
- to find candidates for templates
- to observe the need of semantics
- Indexing
 - to reduce the search space

Similarity Matrix



	s1	s2	s3		sn
s1	1	0.85	0.02		0.15
s2	0	1	0.12		0.96
s3	0	0	1		0.48
	0	0	0	1	0.50
sn	0	0	0	0	1

Template

- generalization of sentences that are translations of each other, where sequences of one or more words are replaced by variables, with alignments between the resulting word sequences and/or variables made explicit
 - \blacksquare E.g (SL)**Tfa** V_i **Tfb** V_{i+1} **Tfc** <---> (TL) V_i **Tfd** V_{i+1} **Tfe**,

where $\mathbf{Tfx} - \mathbf{text}$ fragment \mathbf{x} $V_i - variable i$

Problem description

 Given a sentence aligned corpus, find sentences that are similar enough to become candidates for translation templates

- no syntactic annotation of the corpus
- no other linguistic resource
- similarity on the surface form only

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String Similarity Measures

- String Similarity measures are used in applications:
 - Spell check
 - □ Text prediction
 - Translation Memories
 - □ EBMT (matching)

Types

- character-based
 - similarity at the character level
- □ token-based
 - similarity at the token level
- hybrid
 - □ token based similarity first applied, then character based on each similar token

String Similarity Measures under consideration

- 20 string similarity measures
 - □ 18 SymMetrics package*
 - □ 10 character based, 5 token based, 3 hybrid
- \square 2 new
 - token-based
 - Common Words (CW)
 - Adapted Levenshtein Distance (ALD)

*http://www.dcs.shef.ac.uk/sam/simmetrics.html.

New Token based Measures -1-

- □ Common Words (CW):
 - number of common tokens for two given strings *s1* and *s2*

e.g:

- (s1) Writing and sending a multimedia message
- (s2) Reading and replying to a multimedia message

CW = 4 [and a multimedia message]

New Token based Measures

- 2-

- □ Adapted Levenshtein Distance (ALD)
 - □ For the given two strings s1 and s2:
 - □ Token Levenshtein Distance (TLD) is the traditional Levenshtein Distance, but on token level;
 - □ The maximal number of tokens of s1 and s2 is determined;
 - ☐ The obtained value is normalized to get values between 0 and 1.

ALD (example)

$$ALD(s_1,s_2)=1-\frac{TLD}{2*\max(Length(s_1),Length(s_2))}$$

- (s1) Writing and sending a multimedia message
- (s2) Reading and replying to a multimedia message

$$TLD = 3$$

 $max(length(s1), length(s2)) = 7$
 $ALD = 1-(3/14) = 0.78$

Thresholds

- experimentally established
 - □ identical strings (1)
 - completely different strings (0)
 - substrings
 - □ word order
 - length of strings

Thresholds: Character-based

- \Box TagLink Token = 0.5
- □ Euclidean Distance = 0.5
- \square Smith-Waterman = 0.6
- \square Smith-Waterman-Gatoh = 0.6
- \Box Jaro = 0.7
- □ Jaro Winkler=0.7
- □ Needlemann-Wunch= 0.7
- \Box Levenshtein Distance = 0.75
- □ Dice Similarity=0.75
- □ Cosine Similarity= 0.75

