

About me

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Profile | **Q&A** | **Recommendations** | **Connections**

Current

- Associate Professor at **Leuven** [Edit]

Past

- Assistant Professor at Humboldt Univ. Berlin, U Hamburg

Education

- Humboldt-Universität zu Berlin
- Universität Hamburg
- The University of Edinburgh
- Freie Universität Berlin
- University of Cambridge

Data Mining for Enlightenment
Bettina Berendt
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Thanks to ...

Daniel Trümper
Tool at <http://www.cs.kuleuven.be/~berendt/PORPOISE/>

Ilija Subašić
Tool forthcoming;
all beta testers and experiment participants welcome!

First motivation: Global+local interaction; beyond "similar documents"

with respect to what?

similarity - Google Search - Mozilla Firefox

Web Images Maps News Shopping Gmail more

Google similarity

Search: the web pages

Web Results 1 - 10 of about 23,200,000 for similarity

Similarity - Wikipedia, the free encyclopedia

Similarity is some degree of symmetry or either analogy and resemblance between two or more objects or objects. The notion of similarity is central to many areas of science and philosophy.

Similar pages

Suchen: | Abwärts | Aufwärts | Hervorheben

Fertig

- Similarity Search (15)
- Self-Similarity (13)
- Geometry (3)

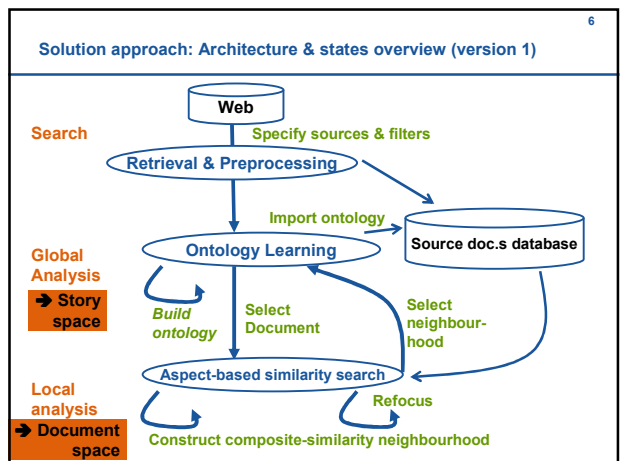
3. Protein ranking: From local to global structure in the protein network. Genetics. Protein ranking: From local to global structure in the protein similarity network.

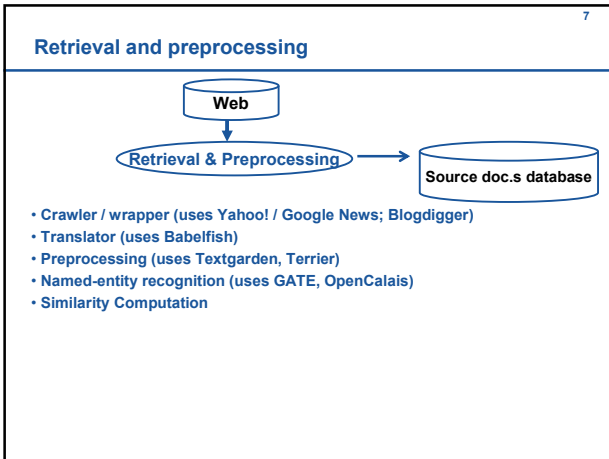
Solution vision: Sailing the Internet

Search

Global Analysis

Local analysis





Ontology learning (1)

The screenshot shows the 'OntoGen -- Text Garden' application. The 'Retrieval & Preprocessing' step is highlighted, showing options for 'Specify sources & filters', 'Import ontologies', 'Build ontology', 'Select neighbourhood', 'Aspect-based similarity search', and 'Refocus'. A 'Source docs' list is visible on the right. At the bottom, it says 'Tool: Blaž Fortuna: <http://blazfortuna.com/projects/ontogen>'.

Ontology learning (2)

The screenshot shows the 'Ontology visualization' step. A hierarchical tree structure is displayed, with nodes like 'root', 'top, front, newspaper', 'subConcept', 'senator, sen, republican', 'general, camera, percent', 'postdoc, bush, farn', 'military, government, election', 'studies, research, patient', 'veteran, research, patient', and 'military, research, patient'. The 'Concept properties' panel on the left shows details for the 'senator, sen, republican' concept, including keywords, SVM keywords, and all documents.

