Using Wikipedia and Wiktionary in Domain-Specific Information Retrieval

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Ubiquitous
Knowledge
Processing

Domain-Specific Track

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- Monolingual tasks
  - English
  - German
  - Russian

- Bilingual tasks
  - English topics – German documents

Objectives

- Overcome vocabulary gap between queries and documents

- Methods
  - Query Expansion
  - (Blind) Relevance Feedback

- Our approach:
  - **Semantic Relatedness (SR)** of query and document terms

Outline

- IR models based on SR

- Preprocessing of queries and documents

- Results for monolingual runs: English, German, Russian

- Approach for bilingual IR

- Results for bilingual runs: English – German

- Additional experiments concerning efficiency

- Summary and future work

Semantic Relatedness Measure

0.9

car ⟷ drive

0.1

car ⟷ flower

SR Measures and Knowledge Bases

- Various SR measures
  - Path based, information content based, dictionary based
- Using linguistic knowledge bases like WordNet

- Our approach:
  - Using **collaborative knowledge bases** like Wikipedia and Wiktionary
    - +: coverage, domain-specific terms, up-to-dateness
    - -: quality, accessibility

  - **Concept vector based** SR measure
    - Represent terms and texts as concept vectors
    - Compare concept vectors to compute SR
    - **Explicit Semantic Analysis** (Gabrilovich & Markovitch, 2007)

Constructing Concept Vectors from Knowledge Sources

Knowledge Sources

| Concepts | Article Titles |
| Textual Representation | Article Text |

Representing Words as Concept Vectors

taxicab

0.8	automobile
0.7	drive
0.6	fast
0.8	hire
0.6	New York
0.8	passenger
0.1	SUV
0.9	taxi
0.8	transport
0.8	yellow

In some countries, taxicabs are commonly yellow. This practice began in Chicago, where taxi entrepreneur John Hertz painted his taxis yellow based on a University of Chicago study alleging that yellow is the color most easily seen at a distance.

Concept Vector Measure

taxicab

automobile	0.8
drive	0.7
fast	0.6
hire	0.8
New York	0.6
passenger	0.8
SUV	0.1
taxi	0.9
transport	0.8
yellow	0.8

cosine similarity

0.7	automobile
0.8	drive
0.2	fast
0.1	hire
0.0	New York
0.1	passenger
0.0	SUV
0.0	taxi
0.9	transport
0.1	yellow

truck

$$\overline{V}_{taxicab} \times \overline{V}_{truck} = \text{SR score}$$

New Resource for IR:
Wiktionary – Wikipedia's lexical companion

- Language
- Etymology
- Pronunciation
- Part-of-speech
- Word senses
- Synonyms
- Derived Terms
- Translations

- Abbreviations, Antonyms, Categories, Collocations, Examples, Glosses, Hypernyms, Hyponyms, Morphology, Quotations, Related terms, Troponyms

Constructing Concept Vectors from Knowledge Sources

Knowledge Sources

	WIKIPEDIA The Free Encyclopedia	Wiktionary

Concepts	Article Titles	Entry Titles
Textual Representation	Article Text	Entry Information

(Zesch et al., 2008)

Constructing Concept Vectors from Knowledge Sources

Knowledge Sources

| | Article Titles | **+** | Entry Titles |
| Concepts | | | |

| Textual Representation | Article Text | | Entry Information |

(Zesch et al., 2008)

Combination of Wikipedia and Wiktionary

taxicab

passenger	0.8
SUV	0.1
taxicab	0.9
transport	0.8
yellow	0.8
passenger	0.4
taxicab	1.0
transport	0.7
yellow	0.6

Wikipedia

Wiktionary

$$\overline{v}_{taxicab}$$

SR Based IR Models

- **SR-Text** (Gabrilovich & Markovitch, 2007)
  - Compare concept vectors of query and document

SR-Text

Query

Document

car   **ferry**

… Atlantic … transport …

concepts		car		ferry		=	Document	... +			+		+ ...

concepts

passenger	0.4	0.5	0.6
ship	0.3	0.9	0.8
transport	0.8	0.8	1.0
ocean	0.0	0.6	0.4

+ =

0.4	0.1	0.5
0.8	0.3	0.8
0.7	0.0	0.9
0.8	0.6	0.3

= ... + + + ...

