

Keyphrase Annotation with Graph Co-Ranking

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How to infer the main content of a domain-specific document?

Keyphrases

- Single- or multi-word expressions
- Important topics/concepts
- Useful to multiple Information Retrieval tasks:
 - ▶ Document indexing
 - ▶ Text summarization
 - ▶ Query expansion
 - ▶ etc.

*Toucher : le tango des sens. Problèmes de **sémantique lexicale***

*À partir d'une hypothèse sur la sémantique de l'unité lexicale 'toucher' formulée en termes de forme schématique, cette étude vise à rendre compte de la **variation sémantique** manifestée par les emplois de ce **verbe** dans la construction **transitive** directe 'C0 toucher C1'. Notre étude cherche donc à articuler variation sémantique et invariance fonctionnelle. Cet article concerne essentiellement le mode de variation co-textuelle : en conséquence, elle ne constitue qu'une première étape dans la compréhension de la construction des valeurs référentielles que permet 'toucher'. Une étude minutieuse de nombreux exemples nous a permis de dégager des constantes impératives sous la forme des 4 notions suivantes : sous-détermination sémantique, contact, anormalité, et contingence. Nous avons tenté de montrer comment ces notions interprétatives sont directement dérivables de la forme schématique proposée.*

Reference keyphrases (French):

*Français; modélisation; analyse distributionnelle; interprétation sémantique; **variation sémantique**; **transitif**; **verbe**; syntaxe; **sémantique lexicale**.*





Reference keyphrases (English):

French; modelling; distributional analysis; semantic interpretation; **semantic variation**; **transitive**; **verb**; syntax; **lexical semantics**.

- Silence
- Domain consistency
- Free syntax (*e.g. syntax, semantic variation, transitive, etc.*)
- Risks of semantic redundancy (over generation)

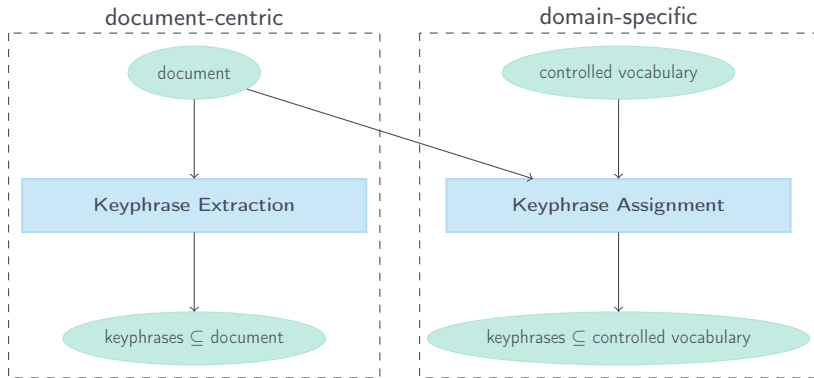
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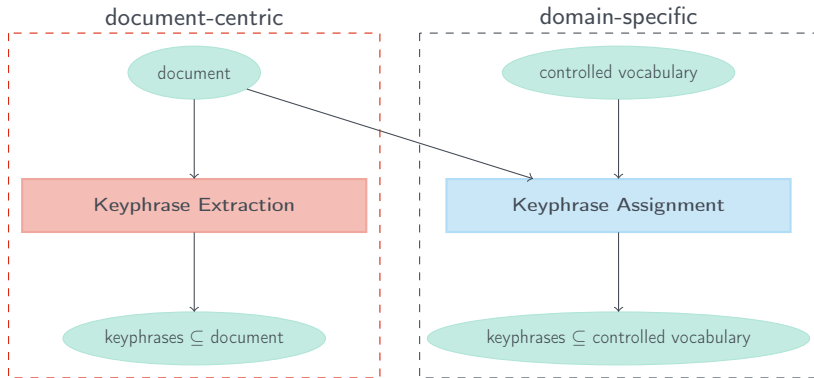


- **Silence** 
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- Free syntax (*e.g. syntax, semantic variation, transitive, etc.*) *
- Risks of semantic redundancy/over generation *

* (Bougouin et al., 2013, TopicRank)

