

XTRACT: A System for Extracting Document Type Descriptors From XML Documents

Kyuseok Shim KAIST

http://cs.kaist.ac.kr/~shim

(Joint work with Minos Garofalakis, Aristides Gionis, Rajeev Rastogi and S. Seshadri)

XML



- A W3C standard to complement HTML
- An instance of semistructured data [Abi97]
 - Document Type Descriptor (DTD)
- Origin: SGML
- Tags describe the semantics of the data
 - HTML simply specify how the data time is to be displayed
- An element can contain a sequence of nested subelements
- Sub-elements may themselves be tagged elements or character data

XML



- Standard for data representation and data exchange
- Looks like HTML but it isn't
- Collection of elements
 - Atomic (raw character data)
 - Composite (sequence of nested sub-elements)

Example

```
<br/><book>
<title> A Relational Model for Large Shared Data Banks </title>
<author>
<name E.F. Codd </name>
<affiliation> IBM Research </affiliation>
</book>
```



Document Type Definition

- A part of XML specification
- An XML document may have a DTD
- Grammar for describing the structure of XML document
- The structure of an element is specified by a regular expression
- Terminology for XML
 - well-formed: if tags are correctly closed
 - valid: if it has a DTD and conforms to it
- For exchanges of data, validation is useful



Document Type Definition

- Syntax
 - comma: sequence
 - |: or
 - (): grouping
 - ?, *, +: zero or one, zero or more, one or more occurrences
 - ANY: allows an arbitrary XML fragment to be nested within the element



An Example DTD

- <!ELEMENT article (title, author*)>
- <!ELEMENT title (#PCDATA)>
- <!ELEMENT author (name, affiliation)>
- <!ELEMENT name (#PCDATA)>
- <!ELEMENT affiliation (#PCDATA)>



An Example XML Document

```
<article>
<title> A Relational Model for Large Shared Data Banks </title>
<author> E. F. Codd </name>
<affiliation> IBM Research </affiliation>
<author>
</article>
```



XTRACT System

Goal:

Infer DTDs from XML Documents



Why are DTDs useful?

- Plays a crucial role in efficient storage of XML data
 - [Deutsch, Fernandez, Suciu '99]
 - [Shanmugasundaram et al '99]
 - DTD is exploited to generate effective relational schema
- Optimization of XML queries
 - [Goldman, Widom '97]
 - [Fernandez, Suciu '97]
 - DTD allows to restrict the search only relevant portions of the data
- Aids users to form meaningful queries over the XML database



Why need to infer DTD?

- DTDs are not mandatory => might be missing
- XML databases that have been generated automatically
 - Relational databases
 - Flat files
 - HTML documents
 - Bibliographies



Related Work I

- Mining DTDs from a collection of XML documents has not been addressed in the literature
- Database Community
 - Extraction of schema from semistructured data
 - [Nestrorov, Abiteboul, Motwani '98]
 - [Goldman, Widom '97]
 - [Fernandez, Suciu '97]
- Attempts to find typing for semistructured data
- In DTD, outgoing edges from a type can be described by an arbitrary regular expression
- No ordering is imposed for edges



Related Work II

- [Gold '67], [Gold '78], [Angluin '78], [Pitt '89]
 - Infer formal languages from examples
 - Purely theoretical and focus on investigating the computational complexity of the language inference problem
- [Kipelainen, Mannila, Ukkonen '95]
 - Infers a pattern language from positive examples
 - MDL principle was used
 - Assume the set of simple patterns is available
 - Cannot find general regular expressions
 - Patterns are not known apriori



Problem Simplification

- Element types => alphabet
- Infer DTD for each element type separately
- Example sequences: instances of nested subelements
 - Only one level down in the hierarchy



Problem Formulation

- Given a set example sequences for element e
- Compute a DTD for e

Hard problem since DTDs can be general, complex regular expressions



Example

```
<book>
                                        <book>
  <title> </title>
                                           <title> </title>
 <author>
                                           <author>
   <name> </name>
                                             <name> </name>
   <affiliation> </affiliation>
                                             <address> </address>
 </author>
                                           </author>
</book>
                                           <author>
                                             <name> </name>
                                             <address> </address>
                                           </author>
                                           <editor> </editor>
                                         </book>
<paper>
  <title> </title>
 <author>
   <name> </name>
   <affiliation> </affiliation>
 </author>
  <conference> </conference>
  <year> </year>
</paper>
```



Example (Ctd)

Simplified Example Sequences

Desirable Solution

```
<!ELEMENT book (title, author*, editor?)
<!ELEMENT paper (title, author, conference, year)>
<!ELEMENT author (name, affiliation?, address?)>
```



Naive Approach I

- Factor as much as possible
 - e.g. t, ta, taa, taaa, taaaa
 - t | t (a| a(a | a(a | aa)))
 - much more voluminous and a lot less intuitive

Naive Approach II – DFA Minimization



- Suggestion:
 - Consider all example sequences as positive examples of a language of a DFA
 - Find the minimum DFA that accepts the language
- However:
 - Minimum DFA does not imply an intuitive simple regular expression
 - DFA minimization algorithm finds equivalent language
 - Does not generalize
 - How to infer ta* from ta, taa, taaa, taaaa?



