

Tecto to AMR and translation

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Introduction

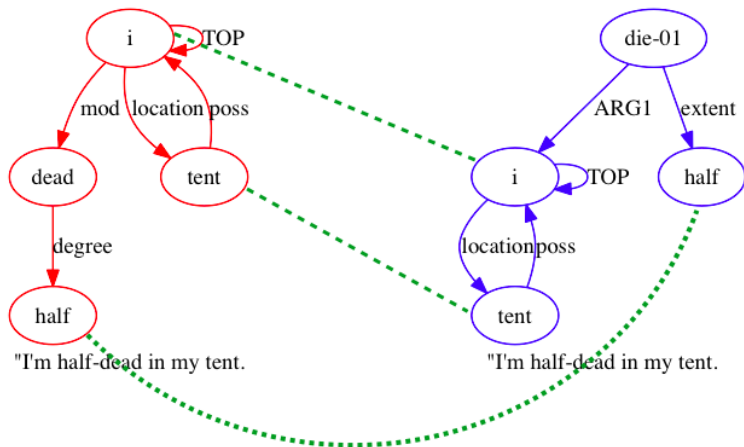
Motivation

- ▶ We are investigating the value of parallel Abstract Meaning Representations (AMRs)
- ▶ Question 1: How similar are AMRs made in different languages? How do you compare them?
- ▶ Question 2: How could we get a large corpus of parallel AMRs?

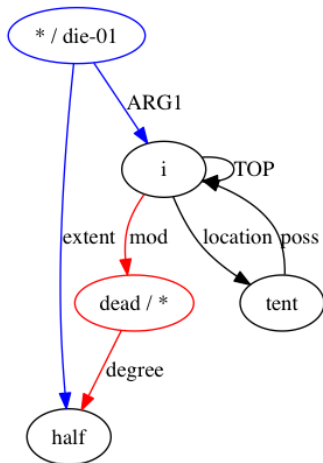
AMRICA

- ▶ (AMR Inspector with Cross-language Alignment)
- ▶ Usual evaluation and alignment methods break across languages.
- ▶ Extension to Smatch (Cai & Knight 2012).

Smatch Classic

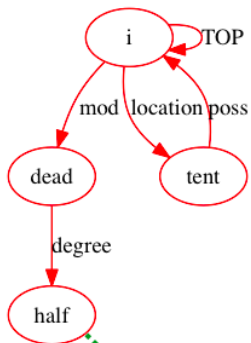


Smatch Classic

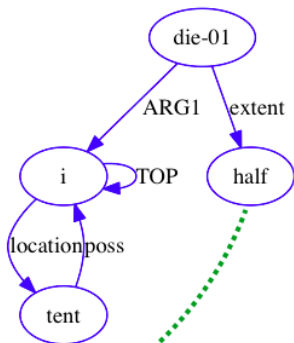


"I'm half-dead in my tent."

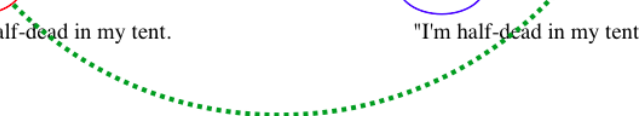
Smatch Classic



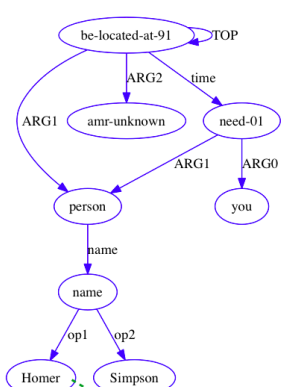
"I'm half-dead in my tent."



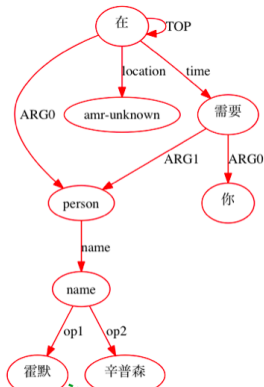
"I'm half-dead in my tent."