1. Related Work
2. Proposal
3. TopicCoRank
4. Evaluation
5. Conclusion





Supervised keyphrase extraction

- Relying on reference keyphrases
- Learning to determine keyphrase likelihood
- Mixing statistical and linguistic features

Unsupervised keyphrase extraction

- Looking for the most important keyphrase candidates
- Using mainly statistics
- Linking keyphrase candidates to each other

Graph-based approach detecting documents most important topics and extracting keyphrases from these topics

- | | | |
|---|--|---------------------------|
| 1 | Keyphrase candidate selection | <i>/(NOUN ADJ)+/</i> |
| 2 | Keyphrase candidates topical clustering | <i>lexical clustering</i> |
| 3 | Topic graph construction | <i>complete graph</i> |
| 4 | Graph-based topic ranking | <i>Google's PageRank</i> |
| 5 | Keyphrase extraction from the important topics | <i>one per topic</i> |

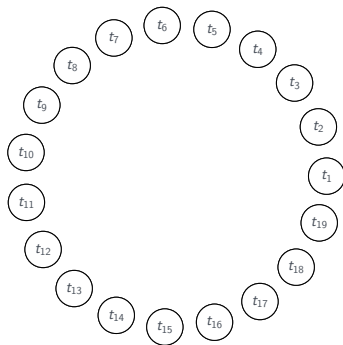
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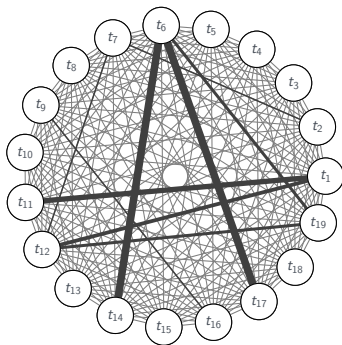
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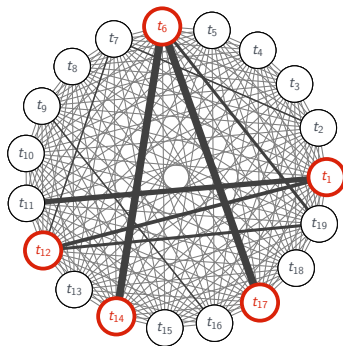
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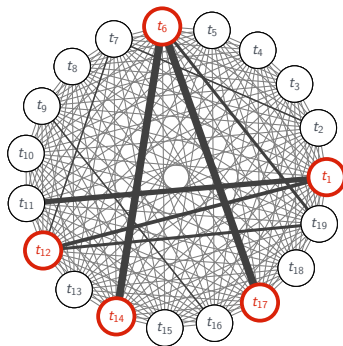
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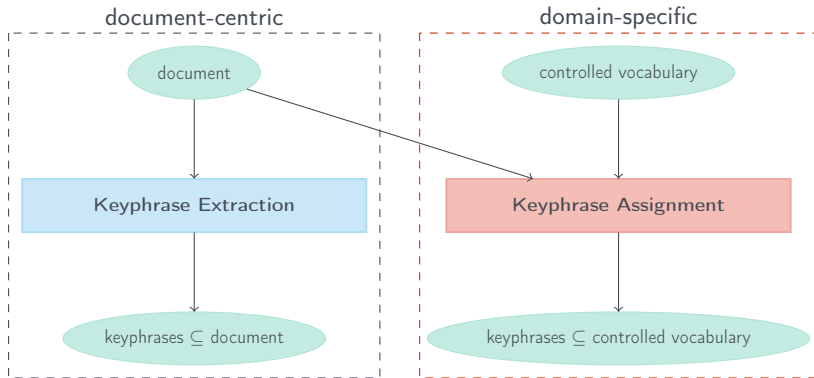
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TopicRank 10 keyphrases (French):

Sémantique lexicale; variation sémantique; problèmes; étude; forme schématique; sens; tango; invariance fonctionnelle; construction transitive directe; article;



- Relying on a domain-specific controlled vocabulary (thesaurus)
- Aiming for consistency across domain

