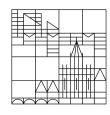


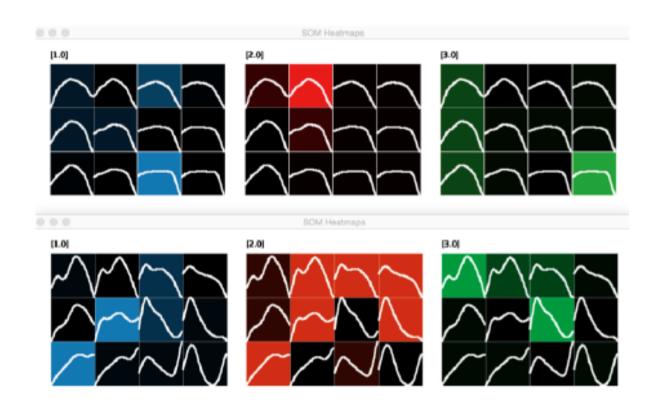
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# Visualizing Linguistic Structure (LingVis)



#### **Miriam Butt**

San Jose, February 15, 2015

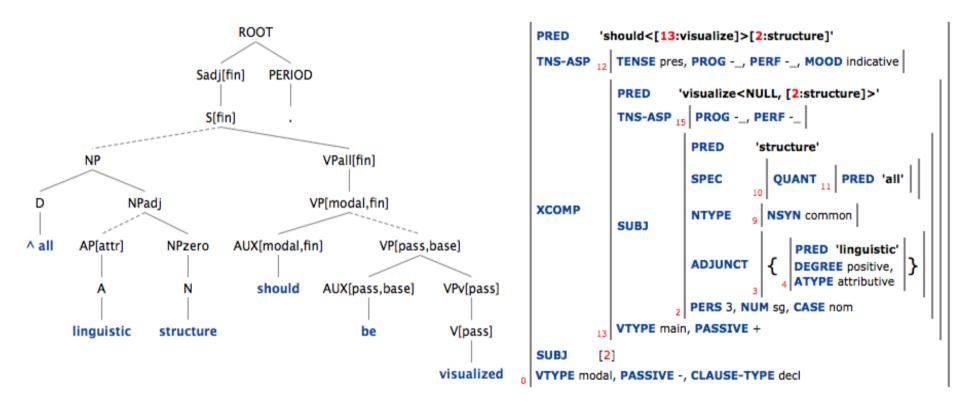
**AAAS Annual Meeting** 

Symposium: Visualizing Verbal Culture: Seeing Language Diversity

# Standard Visualization: Syntax

#### **C-structure**

#### F-structure



Syntactic Analysis with Lexical-Functional Grammar (LFG)

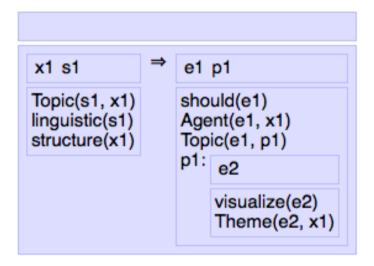
http://iness.uib.no/iness/xle-web (Web Interface for LFG Grammars)

Grammar developed at PARC

# Standard Visualization: Semantics

#### DRS

(Discourse Representation Structure)



Semantic Analysis with Discourse Representation Theory (Boxer)

http://gmb.let.rug.nl/webdemo/demo.php (web interface for CCG/DRT)

Grammar/Semantics developed by Johan Bos and colleagues (Groningen)

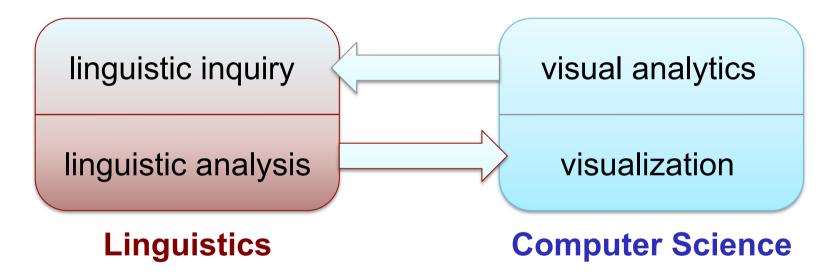
# **Mining Linguistic Data**

#### **Methodological Challenge/Opportunity**

- Use of new technology to detect distributional patterns in language data.
- Ever increasing sources of digital data
  - Wikipedia, social media
  - constructed corpora (raw, annotated: morphology/syntax/semantics)
- Specialized query and search tools (KWIC, COSMAS, DWDS, ANNIS)
- Programming languages specialized for text processing and statistical analysis (Python, R)
- Problem: meaningful patterns difficult to see in the forest of numbers
- Opportunity: Visual Analytics for Linguistics (LingVis)

