# Query Evaluation for XML Databases using Partial Decompression

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**Abstract:**This research work demonstrates the Extraction, compression and query processing of XML documents for Adaptive Compression Techniques and Efficient Query Evaluation. Proposed here are the algorithms for xml compression and Efficient Query Evaluation as - Feasible XML compression using data compression algorithm. Qurey Processor using Sax parsing and Interfaces. It is shown that using the proposed techniques for xml data compression will pave a way for better compression and improve the compression ratio and performance of compressor system

**Keywords:** XML compression, partial decompression, SAX parser, compressor Systems, Query processor, Ziv-Lempel algorithm.

# INTRODUCTION

I.

The proposed research aims at the investigation of efficient Partial query processing techniques on compressed xml dataset in database systems. Mainly our focusing on query proceeding time and memory space. There are a number of XML compression techniques available today which were developed, and tested over the last few years. The problem with XML is that it is text-based, and verbose by its design (the XML standard explicitly states that terseness in XML markup is of minimal importance"). As a result, the amount of information that has to be transmitted, processed and stored is often substantially larger in comparison to other data formats. This can be a serious problem in many occasions, since the data has to be transmitted quickly and stored compactly. Large XML documents not only consume transmission time, but also consume large amounts of storage space.[ H. Liefke and D. Suciu. (2000)]. This section enumerates a few of the most promising techniques namely XGrind, XQC and The evaluation of these compressors considers three main aspects: compression speed, compression ratio, [Al-Hamadani, Baydaa (2011)]

A very large number of XML compressors have been proposed in the literature of recent years. These XML compressors can be classified with respect to two main characteristics. The first classification is based on their awareness of the structure of the XML documents. The main objective of this research work is to initiate an enquiry XML compression in real-time database systems (RTDBS). Xml documents are identified as the major resources that need to be compressed and effectively partial decompression even at the compressed database level with support from underlying operating systems. The main criteria in assessing any XML compressor and decompressor, the success ratio in terms of the CR1 and CS( compression ratio and compressed size).

# II. PRINCIPLES OF THE PROPOSED XML COMPRESSION SCHEME

Several compression techniques have been adopted in these tools, and with varying results. Unfortunately, it is not easy to judge which compressor is the best, or which is the worst. Such a ranking would be necessarily unfair, since each of the compressors that we are aware of differs from the others in some way, and is suited for different purposes. For example, some of the compressors allow transparent access to the compression results. On the other hand, there are XMill and XMLPPM that compress very well, but there is no way to parse the compressed data unless it is decompressed first. Also, some compressors work incrementally and allow on-line processing of the data (XMLPPM), while some are offline by their design (XMill). There is no clear winner |no \ideal" XML compressor|, and one has to formulate his requirements first before he picks and applies the compressor that seems to be the most suitable.

We believe that there are still many paths to be explored in the field of XML data compression. In this work, we present a syntactical compression scheme that is based on probabilistic modeling of XML structure. It does not need the DTD since it infers all the necessary information directly from the input XML data. Moreover, it works incrementally, making on-line compression and decompression possible. Transparent parsing of the compressed data using the SAX interface is possible.

#### III. ALGORITHM AND CORPUS SET

# **3.1Lempel-Ziv Codes**

Lempel-Ziv coding represents a departure from the classic view of a code as a mapping from a fixed set of source messages (letters, symbols or words) to a fixed set of codewords. We coin the term free-parse to characterize this type of code, in which the set of source messages and the codewords to which they are mapped are defined as the algorithm executes. While all adaptive methods create a set of codewords dynamically, defined-word schemes have a fixed set of source messages, defined by context (eg., in text file processing the source messages might be single letters; in Pascal source file processing the source messages might be tokens). Lempel-Ziv coding defines the set of source messages as it parses the ensemble.

