

An Analysis of Approaches to XML Schema Inference*

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Abstract

In this paper we focus on the problem of automatic inferring an XML schema for a given sample set of XML documents. We provide an overview and analysis of existing approaches and compare their key advantages. We conclude the text with a discussion of open issues and problems to be solved as well as their possible solutions.

1 Introduction

Without any doubt the XML [8] is currently a de-facto standard for data representation. Its popularity is given by the fact that it is well-defined, easy-to-use and, at the same time, enough powerful. To enable users to specify own allowed structure of XML documents, so-called *XML schema*, the W3C has proposed two languages – DTD [8] and XML Schema [7,33]. The former one is directly a part of XML specification and due to its simplicity it is one of the most popular formats for schema specification. The latter language was proposed later, in reaction to the lack of constructs of DTD. The key emphasis is put on simple types, object-oriented features (such as user-defined data types, inheritance, substitutability etc.) and reusability of parts of a schema or whole schemas.

On the other hand, statistical analyses of real-world XML data show that a significant portion of XML documents (52% [21] of randomly crawled or 7.4% [24] of semi-automatically collected) still have no schema at all. What is more, XML Schema definitions (XSDs) are used even less (only for 0.09% [21] of randomly crawled or 38% [24] of semi-automatically collected XML documents) and even if they are used, they often (in 85% of cases [4]) define so-called *local tree grammars* [27], i.e. languages that can be defined using DTD as well.

Consequently a new research area of automatic construction of an XML schema has opened. The key aim is to create an XML schema for the given sample set of XML documents that is neither too general, nor too restrictive. It means that the set of document instances of the inferred schema is not too broad in comparison with the sample data but, also, it is not equivalent to them. Currently, there are several proposals of respective algorithms, but there is still a space for further improvements. In this paper we provide an analysis and overview of existing approaches and compare their advantages and disadvantages. In particular, we deal with the problems that have already been solved and solutions used. We conclude with a discussion of remaining open issues as well as their possible solutions.

The paper is structured as follows: Section 2 introduces the relation of XML schemata to theory of languages and automata. In Section 3 we discuss the existing algorithms for schema inference and sum up the key findings. Section 4 discusses the related open issues and Section 5 provides conclusions.

2 XML Schemata and Grammars

The most popular language for description of the allowed structure of XML documents is currently the Document Type Definition (DTD) [8]. For simple applications it is sufficient, but for more complex ones the W3C proposed a more powerful tool – the XML Schema language [7,33].

An XML schema describing the allowed structure of XML documents is a context-free grammar [3], i.e. a grammar where nonterminals can be rewritten without regard to the context in which they occur.

Definition 1 A context-free grammar is quadruple $G = (N, T, P, S)$, where N and T are finite sets of nonterminals and terminals, P is a finite set of productions and S is a non terminal called a start symbol. Each production is of the form $A \rightarrow \alpha$, where $A \in N$ and α is a regular expression over alphabet $(N \cup T)^*$.

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The language generated by grammar G is denoted $L(G)$.

Definition 2 Given the alphabet Σ , a regular expression (RE) over Σ is inductively defined as follows:

- \emptyset (empty set) and ϵ (empty string) are REs.
- $\forall a \in \Sigma : a$ is a RE.
- If r and s are REs of Σ , then (rs) (concatenation), $(r|s)$ (alternation) and (r^*) Kleene closure are REs.

Note that the DTD language adds two abbreviations: $(s|\epsilon) = (s?)$ and $(ss^*) = (s^+)$. Also the concatenation is expressed via the ‘,’ operator. The XML Schema language adds (among other extensions) another one, so-called *unordered sequence* of REs s_1, s_2, \dots, s_k , i.e. an alternation of all possible ordered sequences of s_1, s_2, \dots, s_k .

A language specified by a grammar can be accepted by an automaton, in our case a finite state automaton.

