Semantic Typology and Parallel Corpora Something about Indefinite Pronouns

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- Modeling meaning requires representation space
- Typology: the more languages co-categorize two entities, the more conceptually similar they are (Gentner & Bowerman 2009; Beekhuizen et al. 2014).
- How to obtain such data?



Semantic typology: data acquisition

- Elicitation (Berlin & Kay 1969),
- Secondary sources (Haspelmath 1997),
- Primary text (Cysouw & Wälchli 2009)
 - translated parallel data (subtitles, bibles)
 - reflects actual usage patterns
 - can be used for more abstract domains

Semantic typology: data acquisition

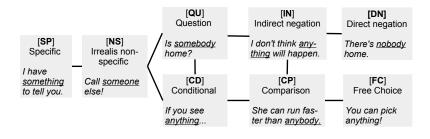
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Our goals:

- contributing to pipeline of extracting verbalization in many languages from parallel text
- compare text-based representations to representations from secondary sources

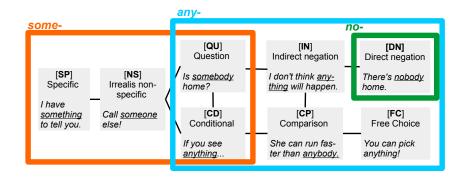
Case study: indefinite pronouns

- Cross-linguistic variation in term extensions
- Formalized using semantic map method (Haspelmath 1997)



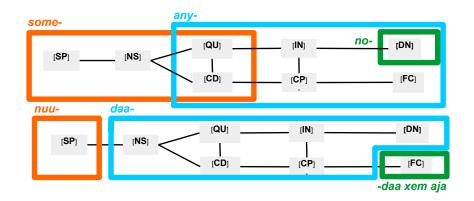
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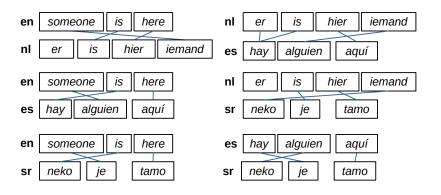


Questions

- Despite great insight, limitations of approach
- Questions better answered with primary texts:
 - Q1 Are all functions equally frequent?
 - Q2 Are functions defined at the right level of granularity?
 - Q3 Do functions display discrete or fuzzy boundaries?
 - Q4 Are functions internally homogenous or do they display further internal structure?

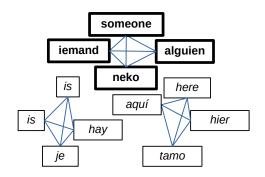
Method

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- Extracted clusters of mutually aligned words
- Linearized clusters and annotated functions

Utterance	en	nl	es	sr	function
someone is here	someone	iemand	alguien	neko	SP
anyone got 5 billion?	anyone	iemand	alguien	neko	QU
she could beat anyone	anyone	iedereen	qualquier	neko	FC

Q1: Frequency of functions

- Split over PEOPLE (e.g., anyone, somebody) and THINGS (e.g., nothing, anything)
- What is the relative frequency per function?

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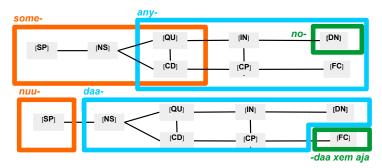
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	SP	NS	CD	QU	IN	DN	СР	FC
PEOPLE THINGS								
Overall	.24	.17	.06	.11	.03	.33	.00	.06

Table: Distribution of functions given ontological category.

SP specific CD conditional IN indirect neg. CP comparison NS DN FC. free choice non-spec. QU question direct negation

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- Is 8 the right number of functions?
- Evaluate with automatic clustering:
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- Evaluate with automatic clustering:
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	k =	2	3	4	5	6	7	8	9	10
PEOPLE		.20	.25	.41	.35	.34	.34	.32	.30	.32
THINGS		.30	.38	.47	.36	.35	.35	.33	.39	.33

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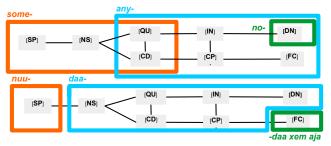


Cluster	SP	NS	CD	QU	IN	DN	СР	FC
1	18	24	6	3	0	2	0	0
2	1	0	2	15	1	4	0	2
3	0	0	1	0	5	27	0	0
4	0	0	0	0	0	0	1	7

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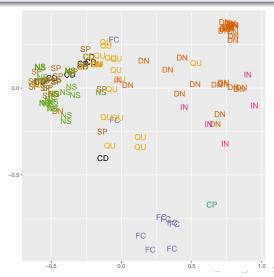


Q3: boundaries between clusters

Optimal Classification MDS (Croft & Poole 2008)

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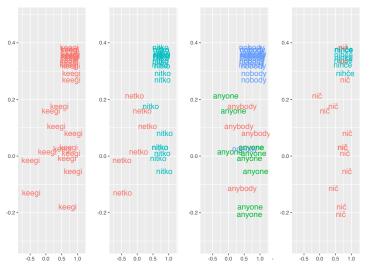
- Optimal Classification MDS (Croft & Poole 2008)
- Clear clusters, but with 'bridges' between them



Q4: internal homogeneity

• Direct negation for PEOPLE in Estonian, Croatian, English, Slovene.

Internal scale: Emphatic > Subjects > Other functions



Recap

- cross-linguistic patterns of co-categorization cognitive representation
- studies indefinite pronouns in parallel usage data (subtitles)
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- usage data allows for fine-grained exploration of semantic contrasts

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Technical extensions

- Scalability: pairwise alignments
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Cognitive plausibility

- E.g., ease of acquisition/order of acquisition
- Similarity/acceptability judgments of language users
- . . .

Thank you!