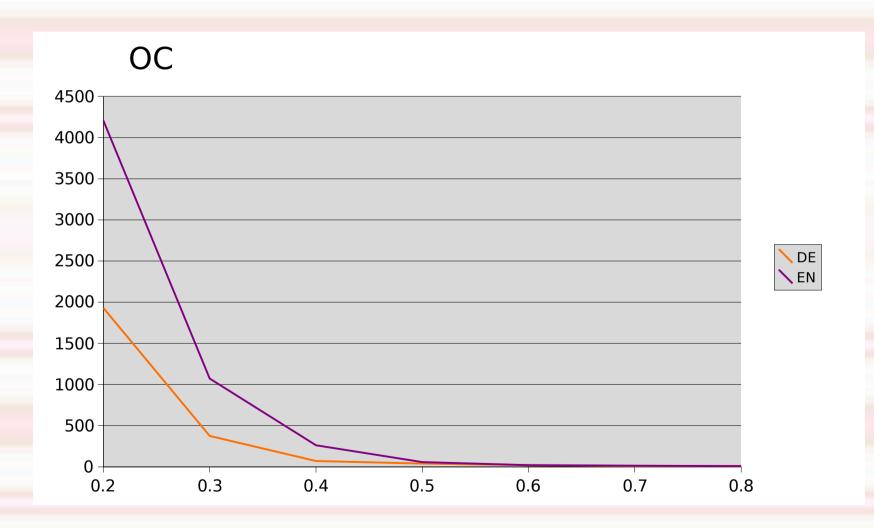
Thresholds: Token-based

- \square Common Words (CW) = 5
- □ Adapted Levenshtein Distance = 0.7
- □ Matching Coefficient = 0.55
- \square Block Distance = 0.6
- □ Jaccard Similarity = 0.45
- \Box Overlap Coefficient (OC) = 0.66
- \Box Q-Grams Distance = 0.65

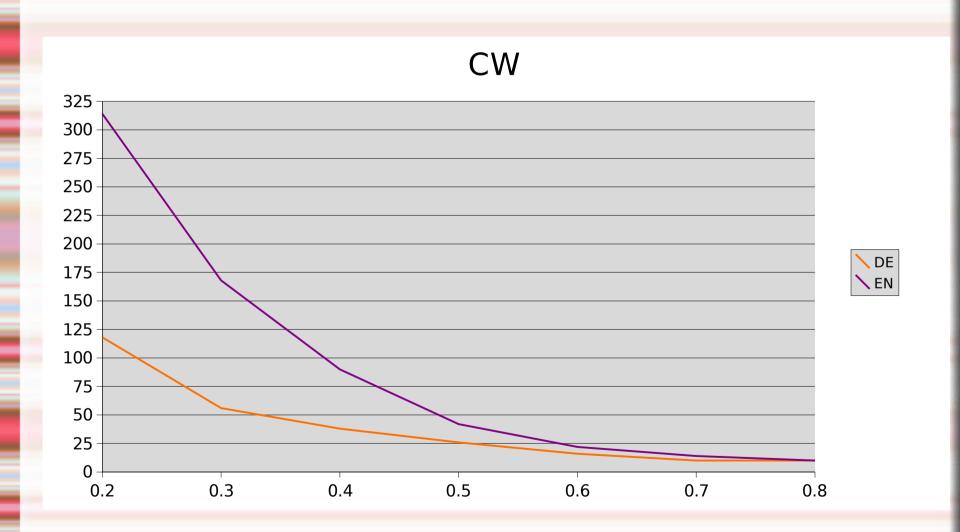
Thresholds: Hybrid

- \square Monge-Elkan = 0.9
- □ Chapman Ordered Name Compound Similarity = 0.75
- \Box TagLink = 0.7

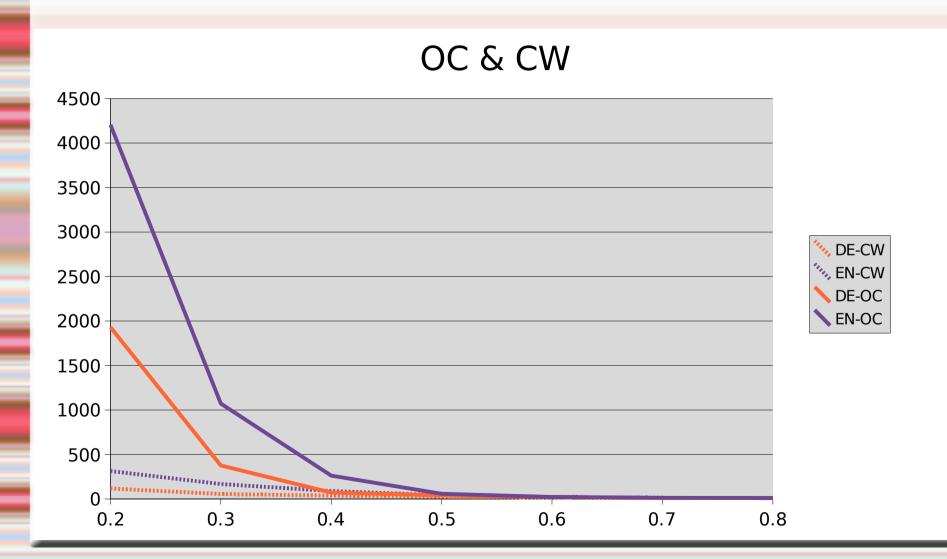
Thresholds/Candidates for Templates (OC)