Definition 3 A finite state automaton (FSA) is quintuple $A = (Q, \Sigma, \delta, S, F)$, where Q is a set of states, Σ is a set of input symbols (alphabet), $\delta : Q \times \Sigma^* \rightarrow Q$ is the transition function, $S \in Q$ is the start state and $F \subseteq Q$ is the set of final states.

The language recognized by an automaton A is denoted $L(A)$.

Note that for each RE we can construct a FSA and vice versa.

3 Analysis of Existing Approaches

The studied problem can be described as follows: Being given a set of XML documents $\Delta = \{D_1, D_2, \dots, D_n\}$ (i.e. words over an alphabet T), we search for an XML schema S_Δ (i.e. a grammar G_Δ) s.t. $\forall i \in [1, n] : D_i$ is valid against S_Δ (i.e. $\Delta \subseteq L(G_\Delta)$). In particular, we are searching for S_Δ that is enough concise, precise and, at the same time, general.

From the point of view of the result, we can distinguish methods which output DTDs or XSDs. The problem is that some of the methods claim to produce XSDs, but their expressive power is not beyond the expressive power of DTD. Since most of the DTD constructs are intended for specification of content models of elements, the existing approaches focus mainly on them. Consequently, they often ignore attributes, mixed content or special data types, such as ID, IDREF(S). In case of XSD constructs the existing methods focus on simple data types, elements

having various contexts and “syntactic sugar” such as unordered sequences.

On the other hand, the types of the inference process can be divided into *heuristic* and *grammar-inferring*. In the former case the result does not belong to any class of grammar and, hence, we cannot say anything about its features. In the latter case the algorithms output particular class of languages with specific characteristics. Although grammars accepting XML documents are context-free, the problem can be reduced to inferring a set of REs, each for a single element. But, since according to Gold’s theorem [15] regular languages are not identifiable only from positive examples (i.e. sample XML documents which should conform to the resulting schema), the existing methods need to exploit either heuristics or restriction to an *identifiable* subclass of regular languages.

Most of the existing approaches use the following strategy: For each occurrence of element $e \in \Delta$ and its subelements e_1, e_2, \dots, e_k we construct a production $e \rightarrow e_1 e_2 \dots e_k$. The left hand side is called *element type*, the right hand side is called a *content model* of the element type. The productions form so-called *initial grammar* (IG). For each element type the productions are then merged, simplified and generalized using various methods and criteria. A common approach is so-called *merging state algorithm*, where a *prefix tree automaton* (PTA) is built from the productions of the same element type and the automaton is generalized via merging of its states. Finally, the generalized automaton/grammar is expressed in syntax of the respective XML schema language.

An example of IG and PTA is depicted in Figure 1.

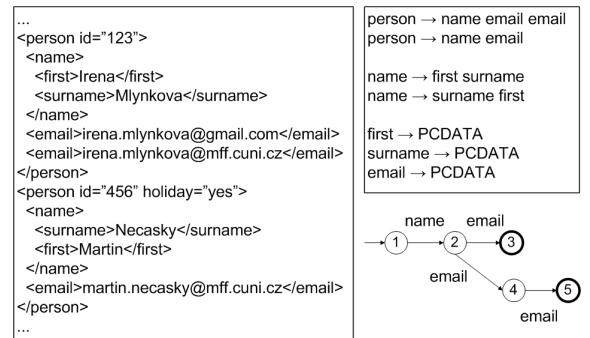


Figure 1. IG and PTA for element person

3.1 Heuristic Approaches

Heuristic approaches are based on experience with manual construction of schemas. Their results do not belong to any special class of grammars and they are

based on generalization of IG using a set of predefined heuristic rules, such as, e.g., “if there are more than three occurrences of an element, it is probable that it can occur arbitrary times”. These techniques can be further divided into methods which generalize IG until a satisfactory solution is reached (e.g. [25, 30]) and methods which generate a number of candidates and then choose the optimal one (e.g. [14]). While in the first case the methods are threatened by a wrong step leading to a suboptimal schema, in the latter case they have to cope with space overhead and specifying a reasonable function for quality evaluation.

