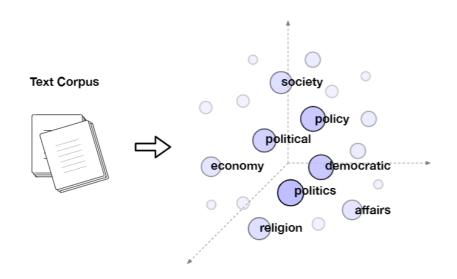


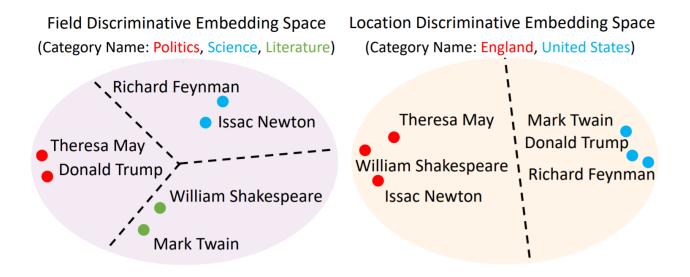
Over 80% of Big Data is Unstructured Text Data

- Ubiquity of big unstructured, text data
 - Big Data: Over 80% of our data is from text (e.g., news, papers, social media): unstructured/semi-structured, noisy, dynamic, inter-related, high-dimensional, ...
- How to mine/analyze such big data systematically?
 - Text Representation (i.e., computing vector representations of words/phrases/sentences)
 - Basic Structuring (i.e., phase mining & transforming unstructured, typed entities/relationships (IE))
 - Advanced Structuring: Discovering Hierarchies/taxonomies, exploring in multi-dimensional space

Text Representation: Embeddings & Language Models

Word embeddings map words into a vector space which reflects semantic similarity



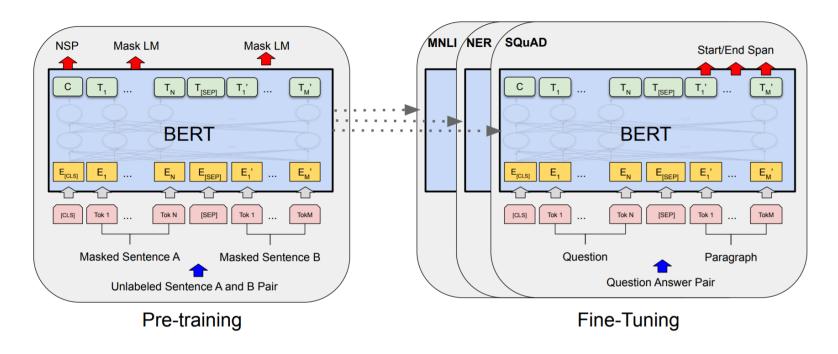


Unsupervised word embeddings: learned from corpus statistics

(Weakly-)supervised word embeddings: learned from corpus statistics & user guidance

Text Representation: Embeddings & Language Models

 Language models are pre-trained on large-scale general-domain corpora to learn universal/generic language representations that can be transferred to downstream tasks via fine-tuning

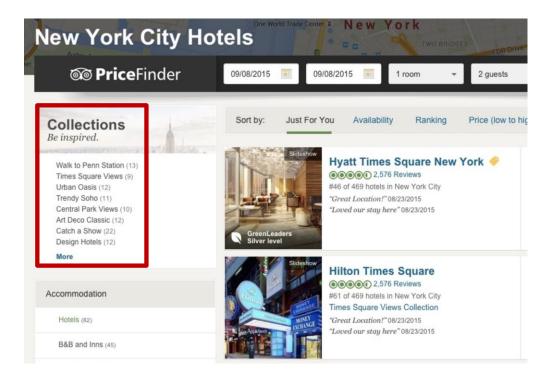


Unsupervised/Self-supervised;
On large-scale general domain corpus

Task-specific supervision;
On target corpus

Basic Structuring: Phrase Mining and Information Extraction

Example: Finding "Interesting Hotel Collections"



Grouping hotels based on structured facts extracted from the review text

Different Dimensions of Information

Features for "Catch a Show" collection

- 1 broadway shows
- 2 beacon theater
- 3 broadway dance center
- 4 broadway plays
- 5 david letterman show
- 6 radio city music hall
- 7 theatre shows

Features for "Near The High Line" collection

- 1 high line park
- 2 chelsea market
- 3 highline walkway
- 4 elevated park
- 5 meatpacking district
- 6 west side
- 7 old railway

Basic Structuring: Automated Named Entity Recognition & Typing

Angiotensin-converting enzyme 2 GENE_OR_GENOME (ACE2 GENE_OR_GENOME) as a SARS-CoV-2 CORONAVIRUS receptor CHEMICAL: molecular mechanisms and potential therapeutic target.