Thresholds/Candidates for Templates (CW)



Thresholds/Candidates for Templates



Experimental settings

- corpus: technical
- □ languages: De, En, Ro
- □ 100 sentences
 - to make observations, assumptions
 - manual evaluation

Experiments -1-

Token-based	Ge	En	Ro
CW	4	11	11
Matching coefficient	12	10	9
Block Distance	13	12	13
Jaccard Similarity	12	10	9
OC	24	19	25
Q-Grams Distance	9	9	6
Total	74	71	73
Unique pairs	26	30	31

Experiments -2-

Character-based	Ge	En	Ro
Levenshtein Distance	1	3	2
Dice Similarity	5	4	3
Cosine Similarity	5	4	3
Euclidean Distance	5	4	3
Jaro	35	32	56
Jaro-Winkler	86	72	109
Needleman-Wunch	24	40	22
SW	83	82	49
SW-Gotoh	107	103	73
Tag Link Token	70	67	62
Total	421	411	382

Experiments -3-

Hybrid	Ge	En	Ro
CONC	48	48	29
Tag Link	19	17	19
Total	67	65	48
Unique pairs	58	59	40

Observations

- Character-based measures too slow and depend very much on the length of the strings to be compared
 - □ e.g. 300 sentences (De,Ro) ~ 7 minutes
- ☐ Hybrid methods perform not so well in case of German compound nouns
- □ Token-based the most useful for the template extraction
 - Common Words and Overlap Coefficient

Observations -2-

- Common Words the number of common tokens two strings have
 - no word order is taken into account
- Overlap Coefficient (OC) the metric which determines to what degree is one string a substring of another:

 $OC(s_1,s_2) = \frac{(|s_1 \wedge s_2|)}{\min(|s_1|,|s_2|)}$

where: |s| - number of tokens in s, $|s_1 \wedge s_2|$ number of common tokens in s_1 and s_2

Observations

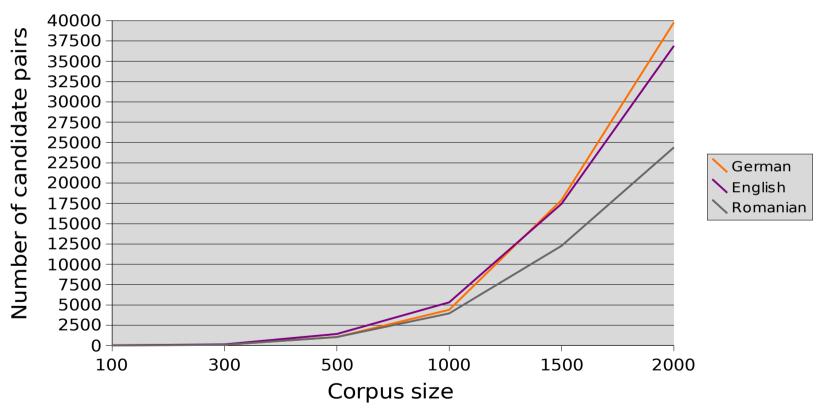
- □ CW + OC used to build the Similarity Matrix
 - \square Thresholds: CW =3; OC = 0.5;
 - Experiments made on sets
 - in different languages
 - of different size
 - of different corpus type

Experiments

- goal: for each language, see how the number of similar sentences changes with the size of the corpus
 - corpus type: technical
 - □ corpus size: up to 2000 sentences
 - □ languages: De, En, Ro