GB-engine Probably the first true XML approach to automatic DTD inference is system called *grammar builder engine* (*GB-engine*) [30]. In this particular case it outputs SGML DTDs – a predecessor of XML DTDs. The approach is relative simple and straightforward. Firstly, the IG is created. Then the productions with the same element type are combined using the alternation into a common production:

$$\begin{aligned} A &\rightarrow B \\ A &\rightarrow C \quad \Rightarrow A \rightarrow B \mid C \mid D \mid E \\ A &\rightarrow D \mid E \end{aligned}$$

In the next step, using a set of heuristic rules the content models of the productions are simplified and generalized until any of the rules can be applied. Apart from simple rules that solve trivial cases, the set involves rules dealing with repetitions, identical bases, redundancy etc.:

$$\begin{aligned} A &\rightarrow B \mid B \mid B \quad \Rightarrow A \rightarrow B^+ \\ A &\rightarrow (B \mid C)^* \mid B? \mid C? \Rightarrow A \rightarrow (B? \mid C?)^* \\ A &\rightarrow B \mid C \mid B \mid C^* \quad \Rightarrow A \rightarrow B \mid C^* \end{aligned}$$

Finally, the productions are rewritten into DTD syntax. Note that apart from the content models, the system supports attributes, #PCDATA and mixed content.

DTD-miner *DTD-miner* [25] is one of the first approaches inferring an XML DTD. It is very similar to the previous case, the key difference is in the representation of the input XML documents and in the heuristic rules. Firstly, the so-called *spanning graph*, i.e. an equivalent of the well-known *data guide* [16], is built for the input XML documents. Each node of the graph represents a unique element $e \in \Delta$ and bears information on all its attributes and textual data. The edges of the graph represent element-subelement relationships occurring in Δ and their occurrence.

The set of rules involves optionality, repetition and grouping. Optionality identifies elements that do not

occur in all the input documents. Repetition identifies adjacent elements that occur multiple times. And grouping rule identifies repeating groups of adjacent elements. Finally, the generated DTD is refined to gain less complex structures using further rules such as:

$$\begin{aligned} A?, B?, C? &\Rightarrow A \mid B \mid C \\ A, B, C, D, A, D, B, C &\Rightarrow (A \mid B \mid C \mid D)^+ \end{aligned}$$

Note that while in the previous case the alternation was the first to be involved in the result, in this case it is the last one.

XTRACT The *XTRACT* [14] system is a classical representative of a merging state algorithm. It differs in two aspects: The approach produces a set of possible solutions and selects the optimal one, i.e. it is able to evaluate quality of a schema generalization. The possible solutions are created using heuristic generalization rules for optionality, repetition and grouping similar to those proposed in DTD-miner.

For the purpose of schema evaluation the authors exploit so-called *minimum description length (MDL) principle*. It expresses the quality of a DTD candidate using two aspects – conciseness and preciseness. *Conciseness* of a DTD is expressed using the number of bits required to describe the DTD (the smaller, the better). *Preciseness* of a DTD is expressed using the number of bits required for description of the input data using the DTD. In other words, the more accurately the structure is described, the fewer bits are required. Since the two conditions are contradictory, their balancing brings reasonable and realistic results.

sk-ANT The *sk-ANT* [35] method extends the previous approach in two aspects. Firstly, the searching for the optimal solution is performed using the *Ant Colony Optimization (ACO)* heuristics and a new merging method called *sk-strings* is introduced.

The ACO heuristics is a kind of general heuristics that enables one to find a suboptimal solution. A set of artificial “ants” $B = \{a_1, a_2, \dots, a_l\}$ search a space S of possible solutions (i.e. DTD generalizations) trying to find the optimal solution. The quality of a solution is evaluated using the MDL principle. In i -th iteration each $a \in B$ searches a subspace of S for a local suboptimum until it “dies” after performing a predefined amount of steps. A step of an ant represents an application of any of the merging criterions on the current DTD. While searching, an ant a spreads a certain amount of “pheromone”, i.e. a positive feedback which denotes how good solution it has found so far. This information is exploited by ants from the following iterations to choose better search steps.

