Towards automatically extracting morphosyntactical error patterns from L1-L2 parallel dependency treebanks

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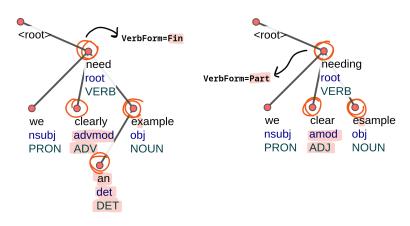
L1-L2 treebanks



- learner sentences || correction hypotheses
- no explicit error labelling, just morphosyntactical annotation
- ightharpoonup main design goal: interoperability ightarrow UD

Example

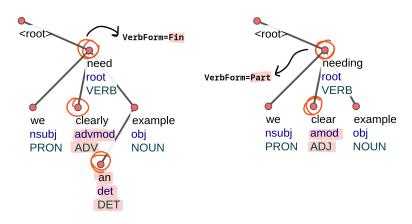




(we clearly needing an example, we clear needing _ esample)

Example

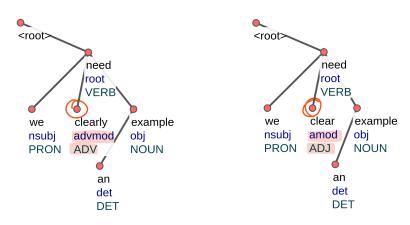




(we clearly needing an example, we clear needing $\underline{}$ example)

Example





(we clearly need an example, we $\underline{\text{clear}}$ need an example)

Tasks



- find instances of specific error patterns → L2-UD query engine¹
- $lap{f \vdash}$ automatically classify syntactical errors ightarrow SErCL 2

¹ Masciolini, 2023

² Choshen et al., 2020

Tasks



- find instances of specific error patterns → L2-UD query engine¹
- automatically classify syntactical errors → SErCL²
- extract machine-readable error patterns

¹ Masciolini, 2023

² Choshen et al., 2020

Error pattern extraction

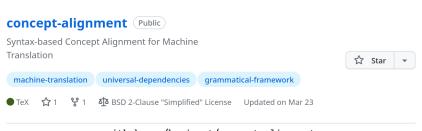


- 1. **error detection**: align L1-L2 sentences and filter discrepant alignment
- 2. **pattern generation**: convert pairs of UD subtrees into machine-readable error patterns

Step 1: error detection

concept-alignment



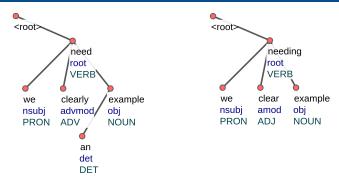


 ${\tt github.com/harisont/concept-alignment}$

- extracts **subtree alignments** from parallel UD treebanks
- syntax-based but language-agnostic
- designed to generate translation lexica, but easy to adapt to the L1-L2 case

Alignments

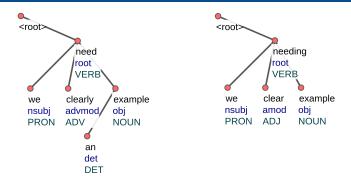




- \(\text{we clearly need an example}\), \(\text{we clear needing example}\), \(\text{need, needing}\)
- we, we
- ⟨clearly, clear⟩
- · 〈an example, example〉, 〈example, example〉

Errors





- \(\text{we clearly need an example}, we clear needing example} \(\text{*}, \) \(\text{need, needing} \(\text{*} \)
- \(\square\) \(\square\) we, we\(\rangle\)
- **→** ⟨an example, example⟩*, ⟨example, example⟩

Step 2: pattern generation

Query languages for UD trees



- several options to choose from
 - PML-TQ, Grew-match, UDAPI...
- decided on gf-ud's embedded query language
 - sufficiently expressive and user-friendly
 - easy to use as a library

UD patterns in gf-ud

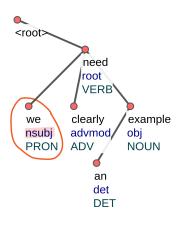


pattern type	example
single-token patterns tree patterns sequence patterns logical operators ³	DEPREL "nsubj" TREE (POS "NOUN") [DEPREL "det"] SEQUENCE [DEPREL "advmod", POS "VERB"] OR [POS "NOUN", POS "PRON"]

³ AND, OR, NOT

Single-token patterns

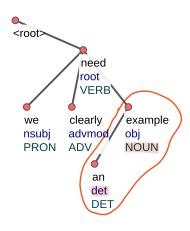




DEPREL "nsubj"

Tree patterns

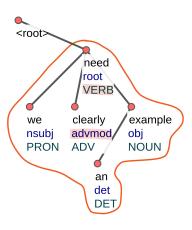




TREE (POS "NOUN") [DEPREL "det"]

Sequence patterns

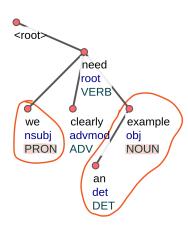




SEQUENCE [DEPREL "advmod", POS "VERB"]

Logical operators

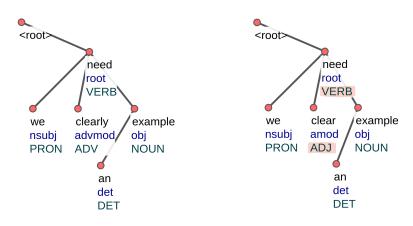




OR [POS "NOUN", POS "PRON"]