The Lempel-Ziv algorithm consists of a rule for parsing strings of symbols from a finite alphabet into substrings, or words, whose lengths do not exceed a prescribed integer L(1); and a coding scheme which maps these substrings sequentially into uniquely decipherable codewords of fixed length L(2) [Ziv and Lempel 1977]. The strings are selected so that they have very nearly equal probability of occurrence. As a result, frequentlyoccurring symbols are grouped into longer strings while infrequent symbols appear in short strings. This strategy is effective at exploiting redundancy due to symbol frequency, character repetition, and high-usage patterns. Figure 3.1 shows a small Lempel-Ziv code table. Low-frequency letters such as Z are assigned individually to fixed-length codewords (in this case, 12 bit binary numbers represented in base ten for readability). Frequently-occurring symbols, such as blank (represented by \_) and zero, appear in long strings. Effective compression is achieved when a long string is replaced by a single 12-bit code.

, sumg is it	praceu 0
Symbol strin	g Code
А	1
Т	2
AN	3
TH	4
THE	5
AND	6
AD	7
_	8
	9
	10
0	11
00	12
000	13
0000	14
Z	15
###	4095

Fig 3.1 -- A Lempel-Ziv code table.

The Lempel-Ziv method is an incremental parsing strategy in which the coding process is interlaced with a learning process for varying source characteristics [Ziv and Lempel 1977]. The Lempel-Ziv algorithm parses the source ensemble into a collection of segments of gradually increasing length. At each encoding step, the longest prefix of the remaining source ensemble which matches an existing table entry (alpha) is parsed off, along with the character (c) following this prefix in the ensemble. The new source message, alpha c, is added to the code table. The new table entry is coded as (i,c) where i is the codeword for the existing table entry and c is the appended character. For example, the ensemble 010100010 is parsed into { 0, 1, 01, 00, 010 } and is coded as  $\{(0,0), (0,1), (1,1), (1,0), (3,0)\}$ . The table built for the message ensemble EXAMPLE is shown in Figure 3.1. The coded ensemble has the form: { (0,a), (1,space), (0,b), (3,b), (0,space), (0,c), (6,c), (6,space), (0,d), (9,d), (10,space), (0,e), (12,e), (13,e), (5,f), (0,f), (16,f), (17,f), (0,g), (19,g), (20,g), (20) }. The string table is represented in a more efficient manner than in Figure 5.1; the string is represented by its prefix codeword followed by the extension character, so that the table entries have fixed length. The Lempel-Ziv strategy is simple, but greedy. It simply parses off the longest recognized string each time rather than searching for the best way to parse the ensemble.

TABLES 5.1(COM US DETAILS)										
File Name	Discription size	Elements	attributes	Max Depth	Avg-Depth					
Mondial	1 MB	22423	47423	5	3.59274					
Swissport	109 MB	2977031	2189859	5	3.55671					
Treebank	82 MB	2437666	1	36	7.87279					
Dblp	127 MB	3332130	404276	6	2.90228					
Yahoo	24KB	342	0	5	3.76608					
UWN	2 MB	66729	6	5	3.95243					

#### TADIES 31(CODDUS DETAILS)

Ebay	34.7 KB	156	0	5	3.7564
Ubid	19.8 KB	342	0	5	3.7661
321gnoe	24.9 KB	311	0	5	3.7653

# OUR CORPUS

#### 1.SwissProt

SWISS-PROT is a curated protein sequence database which strives to provide a high level of annotations (such as the description of the function of a protein, its domains structure, post-translational modifications, variants, etc.), a minimal level of redundancy and high level of integration with other databases.

#### 2. Treebank (partially encrypted)

English sentences, tagged with parts of speech. The text nodes have been encrypted because they are copywritten text from the Wall Street Journal. Never the less, the deep recursive structure of this data makes it an interesting case for experiments.

#### **3.Mondial**

World geographic database integrated from the CIA World Factbook, the International Atlas, and the TERRA database among other sources.