Inspection of ontology and instances

The screenshot shows the 'Porpoise' interface displaying a table of instances for the 'Purpose' cluster. The table has columns for Name, Title, Date of publication, Type, and Language. The instances listed include various news articles and documents related to political events and military actions.

Inspection of documents

The screenshot shows the 'Porpoise' interface displaying a document view for 'Action: The Capitulation Caucus'. The document content includes the following text:

Action: The Capitulation Caucus Tue Sep 11, 2007 at 07:25:23 AM PST: What is The Capitulation Caucus? That's what we want to know and we need your help. In the coming days, Congress will once again take up legislation on Iraq. As we see the first indication of fissures in Republican support for George Bush's endless war, now is the time for Democrats to stand firm. Yet one of the first bills the House may be voting on is the Lieberman/Danahy Bill, H.R. 3087. Toothless doesn't even begin to describe it.

(a) Strategy Required- Not later than 60 days after the date of the enactment of this Act, the President, in coordination with the Secretary of State, the Secretary of Defense, the Joint Chiefs of Staff, and other senior military leaders, shall develop and transmit to Congress

More on documents

The screenshot shows the 'Porpoise' interface displaying a document view for 'Action: The Capitulation Caucus'. The document content is the same as in the previous screenshot. Below the document content, there is a list of 'Named Entities' extracted from the text, including: George Bush, House of Representatives, Secretary of Defense, the House, President, House, Representative, Secretary of State, Congress, Compression, Iraq, and Republican.

The neighbourhood of a document

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Constructing the similarity measure & neighbourhood (I)

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Constructing the similarity measure & neighbourhood (II)

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Constructing the similarity measure & neighbourhood (III)

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Comparing documents

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Comparing documents; utilizing multilingual sources

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Refocusing

Porpoise

File | Export | Help

Similarity Dimensions

Current: 0.02000

Change similarity: [0.1] [+0.1] [-0.1]

Named Entities

Current: 0.10000

Change similarity: [0.1] [+0.1] [-0.1]

Date

Current: 0.02000

Change similarity: [0.1] [+0.1] [-0.1]

Cluster: Neighbours

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Structuring a neighbourhood

Porpoise

File | Export | Help

Similarity Dimensions

Current: 0.02000

Change similarity: [0.1] [+0.1] [-0.1]

Named Entities

Current: 0.10000

Change similarity: [0.1] [+0.1] [-0.1]

Date

Current: 0.02000

Change similarity: [0.1] [+0.1] [-0.1]

Cluster: Neighbours

Ex.: Finding a "story"

Document: Arizona congressman to retire ...

Porpoise

File | Export | Help

Similarity Thresholds

Current: 0.10000

Adjust threshold: []

Named Entities

Current: 0.02000

Adjust threshold: []

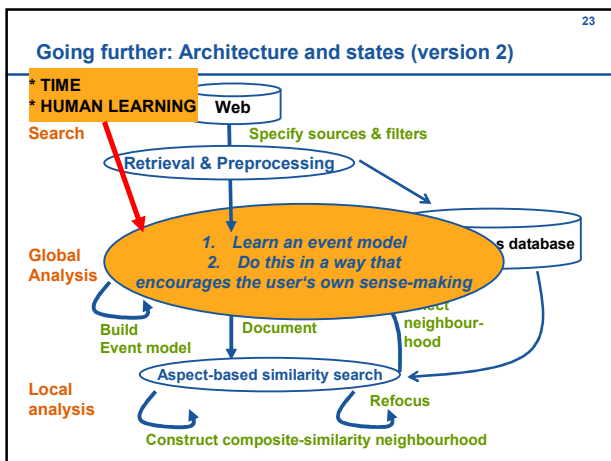
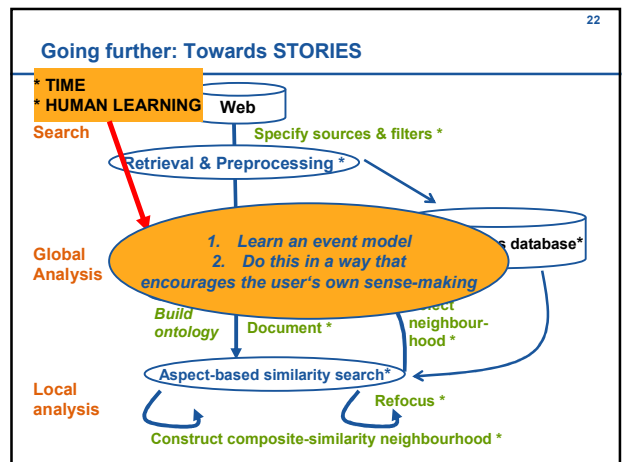
Date

Current: 0.5

Adjust threshold: []

Similarity 0.5 equals 4 days before and after the selected document.