Handful of attempts

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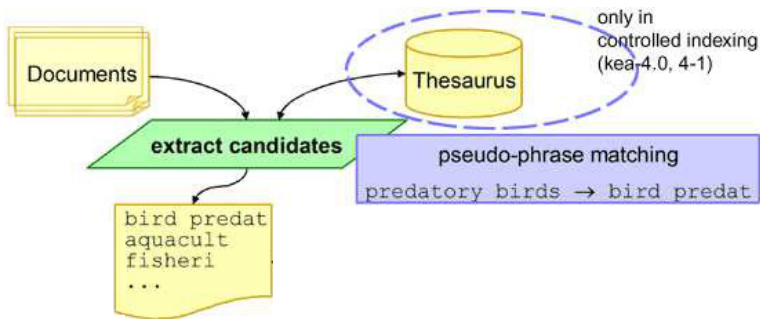


Figure from Medelyan and Witten (2006)

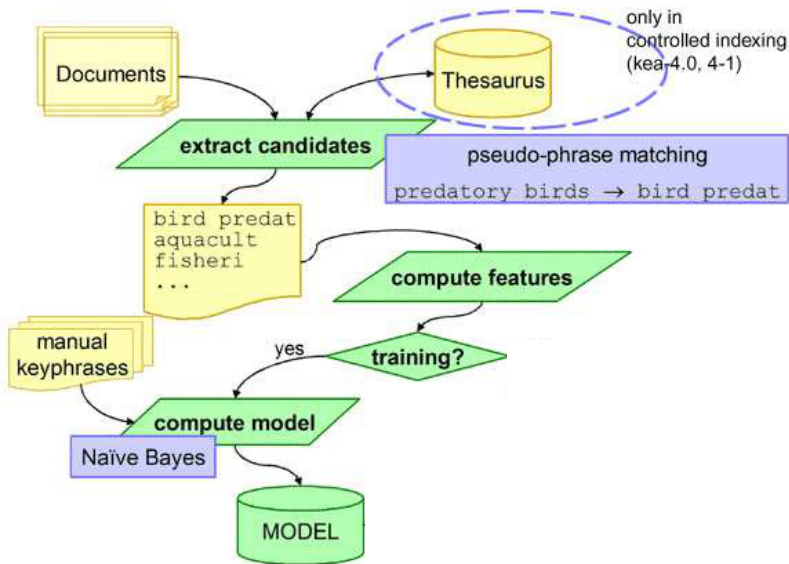


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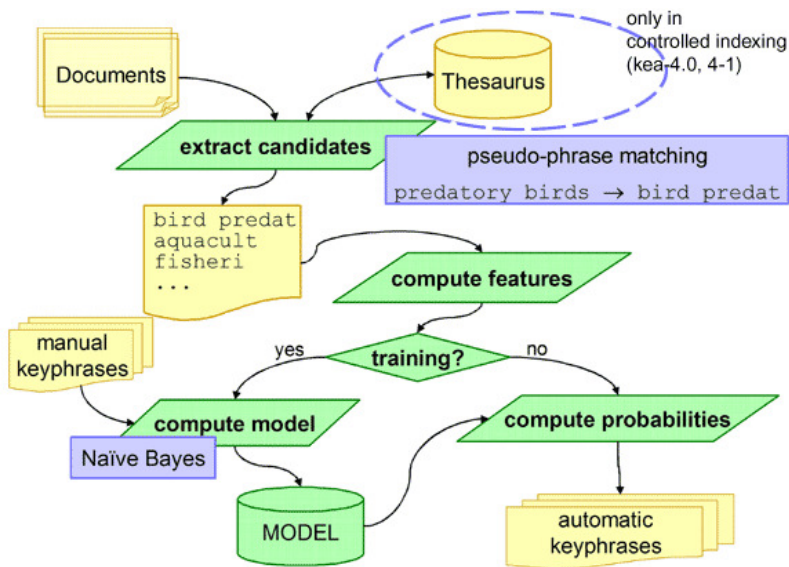


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Kea++ 10 keyphrases (French):

Toucher; *variation sémantique*; *sémantique*; *tangoa*; *direction*; *formant*; *sémantique lexicale*; *impératif*; *invariant sémantique*; *transitif*;

