Grammar-Based Concept Alignment for Domain-Specific Machine Translation

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- grammar-based pipelines (cf. GF) provide strong guarantees of **grammatical correctness**
- **lexical exactness** is as important as grammaticality
    - need for high-quality **translation lexica** preserving semantics *and* morphological correctness

Translation lexica

- Often built **manually**
  - **time** consuming
  - significant **linguistic knowledge** required

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- Often built **manually**
  - **time** consuming
  - significant **linguistic knowledge** required
- need for at least partial **automation**
  - **example parallel data** required

A parallel corpus

Alice thought she might as well wait, as she had nothing else to do, and perhaps after all it might tell her something worth hearing.

For some minutes it puffed away without speaking, but at last it unfolded its arms, took the hookah out of its mouth again, and said, 'So you think you're changed, do you?'

'I'm afraid I am, sir,' said Alice; 'I can't remember things as I used--and I don't keep the same size for ten minutes together!'

Alice pensò che poteva aspettare, perchè non aveva niente di meglio da fare, e perchè forse il Bruco avrebbe potuto dirle qualche cosa d'importante.

Per qualche istante il Bruco fumò in silenzio, finalmente sciolse le braccia, si tolse la pipa di bocca e disse: — E così, tu credi di essere cambiata?

— Ho paura di sì, signore, — rispose Alice. — Non posso ricordarmi le cose bene come una volta, e non rimango della stessa statura neppure per lo spazio di dieci minuti!

From Lewis Carroll, *Alice's Adventures in Wonderland*. Parallel text at `paralleltext.io`

Types of alignment

Word alignment:

Alice thought she might as well wait, as she had nothing else to do, and perhaps after all it might tell her something worth hearing.

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Phrase alignment:

Alice thought she might as well wait, as she had nothing else to do, and perhaps after all it might tell her something worth hearing.

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Approaches to automation

statistical (e.g. IBM models)	**syntax-based**
require **large amounts of data**	work consistently well even on **individual sentence pairs**
works with **raw** data	requires the data to be **analyzed**
correspondences between **strings**	correspondences between **grammatical objects**
"fixed" level of abstraction (**word** or **phrase**)	all levels of abstraction $\rightarrow$ **concept** alignment

Our approach

- Inconsistencies between different grammar formalisms $\rightarrow$ translation lexicon implemented in **GF**