Cluster: Neighbours



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Solution approach 1: Find latent topics

Document Atlas 2.0 - Text Garden

File | About

Map Properties

Document names

Common words

Map style: plain

Gradient

Font sizes: [10] [15] [20]

Categories

Category: [outlet]

Thresholds

Documents: []

Relations: []

Selected Document

File: []

Content: []

Tool: Blaž Fortuna : <http://docatlas.ijs.si>

temporal development only by comparative statics

no „drill down“ possible

no fine-grained relational information

→ lacks structure

Solution approach 2: Temporal latent topics

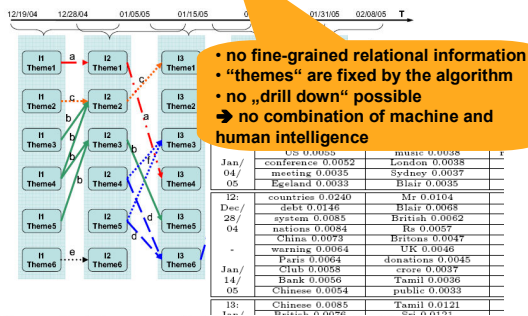


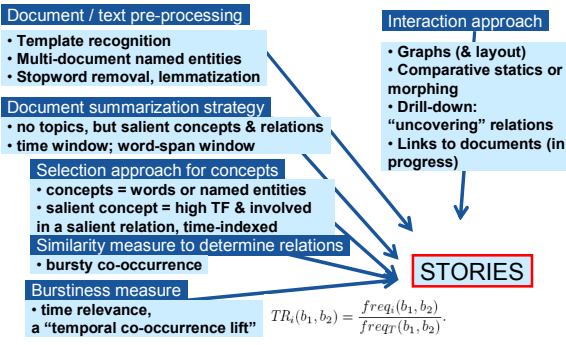
Figure 6: Theme evolution graph for Asia Tsunami
Mei & Zhai, PKDD 2005

The ETP3 problem

Evolutionary theme patterns discovery, summary & exploration

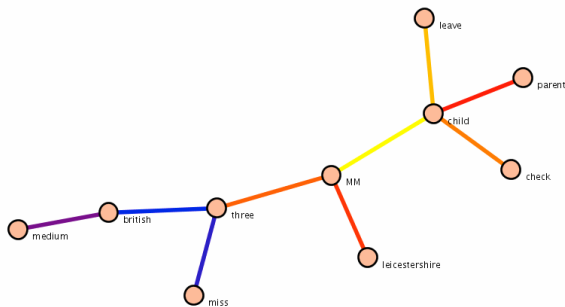
1. identify topical sub-structure in a set (generally, a time-indexed stream) of documents constrained by being about a common topic
2. show how these substructures emerge, change, and disappear (and maybe re-appear) over time
3. give users intuitive and interactive interfaces for exploring the topic landscape and the underlying documents
and for their own sense-making
– use machine-generated summarization only as a starting point!

Ingredients of a solution to ETP3 = Evolutionary theme patterns discovery, summary and exploration