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Leveraging document content	<u>Vs.</u>	Ignoring inner-document relations
Ignoring domain vocabulary	<u>Vs.</u>	Mapping document to its domain
Limited to document content	<u>Vs.</u>	Limited to vocabulary coverage

Professional indexer point of view

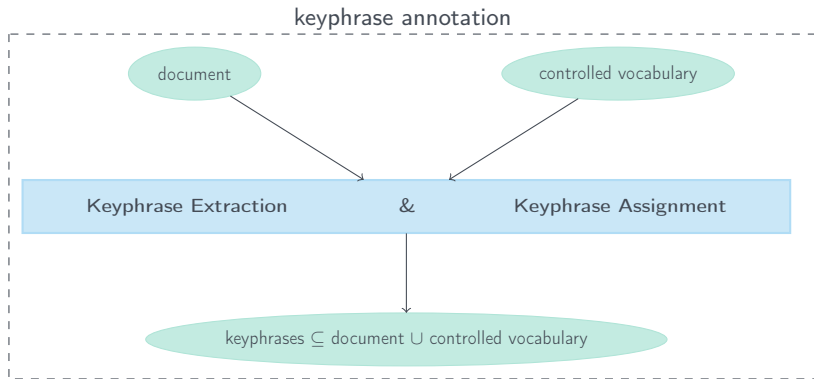
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Supervised extension of TopicRank to add assignment capabilities alongside extraction

Hypothesis

- The domain supplements the document
- ⇒ improves keyphrase extraction
- The training documents represent the domain
- ⇒ keyphrases circumvents use of controlled vocabulary

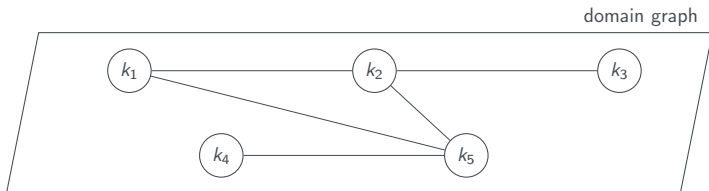
Undirected graph of keyphrases annotated to training documents

- Each keyphrase is represented by a vertex
- Keyphrases of the same documents are connected
- Edges are weighted by the number of documents containing both keyphrases