On the other hand, the *sk*-strings merging rule is based on a relaxed variant of Nerode equivalence. *Nerode equivalence* assumes that two states p and q are equivalent if sets of all paths leading from p and q to terminal state(s) in F are equivalent. But as such condition is hardly checked, we can restrain to *k*-strings, i.e. only paths of length of k or paths terminating in a terminal state. The respective equivalence of states then depends on equivalence of sets of outgoing *k*-strings. In addition, for easier processing we can consider only s most probable paths, i.e. we can ignore singular special cases.

While the ACO heuristics enables one to avoid constructing multiple solutions concurrently, the *sk*-string provides a better merging criterion.

ECFG The so far described approaches focus on inferring DTDs, in particular respective content models. But, since the XML Schema language offers wider range of constructs, there also appear heuristic approaches dealing with pure XSD structures. Probably the first representative is proposed in paper [9].

The approach focuses on inference of content models consisting of complex types, sequences and choices, simple data types and exact occurrence ranges. Using an XSD formalism – so-called *extended context-free grammars (ECFG)* – the authors extend a classical merging state algorithm with preserving the exact ranges of occurrences and adding a step which infers simple data types. For this purpose each set of values of an element/attribute is analyzed to identify the minimal data type which contains all of them. Nevertheless, the authors focus only on numeric data types (such as `decimal`, `float`, `long`, `negativeInteger`), `date`, `binary` and `string`.

SchemaMiner The *SchemaMiner* system proposed in [34] is another representative of an inference approach that deals with true XSD constructs. It focusses on inferring elements with the same name but different structure and unordered sequences. For this purpose the authors exploit ideas from the previously described works, such as ACO heuristics, *sk*-strings, (k, h) -context [2] or MDL principle in combination with exploitation of tree and graph similarity and clustering.

The elements with the same name but different content are supported only in XSDs and their inference requires exploitation of more sophisticated approaches than combining productions with the same element type. On the other hand, although the unordered sequences are a classical example of XSD “syntactic sugar”, their exploitation enables to define simple and, hence, realistic and usable schemas.

3.2 Inference of a Grammar

Methods *inferring a grammar* exploit the theory of languages and grammars and thus ensure a certain degree of quality of the result. They are based on the idea that we can view an XML schema as a grammar and an XML document valid against the schema as a word generated by the grammar. As we have mentioned, since the class of the regular languages is not identifiable from positive examples, the grammar-inferring methods focus on its identifiable subclasses. All the approaches are classical merging state algorithms, whereas the merging criteria are mostly directly defined by the characteristics of the output class of the language.

(k, h) -contextual languages Paper [2] is probably the first approach dealing with inference of a particular class of XML languages. The approach is based on the observation that if a sufficiently long sequence of terminals occurs in two places in the examples, the components that follow are independent on the position of the sequence within the document.

Definition 4 A regular language L is *k*-contextual, if there exists a finite automaton A s.t. $L = L(A)$ and for any two states p_k, q_k of A and all input symbols $a_1 a_2 \dots a_k$: if there are two states p_0, q_0 of A s.t. $\delta(p_0, a_1 a_2 \dots a_k) = p_k$ and $\delta(q_0, a_1 a_2 \dots a_k) = q_k$, then $p_k = q_k$.

Definition 5 A regular language L is (k, h) -contextual, if there exists a finite automaton A s.t. $L = L(A)$ and for any two states p_k, q_k of A and all input symbols $a_1 a_2 \dots a_k$: if there are two states p_0, q_0 of A s.t. $\delta(p_0, a_1) = p_1, \delta(p_1, a_2) = p_2, \dots, \delta(p_{k-1}, a_k) = p_k$ and $\delta(q_0, a_1) = q_1, \delta(q_1, a_2) = q_2, \dots, \delta(q_{k-1}, a_k) = q_k$, then $p_i = q_i$ for every i s.t. $0 \leq h \leq i \leq k$.

