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Efficient Keyword Search over Virtual XML Views

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Applications - Personal Portal





Keyword Search over XML View

- Materialized XML Views?
 - Similar to keyword search over XML documents
 Many well-studied algorithms
 Materialize views when loading documents
 - Not applicable in emerging applications!
 Overlap/Duplicate/Update overhead
 View definitions not known a-priori
- ► →Keyword Search over *Virtual* XML Views

Related Work

- Scoring and Indexing in IR community
 - DBXplorer [Agrawal02], Banks [Bhalotia02], ObjectRank [Balmin04], XRank [Guo02], Discover [Hristidis 02]
 - Work with materialized documents
- Integrating keyword search and structural queries
 - GTP [Chen 03], TermJoin [Khalifa 03]
 - Access base data to evaluate the view
- Projecting XML documents [Marian 03]
 - Access base data; not leveraging indexes

Outline

Motivation

- Problem Definition
- High-level Overview
- PDT Generation Algorithm
- Experimental Results
- Conclusion

Problem Definition

- Ranked Keyword Search over Virtual XML Views
 - Input: a set of keywords Q = {k1, k2, ..., kn}, an XML view definition V over an XML database D
 - Output: k view elements with highest scores
 - TF-IDF scores
 - TF(k, e): # occurences of the keyword k in an element e
 - *****IDF(k): the inverse of # of elements containing k

 $\text{Score}(e, Q) = \sum_{i} TF(k_{i}, e) * IDF(k_{i})$

Score(e, Q) is further normalized by the length of the view elements

Running Example





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Our Approach



Our Approach



Our Approach -- Challenges



. Joining books & reviews requires isbn (data value)

-- how to get data values without accessing the base data?

2. Scoring view elements requires aggregate statistical data (e.g., tf from book and review)?

-- How to collect them without materializing the view elements?

Dewey ID and Index



PathID	Value	IDList
/books/book/isbn	"111-111-1111"	1.1.1,1.2.1
/books/book/isbn	"222-222-2222"	1.2.1
/books/book/authoi /fn	· "Jane"	1.2.3, 1.7.3

Inverted Index: [Guo03],[Botev05]



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XML View \rightarrow Query Pattern Tree (QPT)

- Similar to GTP, proposed by Chen 2003 for normal query evaluation
 - Captures the structural parts required by queries
 - Mandatory/Optional edges
- New features
 - Node annotations
 - ♦ V: value required to evaluate the view
 - ♦C: content used in the view



PDT Intuition



PDT Generation



Step 1: Get List of IDs



B+-Tree			
Value	IDList		
"111-11-111"	1.1.1		
"121-23-1321"	1.2.1		
"Jane"	1.2.3, 1.7.3		
	ee Value "111-11-111" "121-23-1321" "Jane"		

Key idea: for each node without mandatory child edges, obtain the corresponding list of ids

books//book/isbn: (1.1.1:"111-11-111"),(1.2.1,"121-23-1321")

books//book/title: 1.1.4, 1.2.3, 1.9.3

books//book/year: (1.2.6, 1.5.1:"1996"), (1.6.1:"1997"))





PDT Generator – Merging IDs

(books//book/isbn, (1.1.1: "111-11-1111"), (1.2.1: "121-23-1321"),...) (books//book/title,1.1.4, 1.2.3, 1.9.3, ...) (books//book/year, (1.2.6, 1.5.1: "1996"), (1.6.1:"1997"), ...) **PDT IDs** Candidate

Idea: a loop that merges ids in the lists, and creates the CT nodes in dewey id order

At each step, we check the min id in the CT

if satisfies all restrictions \rightarrow PDT

if satisfies descendant restriction and not ancestor \rightarrow PDT Cache if not satisfies descendant restriction and does not have child node in the CT \rightarrow Discard

Adding CT Nodes from Top Down



Removing CT Nodes from Bottom Up

- Try to determine if a node should be in the PDT: check ancestor constraints
 - Remove IDs known to be non-PDT nodes
 - Nodes in the PDT cache defer checking ancestor restrictions



Correctness and Complexity

- Theorem (Informal)
 - Given a set of keywords, an XQuery view and a database,
 - The result sequence, after being materialized, are identical to as if the view was materialized
 - The byte lengths of each element are identical
 - The TFs of each keyword in each element are identical
 - Formal proof in the technical report
- Complexity: polynomial with respect to the number of IDs, the length of paths, the depth of the documents, and the number of keywords

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Experiments

- Real-world INEX data
 - 500MB
 - Publications with author information and others
 - View: nested articles under authors.
 - Only require author names when evaluating the view
 - Article content (huge) only required after the top k results are identified



Experiments

- Setup
 - 3.4Ghz CPU, 2GB Mem
 - Windows XP
 - Implemented in C++
- Alternatives
 - Baseline: materialize all view results on the fly
 - Timber (GTP [Chen 03] + TermJoin [Khalifa 03])
 not tokenized but still access base data to evaluate the view
 - Proj [Marian 03] : access base data to produce PDT

Varying size of data



Varying size of data



3 100 200 300 400 500 Size of Data(MB)

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Conclusion

- A system architecture for keyword search over virtual XML views
- Novel algorithms to generate pruned data relevant to XML view
- Implemented, and experimentally evaluated
 - 10 times faster than other alternatives
- Future work
 - Top-K keyword search queries
 - Our approach returns pruned version of "all" elements, which is unnecessary
 - Returns most relevant results only
 - QPT/PDT may be adapted for normal query evaluations

Optimizations and Extensions

Extensions

- One ID corresponds to more than one QPT nodes
 - $*//a//a \rightarrow /a/a/a$
 - QNode \rightarrow QNodeSet

Optimizations

- Currently lazy checking of ancestor restrictions
 - Can check in top down phase, and save memory usage of pdt cache
- PDT nodes are output not in document order
 Can enforce document order

Complexity

- $O(Nqdf+Nqd^2+Nd^3+Ndkc))$
 - N: # of IDs in the lists
 - q: the depth of the paths
 - d: the depth of the documents
 - k: the number of keywords
 - c: unit cost of inverted list access
 - Nqdf+Nqd²: cost of top down processing
 - Nd³: cost of bottom up processing
 - Ndkc: cost of inverted list access