# **Overall Interdisciplinary Goal**

- Integrate methods from visual analytics into domains of linguistic inquiry.
- Explore challenges based on the needs of linguistic analysis for visualization methods.



#### **Visual Analytics:**

- Interactive exploratory visual access to data.
- Iterations of hypothesis-formation and hypothesis-testing.

# **Example: Identifying N-V complex predicates in Hindi/Urdu**

- Goal: identify sequences of Noun+Verb for understanding complex predicate patterns
  - phone-do, use-do, memory-come, begin-do/ come
- Data: 7.9 million word raw (unannotated) corpus of Urdu (BBC Urdu)

ایک 172 373 7027 تباث 147 394 588 تقو 142 105 235 ادىپ 103 754 956 کالہ 102 1501 3609 دمآرب 210 96 اهكر 263 ىمغز 59 1161 1213 315 75 59 2267 0

197 262

165 13

#this file lists X in X+kar, X+ho, X+hu, X+rakh sequences with corr

#X = word occurring directly to the left of LV (LV: kar, ho, hu, rakh)

0

0

0

0

esponding occurrences in the (candidate) CP sequences

#kar: # of occurrences of X with kar #ho: # of occurrences of X with ho

#hu: # of occurrences of X with hu

674 466 524

366 254 609

227 1232 100

359 135 44

378 2336 1691

#rakh: # of occurrences of X with rakh
X #hu #kar #ho #rakh

[Butt et al., Coling 2012]

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مفاض

عورش

مولعم

حلمح

رثاتم

ناصقن

مكامهد

# **Example: Pixel Visualization**

#### Statistical Data:

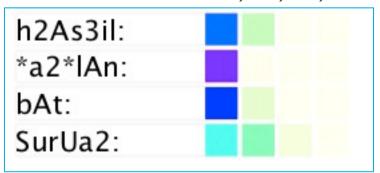
ID	Noun	Rel. freq. with $kar$	Rel. freq. with ho	Rel. freq. with hu	Rel. freq. with $rak^h$
1	حاصل	0.771	0.222	0.007	0.000
2	اعلان	0.982	0.011	0.007	0.000
3	بات	0.853	0.147	0.000	0.000
4	شروع	0.530	0.384	0.086	0.000

Table 2: Relative frequencies of co-occurrence of nouns with light verbs

do, be, become, put kar, ho, hu, rakh

'achievement'
'announcement'
'talk'
'beginning'

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#### Color Scale

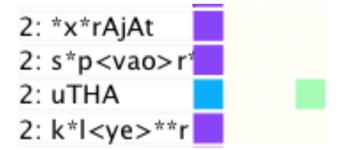
0.25	
0.5	
0.75	
1.0	

# **Example: Identifying N-V complex predicates in Hindi/IUrdu**

Tool facilitates zooming and mousing over to see the underlying data set



Outliers/Errors are easily identified



# **Example: V1 in the History of Icelandic**

## **V1 (Verb Initial or Verb First)**

- Verb initial structures were common in matrix declaratives in Germanic.
- In German (and English) they mostly survive in narrative/joke contexts

Walked a man into a pub...