The *k*-contextual and (k, h) -contextual languages form two identifiable subclasses of regular languages which assume that the context of elements is limited. The algorithm is a classical merging state approach starting with a prefix tree automaton, but the merging is not made on the basis of heuristics, but on the basis of the respective features of the languages. The merging criterion is based on an assumption that two states p_k and q_k of the automaton are identical (i.e. can be merged) if there exist identical paths of length k terminating in p_k and q_k . In case of (k, h) -context, also h preceding states in these paths are then identical. The resulting grammar is finally refined to acquire more realistic and concise result.

***f*-distinguishable languages** A different class of identifiable regular languages is inferred in [13]. These are so-called *f-distinguishable languages*.

Definition 6 Let T be a set of terminals and F some finite set. A mapping $f : T^* \rightarrow F$ is called a distinguishing function, if $f(w) = f(z)$ implies $f(wu) = f(zu)$ for all $u, w, z \in T^*$.

Language $L \in T^*$ is called *f-distinguishable* if, for all $u, v, w, z \in T^*$ with $f(w) = f(z)$, we have $zu \in L \Leftrightarrow zv \in L$ whenever $\{wu, wv\} \in L$.

Being given a set of positive examples Δ and the distinguishing function f , the authors propose a merging state algorithm that constructs an automaton A accepting that smallest *f*-distinguishable language that contains Δ .

1-unambiguity An important aspect of XML schemas is so-called *1-unambiguity*. According to the W3C specification, all content models in an XML schema must be 1-unambiguous (deterministic), i.e. they can be matched without looking ahead. A simple example of an ambiguous (non-deterministic) content model is $(e_1, e_2)|(e_1, e_3)$, where while reading e_1 we are not able to decide which of the alternatives to choose unless we read the following element. Though this topic is for some research groups controversial and there exist several studies dealing with (un)necessity of this constraint [19], this condition still remains valid. On the other hand, we can find XML parsers and validators that are able to process also ambiguous content models.

Probably for the first time this problem has been faced in paper [22]. For the purpose of preserving the 1-unambiguity, the authors restrict to so-called *single-occurrence* property of all derived content models which ensures the 1-unambiguity.

Definition 7 A single-occurrence regular expression (SORE) is a regular expression α over Σ s.t. each $s \in \Sigma$ occurs at most once in α .

The authors propose a set of heuristic transformation rules that modify and generalize the IG so that the single-occurrence property is fulfilled in the result. An extension of the proposed approach for XSDs involving simple data types and attributes (which are not supported in the original method) has recently been implemented in system *XStruct* [17].

SOREs and CHAREs The strategy of paper [5] – to define an identifiable class of regular languages and

respective inference algorithm – is similar to the previous ones, but the motivation is slightly different. The authors result from their analysis of real-world XML data and XML schemas and define the classes so that they cover most of the real-world examples. Hence, contrary to the previous works based purely on results of theory of languages, the usability of this approach is undeniable.

The authors focus on two classes of identifiable REs to be inferred – the previously defined SOREs and new, so-called *chain regular expressions (CHAREs)*.

Definition 8 A chain regular expression (CHARE) over Σ is a SORE over Σ that consists of a sequence of factors $f_1 f_2 \dots f_n$, where every factor is an expression of the form $(a_1 | a_2 | \dots | a_k)$, $(a_1 | a_2 | \dots | a_k)?$, $(a_1 | a_2 | \dots | a_k)^+$ or $(a_1 | a_2 | \dots | a_k)^*$, where $k \geq 1$ and every $a_i \in \Sigma$.

The motivation for focusing on CHAREs results from authors' experience with inferring DTDs for real-world XML data. They discover that for small data sets the SOREs are too rich and inference of CHAREs provides more realistic and concise results. Similarly to the previous cases, both the algorithms are based on merging states of a prefix tree automaton using rules that ensure that the result belongs to the required class.

