# Towards Universal Semantic Tagginge

Lasha Abzianidze (joint work with Johan Bos) 29.05.2018





From *lexical* semantics to phrasal semantics

Semantic lexicon is usually large

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- 07/1937 I have gone to the cinema
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- <u>00/2206</u> I have to warn him

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Can POS tags help?

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Can POS tags help? NO as all the three gets VBP

### **More examples**

- He himself<sup>PRP</sup> tried it, Tom cut himself<sup>PRP</sup> while shaving
- and<sup>CC</sup>, or<sup>CC</sup>, but<sup>CC</sup>
- $\circ$  ... **to<sup>TO</sup>** write ..., ... **to<sup>TO</sup>** cinema ...
- does not like any<sup>DT</sup> X. Give me any<sup>DT</sup> X
- a(n)/every/no/the/some/each/that/these/(n)either...<sup>DT</sup>
- ill<sup>JJ</sup> / skillful<sup>JJ</sup> / fake<sup>JJ</sup> professor
- Google, New York; Ann, Bill and Mary; Ann, a director,...

# Outline

- Groningen/Parallel Meaning bank
- UNIversal SEmantic Tagset
- Results & Challenges
- Conclusion

# Formal compositional semantics in Parallel Meaning Bank

- Heavy lexical units: DRSs
- Few combining rules: Rules of CCG
- $\lambda$ -calculus for computation:  $\lambda$ -DRS

e1 t1 t2 male.n.02(x1)leave.v.01(e1) Time(e1, t1)Theme(e1, x1) time.n.08(t1) t1 X t2  $t1 \prec now$ measure.n.02(t2) t2 X now Unit(t2, day) Theme(t2, 3)

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# Compositionality

#### Projection

Time(v5, t1)

time(t1) t1 < now



(PAT)

















x1 e1 t1 t2 male.n.02(x1)leave.v.01(e1)Time(e1, t1)Theme(e1, x1) time.n.08(t1) t1 X t2  $t1 \prec now$ measure.n.02(t2) t2 X now Unit(t2, day) Theme(t2, 3)

<b>ago</b> λv1.λv2.λv3	.λν4. ( (ν2 @ v3) ĝ	) λν5. (v1 @ λν6. ( t <b>???</b> <sup>ti</sup>	1 : (v me(t1) t1 x v6 v6 x now Time(v5, t1)	4 @ v5) )))
x1 male(x1) (v1 @ x1) )	days λv1. Peacur(v1) Unit(v1. day)	three λv1.λv2.( <b>??</b> Theme		►
	left λv1.λv2. (v1 @ λv3. (e1 t1 : (v2 @ e1 Pave(e1) Theme(e1, v3)			
He MU1. (		time(t1) t1 ~ now		o Av1.Av2. ( x1

Goal

x1 e1 t1 t2 male.n.02(x1)leave.v.01(e1) Time(e1, t1)Theme(e1, x1) time.n.08(t1) t1 X t2  $t1 \prec now$ measure.n.02(t2) t2 X now Unit(t2, day) Theme $(t_2, 3)$ 

## **GMB: Before UNISET**



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#### POS tags are not enough



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left λv1.λv2. (v1 @ λv3. ( e1 t1 : (v2 @ e1) )) lea me(e1, t1) Theme(e1, v3) time(t1) t1 = now

#### POS tags are not enough



λν1.λν2. (v1 @ λν3. ( e1 t1 :(v2@e1))) Theme(e1, v3). time(t1) t1 - now

# Something else is needed

- POS tags lack fine-grained semantic information
- Relying on lemmas  $\rightarrow$  not language neutral
- Relying on CCG categories  $\rightarrow$

framework/language dependent

 Sometimes even a CCG category, lemma, and a POS-tag do not suffice: and, any











# **SEM tags for semantics**

- Schema of lexical semantics is determined by a sem-tag and a syntactic category (SEM, CAT)
- Less sensitive to syntax (vs POS tags)
- Semantic info complements thematic roles, syntax and *lemma*.
- Generalizes over POS tags and Named Entity classes