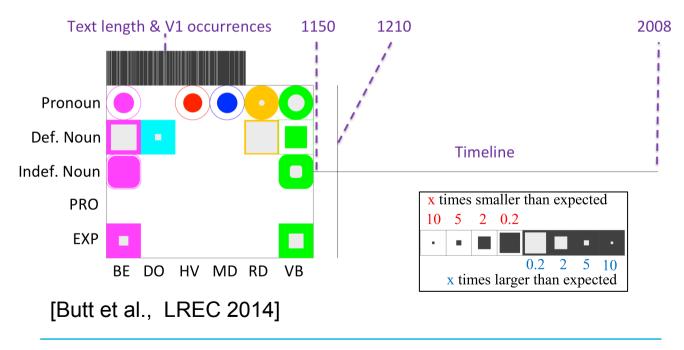
#### **Questions**

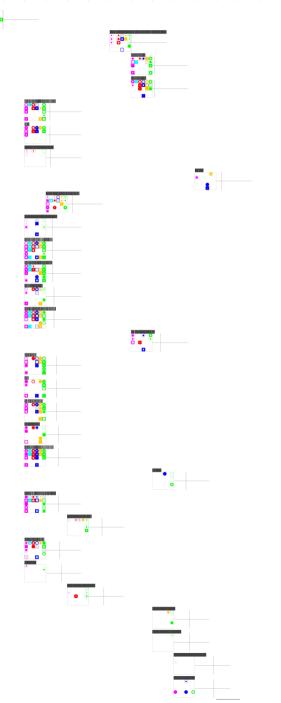
- What determines the appearance of V1?
- How did this change over the history of Germanic?
- Case Study: V1 in the history of Icelandic
- Corpus: IcePaHC
  - syntactically annotated (Penn Treebank style)
  - 60 texts
  - 12th century CE to 21st century CE

# **Example: V1 in Icelandic**

## **Visual Analytic Access to Data**

- Glyph Visualization of likely factors
- Overview of all 60 texts at once
- Can drill down to individual data points interactively
- Keim's Mantra: Overview First Details on Demand

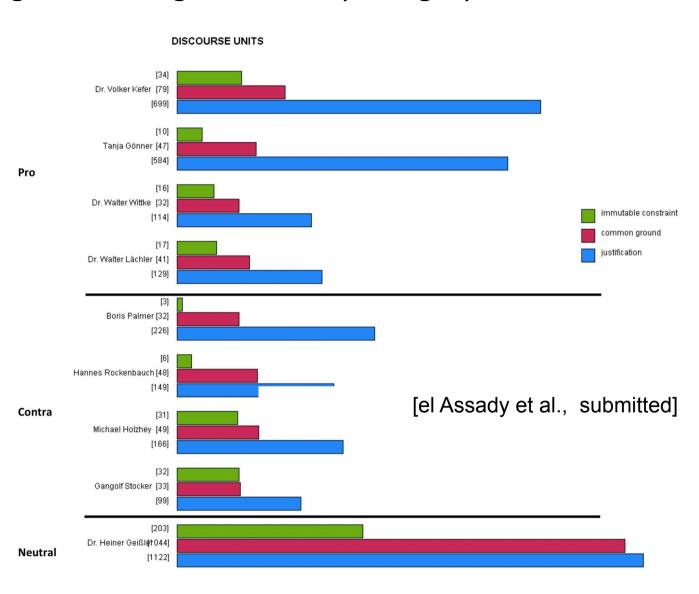




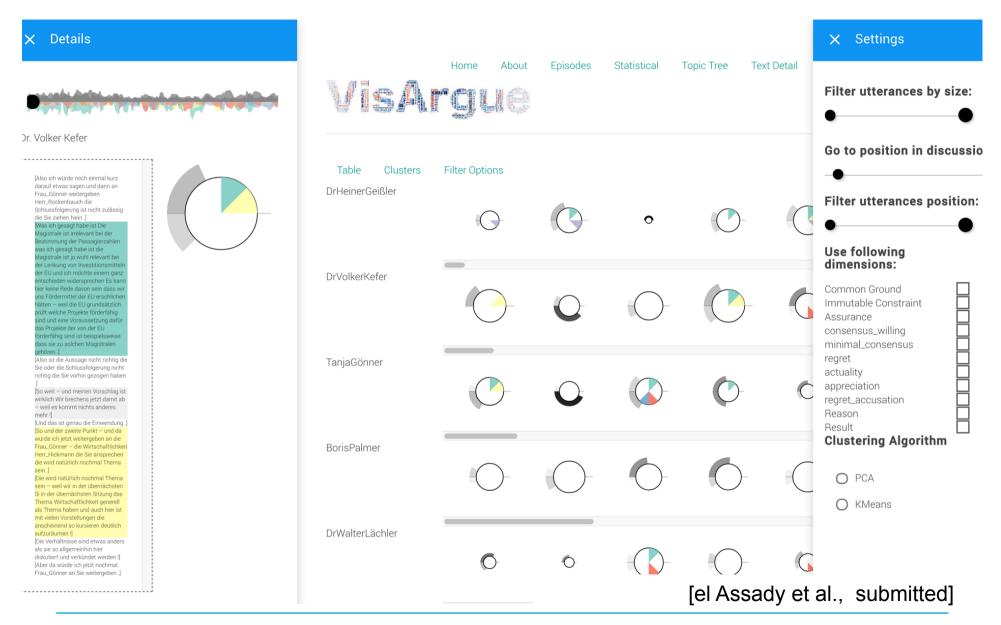
# **Example: Analyzing Political Argumentation (VisArgue)**