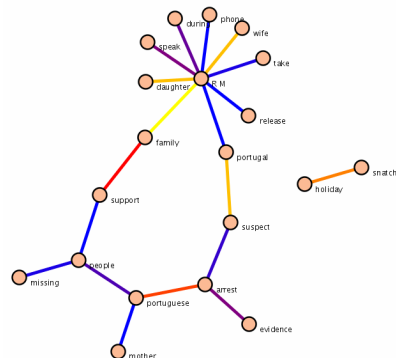


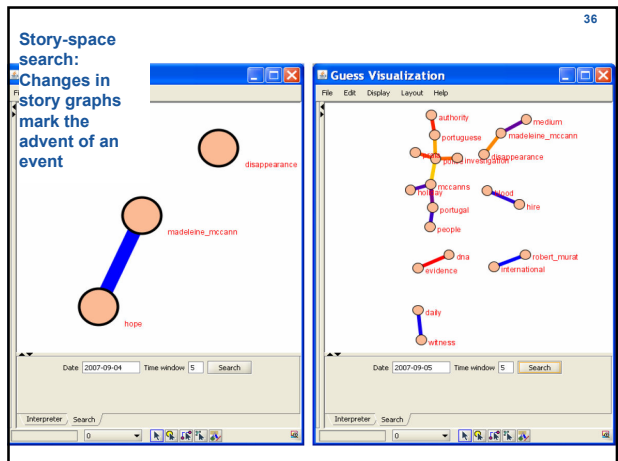
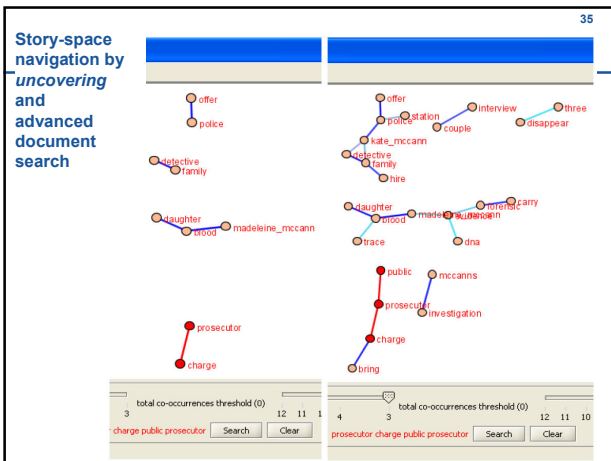
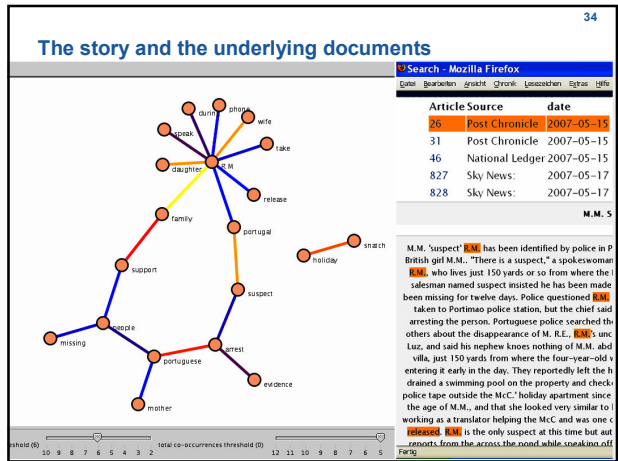
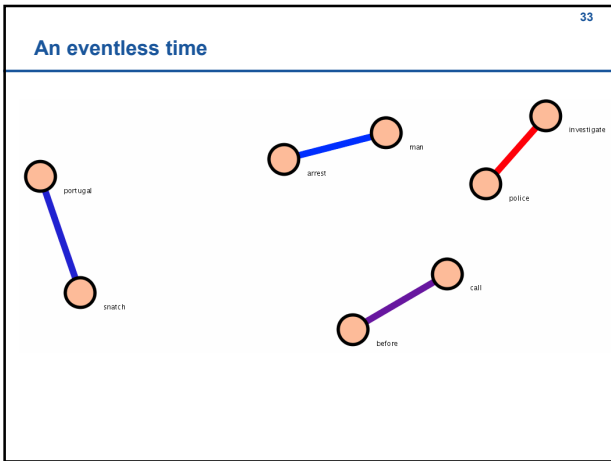
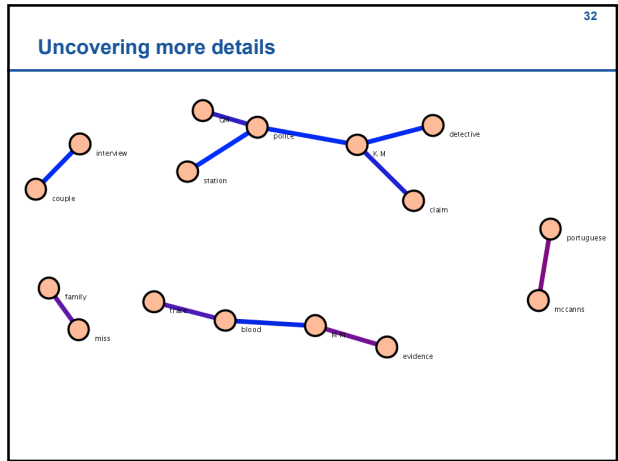
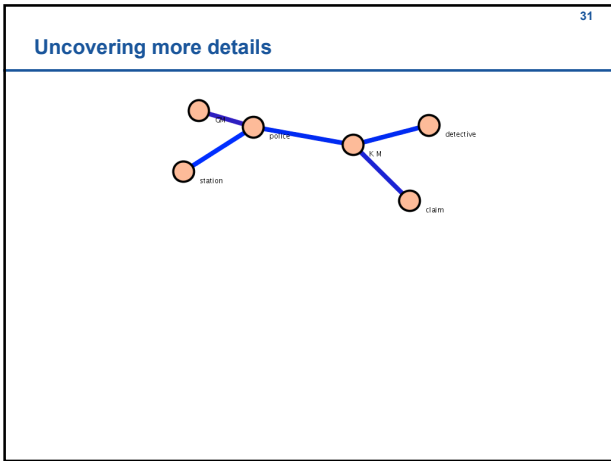
"Powerpoint demo"