SARS-CoV-2 CORONAVIRUS has been sequenced [3]. A phylogenetic EVOLUTION analysis [3,4] found a bat WILDLIFE origin for the SARS-CoV-2 CORONAVIRUS. There is a diversity of possible intermediate hosts NORP for SARS-CoV-2 CORONAVIRUS, including pangolins WILDLIFE, but not mice EUKARYOTE and rats EUKARYOTE [5] . There are many similarities of SARS-CoV-2 CORONAVIRUS with the original SARS-CoV CORONAVIRUS. Using computer modeling, Xu et al **PERSON**. [6] found that the spike proteins GENE_OR_GENOME of SARS-CoV-2 CORONAVIRUS and SARS-CoV CORONAVIRUS have almost identical 3-D structures in the receptor binding domain that maintains Van der Waals forces PHYSICAL_SCIENCE . SARS-CoV spike proteins GENE_OR_GENOME has a strong binding affinity DISEASE OR SYNDROME to human ACE2 GENE OR GENOME, based on biochemical interaction studies and crystal structure analysis [7]. SARS-CoV-2 CORONAVIRUS and SARS-CoV spike proteins GENE_OR_GENOME share identity in amino acid sequences and, importantly, the SARS-CoV-2 CORONAVIRUS and SARS-CoV spike proteins GENE OR GENOME have a high degree of homology [6, 7]. Wan et al PERSON. [4] reported that residue 394 CARDINAL (glutamine CHEMICAL) in the SARS-CoV-2 CORONAVIRUS receptor-binding domain

Adv. Structuring: Multidimensional Nature of Texts

The same document can naturally describe things across multiple dimensions

- Example:
 - A technical review may cover
 - Brands
 - Products
 - Aspects
 - Years

Apple's 10th anniversary iPhone X sets a new gold standard for the next decade of iPhones. Coming hot on the heels of the iPhone 8 and iPhone 8 Plus, the iPhone X stole the show despite sharing nearly identical internal hardware. The X (pronounced "ten," like the Roman numeral) is a beautiful, modern sculpture, and iPhone owners finally have a reason to show off their phones again.

As we're now about four months from Apple's next iPhone launch, we're revisiting the iPhone X to see if it's still worth the high price tag.

