

# Towards Optimized Multimodal Concept Indexing

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#### TU

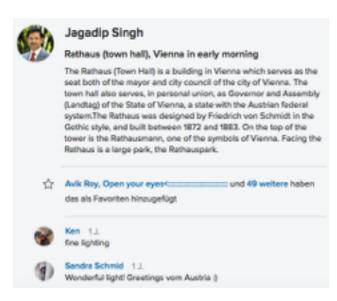
# Agenda

- Multimodal Retrieval
  - Social Image Retrieval
  - Mucke Framework
- Concept-based Text Retrieval
  - Semantic Similarity
  - Methodology
  - Experimental Results
- Optimization
  - Two-Phase Process
  - Approximation Nearest Neighbors
- Conclusion



#### **Multimodal Retrieval**

- Social Image Retrieval → our focus
  - Images
  - Tags, title, and description
  - Meta-data i.e. user profile and Wikipedia page
  - Key-word search

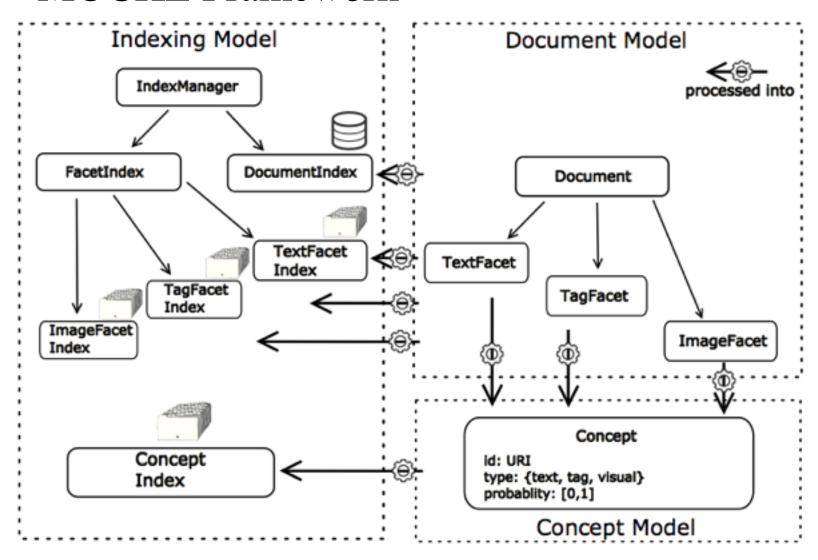




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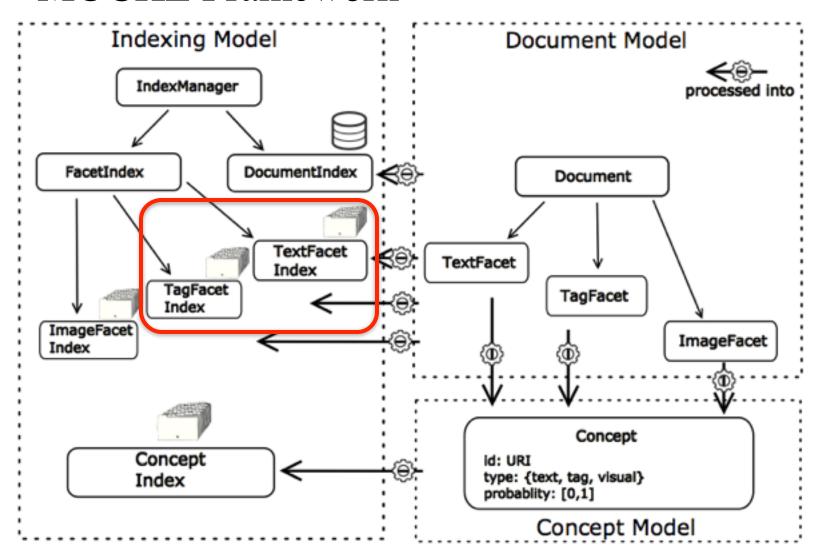
#### **W** Concept-based Multimodal Indexing

