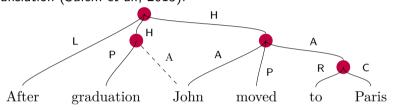
SemEval 2019 Task 1: Cross-lingual Semantic Parsing with UCCA

Daniel Hershcovich, Leshem Choshen, Elior Sulem, Zohar Aizenbud, Ari Rappoport and Omri Abend



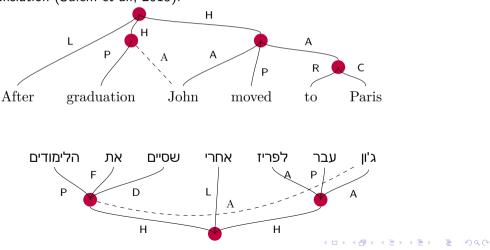
June 6, 2019

Cross-linguistically applicable semantic representation (Abend and Rappoport, 2013). Builds on Basic Linguistic Theory (R. M. W. Dixon). Stable in translation (Sulem et al., 2015).

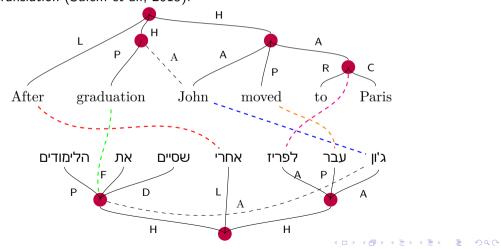


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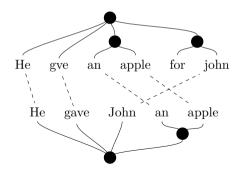


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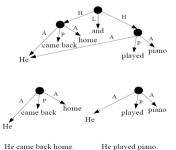


Applications

- Semantics-based evaluation of
 - Machine translation (Birch et al., 2016)
 - Text simplification (Sulem et al., 2018a)
 - Grammatical error correction (Choshen and Abend, 2018)
- Sentence splitting for text simplification (Sulem et al., 2018b).





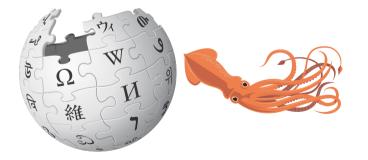


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Intuitive annotation interface and guidelines (Abend et al., 2017). ucca-demo.cs.huji.ac.il

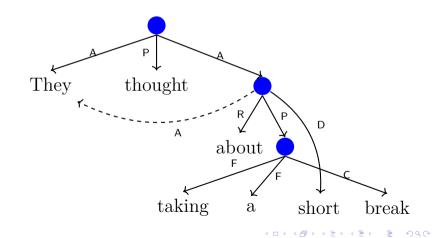
Linker (L) Ground (G) Participant (A) State (S)	i i i	school Julie N great r	<u>m Bradley Pitt was born in Shawnee, Oklahoma</u> , to William Alvin Pitt, who ran a trucking company, and Jane Etta (r I counsellor. The family soon moved to Springfield, Missouri, where he lived together with his younger siblings, Doug Neal (born 1969). Born into a conservative household, he was raised as Southern Baptist, but has since stated that he relationship with religion " and that he " oscillates between agnosticism and atheism." Pitt has described Springfield as e James country ", having grown up with " a lot of hills, a lot of lakes ".	las (born 1966) and e does not " have a
Process (P)	i	= 1 F	1 William Bradley Pitt was born in Shawnee , Oklahoma	+ 💬 F 🗙
Adverbial (D)	i	1	-1 A William Bradley Pitt	+ 💬 F 🗙
Time (T)	i	1	-2 F was	+ 💬 F 🗙
Center (C)	i		-3 P born	+ 🙂 F 🗙
Elaborator (E)	i			
Connector (N)	i		-4 A in Shawnee , Oklahoma	+ 💬 F 🗙
Relator (R)	i		1-4-1 R in	+ 💬 F 🗙
Uncertain (UNC)	i		1-4-2 C UNA Shawnee , Oklahoma	+ 💬 F 🗙
Unanalyzable (UI	i			
Function (F)	i			

The Task: UCCA parsing in English, German and French in different domains.



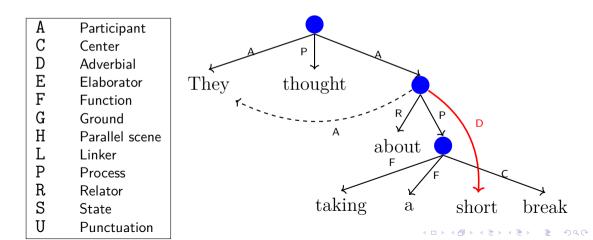
Graph Structure

Labeled directed acyclic graphs (DAGs). Complex units are non-terminal nodes.



Graph Structure

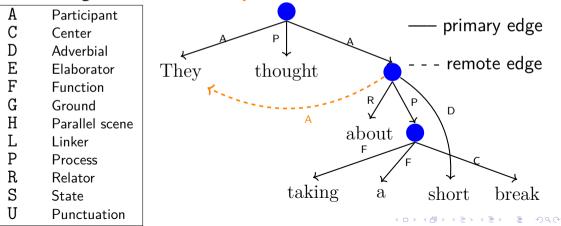
Labeled directed acyclic graphs (DAGs). Complex units are non-terminal nodes. Phrases may be discontinuous.



Graph Structure

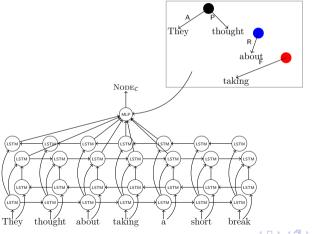
Labeled directed acyclic graphs (DAGs). Complex units are non-terminal nodes. Phrases may be discontinuous.