AMERICA

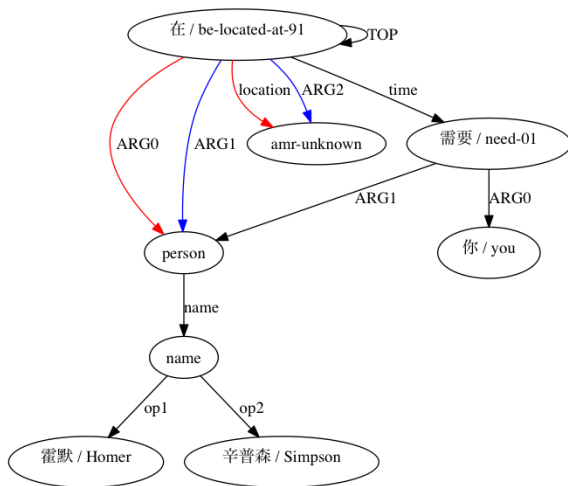


Where's Homer Simpson when you need him ?



当你需要他时，霍默辛普森在哪里？

AMRICA



Where's Homer Simpson when you need him?
当你需要他时，霍默辛普森在哪里？

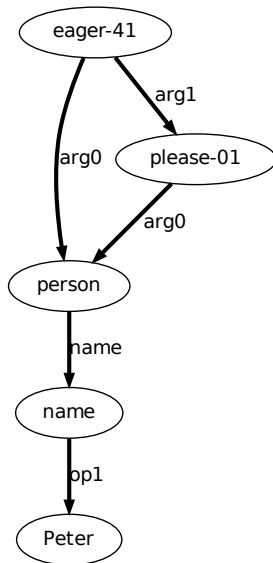
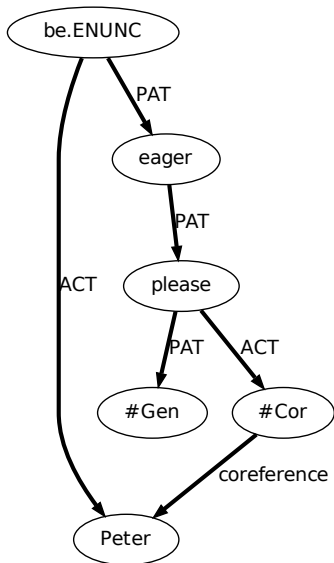
T-layer to AMR conversion

- ▶ PCEDT: Large parallel corpus (WSJ) annotated with t-layer for English and Czech
- ▶ T-layer to AMR conversion would provide a large static parallel AMR corpus.
- ▶ Could be used dynamically to turn a "t-layer" parser into an AMR parser.

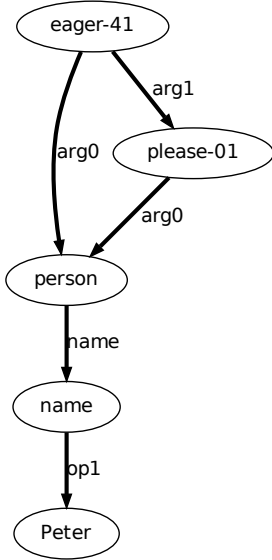
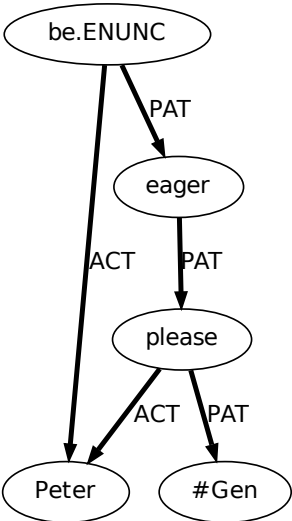
Why this might work

- ▶ AMR and t-layer are very similar:
 - ▶ Both abstract away from syntax.
 - ▶ Both make all semantic links in a sentence in a graph format.
 - ▶ Both do coreference
- ▶ Various minor structural differences.
- ▶ AMR is more abstract, makes more inference.