cosine similarity

*   *

tf.idf

0.6   0.7

*   *

0.5   0.9

SR Based IR Models

- **SR-Text** (Gabrilovich & Markovitch, 2007)
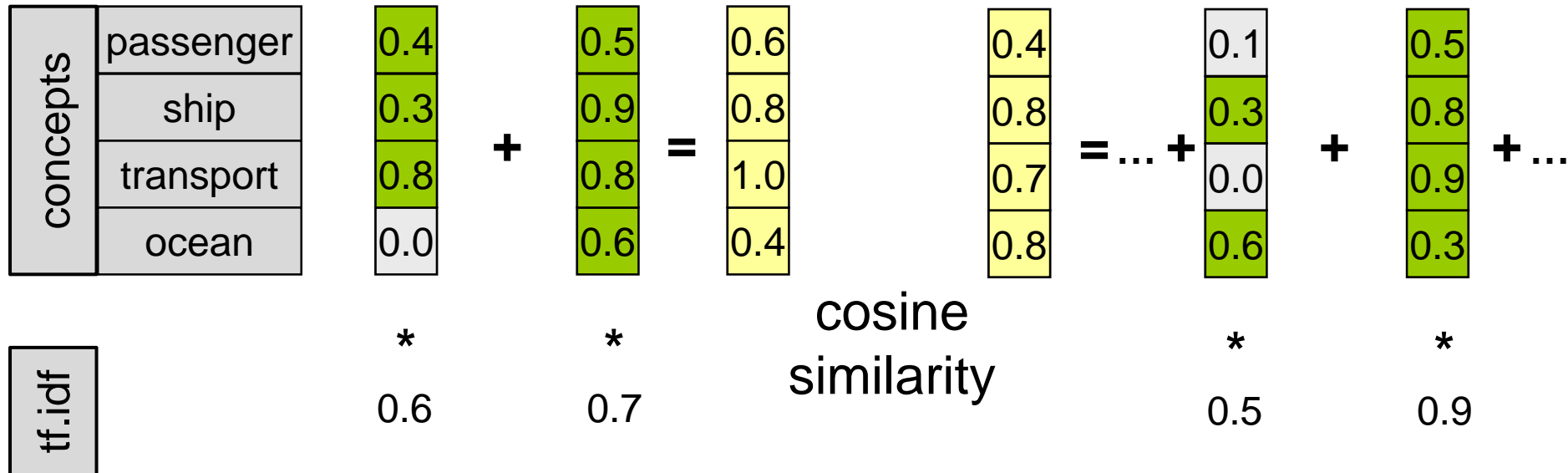 - Compare concept vectors of query and document

- **SR-Word** (Müller & Gurevych, 2006)
  - Compare concept vectors of query and document term pairs

SR-Word

Query   **Document**

	car	ferry		... Atlantic ...	transport ...
passenger	0.4	0.5		0.1	0.5
ship	0.3	0.9		0.3	0.8
transport	0.8	0.8		0.0	0.9
ocean	0.0	0.6		0.6	0.3

concepts

cosine similarity

tf.idf   0.6   0.7   0.5   0.9

score

$$0.1 * 0.6 * 0.5 + 0.8 * 0.6 * 0.9 + 0.7 * 0.7 * 0.5 + 0.8 * 0.7 * 0.9 + \ldots$$

Query

Document

car   **ferry**

… Atlantic … transport …

concepts		car	ferry		Document	
	passenger	0.4	0.5		0.1	0.5
	ship	0.3	0.9		0.3	0.8
	transport	0.8	0.8	...	0.1	1.0
	ocean	0.0	0.6		0.0	0.3

tf.idf

0.6   0.7

0.5   0.9

threshold for SR values: 0.2

score

$\boxed{0.1} * 0.6 * 0.5 + 0.8 * 0.6 * 0.9 + 0.7 * 0.7 * 0.5 + 0.8 * 0.7 * 0.9 + …$