Experiment -1-

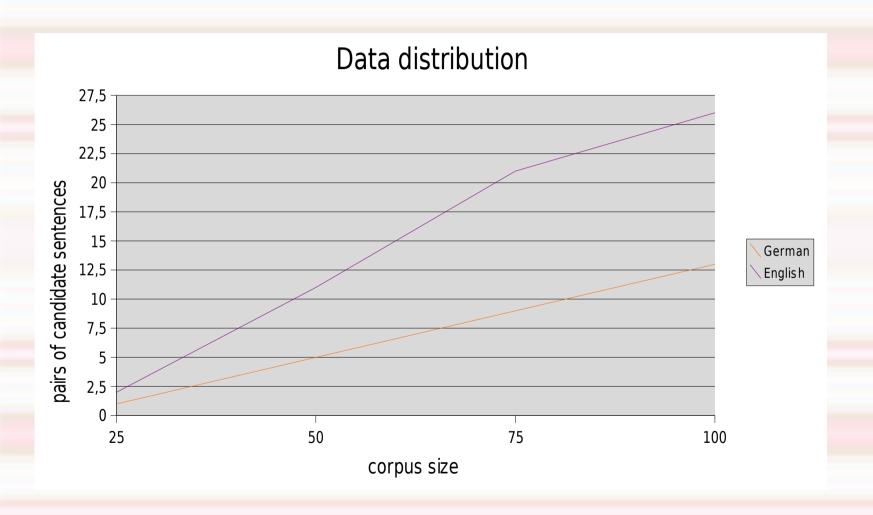




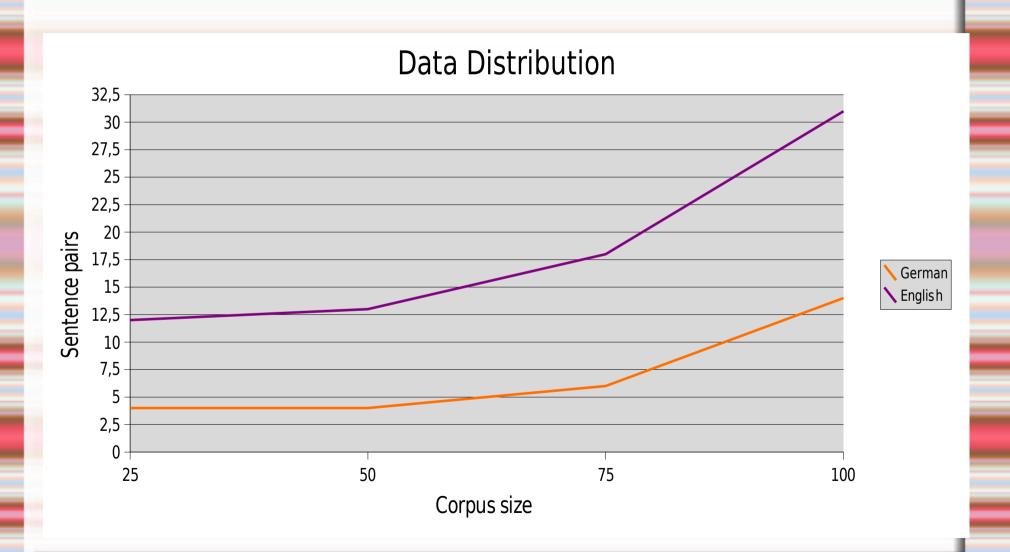
Experiment -2-

- corpus dependency
 - □ up to 100 sentences
 - news and technical corpora
 - □ languages: De, En

Experiment -2-(News)



Experiment -2- (Technical)



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Index vs Similarity Matrix

- Search for similar sentences
 - n*(n-1)/2 comparisons have to be made, where n is the number of sentences in a corpus
 - -e.g: corpus of 100 sentences 4950 comparisons

Index

Corpus type	Language	Corpus size	Search space
News	En	100	2001
News	De	100	1390
Technical	En	100	479
Technical	De	100	456

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Baseline System

Language neutral recursive machine learning algorithm based on principle of similar distributions of strings:

Source Language and Target Language strings that cooccur in two (or more) sentence pairs of a bilingual corpus are likely to be translations of each other

Problems

Proved to have serious limitations:

- -the templates obtained are often not translations;
- -no template is learned if different lexical items are used semantics would be extremely useful in this case;
- -big memory problems for a small corpus of 400 sentences;
- -useful information is lost.