***k*-local single-occurrence grammars** Following their previous work [5], the authors have recently focussed on features and properties of real-world XSDs [6]. Using a similar strategy, they first discover a subclass of XSDs that is most common in real-world XML data (occurs in 98% cases) and, at the same time, that can be identified only from positive examples – so called *k-local, single-occurrence XSDs*.

Definition 9 An XSD is *k-local*, if its content models depend only on labels up to the *k*-th ancestor.

The authors then propose a theoretically complete merging state algorithm called *iXSD* that enables one to infer *k*-local, single-occurrence XSDs.

4 Open Issues

The key characteristics of the described approaches are summed up in Tables 1 and 2. Although each of the existing approaches brings certain interesting ideas and optimizations, there is still a space of possible future improvements.

Table 1. Key characteristics of heuristic methods

Name	Schema	Key Advantages
GB-engine [30]	SGML DTD	First simple heuristic rules.
DTD-miner [25]	DTD	Spanning graph, heuristic rules for optionality, repetition and grouping.
XTRACT [14]	DTD	Set of candidate solutions, MDL principle.
<i>sk</i> -ANT [35]	DTD	<i>sk</i> -string merging (based on Nerode equivalence), ACO heuristics.
ECFG [9]	XSD	Precise occurrence ranges, simple data types.
SchemaMiner [34]	XSD	Unordered sequences, elements with the same name, but different structure.

Table 2. Key characteristics of grammar-inferring methods

Name	Schema	Key Advantages
[2]	DTD	k -contextual and (k, h) -contextual languages.
[13]	DTD	f -distinguishable languages.
[22]	DTD	1-unambiguity.
[5]	DTD	Single-occurrence and chain REs, based on knowledge of real-world data.
[6]	XSD	k -local single-occurrence XSDs, based on knowledge of real-world data.

User Interaction In all the existing papers the approaches focus on automatic inference of an XML schema. The problem is that the resulting schema may be highly unnatural. Although e.g. the MDL principle evaluates the quality of the schema using a realistic assumption that it should tightly represent the data and, at the same time, be concise and compact, users’ preferences can be quite different. (Note that this is not the same motivation as in case of papers [5,6] that focus on real-world DTDs and XSDs.) Hence, a natural improvement may be exploitation of user interaction.

For instance, the user may influence the process of merging by proposing preferred merging operations/target constructs, clustering similar elements etc. Such approach will not only enable to find more concise result, but to find it more efficiently as well. Some of the existing papers (e.g. [2]) mention the aspect of user interaction, typically in the final step of refinement of the result, but there seems to be no detailed study and, in particular, respective implementation.

Other Input Information In all the existing works the XML schema is inferred on the basis of a set of positive examples, i.e. XML documents that should conform to the inferred schema. The Gold’s theorem highly restricts the existing solutions and, hence, the authors focus on heuristics or limit the methods to identifiable classes of languages. But another natural solution to the problem is to exploit additional information, such as XML schema or XML queries.

In the former case we can find the motivation in typical situation [24] when a user creates an XML schema of XML documents but then updates only the data, whereas the schema is considered as a kind of doc-

umentation. Consequently, the schema does not describe the current structure of the data anymore, however it can be used as a source of information because certain matching can be still found. Note that an opposite problem is being currently solved in the area of *schema evolution* (e.g. [20]).

In case of exploitation of XML queries the motivation is similar. In general, the queries restrict only parts of the data structure (those that should appear at output), however even this partial information can be exploited for schema inference. Similarly to the previous case, a related problem is being solved in the area of *XML views* (e.g. [29]).

In addition, there seems to be no approach that would exploit negative examples (i.e. XML documents that should not conform to the schema).

XML Schema Simple Data Types One of the biggest advantages of the XML Schema language in comparison to DTD is its wide support of simple data types [7]. It involves 44 built-in data types, as well as user-defined data types derived from existing simple types using `simpleType` construct. Hence, a natural improvement of the existing approaches is a precise inference of simple data types. Most of the existing approaches omit the simple data types and consider all the values as strings; the only two exceptions [9,17] focus only on selected built-in data types.