SR-Word: Term Not Semantically Related

	Query		Document

Query

car   **ferry**

Document

… Atlantic … transport …

concepts		car	ferry		Atlantic		transport	
	passenger	0.4	0.5		0.1		0.5	
	ship	0.3	0.9		0.3		0.8	
	transport	0.8	0.8	…	0.1	…	1.0	…
	ocean	0.0	0.6		0.0		0.3	

tf.idf

0.6   0.7

0.5   0.9

score

sr(car, Atlantic) < threshold   sr(car, transport) < threshold   …

decrease

SR-Word: No Match on String-Level

Query

Document

car   ferry

~~car~~ … Atlantic … transport …

concepts	car	ferry			
passenger	0.4	0.5	0.1		0.5
ship	0.3	0.9	0.3		0.8
transport	0.8	0.8	0.1		1.0
ocean	0.0	0.6	0.0		0.3

tf.idf   0.6   0.7   0.5   0.9

score

query term not contained in document

decrease

Retrieval Models

- **Semantic Relatedness**
  - SR-Text (SRT)
  - SR-Word (SRW)

- **Baseline: Vector Space Model**
  - Apache Lucene (LUC)

- **Combination using CombSUM (Fox & Shaw, 1994)**
  - For each document the similarity scores of the models are normalized and added up

Outline

- IR models based on SR

- **Preprocessing of queries and documents**

- Results for monolingual runs: English, German, Russian

- Approach for bilingual IR

- Results for bilingual runs: English – German

- Additional experiments concerning efficiency
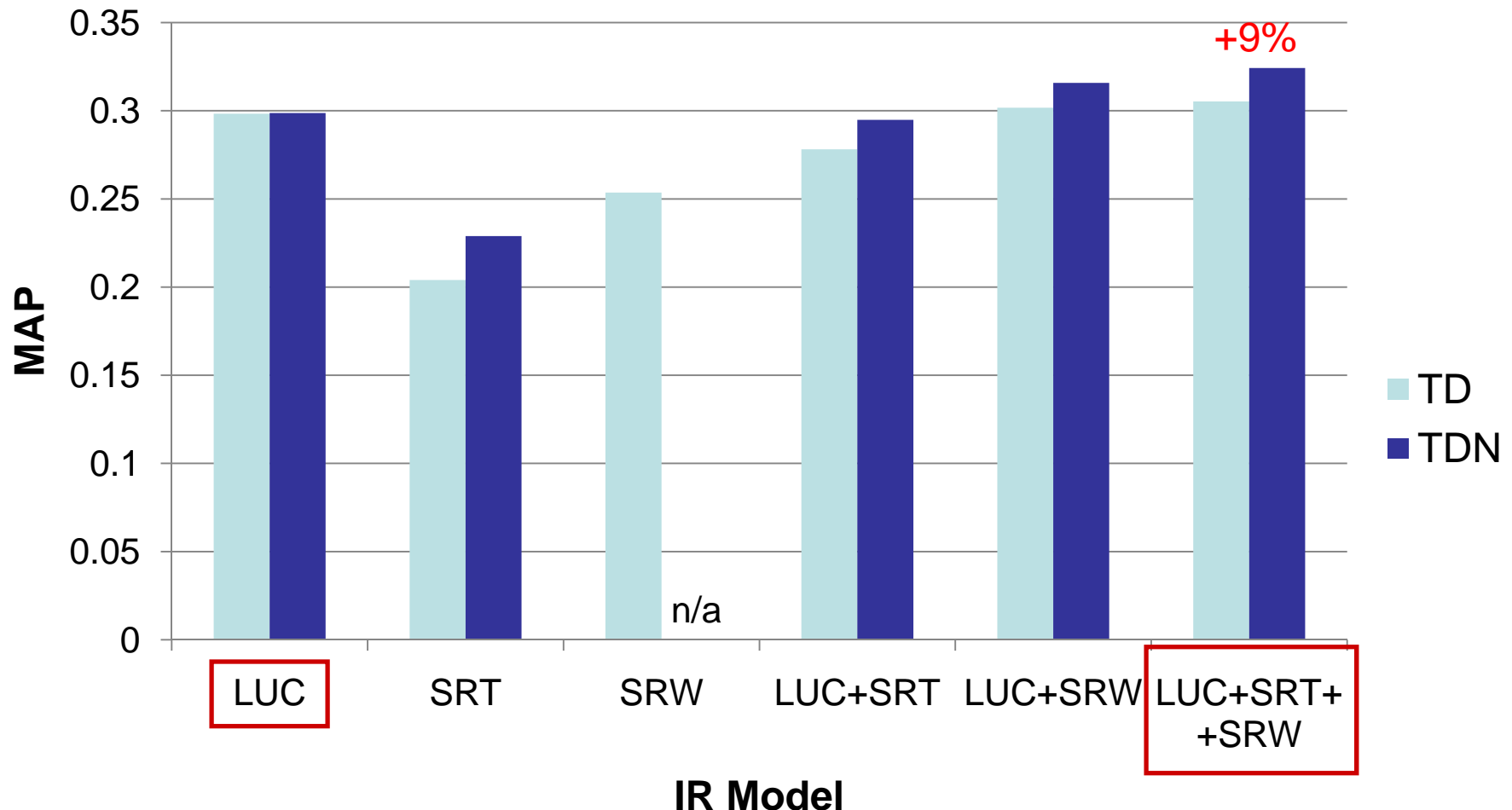
- Summary and future work