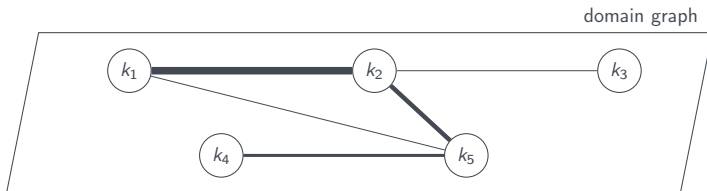
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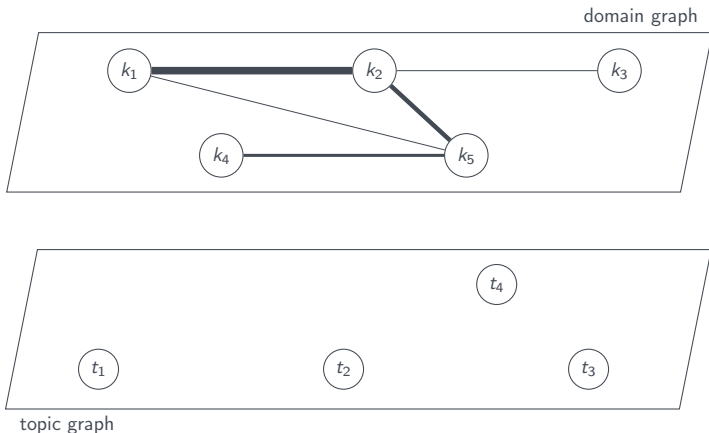
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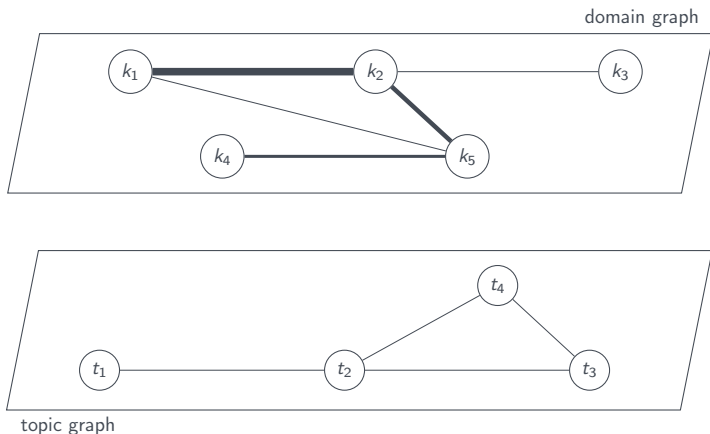
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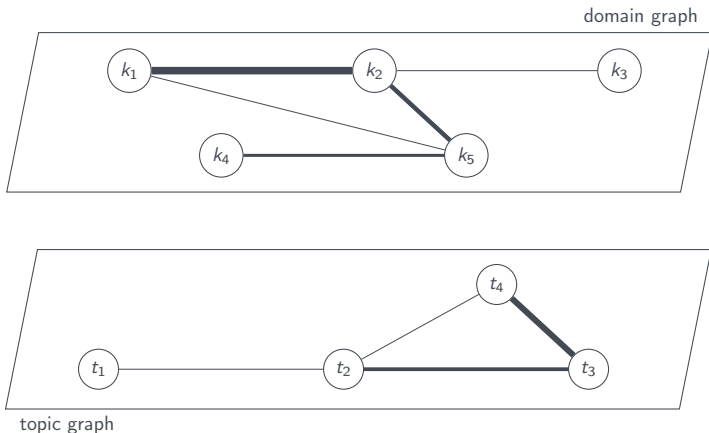
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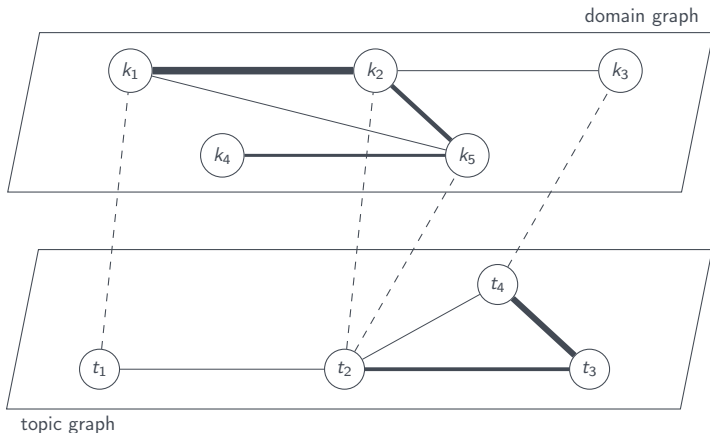
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- Domain keyphrases k_i are as much important as they are strongly connected to as much other important keyphrases k_j as possible

$$R_{in}(k_i) = \sum_{k_j \in E_{in}(k_i)} \frac{w_{ij} S(k_j)}{\sum_{k_k \in E_{in}(k_j)} w_{jk}}$$

- Document topics t_i are as much important as they are strongly connected to as much other important topics t_j as possible

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- Domain keyphrases and document topics v_i gain importance from each other

$$R_{out}(v_i) = \sum_{v_j \in E_{out}(v_i)} \frac{S(v_j)}{|E_{out}(v_j)|}$$

- Both inner- and outer- recommendation are combined with empirically tuned damping factors