- Public mediation on S21 (controversy around Stuttgart train station)
- Speakers are either Pro or Contra.
- Mediator is supposed to be neutral
- Data is annotated (rule based)

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## **Glyph Visualization of Utterance Content**



# **Example: Speech Data**

- Japanese Native and German L2 Learner data (pitch contours and meta data)
- F0 contours are smoothed and normalized into pitch vectors
- The pitch vectors are visualized via self-organizing maps (SOM)

[Sacha et al., submitted]

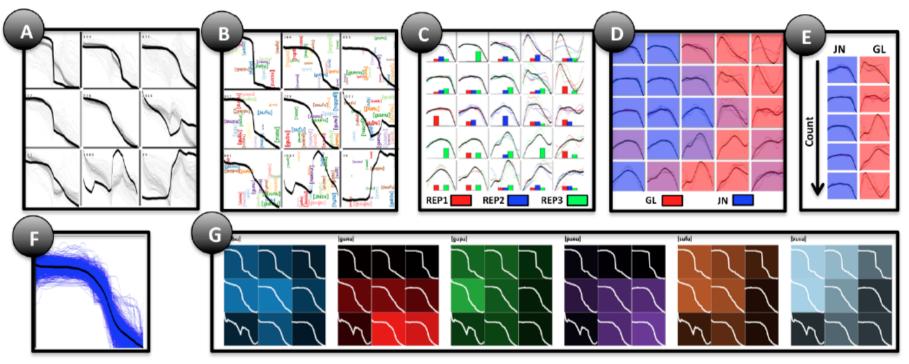


Figure 3: Different approaches to visualize SOM-results according to available meta-data. (A) Grid visualization, (B) word cloud, (C) bar charts, (D) mixed color cells, (E) ranked group clusters, (F) one single cell that visualizes contained vectors and the cluster prototype, (G) separated heatmaps for all values of a categorical attribute.

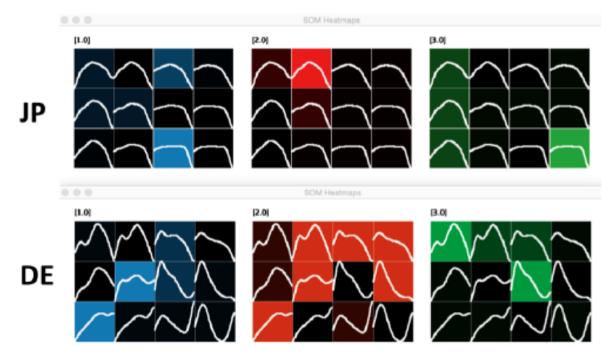
## **Example: Speech Data**

Speakers pronounced "sorry/excuse me" in ever more exasperating circumstances

- Japanese natives do not vary the pitch contour
- German learners do vary the pitch
- German beginner learners do so more

#### **Interactive Exploration:**

- individual cells can be merged
- meta data can be inspected



[Sacha et al., submitted]

#### **Outlook**

## **Further Exploration of Possibilities offered by Visual Analytics**

- The systems illustrated here are very new.
- Interactive exploratory linguistic analysis is on-going.
- Systems are being fine-tuned.

#### Workflow

- Use cases for Digital Humanities /eHumanities are being developed.
- Infrastructure Platforms (mix and match the available tools)

#### **Measuring Success**

- Development of Evaluation Metrics for LingVis.
- Use cases, work flow and result comparison.