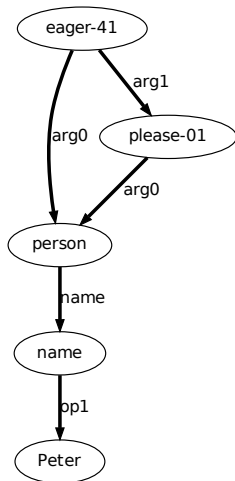
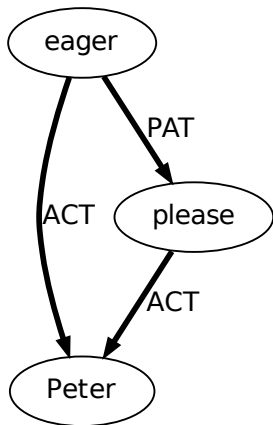
“Peter is eager to please”



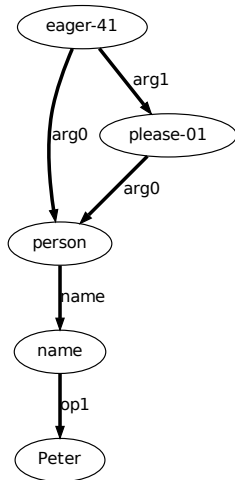
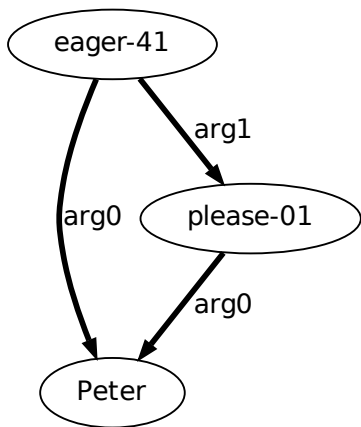
Merging of Coreferent Nodes



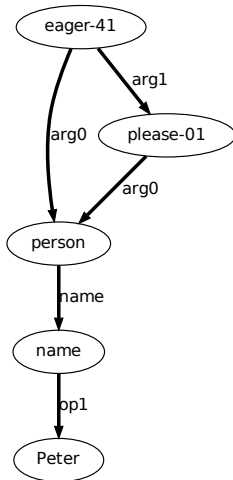
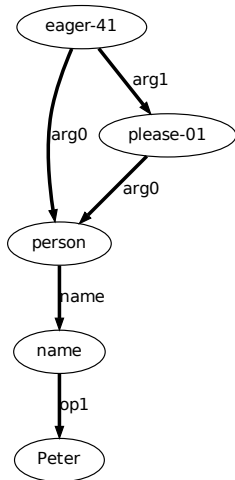
Elimination of semantically light words



Semantic Roles and Senses



Add Named Entities



Conversion Procedures

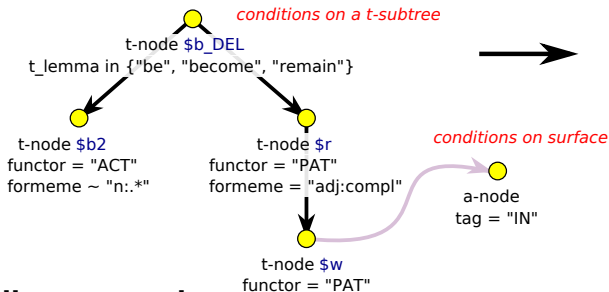
- ▶ Converted t-trees to AMR format
- ▶ Added named entities using NER systems (Stanford and NameTag)
- ▶ Tried two strategies for doing more complex changes to the graphs:
 - ▶ PML-TQ
 - ▶ Tsurgeon
- ▶ List-based verbalization and semantic role mapping

PML-TQ rules

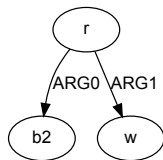
- ▶ Based on AMR guidelines (generalized)
- ▶ For copula, attributes, non-core roles ...