Preprocessing

- Lemmatization
  - Using the probabilistic part-of-speech tagging system TreeTagger (Schmid, 1994)
  - English, German, Russian

- Decompounding (Langer, 1998)
  - German
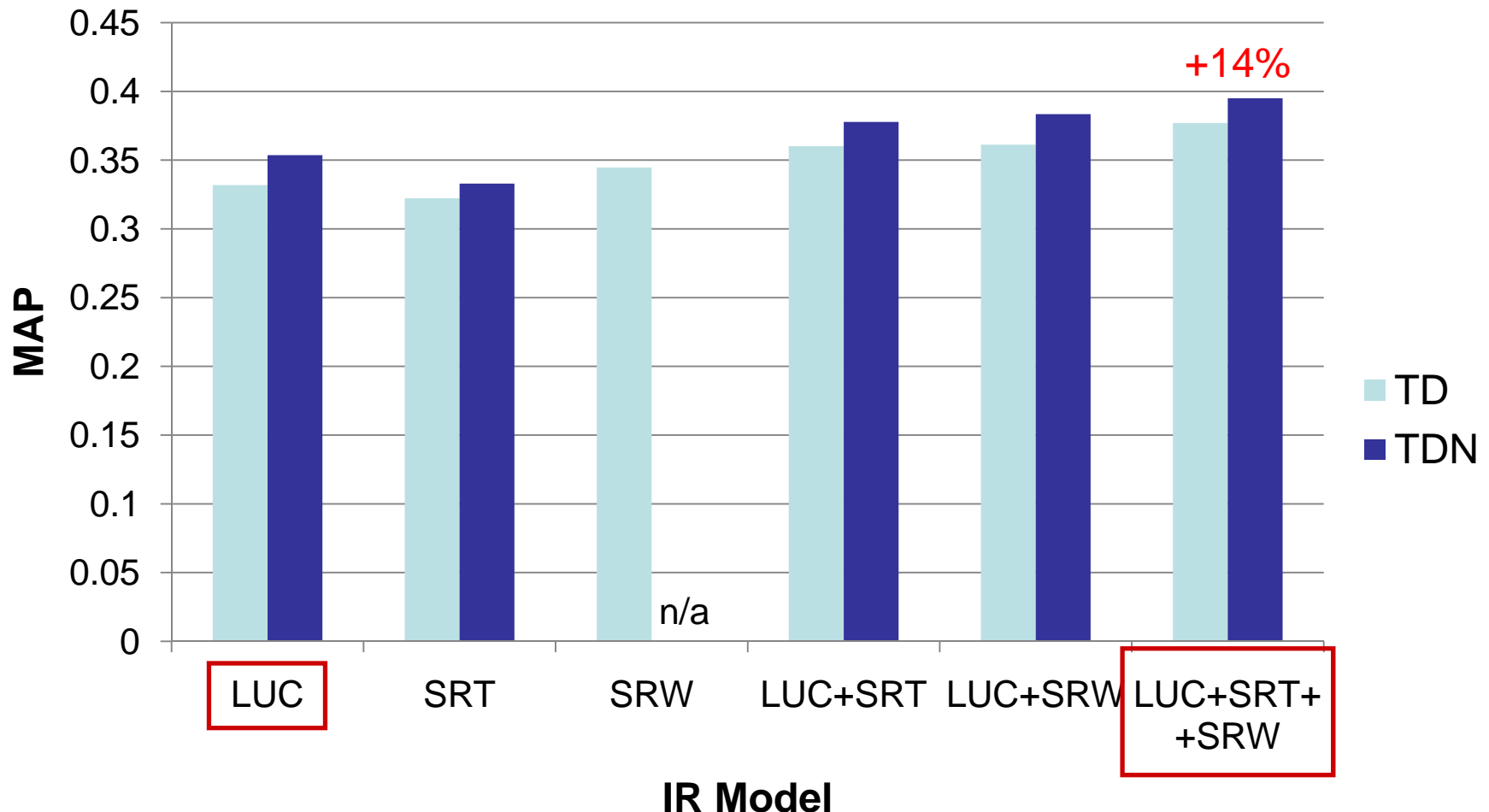 - Using compound words and their elements