#### MUCKE Framework



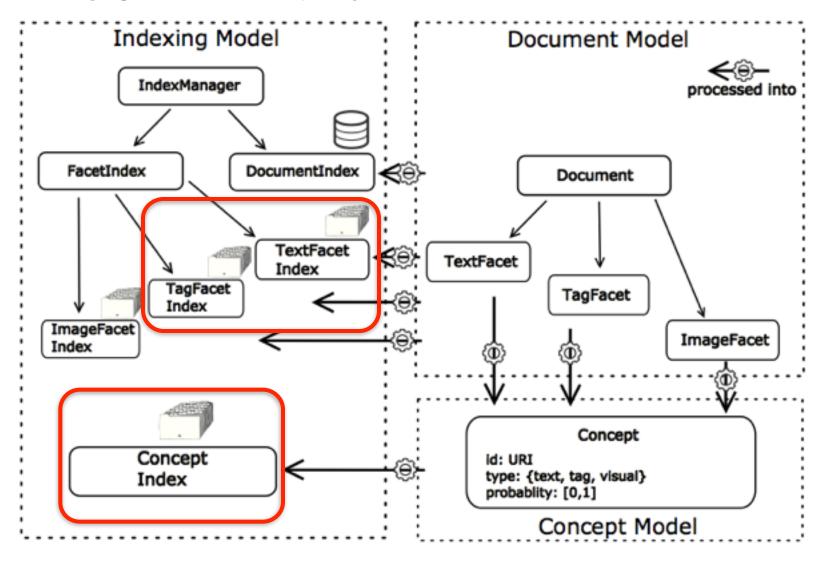
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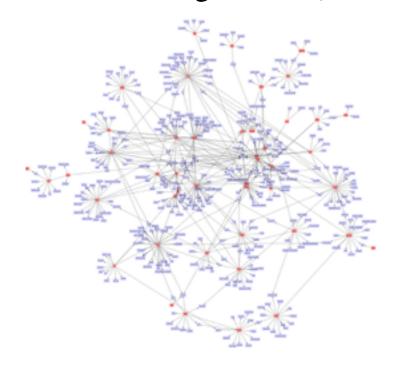


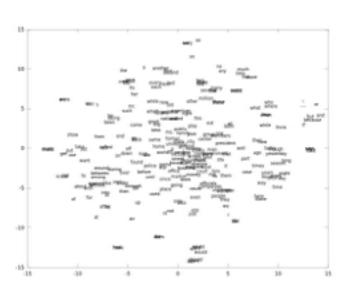


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  - synonyms (bank, trusted company)
  - hyponym/hypernym (skyscraper, building)
  - antonym (cold, warm) etc.



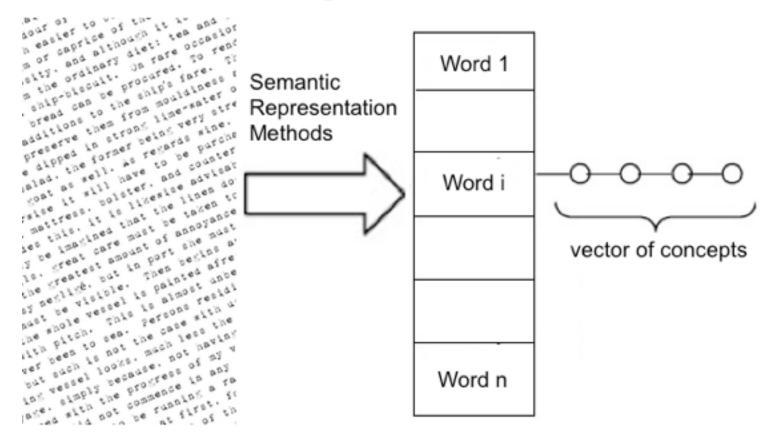
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  - Knowledge-based (*WordNet*) vs. Statistical methods







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  - Semantic Word Representation (word embedding)

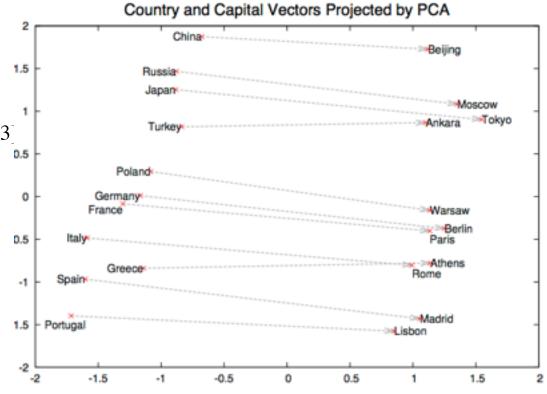




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  - Adding word vectors
- Word2Vec [Mikolov 2013]
  - Neural Networks
  - Skip-Gram model





