

# Exploiting Information Needs and Bibliographics for Polyrepresentative Document Clustering

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# Outline

- 1 Introduction
- 2 Polyrepresentation and Clustering
- 3 Evaluation
- 4 Conclusion



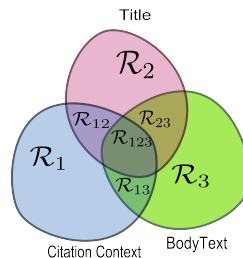
# Introduction

- Principle of Polyrepresentation in IIR
- Representations of information need and information object
- Helps to minimize the gap between user's space and information space
- Document Clustering



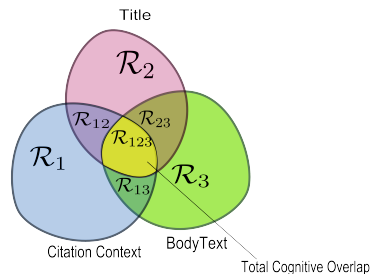
# Polyrepresentation and Clustering

- Mapping of clusters to polyrepresentation
- Search strategy:
  1. User investigates total cognitive overlap cluster
  2. User jumps to different cluster based on preferences



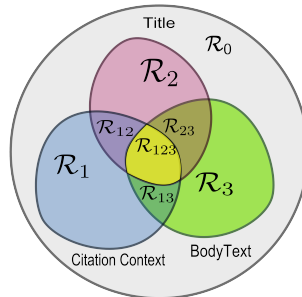
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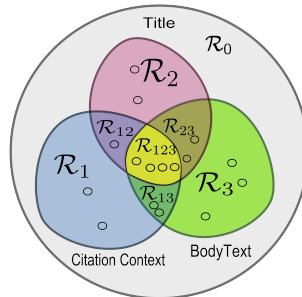
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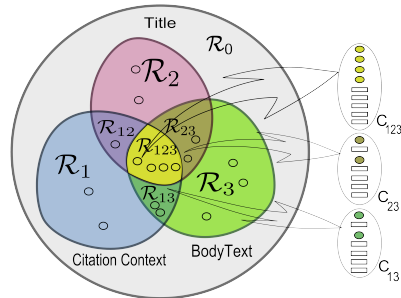
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# Simulated User and Cluster-based Ranking

- Rough simulation of search strategy
- Creates a ranking that we evaluate against baseline

**Require:** Clustering  $\mathcal{C}$ ,  $k$

$r \leftarrow ()$  {The ranking, initially an empty list}

$C \leftarrow$  ranked list of clusters in  $\mathcal{C}$  (using eF or SD)

**for all** cluster  $c \in C$  **do**

$l \leftarrow$  ranked list of documents in  $c$  {process  $C$  in descending weight order}

**for**  $i = 1$  to  $k$  **do**

$r \leftarrow r + l[i]$  {append document at rank  $i$  to  $r$ }

**end for**

**end for**

**return**  $r$



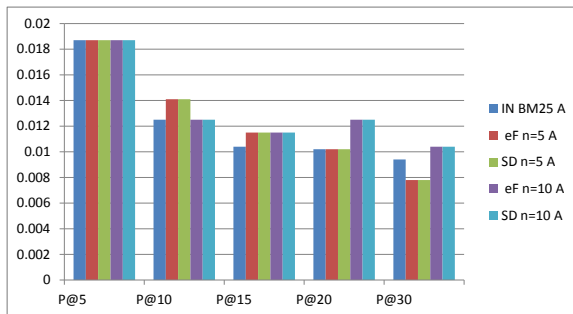
# Experiment Setup

- PF (full text) sub collection of iSearch collection
- Collection's citation information is used for context extraction
- Terrier3.5 Search Engine



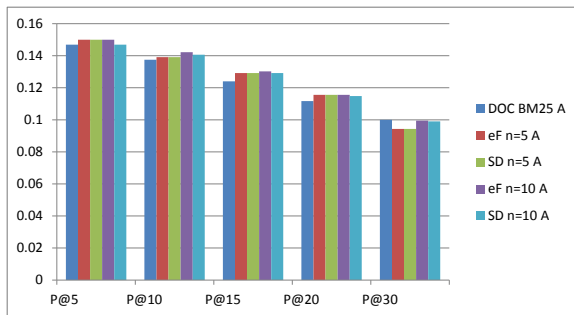
# Evaluation Results

## Results for All Queries IN



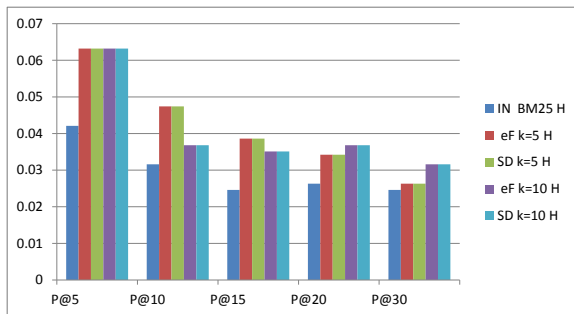
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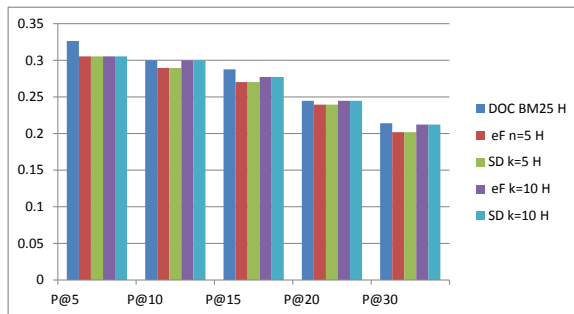
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## Queries with High Relevance Information IN



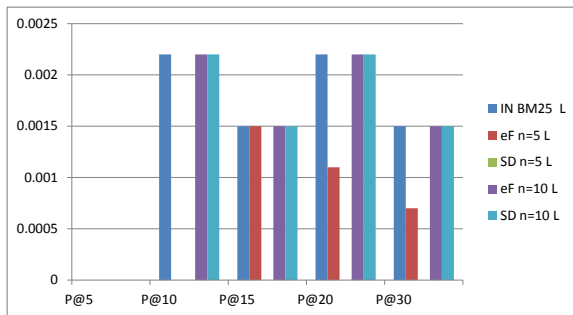
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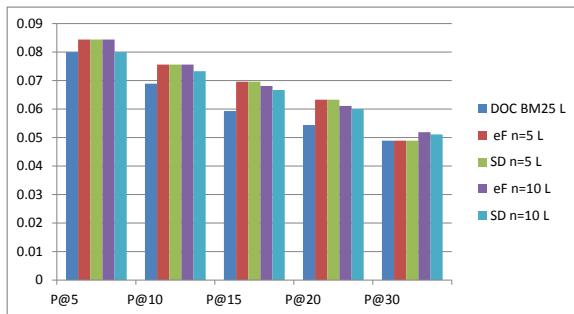
# Evaluation Results

## Queries with Low Relevance Information IN



# Evaluation Results

## Queries with Low Relevance Information Doc





# Conclusion

- Cluster ranking and cluster-based ranking have potential
- Bibliometric information i.e citation context and references show improvement on IR performance when combined with clustering
- Simulated user based evaluation of interactive systems can be enhanced





# Bibliographic context