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Monolingual Results: English
IR Models and Query Types

Monolingual Results: German IR Models and Query Types

Monolingual Results: Russian
IR Models and Query Types

Outline

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- Results for monolingual runs: English, German, Russian

- **Approach for bilingual IR**

- Results for bilingual runs: English – German

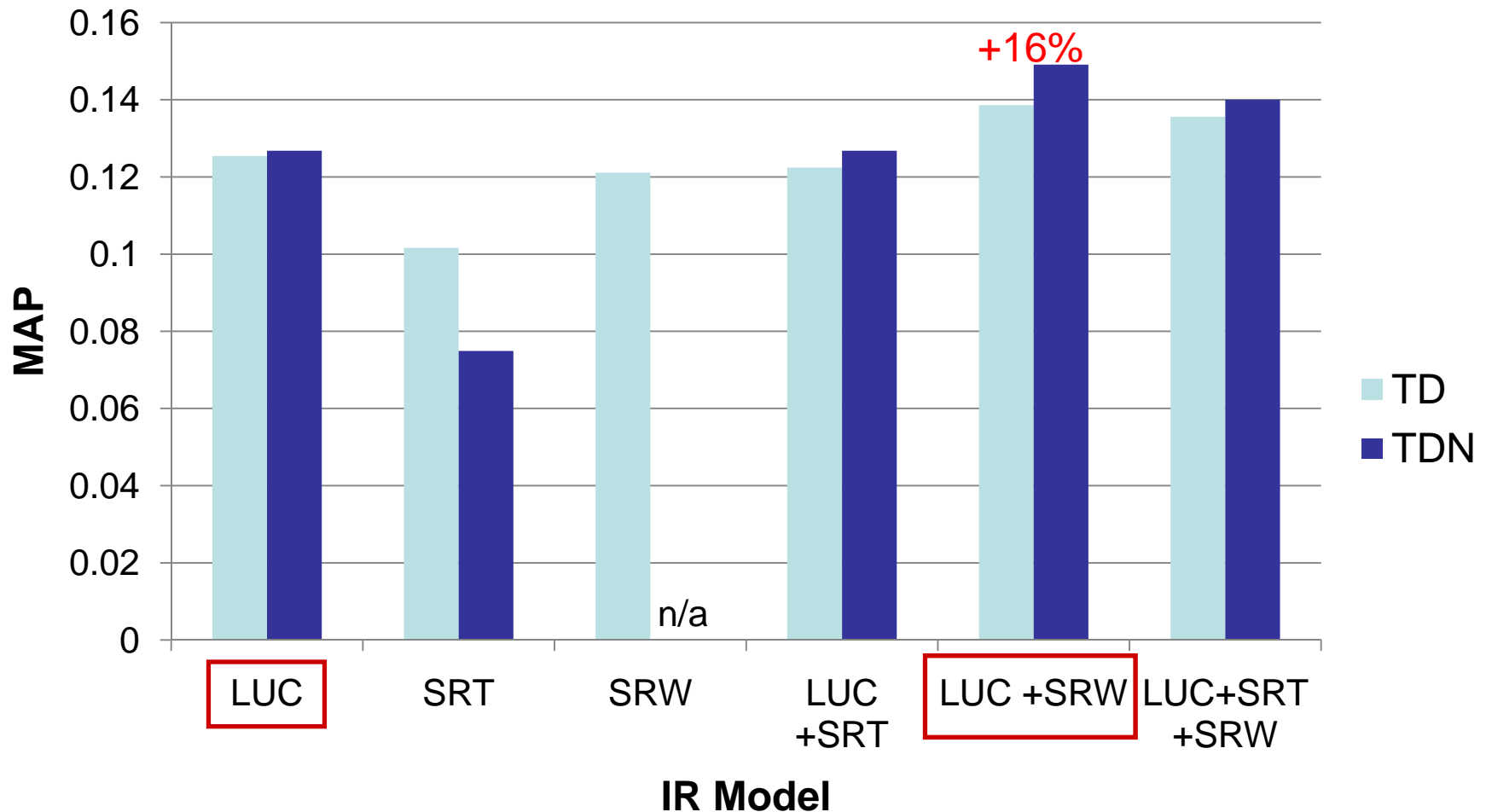- Additional experiments concerning efficiency

