

# Towards the Realistic Natural Language Representations

Examples and Experience

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# Outline and Take-home Message

Overview

Outline

Natural Language Processing Successes and Failures

Human Computing Inspiration

Psycholinguistic evidence

Unreasonable Effectiveness of Data

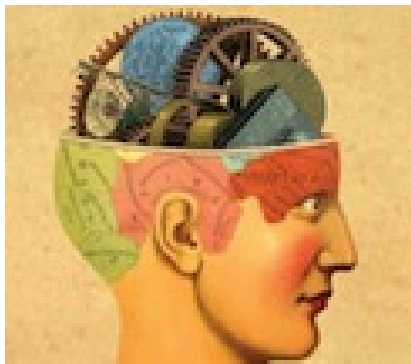
Examples

Exploratory [Math] Search

Summary

# An Example of Natural Language Processing

zuzana en-gram neverit vhlava mit



# Levels of Natural Language Processing

1. strings as information bearing chunks (Vítek)
2. words (morphology)
3. phrase and sentence parsing (syntax)
4. word usage in context, word sketches, collocations, named entity recognition, word meaning, phrase and sentence meaning, narrative representations (semantics)
5. natural language understanding and dialogue, information (knowledge) retrieval (pragmatics)

Mostly *discrete, non-distributional* representations: grammars, dictionaries, thema/rhema segmentations, word sketches.

# Failures of Natural Language Processing

- ▶ Machine translation
- ▶ Semantic web
- ▶ Turing test

# Successes of Natural Language Processing

- ▶ Watson, Siri: **machine learning** of **continuous** representations
- ▶ **distributional** semantics (LSI, LDA)
- ▶ **random** walking in texts (Page's rank), availability of computational resources (global, data intensive and driven computation)
- ▶ **human** computation with fun (Luis von Ahn)!



# Inspiration from Human Computers (Psycholinguistics Evidence)

- ▶ evidence for **priming**, **semantic priming**, “on-the-fly computation from distributed linked representation”.
- ▶ Hoey’s **lexical priming** theory: each occurrence of lexical item enforces ‘priming’ of it given a co-**locational** context
- ▶ “new encounter either reinforces the priming or loosens it”: Hebbian learning principle of synaptic neural net learning
- ▶ **locality** principle (by magnetic resonance)
- ▶ **subconsciousness** processing (during sleep); one-frame a advertisement
- ▶ functionality, expressiveness, behavioural **change** depending on “spike trains” of internal (beer, marijuana, engagement) or external stimuli (conscious or unconscious learning): nothing carved forever as insect into amber – **equilibrium dynamics**
- ▶ language **subjectivity** (entity meaning)

# Towards Computational Lexical and Semantic Primings?

- ▶ what representations should be used?
- ▶ appropriate data structures for given algorithm and application is the key (c.f. OOP)
- ▶ empirical linguistics, natural language data (e.g. text or speech corpora) as valuable asset and data to deliver natural language application functionality

Doug Laney (Chief Data Officer): “We have only two assets: money and data. If we don’t treat data as an asset, then it’s an expense.”

Peter Norvig (Google Research President): “We should stop acting as if our goal is to author extremely elegant theories, and instead embrace complexity and make use of the best ally we have: the **unreasonable effectiveness of data.**”



Conscious  $\rightarrow$  Unconscious

Discrete  $\rightarrow$  Continuous

# Unreasonable Effectiveness of Language Data Representations Computed from Corpora

- ▶ introvert/extrovert evaluation from tweets or facebook status changes
- ▶ opinion mining from texts by random walking in texts (Sebastiani)

from “scholastic” **discrete** language representations (word lists, word nets, word sketches, graphs, logics, parsing trees) **into continuous** representations: syntax ‘only’ encodes information structure, is secondary; meaning is primary

as natural language is **subjective**, so it make sense to use personal/thematic corpora for choosing and interpreting data

Random walking with restart (from PageRank to WordMeaningRank):

- ▶ a document is supported when cited, and credit propagates (iterative computation over graph/matrix) ending in continuous credit evaluation computed from graph/matrix cluster
- ▶ a word meaning computed from corpora is supported when used, and word meaning propagates, ending in continuous word meaning evaluation from given (personal, domain) corpora

Although computationally extensive, use cases in [math] IR, word and formulae disambiguation, . . .

# Coping with Information Overload by Filtering of *Big Data*



Life is searching, searching is *killer app* in information society: group **similar** and narrow focus of (faceted) search in [your] Big Data using natural language knowledge representation.

Different types of representation useful: for **plagiarism** word/entity/citation  $n$ -grams, for narrative search trains of thought signatures (similarity of semantic, meaning representation in space), for topical exploratory search **distributional semantics**.