$$S(k_i) = (1 - \lambda_k) R_{out}(k_i) + \lambda_k R_{in}(k_i)$$

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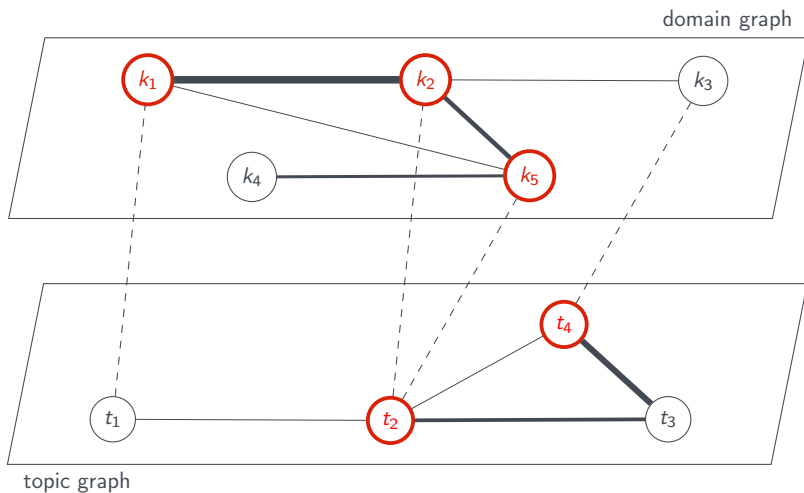
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TopicCoRank 10 keyphrases (French):

Sémantique lexicale; sémantique; verbe; variation sémantique; français; hypothèse; syntaxe; pragmatique; interprétation sémantique; analyse distributionnelle;

Benefits

- + Combines extraction and assignment: ↗ recall & ↗ precision
- + Circumvents the need of a controlled vocabulary

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3 French corpora covering 3 domains of Humanities and Social Sciences

- Manually annotated by professional indexers
- Provided with controlled vocabularies
- Annotated based on both controlled vocabulary and content

Corpus	Linguistics			Information Science			Archaeology		
	train	dev	test	train	dev	test	train	dev	test
Doc.	515	100	200	506	100	200	518	100	200
Tokens/Doc.	161	151	147	105	152	157	221	201	214
Keyphrases	8.6	8.8	8.9	7.8	10.0	10.2	16.9	16.4	15.6
Silence (%)	60.6	63.2	62.8	67.9	63.1	66.9	37.0	48.4	37.4

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Related work:

- TopicRank
- Kea++

Extraction

Assignment

TopicCoRank alternatives:

- TopicCoRank_{extr}
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Stem-based comparisons at the top 10 outputted keyphrases

- Recall
- Precision
- F1-score

Method	Linguistics			Information Science			Archaeology		
	P	R	F	P	R	F	P	R	F
TopicRank	11.82	13.1	11.9	12.1	12.8	12.1	27.5	19.7	21.8
KEA++	11.6	13.0	12.1	9.5	10.2	9.6	23.5	16.2	18.8
TopicCoRank	24.5	28.3	25.9	19.4	19.6	19.0	46.6	31.4	36.7
TopicCoRank _{extr}	15.9	18.2	16.7	15.9	16.2	15.6	39.6	26.4	31.0
TopicCoRank _{assign}	25.8	29.6	27.2	19.9	20.0	19.5	49.6	33.3	39.0

- TopicCoRank outperforms baselines
- Graph-based co-ranking is succesful for extraction alone
- Graph-based co-ranking is succesful for assignment alone
- TopicCoRank_{assign} performs best due to datasets specificities

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Method	Linguistics			Information Science			Archaeology		
	P	R	F	P	R	F	P	R	F
TopicRank	11.82	13.1	11.9	12.1	12.8	12.1	27.5	19.7	21.8
KEA++	11.6	13.0	12.1	9.5	10.2	9.6	23.5	16.2	18.8
TopicCoRank	24.5	28.3	25.9	19.4	19.6	19.0	46.6	31.4	36.7
TopicCoRank _{extr}	15.9	18.2	16.7	15.9	16.2	15.6	39.6	26.4	31.0
TopicCoRank _{assign}	25.8	29.6	27.2	19.9	20.0	19.5	49.6	33.3	39.0

- TopicCoRank outperforms baselines
- Graph-based co-ranking is succesful for extraction alone
- Graph-based co-ranking is succesful for assignment alone
- TopicCoRank_{assign} performs best due to datasets specificities

1. Related Work
2. Proposal
3. TopicCoRank
4. Evaluation
5. Conclusion

TopicCoRank:

- Supervised extension of TopicRank
- Combination of keyphrase extraction and assignment in a mutual reinforcing manner
- Good performances overall

Future work:

- Apply the supervised extension to other range of graph-based methods
- Investigate application/adaptation to non domain-specific documents
- Investigate impact on terminology/controlled vocabulary maintenance

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