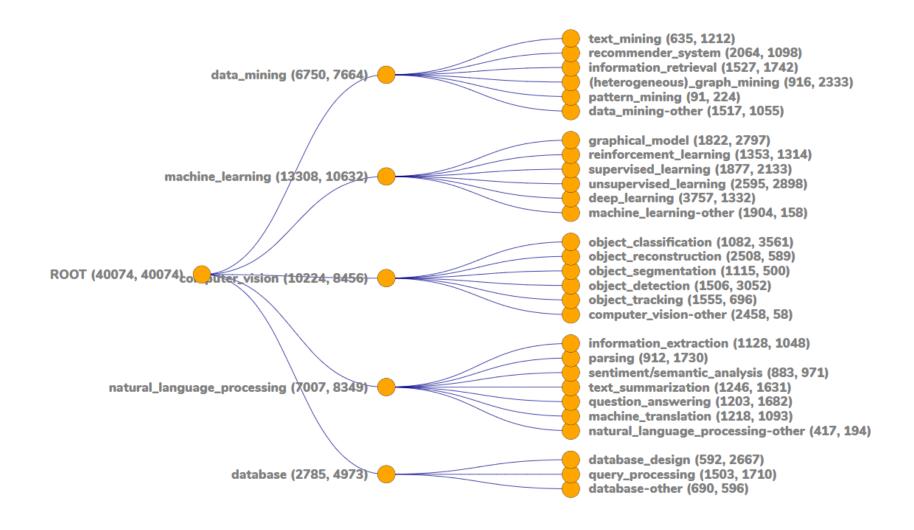
•••

Advanced Structuring: Automatic Taxonomy Generation

Automatically Generated Taxonomy Visualization

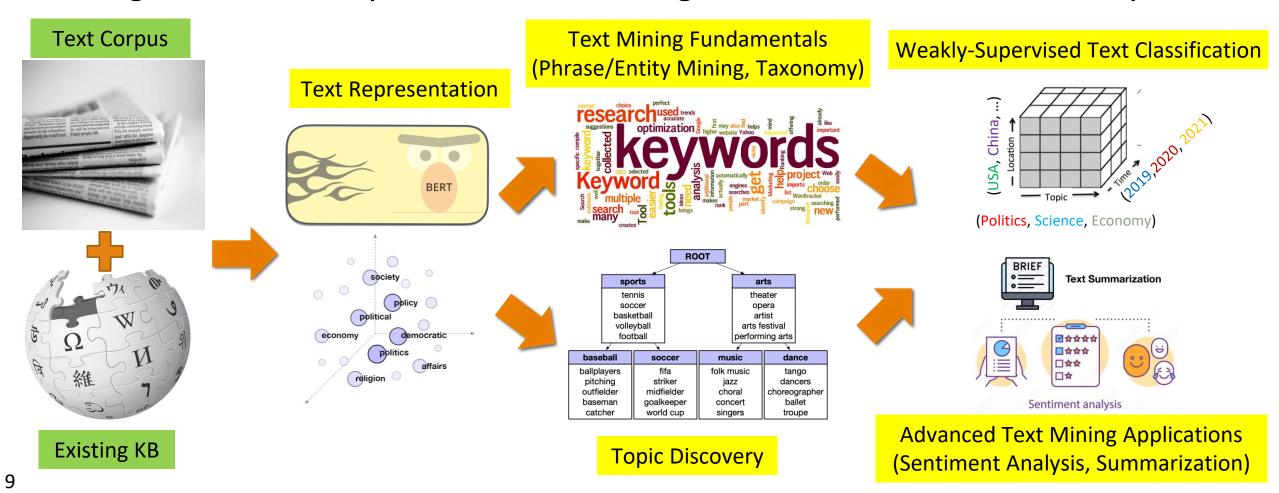
Current Selected: ROOT

Numbers in () from left to right represents the number of main papers and the number of secondary papers respectively.



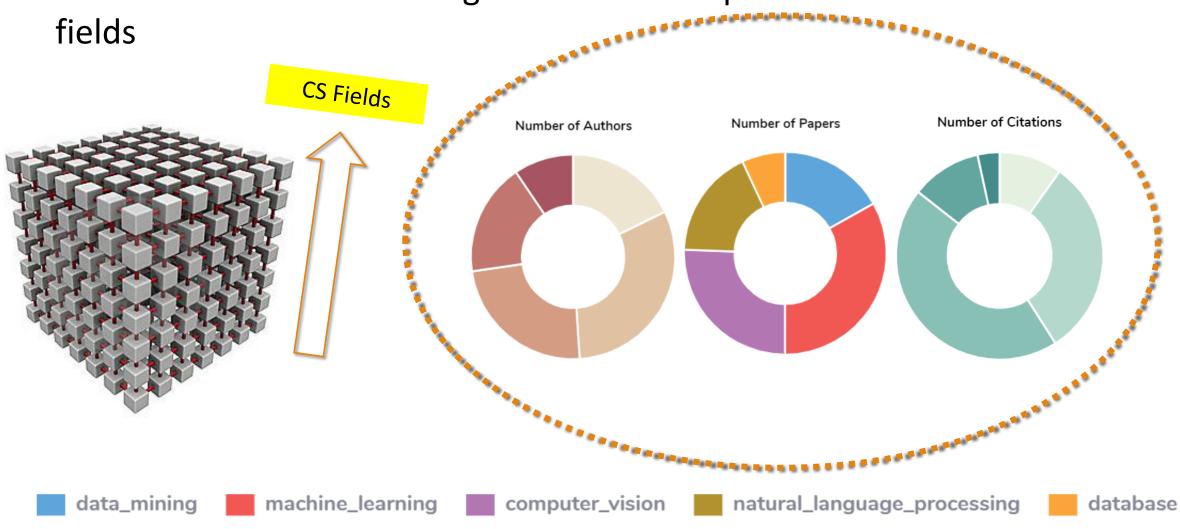
Adv. Structuring: Multi-Dimensional Text Cube Construction

- Understand and Extract Information from Massive Text Corpora
- Organize and Analyze Information using Multidimensional Text Analysis