Note that the necessity to infer simple data types is naturally closely related to the purpose the schema is inferred for. Assuming that the resulting XML schema is used within a kind of XML data editor, the inferring module should propose also simple data types. On the other hand, if the inferred XML schema is used as a

solution for approaches based on existence of an XML schema, e.g. schema-driven XML-to-relational mapping methods (e.g. [31]), the simple data types are of marginal importance and, thus, can be omitted.

XML Schema Advanced Constructs The second big advantage of the XML Schema language are various complex constructs. The language exploits object-oriented features, such as user-defined data types, inheritance, polymorphism, i.e. substitutability of both data types and elements, etc. Although most of these constructs do not extend the expressive power of XML Schema in comparison to DTD (i.e. they are a kind of “syntactic sugar”) [23], they enable one to specify more user-friendly and, hence, realistic schemas. Naturally, their usage is closely related to the previously described problem of user-interaction, since only the user can specify which of the constructs are preferred.

Integrity Constraints Both DTD and XML Schema enable one to specify not only the structure, but also various semantic constraints. Both involve ID and IDREF(S) data types that specify unique identifiers and references to them. The XML Schema language extends this feature using **unique**, **key** and **keyref** constructs that have the same purpose but enable one to specify the unique/key values more precisely. In addition, the **assert** and **report** constructs enable one to express specific constraints on values using the XPath language [10]. Unfortunately, none of the existing approaches focusses on any of these constraints. In addition, there are also more general integrity constraints [28] that could be inferred, though they cannot be expressed in the existing schema specification languages so far. In general, their inference would extend the optimization of approaches that analyze and exploit information on XML data from XML schemas.

Currently there exist several works which focus on constraint inference [12, 32], but they focus on restricted cases of integrity constraints in special situations. There seems to exist no method that would combine schema inference with a more general approach to inference of related integrity constraints.

Other Schema Definition Languages The DTD and XML Schema are naturally not the only languages for definition of structure of XML data, though they are undoubtedly the most popular ones. There are also other relatively popular schema specification languages. The most popular ones are *RELAX NG* [26] and *Schematron* [18] which both have already become ISO standards.

The Relax NG has similar strategy as both XML Schema and DTD, i.e. it describes the structure of XML documents using content models. Contrary to XML Schema it has much simple syntax, whereas contrary to DTD it supports a richer set of simple data types. On the other hand, the Schematron uses completely different strategy. It does not specify a grammar the XML documents should conform to, but a set of conditions, i.e. integrity constraints, the documents must follow. The conditions are expressed using XPath. Hence, while the inference of Relax NG schema can be based on inference of a DTD/XSD without radical modifications, the approach to inference of Schematron constraints will probably require a brand new method. On the other hand, it can be a natural first step towards inference of general integrity constraints as described before.

Data Streams A special type of XML data that have only recently become popular and, hence, the necessity for proposing respective processing approaches is crucial are so-called *XML data streams*. In this particular application the input data are so huge that they cannot be kept in a memory concurrently, they cannot be read more than once or their processing cannot “wait” for the last portion of the data. Hence the situation is much more complicated. All the XML technologies are currently being accommodated to stream processing and, hence, there occur also requirements for efficient XML schema inference for XML streams [11].

5 Conclusion

The XML schema of XML documents is currently exploited mainly for two purposes – data-exchange and optimization. In the former case we usually need the inferred schema as a candidate schema further improved by a user using an appropriate editor. In the latter case the approaches exploit the knowledge of the schema for optimization purposes such as finding the optimal storage [31] or compression [1] strategy. In general, almost any approach that deals with XML data can benefit from the knowledge of their structure, i.e. XML schema. The only question is to what extent.

Our aim was to provide an analytical study of existing approaches to XML schema inference, as well as a discussion of remaining open issues. We showed that the basic aspects of the problem (such as inference of REs) have successfully been solved. However, there still remain open issues and unsolved problems to focus on. This text should serve as a good starting point for readers searching for an existing solution to

their inference problem, as well as those searching for an interesting and practical research topic.

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