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Crosslingual Approach

- Machine translation using **Systran**

- SR-Text: **Cross-language links** in Wikipedia

English	English Wikipedia			German Wikipedia			German Wikipedia		German

Query →

English Wikipedia	
passenger	0.8
SUV	0.1
taxicab	0.9
transport	0.8
yellow	0.8

German Wikipedia	
Passagier	0.8
SUV	0.1
Taxi	0.9
Verkehr	0.8
Gelb	0.8

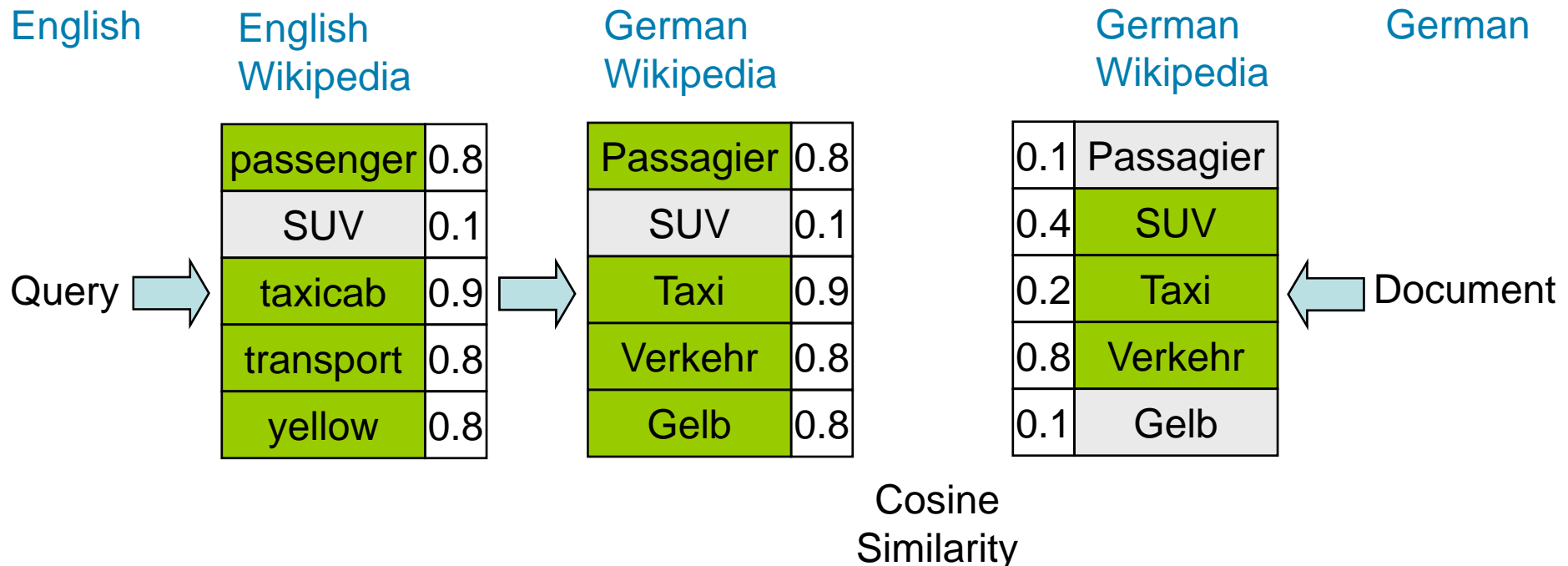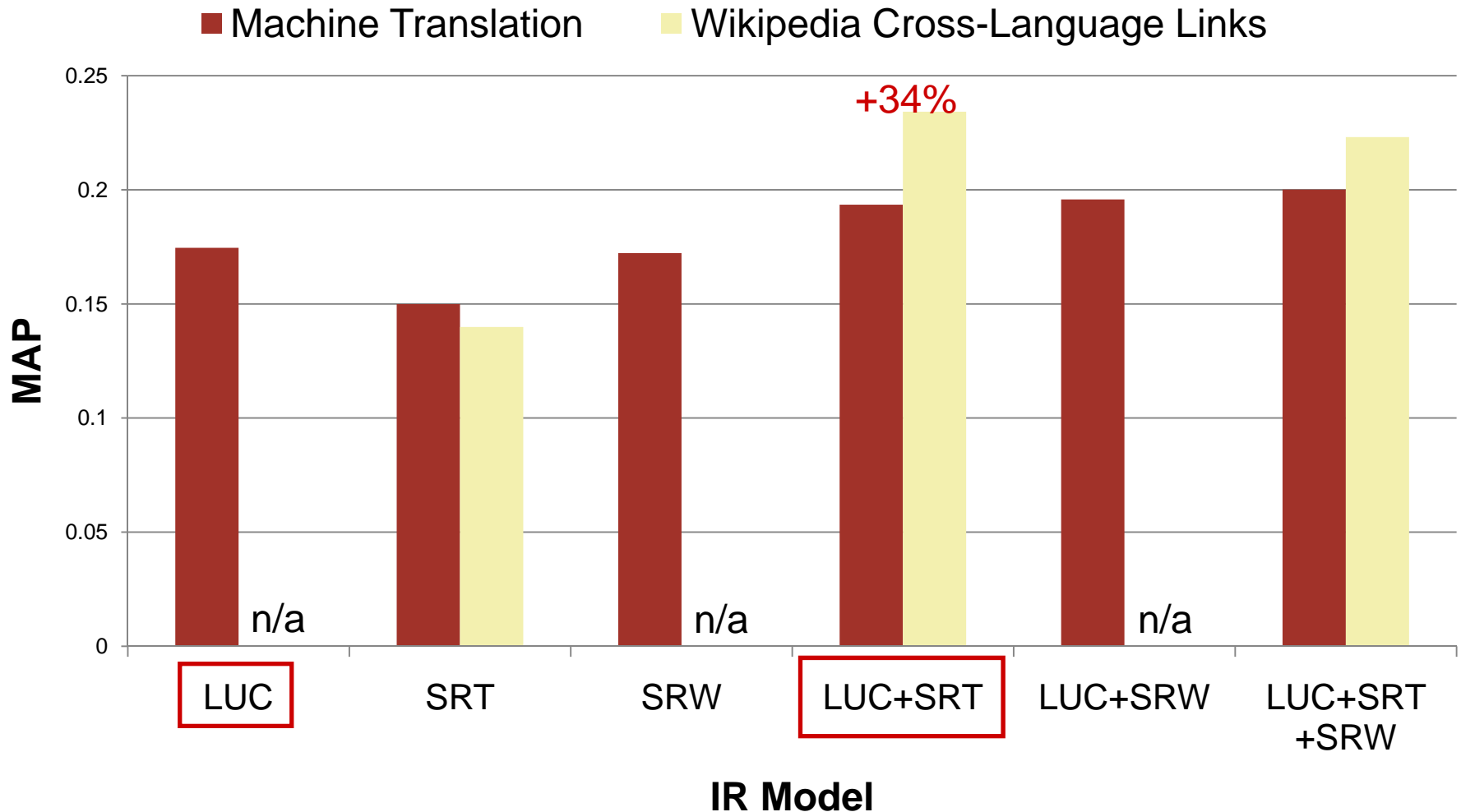|	German Wikipedia
0.1	Passagier
0.4	SUV
0.2	Taxi
0.8	Verkehr
0.1	Gelb

← Document

Cosine Similarity

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- **Results for bilingual runs: English – German**