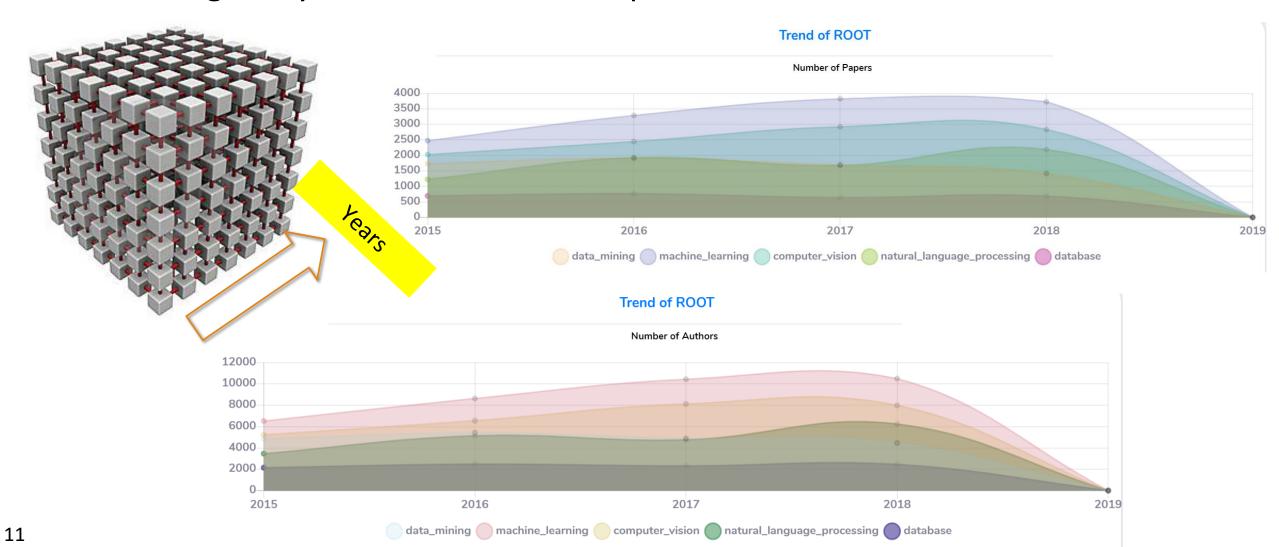
Application: DBLP—Automatic Paper Categorization

Multidimensional text categorization and exploration across different CS

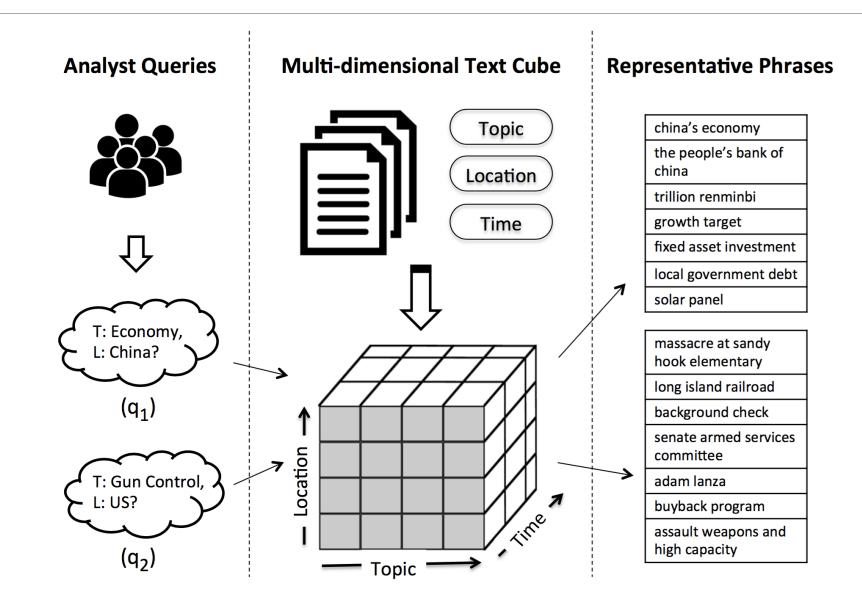


Application: DBLP—Trending Analysis

□ Trending analysis on CS field development



Application: Comparative Summarization



Tutorial Outline

- Introduction
- Part I: Text Embedding and Language Models
- □ Part II: Revisiting Text Mining Fundamentals with Pre-Trained Language Models
- Part III: Embedding-Driven Topic Discovery
- □ Part IV: Weakly-Supervised Text Classification: Embeddings with Less Human Effort
- Part V: Advanced Text Mining Applications Empowered by Embeddings
- Summary and Future Directions

Our Roadmap of This Tutorial

