

• In this talk, I report <u>experiments</u> we conducted on the task of <u>translating lists of keyphrases</u>.

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Translating lists of keyphrases

ABSTRACT Many digital libra

ABSTRACT Many digital libraries are comprised of documents from disparate sources that are independent of the rest of the collection in which they reside. A user's ability to explore is severely curralled whom other, related, documents, or even to tell if such documents exist. We describe a method for automatically introducing topic-based links into documents to support torbwing in digital libraries. Automatic keyphrase-based similarity measures are used to select and rank destinations. Two implementations are described in that applications using standard HTML, and one that uses a wheter range of miefacte techniques to provide more sophisticated linking capabilities. An evaluation throw that keyphrase-based digital compatibilities. An evaluation throw that keyphrase-based digital compatibilities. An evaluation throw that keyphrase-based limiting to conserve word as well as a popular full-late transient of full-formation finding relevant documents.

Keywords: automated hypertext generation, keyphrase extraction, information retrieval, information exploration

This problem is regularly encountered by the New Zealand Digital Library (NZDL, http://www.mzdl.org) [18]. Users cannot navigate between documents that address similar topics because the collections have no evident structure, and lack explicit relationships between their constituent parts. Links to support marigation must therefore be introduced by other means.

navigation must therefore be introduced by other means. This can be done manually, by awking human experts to ident similar documents and introduce links between them. There two problems with this approach, it is time-consuming, so quid becomes impracticable as the number of documents increase [11], and peeple are inconsistent in their selection of link anch and destinations, reducing the coherency of the result hypertext. Semi-automated (or supervised) techniques has process larger numbers of documents, but uliminately suffer the same problems [4]. A third approach, fully autoor the same problems [4]. A third approach, fully autoor destination of the set of the set of the set of the set of the same problems [4]. A third approach fully autoor destination of the set of th Clusty Web+ News Images Shopping Wikipedia Blogs Jobs Custo

In this paper we describe two systems, Kniles and Phras-automatically generate links to support browsing in

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Translating lists of keyphrases

- Related to: •
 - Sub-sentence-level translation (noun phrases, terminology, ...)
 - Query translation (cross-lingual IR); —
 - Cross-lingual summarization (Summary as a list of Keyphrases)



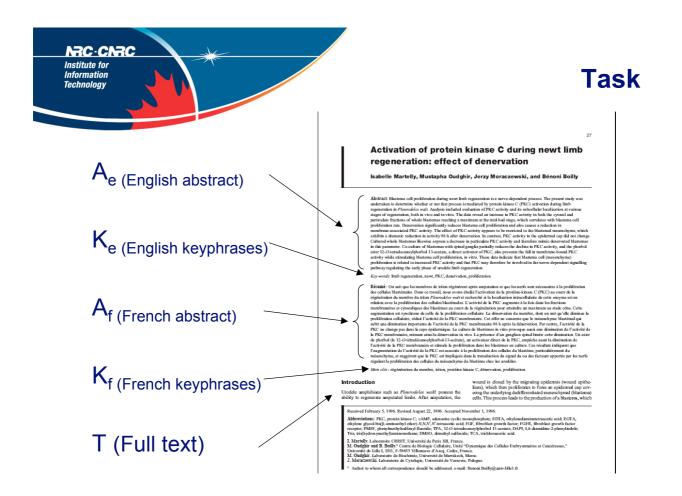
Translating lists of keyphrases

- BIKE (Bilingual Keyphrase Experiments):
 - − Bidirectional French \leftarrow → English keyphrase translation
 - Strategy pieces:
 - Statistical MT
 - Terminological resource
 - Inflectional morphology
 - WSD (list of keyphrases as a context)
 - Experiments in training and combining pieces.



Task

- Collection of 3058 scientific papers from the Canada Institute for Scientific and Technical Information (CISTI) in ten domains (biochemistry, botany, chemistry, civil engineering, environment (x2), genomics, geotechnical, microbiology and pharmacology)
- Each document is a tuple {A_f, A_e, K_f, K_e, T}





Task (performance metric)

- English \rightarrow French
- For each keyphrase, 1 candidate translation is produced
- Exact translation required (reproductibility, lowerbound exp.)
- Accuracy: ratio of correctly translated keyphrases to the total number of keyphrases
- Results are reported for all domains (30% held-out split)
- Significance testing done using 10-domain split



• Baseline (no translation)



Experiment 2

Statistical MT:

- Phrase-based statistical MT [Koehn et al., 2003]
- We calculate the conditional Fr-given-En probabilities, and retain <u>only the most probable translation</u>
- Hansard Model: Canadian Hansard parallel corpora
- CISTI Model: 40% training split (abstracts and keyphrases)
 - CISTI « global »: all journals
 - CISTI « individual » one model per journal



- How to improve statistical MT translation?
- Problem:

English	French
fiber	de fibre
population sizes	des dimensions des populations

• Remove prefixes ("de la ", "le ", "la ", "les ", "l", "du ", "de ", ",") and suffixes (",", " de", " du", " des") from French translation proposed by the MT system.