- Additional experiments concerning efficiency
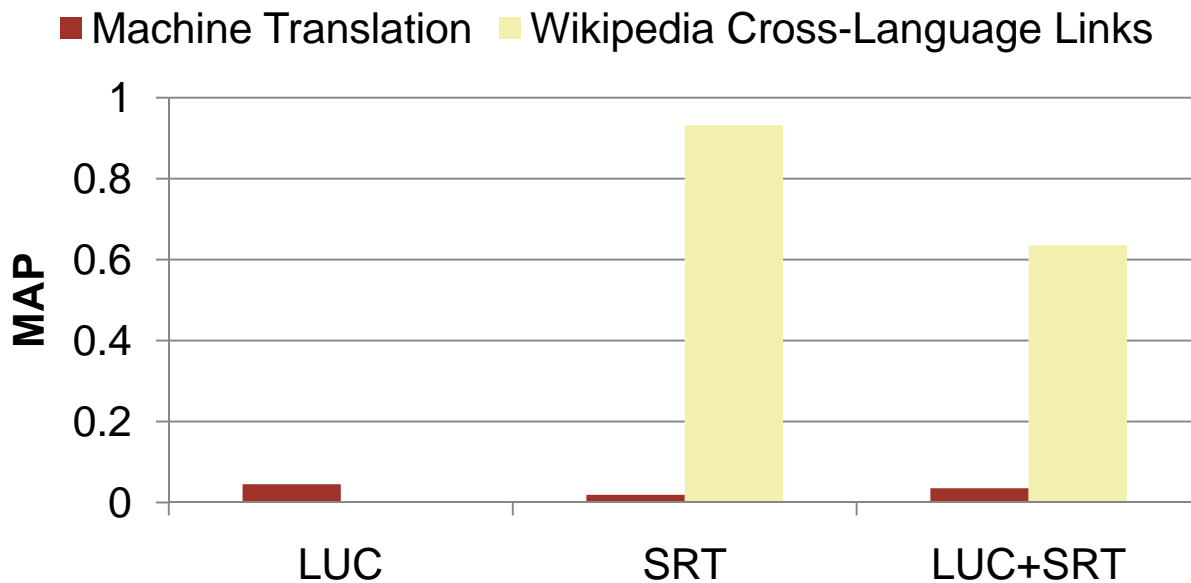
- Summary and future work

Bilingual Runs: English – German
IR Models and Translation Approach

■ Machine Translation   ■ Wikipedia Cross-Language Links

+34%

MAP

IR Model

LUC · SRT · SRW · LUC+SRT · LUC+SRW · LUC+SRT+SRW

n/a

Example Topic

- Topic No. 209
- English title
  - *Doping and sports*
- *German title translated by Systran*
  - *Lackieren und Sport* (engl.: *painting/lacquer and sports*)

■ Machine Translation   ■ Wikipedia Cross-Language Links

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- **Additional experiments concerning efficiency**

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Additional Experiments

- SR based models are computationally expensive

- Reducing dimensions of concept vectors in SR-Text
  - Experiments on CLEF data from previous years
  - Using only the highest 20,000 – 25,000 concepts for the concept vectors of queries and documents yields same MAP values

- Only ranking the documents retrieved by Lucene
  - Experiments on monolingual English data from this year
  - Only a fraction of query – document comparisons necessary
  - No decrease of MAP
    - SR based models might not retrieve additional relevant documents
    - But they help to rank relevant documents higher

Summary

- Introduction of **Wiktionary as new resource** for IR
- **Combination of Wikipedia and Wiktionary** as knowledge base for two SR based IR models: SR-Text and SR-Word
- **Combination of SR based models with Lucene** using CombSUM
- Monolingual:
  - Improvement of combination of models compared to baseline:
    - **English: 9%, German: 14%, Russian: 16%**
  - **SR-Word outperforms SR-Text**
- Bilingual:
  - Machine translation
  - SR-Text: **cross-language links** in Wikipedia
    - Improvement by combining Lucene and SR-Text:
      - **English – German: 34%**
- **Efficiency** of SR based models **can be improved**

Future Work

- Combination of SR based models with other IR models (Okapi BM25, …)

- Different combination methods for models and knowledge bases

- Using cross-language links in Wiktionary for cross-lingual IR

Acknowledgments

Ubiquitous Knowledge Processing Lab

http://www.ukp.tu-darmstadt.de

Wikipedia & Wiktionary API

http://www.ukp.tu-darmstadt.de/software