An event: a missing child



A central figure emerges in the police investigations





Data collection and preprocessing

- Articles from Google News 05/2007 – 11/2007 for search term “madeleine mccann”
 - (there was a Google problem in the December archive)
- Only English-language articles
- For each month, the first 100 hits
- Of these, all that were freely available → 477 documents

- Preprocessing:
 - HTML cleaning
 - tokenization
 - stopword removal

Story elements

- *content-bearing words*
 - the 150 top-TF words without stopwords

$$TF_{ij} = \frac{n_{ij}}{|d_i|}$$

Story stages: co-occurrence in a window

[Sep 7, 2007] Madeleine McCann's mother is being made a formal suspect by police investigating the child's disappearance. The move was revealed as Kate McCann, 35, who emerged early on Friday morning after almost 11 hours of questioning by Portuguese police on Thursday night, prepared to face officers again.

- “mother” and “suspect” co-occur
 - in a window of size ≥ 6 (all words)
 - in a window of size ≥ 2 (non-stopwords only)

Salient story elements

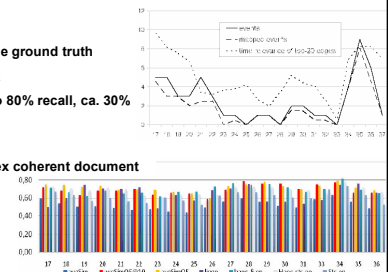
1. Split whole corpus T by week (17 = 30 Apr + until 44 = 12 Nov +)
2. For each week
 - Compute the weights for corpus t for this week
3. Weight =
 - Support of co-occurrence of 2 content-bearing words w_1, w_2 in t = (# articles from t containing both w_1, w_2 in window) / (# all articles in t)
4. Threshold
 - Number of occurrences of co-occurrence(w_1, w_2) in t ≥ θ_1 (e.g., 5)
 - Time-relevance TR of co-occurrence(w_1, w_2) = support(co-occurrence(w_1, w_2)) in t / support(co-occurrence(w_1, w_2)) in T ≥ θ_2 (e.g., 2) *
5. Rank by TR, for each week identify top 2
6. Story elements = peak words = all elements of these top 2 pairs (# = 38)

Salient story stages, and story evolution

7. Story stage = co-occurrences of peak words in t
 - For each week t: aggregate over t-2, t-1, t → moving average
8. Story evolution = how story stages evolve over the t in T

Evaluations (so far ...)

1. Information retrieval quality
 - Challenge: What is the ground truth
 - → Build on Wikipedia
 - Edges – events: up to 80% recall, ca. 30% precision
2. Search quality
 - Story subgraphs index coherent document clusters
3. Learning effectiveness
 - Document search with story graphs leads to averages of
 - 75% accuracy on judgments of story fact truth
 - 3.4 nodes/words per query



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Summary

Navigation in story space → story building

+

Document search

+

Navigation in document space

lead to understandable, useful + intuitive interfaces

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The Future

- Better language processing
- Linkage information!
- Opinion mining

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Thanks!

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References

Subašić, I. & Berendt, B. (2008). Web Mining for Understanding Stories through Graph Visualisation. In *Proceedings of ICDM 2008*. IEEE Press.

Berendt, B. and D. Trümper (in press). Semantics-based analysis and navigation of heterogeneous text corpora: the Porpoise news and blogs engine. In I.-H. Ting & H.-J. Wu (Eds.), *Web Mining Applications in E-commerce and E-services*, Berlin etc.: Springer.

Berendt, B. & Subašić, I. (in press). Measuring graph topology for interactive temporal event detection. To appear in *Künstliche Intelligenz*.

Berendt, B. & Subašić, I. (under review). Discovery of interactive graphs for understanding and searching time-indexed corpora.

Please see <http://www.cs.kuleuven.be/~berendt/> for these papers.