#### 4.DBLP Computer Science Bibliography

The DBLP server provides bibliographic information on major computer science journals and proceedings. DBLP stands for Digital Bibliography Library Projects. Shakespearea collection of the plays of William Shakespeare in XML [5]. The first four datasets given above are regarded as data-centric as the XML documents have a very regular structure, whereas the last one is regarded as document-centric as the XML documents have a less regular structure:

IV. EXPERIMENTAL EVALUATION Table 4.1 Auction Dataset

Auction I	Data KB	CS	CR1	CR2	T1	T2	Т3	TA
Yhoo	24.8	21.7	7	12.5	0.708	0.733	0.725	0.722
Ebay	34.7	41.6	9.590778	-19.8847	0.843	0.84	0.84	0.841
Ubid	19.8	13.5	5.454545	31.81818	0.729	0.717	0.715	0.7203
321gnoe	24.9	22.5	7.228916	9.638554	0.729	0.722	0.72	0.7237

#### Table 4.2: University Courses

Unive	ersity Courses	CS	CR1	CR2	T1	T2	T3	TA
Reed	227	164	5.779736	27.7533	1.143	2.98	1.228	1.7837
Uwm	2.22	1.3	4.684685	41.44144	2.334	2.533	2.3	2.389
WSU	1.57	1.11	5.656051	29.29936	5.901	3.612	3.538	4.3503

# Table 4.3: Transaction Processing Performance Council (TPC)

TF	PC-H	CS	CR1	CR2	T1	T2	T3	TA
Nation	4.47	3.87	6.926174	13.42282	0.0.0.27	0.0.0.17	0.0.0.46	0.0.0.30
customer	502	425	6.772908	15.33865	0.0.1.174	0.0.0.985	0.0.1.280	0.0.1.146
Supplier	28.5	24.6	6.905263	13.68421	0.0.0.95	0.0.0.64	0.0.0.98	0.0.0.85
Partsupp	2.13	1.82	6.835681	14.55399	0.0.4.947	0.0.4.69	0.0.3.965	0.0.4.53
Order	5.12	3.44	5.375	32.8125	0.0.12.716	0.0.7.191	0.0.8.466	0.0.9.45
Part	603	490	6.500829	18.73964	0.0.1.921	0.0.1.453	0.0.1.611	0.0.1.66

Dataset		S	CR1	CR2	T1	T2	T3	TA
dblp1(KB)	1003	438	3.493519	56.33101	0.0.0.30	0.0.0.24	0.0.20	0.0.0.24
Dblp(MB)	131	114	6.961832	12.9771	0.0.10.656	0.0.9.166	0.0.7.135	0.0.8.98
Publication(KB)	5.06	4.99	7.889328	1.383399	0.0.0.69	0.0.0.52	0.0.0.18	0.0.0.46
Authers(KB)	67.3	64.1	7.619614	4.754829	0.0.0.166	0.0.0.154	0.0.0.142	0.0.0.154

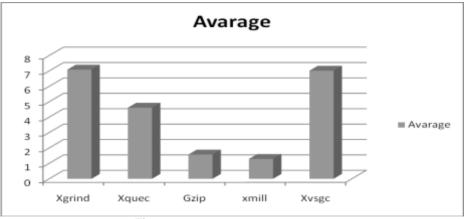


Figure:4.1 CR1 of XVSGC WITH OTHERS

### V. CONCLUSION

To make the XML compressor adaptive and efficient, it is necessary to make the algorithm adaptive and optimal. We started this work with an objective to initiate an enquiry in existing XML compression techniques. The proposed work intent to achieve 'optimal', 'adaptive' and 'flexible', efficient ,'fast' query processing, in XML compression and efficient query evaluation by dividing it into following sub goals -Developing a compression algorithm for XML database.

-Developing an algorithm and query processor, this will help for adaptive /efficient query evaluation and partial decompression. From the results, using the proposed technique of compression of XML databases and query evaluation, it can be concluded that: Compression ration achieved is better than XGrind and comparable with XCQ. Both XMill and XCQ consistently achieve a better compression ratio than gzip.

Compression time is lesser than a Quriable XCQ and XGrind but slightly more compression time than XMill. Decompression time by proposed approach is better than decompression time for XCQ and XGrind and comparable with XMill. A good Query response time and Query speedup factor (QRT and QSF) is achieved with respect to XGrind & XQuec. Finally it can be concluded that using the proposed technique of compression of XML databases and query evaluation over it ,will better improve the performance of XML database systems. During the current research work, several areas have been identified that could be further investigated. The major area of immediate research is an indexing scheme, to aid querying compressed XML databases and development of natural advance compressor. It is anticipated that this will enable further research.

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