Example -1-

Given 2 sentences in English:

- 12: The discussion around the envisaged major tax reform continues.
- 16: **The** head of the FDP parliamentary group, Mr. Solms, however, has deviated from the FDP 's demand to enact **the tax reform** as early as 1998.

The sequence of common elements: [the the tax reform]

Example -1-

```
Generalized template fragments of these 2 sentences:
[The V1 the V2 tax reform V3] (12)
[The V4 the tax reform V5] (16)
Where:
V1 = "discussion around"
V2 = "envisaged major"
V3 = "continues"
V4 = \text{``head of the FDP parliamentary group , } Mr. Solms
  , however , has deviated from the FDP 's demand to
  enact"
```

V5 = "as early as 1998"

Example -1-

The translations into German:

12: Die Diskussion um die vorgesehene grosse Steuerreform dauert an.

16: Der FDP - Fraktionsvorsitzende im Bundestag, Solms, ist von der Forderung der Liberalen abgerueckt, die Steuerreform schon 1998 in Kraft zu setzen.

The sequence of common elements: [die Steuerreform]
The sequence contains only 2 elements --> threshold
established at 3

Solution?

For a given SL corpus:

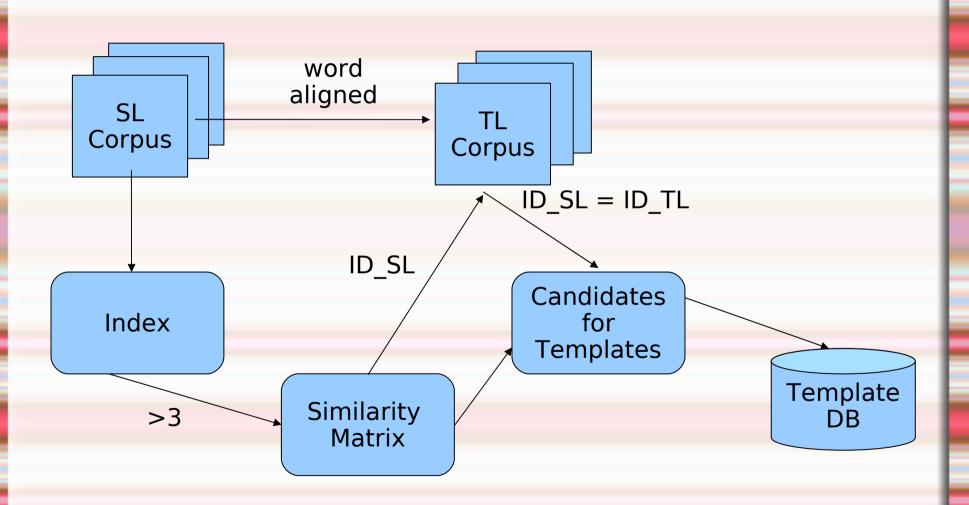
- 1. Index created for each sentence in the corpus.
- 2. Similarity matrix build for the corpus:

 pairs of "similar" sentences with the sequence of
 common elements greater or equal to three are
 extracted;
 - for each pair of similar sentences: their TL (by the sentence ID) counterparts are retrieved for each pair of sentences SL and TL with the same IDs are (word) aligned;

Corresponding TL counterparts of the sequence of common elements are found;

SL and TL parts combined into a template

Algorithm



Another example

For the given sentences in SL (English):

26: Wage conflict in retail business grows

27: The **conflict in** the wage negotiations in the *retail* industry has extended to North Rhine Westphalia.

and the translations into TL (Cerman):

26: Tarifkonflikt im Einzelhandel weitet sich aus

27: Der **Tarifkonflikt im** *Einzelhandel hat sich* auf Nordrhein - Westfalen *ausgeweitet* .