# **UNIvesral SEemantic Tagset**

- 73 sem-tags divided into 13 classes
- Under development (v0.7)
- Designed in a data-driven fashion (EN, NL, IT, DE)
| Attribute | QUC Concrete<br>quantity<br>QUV Vague<br>quantity | two, six million, twice, 5<br>millions, many, enough |  |  |  |
|-----------|---|--|--|--|--|
|           | COL Colour  | red, crimson, light_blue                             |  |  |  |
|           | Ist Intersective                                  | open, vegetarian, quickly                            |  |  |  |
|           | SST Subsective                                    | skillful surgeon, tall kid                           |  |  |  |
|           |   | former, fake   |  |  |  |
| 1         | PRI Privative                                     | former, fake   |  |  |  |
| _         | PRI Privative<br>DEG Degree                       | former, fake<br>2 meters tall, 20 years old          |  |  |  |
| -         |   | · · · ·  |  |  |  |
| -         | DEG Degree  | 2 meters tall, 20 years old                          |  |  |  |



PRO Anaphoric & deictic pronoun

DEF Definite

HAS Possessive pro.



EMP Emphasizing pro.

he, she, I, him the, lo<sup>IT</sup>, der<sup>DE</sup> my, her blamed herself, each other left himself





he, she, I, him the, Io<sup>IT</sup>, der<sup>DE</sup> my, her blamed herself, each other left himself





iS skiing, dO ski, has skied, now Was baked, had gone, did go will, shall has been being treated, aan\_het<sup>NL</sup> has been going/done









Logical	ALT	Alternatives & repetitions	another, different, again			
	XCL	Exclusive	only, just			
	NIL	Empty semantics	to, .,			
Ľ	DIS	Disjunction & existential quantif.	a, some, any, or			
	IMP	Implication	if, when, unless			
	AND	Conjunction & universal quantif.	every, and, who,	, any	/	
not, no, neither, without must, should, have to					NOT Negation	
mi	ght, d		ps, alleged, can	Modality	POS Possibility	

	DAT Full date	27.04.2017, 27/04/17
tity	Dom Day of Month	27th December
Temporal Entity	Yoc Year of century	2017
pora	Dow Day of week	Thursday
Tem	Month of year	April
T	DEC Decade	80s, 1990s
	CLO Clocktime	8:45_pm, 10_o'clock, noon



that, while, because so, ;, and which, but, yet



here, this, above just, later, tomorrow latter, former, above



	PER Person	Axl Rose, Sherlock Holmes		
	GPE Geo-political entity	Paris, Japan		
	GPO Geo-political origin	Parisian, French		
	Geographical location	Alps, Nile		
ity	ORG Organization	IKEA, EU		
Named Entity	ART Artifact	iOS_7		
	HAP Happening	Eurovision_2017		
Na	Unit of measurement	meter, \$, %, degree Celsius		
	CTC Contact info	112, info@mail.com		
		http://pmb.let.rug.nl		
	Literal use of names	his name is John		
	NTH Other names	table 1a, equation (1)		

#### **Tagging & Semantics**

Formal compositional semantics are less favoured:

- Semantics problems
- Difficult to scale up

Make formal semantics study modular

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NLP community loves tagging/labeling tasks

- Conceptually a simple task
- Create an annotated data
- Employ ML techniques for learning

# Data & Results

- Gold EN documents (34.7K)
- Silver EN documents (1.6M)

Universal Semantic Tags

version	# en	# de	# it	<mark># n</mark> l	silver inc.	release date	download
0.1.0	5438	0	0	0	yes	01-05-2018	19 MB ZIP file

- Baseline (UniGram) ~82%
- Stanford tagger ~88.8%
- NN tagger (AUX UPOS) ~92.7% (M. Abdou)

# Challenges

- Account for wide-coverage compositional semantics
- Keep UNISET independent from CCG
- Prevent the number of sem-tags from increasing

### Conclusion

- Facilitates determining lexical semantics
- Contributes to cross-lingual applications
- Useful for other NLP applications
- Useful for other semantic parsers or RTE systems: (ccg2lambda, LangPro, UDepLambda,...)

#### **Future work**

- Cover more semantic phenomena (data-driven)
- Measure an inter-annotator agreement
- Reorganize tagset to simplify learning