Experiment 1 & 2 Results

	Accuracy (%)	
baseline experiment	20.21	
hansard	26.11	
cisti individual	36.50	
cisti global	39.54	
cisti global + correction	41.26	P^{\intercal}

 \dagger statistically significant at the 95% level

 Terminological resource: the Grand Dictionnaire Terminologique (GDT)

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mation

- Try « exact match »
 translation
- Use the first entry





Experiment 3

- **Important problem**: lot of author keyphrases are plural while the ressource contains only singular terms
- We handled limited inflectional morphology:
 - Detection of English plural (e.g.: word ending [^f]ves)
 - Singularization of English term (e.g.: ves \rightarrow f)
 - Pluralization of French term (e.g.: al \rightarrow aux)
 - Simple heuristics for multi-word expressions
 ex.: pomme de terre → pomme<u>s</u> de terre
 pomme de terre frite → pomme<u>s</u> de terre frite<u>s</u>



- How to choose a candidate in a terminological resource?
- Problem: different domain, different translation
 - e.g: marché = market (finance) marché = market place (commerce) marché = contract (law)
- We introduce the Minimal Domain Set (MDS) algorithm
- **Idea**: find the minimal set of *coherent* domains covering all keyphrases.



Experiment 3

- 1. Calculate the frequency of each domain F(D)
- 2. Calculate the number of domain per keyphrase $|D_{\kappa i}|$
- 3. Sort Keyphrases in ascending order of $|D_{Ki}|$

For each keyphrase:

- **4.** Likeliness: From the list of domains D_{Ki}, build a reduced list containing only the domains with the highest frequency F(D)
- **5. Coherence**: From this reduced list, select the domain which has the highest coherence with a member of MDS. Add this domain to MDS.



bar (metallurgy, textile, law, automotive, ...30 other!)
iron (metallurgy, shoe)
cement (shoe, textile)
rubber (shoe, rubber, leasure, graphic)
lace (clothing, shoe, brewing)

(Step 1) F(shoe)	= 4	(Step 2) D _{iron}	= 2
F(metallurgy)	= 2	D _{cement}	= 2
F(textile)	= 2	D _{lace}	= 3
F(other)	= 1	D _{rubber}	= 4
		D _{bar}	= 34

(Step 3) L := iron, cement, lace, rubber, bar.





Experiment 3 Results

	Accuracy (%)	
GDT	35.66	_ +
GDT + morphology	38.30	_₹'.
GDT + morph + MDS	39.22	

 \dagger statistically significant at the 99% level

‡ statistically significant at the 95% level



Experiment 4

Combining all previous modules:

- Cascading modules
 - For instance:





- Combine modules (And optimize by genetic search)
- Use all candidates from phrase tables and GDT
- We used a weighting model involving six features. Features modify the original probabilities.
 - [1.071] candidate proposed by MT cisti model;
 - [0.227] candidate proposed by MT hansard model;
 - [0.477] candidate proposed by GDT;
 - [1.464] candidate proposed by more than one source;
 - [0.853] candidate using morphology;
 - [1.257] GDT candidate(s) in MDS;



Experiment 4 results

	Accuracy (%)
Module cascade	43.83
Best weighted model	47.16

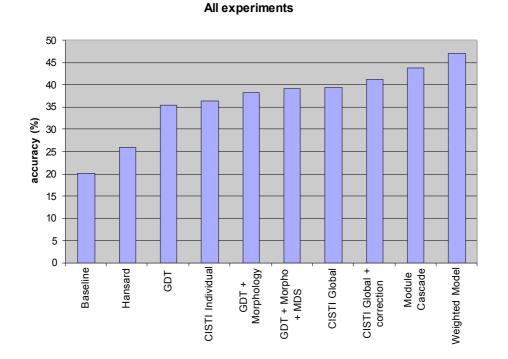


Conclusion

- Overall best solution found by genetic search and 6 features
- Statistical MT and terminological resource are complementary
- We show how to use the keyphrase list to « disambiguate the sense » of a keyphrase



Conclusion





Future Work

Two promising avenues:

- Include Termium, another Fr-En terminological ressource
- Follow (Jayaraman and Lavie, 2005) idea to combine module outputs and *create* candidate translations