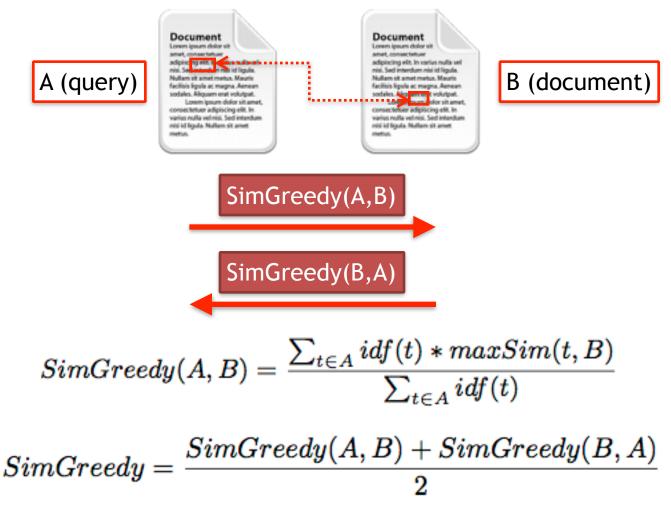
#### **Research Question**

- The use of semantic similarity in Social Image Retrieval
  - tags, title, and description of images
  - normal descriptive language
- From semantical Word-to-Word to Text-to-Text similarity
- How to be efficient?





## Semantic Similarity Method



- Refer to as SimGreedy
- Complexity: O(n\*m)



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  - evaluation metric at P@20



- MediaEval Retrieving Diverse Social Images Task
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- Experiment setup
  - Training models on Wikipedia corpora
  - Models with RI and Word2Vec representation methods
  - 200 and 600 dimensions
  - Solr as baseline



Combination of 2013 and 2014

Representation	Dimension	P@20
Random Indexing	200	†0.788
Random Indexing	600	†0.787
Word2Vec	200	†0.795
Word2Vec	600	†0.793
Solr (Baseline)		0.760



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• Only on 2014

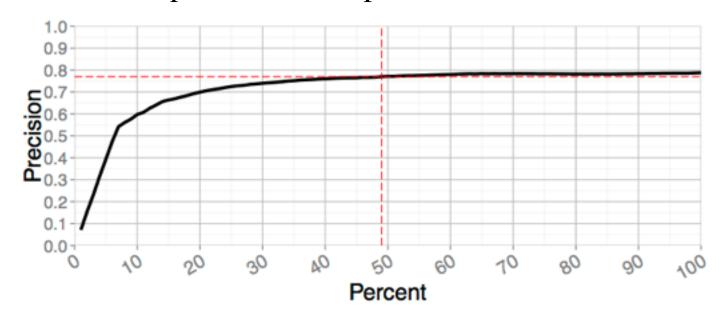
Representation	Dimension	P@20
Random Indexing	200	0.813
Random Indexing	600	0.817
Word2Vec	200	0.833
Word2Vec	600	0.842
Best text (Run1)	0.832	
Best text-visual (I	0.817	
Best all resources	0.876	



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  - combines <u>two</u> retrieval methods
  - *n* percent of the first method is re-ranked by the second one
- Solr as the first, SimGreedy the second
  - checking all the possible values: n=49
  - with same performance, optimizes to almost two times





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- Comparison
  - shorter query time, no parameter tuning

Repres.	Algorithm	Indexing Time	I/O	Query Time	Overall	P@20
W2V	SimGreedy	-	0:16	1:50	2:06	0.795
	SimGreedy + Hybrid	-		0:50	1:06	0.772
	SimGreedy + ANN-Index	0:28		0:17	1:01	0.782
RI	SimGreedy	-	0:14	2:07	2:24	0.788
	SimGreedy + Hybrid	-		1:00	1:14	0.770
	SimGreedy + ANN-Index	0:21		0:19	0:54	0.782

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#### **Conclusion**

- Platform for Concept-based Multimodal Retrieval
- Social Image Retrieval
- Semantic-based Text Retrieval
  - Two term representations: Word2Vec, Random Indexing
  - SimGreedy method
  - Semantic Similarity method more effective than termfrequency methods
- Optimization: Hybrid & ANN-Index
  - both optimized time to half
  - ANN-Index more practical and easy to setup