Variables - 1 -

8. [wage conflict in retail V1]--> [Tarifkonflikt im Einzelhandel V11 sich V21]

```
V1 = "business grows"
V11 = "weitet"
V21 = "aus"
```

Variables - 2 -

9. [V1 conflict in V2 wage V3 retail V4] -->[V11 Tarifkonflikt im Einzelhandel V21 sich V31]

```
V1 = "The"
```

V2 = "the"

V3 = "negotiations in the"

V4 = "industry has extended to North Rhine Westphalia"

V11 = "Der"

V21 = "hat"

V31 = "auf Nordrhein - Westfalen ausgeweitet"

Alignment

8. [wage conflict in retail *V1*]--> [Tarifkonflikt im Einzelhandel V11 sich *V21*]

```
V1 = "business grows"
```

Problems to solve

- tense/aspect:
 - grows vs has extended
- semantics:
 - retail business vs retail industry
 - grows vs extends

Solution to semantics: WordNet -1-

retail business vs retail industry WordNet:

Industry is a direct hyponym of **business** as seen from the WordNet:

- # S: (n) commercial enterprise, business enterprise, **business** (the activity of providing goods and services involving financial and commercial and industrial aspects) "computers are now widely used in business"
 - * direct hyponym / full hyponym
 - o S: (n) **industry**, manufacture (the organized action of making of goods and services for sale) "American industry is making increased use of computers to control production"

Solution to semantics: WordNet -2-

grow/extend - no direct connection found; indirectly - grow -->expand (direct troponym); extend -->expand (verb group);

Problem:

How do I know I chose the right sense of business?
Difficult even for a human to decide which synset is appropriate.

Solution to semantics: FrameNet

FrameNet:

Industry is the lexical unit (LE) belonging to the frame Fields, and LE Business belongs to the Business frame.

Grow/Extend:

LE	Frame
(1) grow.v	Expansion
(2) grow.v	Cause_expansion
(3) grow.v	Becoming
(4) grow.v	Change_position_on_a_scale

Solution to semantics: FrameNet

LE extend contained in the frame Change_event_duration.

Definition: In this frame, an Agent or Cause changes the duration of an Event. The Event will then take place for a New_duration, rather than the Initial_duration. This can be done with by certain Means, in a certain Manner or to a certain Degree.

In my opinion, in our context - the meaning of "extend" does not correspond to the definition of the frame, as certainly an idea of space is expressed by it.

Another example (need of semantics) -1-

Given the two pairs of sentences:

26: Wage conflict in retail business grows

97: Wage dispute in retail sector

26: Der **Tarifkonflikt** *im Einzelhandel* hat sich auf Nordrhein - Westfalen ausgeweitet .

97: Tarifkonflikt des Einzelhandels

Another example (need of semantics) -2-

- WordNet: conflict/dispute the same synset in WordNet:
- S: (n) dispute, difference, difference of opinion, conflict (a disagreement or argument about something important) "he had a dispute with his wife"; "there were irreconcilable differences"; "the familiar conflict between Republicans and Democrats"

business/business sector - the same synset in WordNet

Another example (need of semantics) - 3-

• FrameNet

LE dispute - in Quarrelling frame; LE conflict - in Hostile Encounter frame;

LE business - in Business frame LE sector - in Fields frame

Evaluation -1-

Experiments done with the news corpus (100 sentences)

A total of 53 template fragments were extracted, only 16 of them can be combined in a full template - by the sentence IDs the fragments were extracted from.

Semantics:

Noticed to be useful in 8 template fragments

Evaluation - 2 -

Errors:

- Extracted fragments not translations 4 cases
- □ No fragments learned because of:
 - □ Common Words Threshold (De) 15 cases
 - □ Overlap Coefficient Threshold (En) 5 cases
 - \square Spelling errors 1 case
 - □ Paraphrase 2 cases

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Conclusion

- Similarity matrix used to find candidates for templates
 - Common Words and Overlap Coefficient as similarity criteria
 - ☐ Index used to reduce the search space
- Generalization of similar sentences into translation templates needs semantic information

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Further work

- Decisions on templates:
 - ☐ Generalize on at least two sentences?
 - ☐ If common tokens are in different order, on which sentence should the generalization be made?
 - □ Variables: one token per variable?
- Extract templates without semantics
 - Decide on the source of semantics
- Add semantic information
- Extract templates with semantics

Thank you!

Questions? Suggestions?