# The Effectiveness of *GLOSS* for the Text Database Discovery Problem

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### **Outline**

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- GLOSS: Glossary of Servers Server
- Experimental Framework
- Improving GLOSS
- Conclusions

#### Reference

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## **Introduction**

- Motivation
  - find a scalable solution to the text database discovery problem
  - obvious solutions
    - » forwarding the queries to all known databases
    - » central full index for all of the documents
- · Main idea
  - suggest potentially good databases to search
    - » present the query to server to select a set of promising databases
    - » evaluate the query at the chosen databases
  - estimate by the word-frequency information for each database
    - » how many documents at that database actually contain each word

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### **Introduction**

- Extended semantics
  - exhaustive search
  - all-best search
  - only-best search
  - sample search
- Example 1. find Subject computer

### **GLOSS:** Glossary of Servers Server

- · Query representation
  - atomic subquery is a keyword field-designation pair
  - only consider boolean `and` queries
    - » find Author Knuth and Subject computer
- · Database histograms
  - DBSize(db): the total number of documents in database db
  - -freq(t, db): the number of documents in db that contain t
- Estimate of the result size of a query
  - ESize/Est(q, db) => RSize(q, db)
  - Chosen/Est(q, DB) = { db in DB | ESize/Est(q, db) > 0 ^ ESize/Est(q, db) = max ESize/Est(q, db`), for all db` in DB }

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#### **GLOSS: Glossary of Servers Server**

- Estimators
  - ESize/Ind(find t1^..^tn, db) = [freq(t1,db)/DBSize(db)]\*..\* [freq(tn,db)/DBSize(db)]\*DBSize(db)
    - » keywords appear in the different documents of a database following independent and uniform probability distributions
  - ESize/ $Min(find\ t1^{...}tn,\ db) = min[freq(ti,db)], for i = 1..n$
  - ESize/Binary(find t1^...^tn, db) = 0, if freq(ti,db) = 0 for some i= 1, otherwise
  - example 2. find Author Knuth and Subject computer

### **GLOSS: Glossary of Servers Server**

#### • Evaluation criteria

- compare the prediction of the estimator against what actually is the `right subset` of DB to query
- C/ex : Relevant <= Chosen/est</p>
- C/ab : Best <= Chosen/est
- C/ob: Chosen/est <= Best
- C/sm: Chosen/est <= Relevant

#### • Performance metrics

- $Success(C,Est) = 100 * [ | {q in Q | Chosen/est satisfies C} | / |Q| ]$
- -Alpha(C,Est) = 100 Success(C,Est)
- $Beta(C,Est) = Success(C,Est) 100 * [ | {q in Q | Chosen/est strictly satisfies C} | / |Q| ]$

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### **Experimental Framework**

#### Configuration

- query traces from the FOLIO library IR system
- Relevant(q, DB) = { db in DB | RSize(q, db) > 0 }, Best(q, DB)
- *Ind*: tend to underestimate the result size of the queries

#### • Results

- distinguish two databases
  - » Chosen/Ind = 0 only if Relevant = 0 (or the case of Best = 0)
  - » Success(C/ex, Ind) are much lower than others
  - » the more unrelated subject domains of the databases considered were, the better *Ind* behaved in distinguiching the two databases
- evaluate over six databases

# **Experimental Framework**

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### **Improving GLOSS**

- Elimination of `Subject` index
  - Subject is a compound index built by merging together other `primitive` indexes
  - implicit `or` query : find Subject computer
    - » **find** *Title* computer **or** *Abstract* computer **or** ...
  - two estimates of freq(Subject <w>, <db>)
    - » lower bound : max [freq(index(i) <w>, <db>)]
    - » upper bound : sum [freq(index(i) < w>, <db>)]
- Reduction of histograms
  - threshold : drop the entries of very low frequency
  - classification : define a set of ranges of frequencies
- More flexible definitions

# **Improving GLOSS**

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### **Conclusions**

#### Contributions

- a formal framework for the text database discovery problem
- concept of routing queries to appropriate information sources based on previously collected frequency statistics about sources
- some estimators that may be used to make decisions
- an experimental evaluation according to different semantics

#### • Future research

- hybrid estimator for GLOSS
  - » C/ex: Est/Binary; C/ab, C/ob, C/sm: Est/Ind
- incorporate the cost of charge into the computation of ESize/est
- extend the boolean model to the vector-space retrieval model

Generalizing GLOSS to Vector-Space Databases and Broker Hierarchies *Proceeding of VLDB Conference, 1995*