Desirable DTD

- It generalize to intuitively correct but previously unseen examples
- It should be concise (i.e. small in size)
 - easy to understand and succinct
- It should be precise
 - not cover too many sequences not contained in the set of examples
 - not too general and captures the structure of input sequences

Trade-off!



Example

• p:- t, ta, taa, taaa, taaaa

Candidate DTD	Concise	Precise
ta taa taaa taaaa	no	yes
(t a)*	yes	No
ta*	yes	somewhat



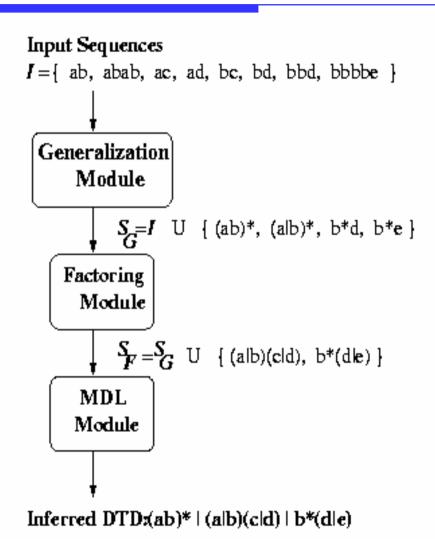
Example

 $I = \{ab, abab, ababab\}$

- (a | b)*
 - a gross over-generalization of the input
 - completely fails to capture any structure inherent in input
- ab | abab | ababab,ab | ab(ab | abab)
 - accurately reflect the structure of the input sequences but do not generalize
- (ab)*
 - succinct and generalizes the input sequence without loosing too much structure information



Overview of XTRACT System





XTRACT System

- Generalization
 - generalizes zero or more candidate DTDs by replacing patters in the input sequence with metacharacters like *
 - e.g. abab => (ab)*, bbbe => b*e
- Factorization
 - factors common subexpressions from the generalized candidate DTDs
 - e.g. b*d | b*e => b* (d | e)
- Minimum Description Length (MDL) Principle
 - MDL ranks each candidate DTD and chooses the minimum cost DTD

MDL Principle



- An information-theoretic measure for quantifying and thereby resolving the tradeoff between the conciseness and preciseness
- MDL principle has been successfully applied in a variety of situations
 - e.g. decision tree classifiers
- Roughly speaking, the best theory to infer from a set of data is the one that minimizes the sum of
 - the length of the theory, in bits (conciseness)
 - the length of the data, in bits, when encoded with the help of the theory (preciseness)



MDL Module

- MDL Principle: minimize the sum of
 - Theory description length, plus
 - Data description length given the theory
- In order to use MDL, need to:
 - Define theory description length (candidate DTD)
 - Define data description length (input sequences) given the theory (candidate DTD)
 - Solve the resulting minimization problem

MDL Module – Encoding Scheme



- Description Length of a DTD
 - Number of bits required to encode the DTD
 - |Size of DTD| x log (# of alphabets)
- Description Length of a sequence given a candidate DTD
 - Number of bits required to specify the sequence given DTD
 - Use a sequence of encoding indices
 - Encoding of a given a is the empty string
 - Encoding of a given (a|b|c) is the index 0
 - Encoding of aaa given a* is the index 3

Example



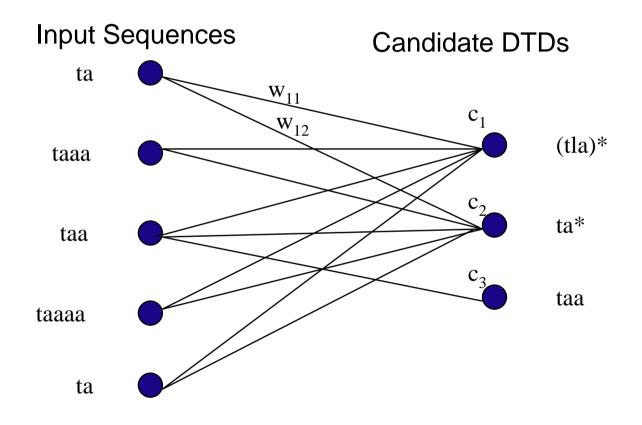
$$I = \{ab, abab, ababab\}$$

- (a | b)*
 - abab: cost of 5 (the number of repetitions (4) + 4 characters to represent chosen character)
 - MDL cost = 6 (encoding DTD) + 3 + 5 + 7 = 21
- ab | abab | ababab
 - MDL cost = 14 + 3 = 17
- ab | ab(ab | abab)
 - MDL cost = 14 + 1 + 2 + 2 = 19
- (ab)*
 - MDL cost = 5 + 3 = 8