L1-L2 UD patterns

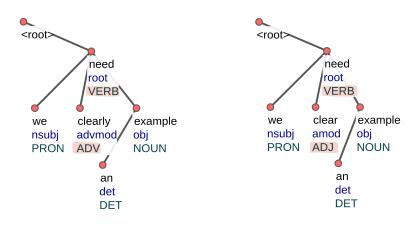




TREE_ (POS "VERB") [POS "ADJ"]

L1-L2 UD patterns

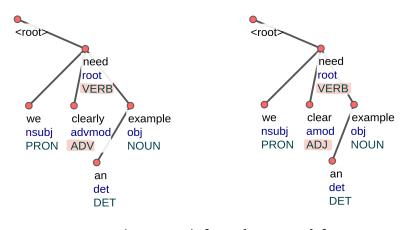




 $\langle {\tt TREE_}$ (POS "VERB") [POS "ADV"], TREE_ (POS "VERB") [POS "ADJ"] \rangle

L1-L2 UD patterns





TREE_ (POS "VERB") [POS "{ADV -> ADJ}"]



0. Automatically generated L1-L2 pattern

```
TREE
   (AND [FORM "need", LEMMA "need", POS "VERB", XPOS "VBP", DEPREL "root",
        FEATS "Mood=Ind|Number=Plur|Person=1|Tense=Pres|VerbForm=Fin"])
   [AND [FORM "we", LEMMA "we", POS "PRON", XPOS "PRP", DEPREL "nsubj"
        FEATS "Case=Nom | Number=Plur | Person=1 | PronType=Prs"],
   AND [FORM {"clearly" -> "clear"}, LEMMA {"clearly" -> "clear"}.
        POS {"ADV" -> "ADJ"}, XPOS {"RB" -> "JJ"},
        FEATS {" " -> "Degree=Pos"}, DEPREL {"advmod" -> "amod"}],
   TREE
      (AND [FORM "example", LEMMA "example", POS "NOUN",
           XPOS "NN", FEATS "Number=Sing", DEPREL "obj"])
      [AND [FORM "an", LEMMA "a", POS "DET", XPOS "DT",
           FEATS "Definite=Ind|PronType=Art", DEPREL "det"]
```



1. Filtering by CoNNL-U field

Ignoring FORM, LEMMA, XPOS and DEPREL:

```
TREE

(AND [POS "VERB",
FEATS "Mood=Ind|Number=Plur|Person=1|Tense=Pres|VerbForm=Fin"])

[AND [POS "PRON", FEATS "Case=Nom|Number=Plur|Person=1|PronType=Prs"],
AND [POS {"ADV" -> "ADJ"}, FEATS {"_" -> "Degree=Pos"}],
TREE

(AND [POS "NOUN", FEATS "Number=Sing"])

[AND [POS "DET", FEATS "Definite=Ind|PronType=Art"]]
]
```



2. Removal of never-discrepant fields

In all alignments, FEATS is either identical both in the L1 and in the L2 or absent in one of the components:

```
TREE
  (AND [POS "VERB"])
  [AND [POS "PRON"],
  AND [POS {"ADV" -> "ADJ"}],
  TREE (AND [POS "NOUN"]) [AND [POS "DET"]]
]
```



3. Elimination of identical subpatterns

Removing identical subtrees⁴:

```
TREE_
(AND [POS "VERB"])
[AND [POS {"ADV" -> "ADJ"}]]
```

 $^{^{\}rm 4}$ optionally, identical roots can be replaced with the wildcard pattern TRUE too



4. Monolingual single-pattern simplifications

AND [p] is equivalent to p:

```
TREE_ (POS "VERB") [POS {"ADV" -> "ADJ"}]
```

Preliminary evaluation

Experimental design



- evaluation through a similar example retrieval task:
 - 1. extract L1-L2 patterns from an error-correction input pair
 - 2. query an L1-L2 treebank with the extracted patterns
- interactive version available as prototype CALL application

Data



2 datasets for linguistic acceptability judgments:

- only one error per sentence
- filtered: only morphosyntactical errors
- automatically parsed with UDPipe 2

name	language	size ⁵	description
BLiMP	English	14 996	artificially generated sentences postprocessed L2 learner sentences
DaLAJ	Swedish	1 198	

⁵ post-filtering

Results



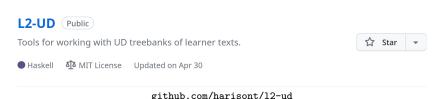
	BLiMP	DaLAJ
R^6	82%	69%
R_+^7	82%	63%

⁶ retrieval rate

⁷ successful retrieval rate

Summarizing





- novel approach to error pattern extraction
- preliminary, bilingual evaluation on LA datasets giving promising results
- interactive similar example retrieval pipeline available as prototype CALL application

Future work



- extraction method:
 - handle nonexistent word forms
 - deal with real-world L2 data:
 - non-morphosyntactical errors (spelling, lexical...)
 - multiple overlapping errors
- example retrieval application:
 - implement pattern selection/ranking
 - build UI
- use L1-L2 patterns from feedback comment generation
- improve automatic annotation of L2 sentences

Thank you!

asynchronous questions/comments: arianna.masciolini@gu.se

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