Our approach

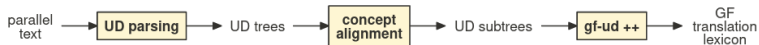- Inconsistencies between different grammar formalisms → translation lexicon implemented in **GF**
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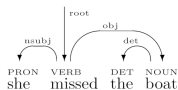

```
# text = she missed the boat
1 she she PRON _ _ 2 nsubj _ _
2 missed miss VERB _ _ 0 root _ _
3 the the DET _ _ 4 det _ _
4 boat boat NOUN _ _ 2 obj _
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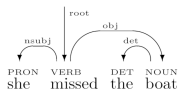
Graphical, CoNNL-U and Rose Tree representation of the same UD tree.

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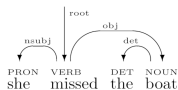
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Graphical, CoNNL-U and Rose Tree representation of the same UD tree.

- Framework for cross-linguistically consistent grammatical annotation
- dependency-labelled links between words (head-dependent pairs)
- cannot be used for target language generation

Grammatical Framework
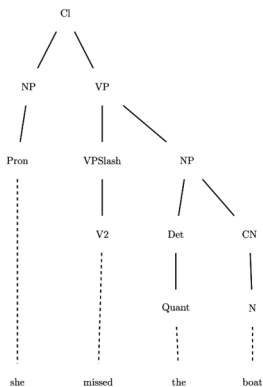
- Constituency grammar formalism for **multilingual grammars** (one abstract syntax + a concrete syntax per language)
- compilation-like translation (parsing + linearization)

Concept Extraction

Definitions

Concept: semantic unit of compositional translation expressed by a word or construction, conceived as a lemma equipped with morphological variations.

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***Concept***: semantic unit of compositional translation expressed by a word or construction, conceived as a lemma equipped with morphological variations.

Alignment: tuple of equivalent concrete expressions in different languages; represents a concept.
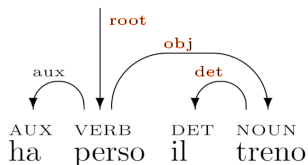
Extraction algorithm

---

procedure EXTRACT($criteria$,(t, u))
    $alignments = \emptyset$
    if (t, u) matches any alignment $criteria$ **then**
        $alignments \mathrel{+}= (t, u)$
        for (t', u') in SORT(SUBTS(t)) $\times$ SORT(SUBTS(u))
do
            extract($criteria$,(t', u'))
    return $alignments$

Matching UD labels
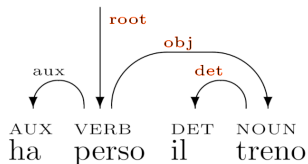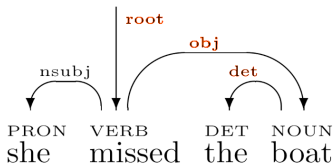
- ⟨*she missed the boat, ha perso il treno*⟩
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- *⟨*the boat, il treno*⟩
- ⟨*the, il*⟩

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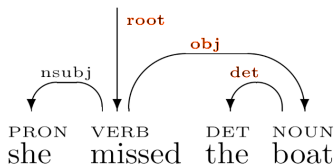
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Simple improvement: aligning heads of matching subtrees

- ⟨*she missed the boat, ha perso il treno*⟩, ⟨*missed the boat, perso il treno*⟩ → ⟨*missed, ha perso*⟩ (including the auxiliary)
- ⟨*the boat, il treno*⟩ → *⟨*boat, treno*⟩

POS equivalence
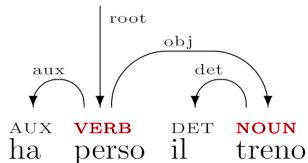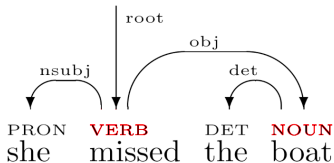
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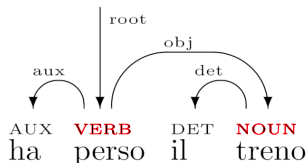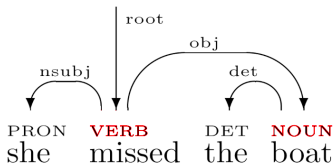
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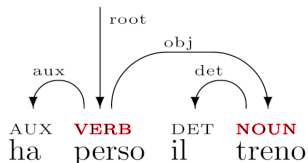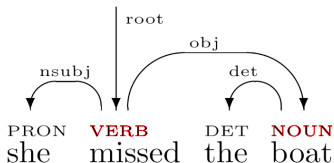
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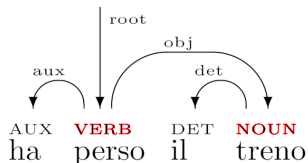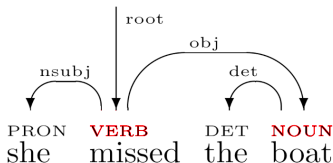
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- can increase recall when labels do not coincide
- can increase precision if used **in conjuncion with labels**

Known translation divergence

Divergence: systematic cross-linguistic distinction.

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- categorial
  - ⟨*Gioara listens **distractedly***, *Gioara lyssnar **distraherad**⟩*
  - ⟨*Herbert completed his **doctoral** thesis*, *Herbert ha completato la sua tesi **di dottorato**⟩*
- conflational
  - ⟨*Filippo is interested in **game development***, *Filippo är intresserad av **spelutveckling**⟩*
- structural
  - ⟨*I called **Francesco***, *Ho telefonato **a Francesco**⟩*
- head swapping
  - ⟨*Anna **usually** goes for walks*, *Anna **brukar** promenera*⟩
- thematic
  - ⟨***Yana** likes **books***, ***A Yana** piacciono **i libri**⟩*

- Allows using CA in conjunction with statistical tools

Known alignment

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- iterative application

Searching for specific patterns

- `gf-ud` pattern matching to look for specific syntactic patterns

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Example predication patterns:

- ⟨*she missed the boat, ha perso il treno*⟩ → ⟨*[subj] missed [obj], ha perso [obj]*⟩
- ⟨*she told you that, hon berättade det för dig*⟩ → ⟨*[subj] told [iobj] [obj],[subj] berättade [obj] för [obl]*⟩

Grammar rules generation

Requirements

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- aligned UD trees
- gf-ud
- **morphological dictionaries**
- **extraction grammar**

Morphological dictionaries

Purely morphological unilingual dictionaries.

Example:

```
...
lin morphologic_A =
  mkAMost "morphologic" "morphologicly" ;
lin morphological_A =
  mkAMost "morphological" "morphologically" ;
lin morphology_N =
  mkN "morphology" "morphologies" ;
...
```

Extraction grammar

Defines the syntactic categories and functions to build lexical entries.

Example (prepositional NPs):

```
PrepNP : Prep -> NP -> PP # case head
```

Lexical rules

Abstract:

```
fun in_the_field__inom_området_PP : PP ;
```

English concrete:

```
lin in_the_field__inom_område_PP =
  PrepNP in_Prep (DetCN the_Det (UseN field_N))
```

Evaluation

Evaluating extraction

UD tree alignments are evaluated:

- independently from the quality of UD parsing (100-sentence subset of the manually annotated PUD corpus)
- on raw text (DMI and CSE course plans corpora)

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Metrics:

- % correct alignments
- % "useful" alignments

Results on PUD corpus

	CE		fast_align (100 sentences)		fast_align (full dataset)	
	en-it	en-sv	en-it	en-sv	en-it	en-sv
distinct alignments	536	638	1242	1044	1286	1065
correct	392 (73%)	514 (80%)	346 (28%)	538 (52%)	540 (42%)	677 (64%)
usable in MT	363 (68%)	503 (79%)	316 (25%)	525 (50%)	510 (40%)	666 (63%)

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- CE module compared with `fast_align`, so extracting only one-to-many and many-to-one alignments
- CE has much higher precision, even when `fast_align` is trained on full 1000-sentence corpus

Results on course plans corpora

	PUD (100 sentences)		course plans	
	en-it	en-sv	DMI (881 sentences)	CSE (539 sentences)
distinct alignments	1197	1325	1823	1950
correct	916 (77%)	1112 (85%)	1205 (66%)	1269 (66%)
usable in MT	880 (74%)	1099 (84%)	1157 (63%)	1248 (64%)

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- reference translations obtained by manual postprocessing of the automatic ones
  - avoid low scores due to different but equally valid lexical and grammatical choices

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BLEU-1 to 4	55	61
BLEU-1 to 3	63	68
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- ▶ sentence-level scores range from 0 (sometimes due to a single semantic error) to 100

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- available as Haskell library + executables

- Concept Propagation

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