MDL Module - Minimization

Facility Location Problem (NP-hard)



Computing the DTD with Minimum MDL Cost



- Facility Location Problem (FLP)
 - Let C be a set of clients and J be a set of facilities
 - c(j): cost of choosing a facility $j \in J$
 - d(j,i): cost of serving client $i \in C$ by facility $j \in J$
 - Choose a subset of facilities $F \subset J$ such that

$$\min_{F\subset J}\{\sum_{j\in F}c(j)+\sum_{i\in C}\min_{j\in F}d(j,i)\}$$

- NP Hard problem
- Approximate algorithm [Charikar, Guha '99]:
 O(n^2logn)

KAIST 한국과학기술원 Korea Advanced Institute of Science & Technology

Generalization Module

- Goal:
 - Introduce metacharacters *, |
 - Produce candidate DTDs of the form
 - a*bc*, (abc)*, (a|b|c)*, ((ab)*c)*, etc.
- Combine two generalization heuristics
 - DISCOVERSEQPATTERN(s,r)
 - S=att...ttb => at*b
 - DISCOVERORPATTERN(s,d)
 - Symbol a1, a2,...am in close proximity => (a1|a2|...|am)
- Candidate DTDs are generated by calling the above functions for appropriate values of r and d



Generalization - Example

DISCOVERSEQPATTERN(s,r)

DISCOVERORPATTERN(s,d)

```
e.g. s = abcbac, d= 2
abcbac => a(b|c)*ac
e.g. s = abcbac, d = 3
abcbac => (a|b|c)*
```



Factoring Module

- Goal: Combine different candidates to derived more compact, factored DTDs forms
 - e.g. ac, ad, bc, bd => (a|b)(c|d)
- MDL modeling justification: Reduce description length of the candidate DTDs
- Intuitive justification: Capture DTDs like
 <!ELEMENT article (title, author*,
 (workshop | conference | journal),
 (computer science | physics | chemistry | ...))>
- Capture optional elements
 - ab,a => a(b | ε) => ab?



Factoring Module

- Adaptation of the factoring algorithms for boolean expressions in logic optimization literature
- Novel heuristic algorithms for selecting subsets of candidate DTDs that give good factored forms



Experimental Result

- Tested Algorithms
 - XTRACT
 - DDbE: IBM alphaworks DTD extraction tool (V 1.0)
 - Java component library for inferring DTD from XML data instances
 - Control parameters: default values are used
- Use both Synthetic and Real-life DTDs
- Data Sets: Generate 1000 random example sequences for each DTD



Experimental Results

- Real-life Data Set
 - Newspaper Association of America (NAA)
 Classified Advertising Standard XML DTD

No.	Original DTD	Simplified DTD
1	< !ENTTTY % included-elements	a b c d e
	"audio-clip blind-box-reply graphic linkpi-char video-clip" >	
2	< !ELEMENT communications-contacts	$(a b c d e)^*$
	(phone fax email pager web-page)* >	
3	< !ELEMENT employment-services(employment-service.type,	ab^*c^*
	employment-service.location * (e.zz-generic-tag)*) >	
4	ENTTTY % location" addr*, geographic-area?, city?, state-province?,</p	a*b?c?d?
	postal-code?, country?" >	
5	< !ELEMENT transfer-info(transfer-number, (from-to, company-id)+,	$(a(bc)^+d)^*$
	contact-info)* >	
6	< !ELEMENT real-estate-services(real-estate-service.type,	$(ab^{?}c^{*}d^{?})^{*}$
	real-estate-service.location?, r-e.response-modes*, r-e.comment?)* >	

Table 4: Real-life DTD Data Set



Experimental Result

- Real-life Data Set
 - XTRACT is able to infer the first five correctly
 - DDbE could obtain the original DTD only the third one