A PML-TQ rule

LHS (PML-TQ Query)



RHS (AMR Subtree)

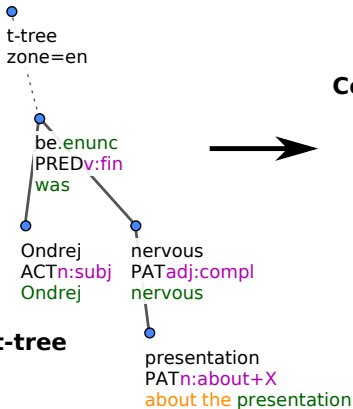


Guidelines example:

The boy is responsible for the work.

PML-TQ rules

Rule application

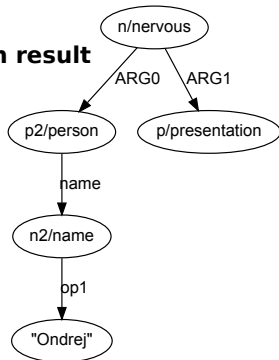


Matching t-tree

Matching sentence:

Ondrej was nervous about the presentation.

Conversion result

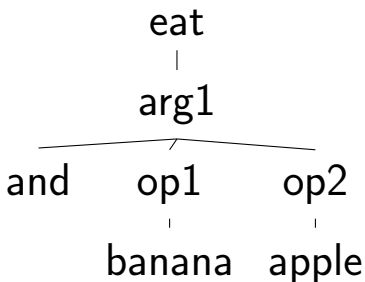
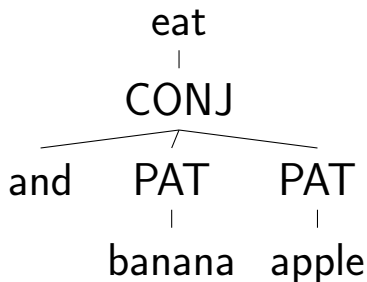


Tsurgeon tree transformation rules

- ▶ We converted to constituency trees so as to use a tree transformation tool, **Tsurgeon** (Levy and Andrews 2006) to quickly implement hand-written rules.

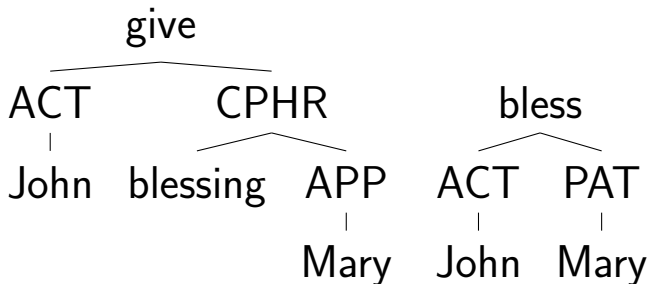
Tsurgeon tree transformation rules

- ▶ Many of the structural differences are just notational differences:



List-based Methods

- ▶ Verbalizations are based on dictionary look-ups:
 - ▶ beekeeper → person :ARG0-of keep-01 :ARG1 bee
- ▶ As are complex predications:



Using Existing Resources

	Vallex	Propbank	Lexicon	Other WSJ annotation	Lexical Lists
Map t-layer roles to AMR roles	X	X	X		
Verbalize nouns/adjectives		X			X
Introduce inferrable predicates					X
Named Entities			X	X	

Results of EN t-to-AMR Conv

	Semantic Role Mapping	Named Entities	Verbalization Lists	Smatch	Smatch w/o senses
Baseline (direct conversion)				20	28
Baseline (direct conversion)	X			33	41
Baseline (direct conversion)	X	X		37	45
Baseline (direct conversion)	X	X	X	40	48
PML-TQ (guidelines-based)	X		X	35	43
PML-TQ (guidelines-based)	X	X	X	38	47
Tsurgeon (rule-based)	X	X	X	44	52
JAMR				44	45