# **Thank You!**

#### More and On-line:

- World Language Atlas Explorer: http://th-mayer.de/wals/#30A/
- PhonMatrix: http://paralleltext.info/phonmatrix/

#### **Interdisciplinary Cooperation (University of Konstanz)**

#### Linguistics

Tina Bögel, Annette Hautli-Janicz, Thomas Mayer, Maike Müller, Frans Plank, Christin Schätzle

#### **Computer and Information Sciences**

Daniel Keim, Menna el Assady, Andreas Lamprecht, Christian Rohrdantz, Dominik Sacha

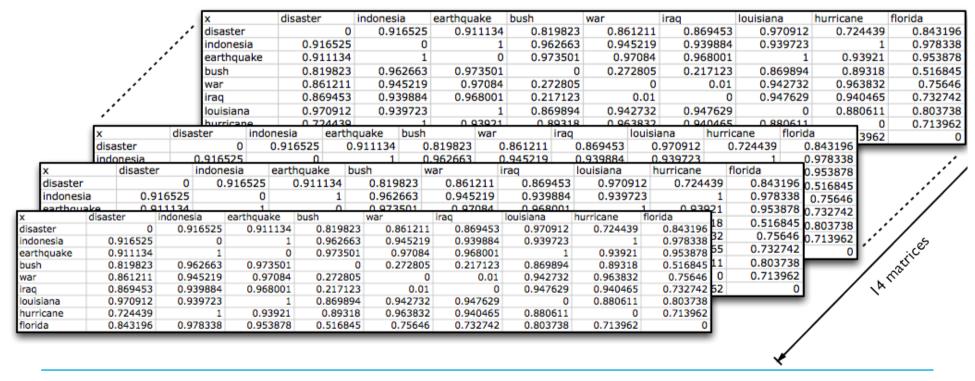
#### **Political Science**

Katharina Holzinger, Valentin Gold

# Example: Correlations between highly frequent words in New York Times articles (2004-2005)

#### **Raw Data**

- 9x9 matrices
- 1 each for each of 14 time slices



# Example: Correlations between highly frequent words in New York Times articles (2004-2005)

**Animated Visualization**(Project Group Oliver Deussen, Univ. of Konstanz)

- Animation of trends/change over time
- Essentials of data easy to access via visualization
- Challenges for Visualization

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- dimensionality reduction: high dimensional distance matrices shown in 2D
- precision vs. stability: a precise visualization for each time step would induce too much confusing movement

earthquake

disaster
louisiana

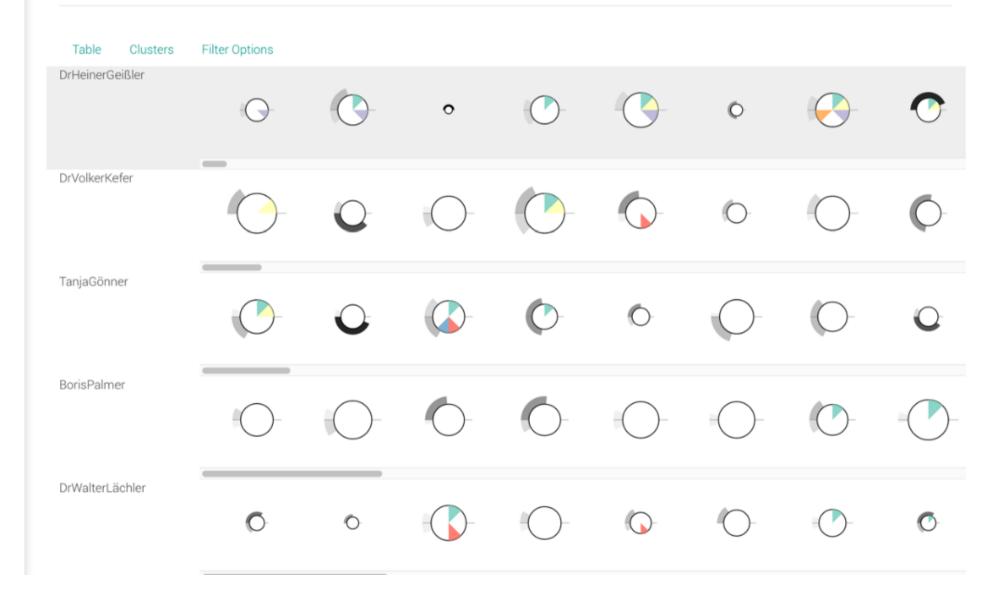
war
iraq
bush

florida
hurricane

2004-11

Home About Episodes Statistical Topic Tree Text Detail Deliberation Argumentation Contact

# VisArgue



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