Remote edges enable reentrancy.



Baseline

TUPA, a transition-based UCCA parser (Hershcovich et al., 2017). bit.ly/tupademo



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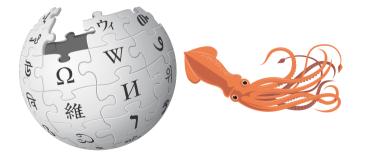
- English Wikipedia articles (Wiki).
- English-French-German parallel corpus from *Twenty Thousand Leagues Under the Sea* (20K).

	sentences	tokens
English-Wiki	5,142	158,573
English-20K	492	12,574
French-20K	492	12,954
German-20K	6,514	144,531

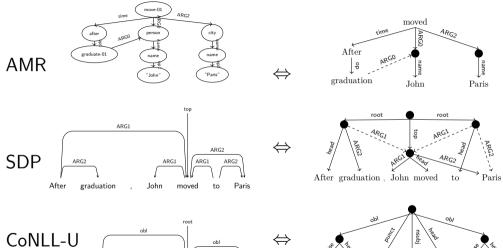


Tracks

- English {in-domain/out-of-domain} × {open/closed}
- German in-domain {open/closed}
- French *low-resource* (only 15 training sentences)



Conversion



case

to Paris

punct

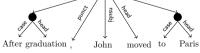
John

After

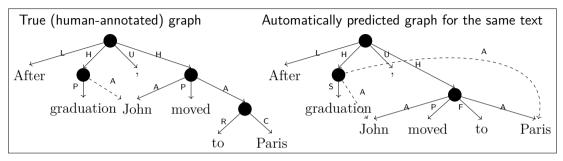
graduation

nsubi

moved

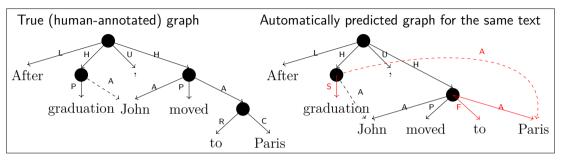


Evaluation



- 1. Match primary edges by terminal yield + label.
- 2. Calculate precision, recall and F1 scores.
- 3. Repeat for remote edges.

Evaluation



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- 2. Calculate precision, recall and F1 scores.
- 3. Repeat for remote edges.

Primary

P
 R
 F1

$$\frac{6}{9} = 67\%$$
 $\frac{6}{10} = 60\%$
 64%

P
 R
 F1

$$\frac{1}{2} = 50\%$$
 $\frac{1}{1} = 100\%$
 67%

Participating Systems

8 groups in total:

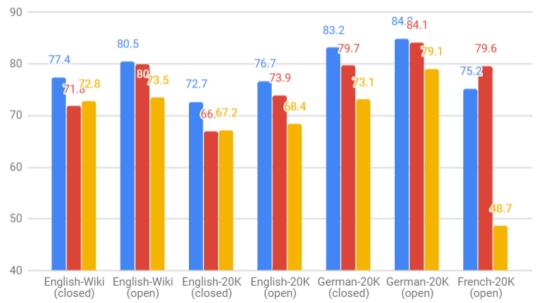
- MaskParse@Deskiñ Orange Labs, Aix-Marseille University
- HLT@SUDA Soochow University
- *TüPa* University of Tübingen
- UC Davis University of California, Davis
- GCN-Sem University of Wolverhampton
- CUNY-PekingU City University of New York, Peking University
- DANGNT@UIT.VNU-HCM University of Information Technology VNU-HCM
- XLangMo Zhejiang University

Leaderboard

Track	1st place		2nd place		3rd place		baseline
English-Wiki closed	HLT@SUDA	0.774	baseline	0.728	Davis	0.722	0.728
English-Wiki open	HLT@SUDA	0.805	CUNY-PekingU	0.800	TüPa	0.735	0.735
English-20K closed	HLT@SUDA	0.727	baseline	0.672	CUNY-PekingU	0.669	0.672
English-20K open	HLT@SUDA	0.767	CUNY-PekingU	0.739	TüPa	0.709	0.684
German-20K closed	HLT@SUDA	0.832	CUNY-PekingU	0.797	baseline	0.731	0.731
German-20K open	HLT@SUDA	0.849	CUNY-PekingU	0.841	baseline	0.791	0.791
French-20K open	CUNY-PekingU	0.796	HLT@SUDA	0.752	XLangMo	0.656	0.487

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HLT@SUDA 📕 CUNY-PekingU 📒 baseline



Main Findings

 HLT@SUDA won 6/7 tracks: Neural constituency parser + multi-task + BERT French: trained on all languages, with language embedding

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Main Findings

- HLT@SUDA won 6/7 tracks: Neural constituency parser + multi-task + BERT French: trained on all languages, with language embedding
- CUNY-PekingU won the French (open) track: TUPA ensemble + synthetic data by machine translation

Surprisingly, results in French were close to English and German

- Demonstrates viability of cross-lingual UCCA parsing
- Is this because of UCCA's stability in translation?

Conclusion

- Substantial improvements to UCCA parsing
- High variety of methods
- Successful cross-lingual transfer

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Thanks!

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Please participate in the CoNLL 2019 Shared Task: Cross-Framework Meaning Representation Parsing SDP, EDS, AMR and UCCA mrp.nlpl.eu

Evaluation Period: July 8-22, 2019

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