NO	Simplified DTD	DTD Obtained by XTRACT	DTD obtained by DDbE
1	a b c d e	a b c d e	a b c d e
2	$(a b c d e)^*$	$(a b c d e)^*$	$(a b c d e)^*$
3	(ab^*c^*)	ab^*c^*	$(ab^+c^*) (ac^*)$
4	$a^*b?c?d?$	a*b?c?d?	$(a^+b(c (c?d))?) ((b a^+)?cd) $
			$((a^+ b)?d) ((a^+ b)?c) a^+ b)$
5	$(a(bc)^+d)*$	(a(bc)*d)*	$(a b c d)^+$
6	$(ab?c^*d?)*$	_	(a b c d)+

Table 6: DTDs generated by XTRACT and DDbE for Real-life Data Set



Experimental Results

- Synthetic Data Set
 - Each DTD is randomly chosen from either
 - A1 | A2 | ... | An or A1 A2 A3 ... An
 - n is randomly chosen between a1 and mb
 - Each Ai has one of the following four forms
 - $-(a_1|a_2|...|a_{mi})$
 - $-(a_1a_2...A_{mi})$
 - $-(a_1|a_2|...|a_{mi})^*$
 - $-(a_1a_2...a_{mi})^*$
 - mi is chosen to be between 1 and ms

KAIST 한국과학기술원 Korea Advanced Institute of Science & Technology

Experimental Results

- Synthetic Data Set
 - abcde | efgh | ij | klm
 - (a | b | c | d | f)*gh
 - (a | b | c | d)* | e
 - (abcde)*f
 - (ab)* | cdef | (ghi)*
 - abcdef(g | h | I | j)(k | I | m | n | o)
 - (a | b | c)d*e*(fgh)*
 - (a|b)(cdefg)*hijklmnopq(r|s)*
 - (abcd)*|(e|f|g)*| h | (I j k I m)*
 - a* | (b | c | d | e | f)* | gh | (I | j | k)* | (Imn)



Experimental Result

Synthetic Data Set

No.	Original DTD	DTD Inferred by XTRACT	DTD Inferred by DDbE
1	abcde efgh ij klm	abcde efgh ij klm	abcde efgh ij klm
2	$(a b c d f)^*gh$	$(a b c d f)^*gh$	$gh(a b c d f)^+gh$
3	$(a b c d)^* e$	$(a b c d)^* e$	$(e(a c d b)^+e)$
4	$(abcde)^*f$	(abcde)*f	$(f(a e d c b)^+f)$
5	$(ab)^* cdef (ghi)^*$	$(ab)^* cdef (ghi)^*$	$cdef(a b g i h)^{+}cdef$
6	abcdef(g h i j)(k l m n o)	abcdef(g h i j)(k l m n o)	abcdef(j(o l m n k) g(o l n m k)
			h(m l n k o) i(o l n m k))
7	$(a b c)d^*e^*(fgh)^*$	$(a b c)d^*e^*(fgh)^*$	$((c b a)d^+e^+ ad^+ bd^+ c(e^+ d^+)? $
			$= ad^* be^*))(f h g)^+((a b c)d^+e^+)$
			$c(e^+ d^+)? a(e^+ d^+)? b(e^+ d^+)?)$
8	$(a b)(cdefg)^*$	$(a b)(cdefg)^*$	((((a b)hijabedefg) b a)
	$hijklmnopq(r s)^{st}$	$hijklmnopq(r s)^*$	$ (c g f e d s r)^+((b a)?hijkamnopq)) $
9	$(abcd)^* (e f g)^* h (ijklm)^*$	$(abcd)^* (ijklm)^* h (e f g)^*$	$h(a d c b e g f i m l k j)^+h$
10	$a^* (b c d e f)^* gh (i j k)^* $	$a^* (b c d e f)^* gh (i j k)^* $	$(a^+ gh)(e f d i j l n m k e b)^+$
	$(lmn)^*$	$(lmn)^*$	$(a^+ gh)$

Table 5: DTDs generated by XTRACT and DDbE for Synthetic Data Set



Experimental Result

- Synthetic Data Set
 - XTRACT finds all of original DTDs
 - DDbe discovers only the first DTD accurately
 - DDbe may not cover every input sequences
 - e.g. gh is not covered for Dataset 2
 - Note: Dataset 4
 - f is appended to both the front and the back
 - symbols repeated frequently are all or'd together
 - Ddbe is not very good at factoring
 - e.g. Dataset 6



Conclusions

- Inference of DTDs can be used for improving database storage and for query optimization
- Hard problem! Naive approaches fail to produce intuitive, meaningful DTDs
- Presented the architecture of the XTRACT system for inferring a DTD for a database of XML documents
 - generalization step
 - factorization step
 - MDL cost selection step
- Demonstrated the effectiveness of XTRACT