# Exploratory [Math] Search Trends (Daniel Tunkelang)



- ▶ Entity-oriented search (NER, word  $\rightarrow$  entity)
- ▶ Knowledge graphs (KG (Google), Satori (Microsoft), Freebase), LinkedIn, Facebook)
- ▶ Search assistance, suggestions, dialogue

# NER Search Example: Project Ottův Slovník naučný

Zde zadejte hledané slovo nebo slova oddělená čárkou (= operátor ACCRUE) nebo jiným operátorem (lze použít tlačítek Volba operátorů). Slova budou hledána ve všech přípustných tvarech (kromě změny kmenové souhlásky) bez zřetele na velká/malá písmena. Při hledání fráze (např. univerzita karlova) se tato dvě slova zadají neoddělená čárkou. Při hledání slova v přesném tvaru (bez skloňování/časování) se slovo uvede v uvozovkách, např. „VŠE“.

Hledaný text

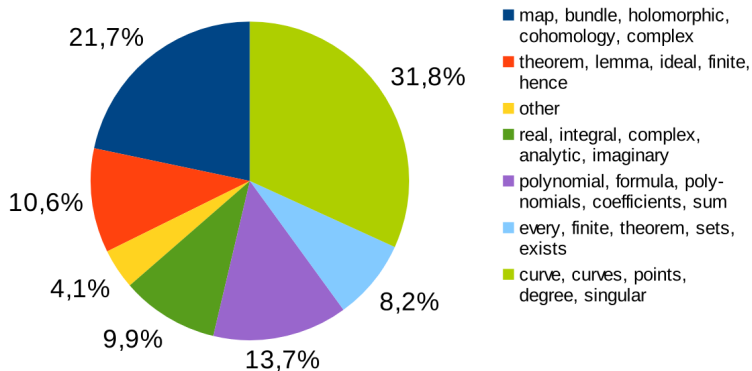
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 hledat v plných textech hesel     pouze v názvech hesel     volný text

Vazby mezi výrazy: ACCRUE (čím více, tím lépe)    AND (a)    OR (nebo)    NOT (ne)

Grabbing the essence (content) of documents → **topical modelling**.

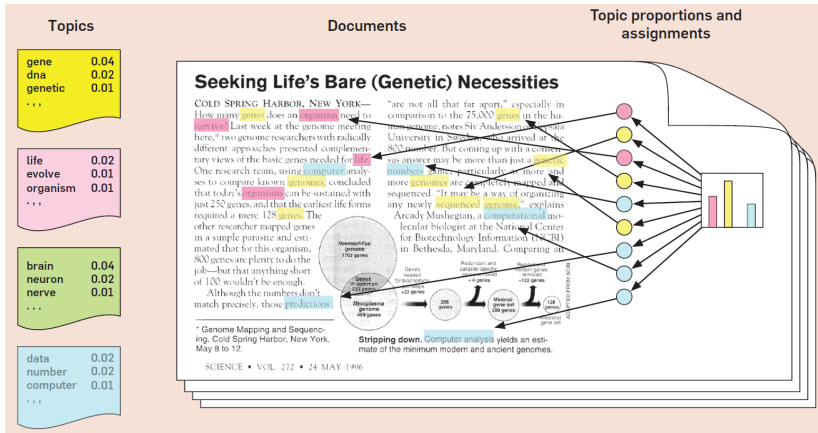
# Leading Edge Example: Automated Meaning Picking from Texts

LDA Topics Pie Chart for math.0406240



# Probabilistic Topical Modeling: Latent Dirichlet Allocation

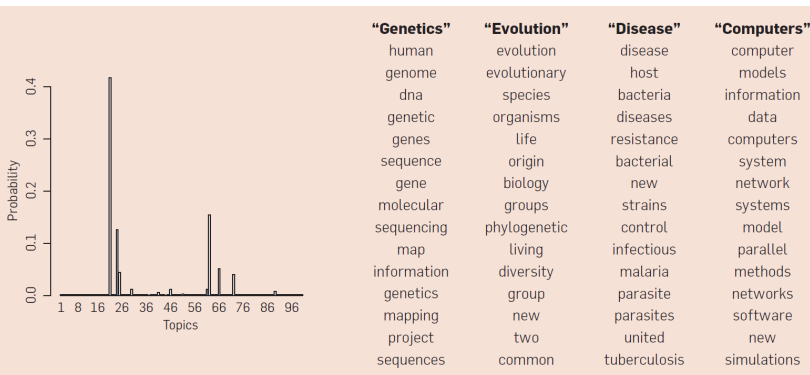
- ▶ topic: weighted list of words
- ▶ document: weighted list of topics

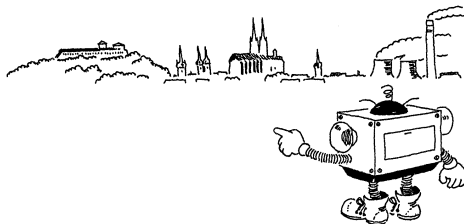




# Topical Modeling: Latent Dirichlet Allocation II

- ▶ all topics computed automatically from document corpora





- ▶ evidence for better NL representation from psycholinguistics
- ▶ from discrete to continuous language representations
- ▶ data-driven, empirical linguistics: the unreasonable effectiveness of distributive, continuous, language representations computed from linguistic data like corpora
- ▶ use cases for exploratory search, disambiguation in information retrieval

Credits: Jiří Franek, (illustrations); Daniel Tunkelang, David Blei (pictures)