Probabilistic XML functional dependencies based on possible world model

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Abstract

With the increase of uncertain data in many new applications, such as sensor network, data integration, web extraction, etc., uncertainty both in relational databases and XML datasets has attracted more and more research interests in recent years. As functional dependencies (FDs) are critical and necessary to schema design and data rectification in relational databases and XML datasets, it is also significant to study FDs in uncertain XML datasets. This paper first proposed XML functional dependencies (XFDs) of deterministic XML dataset based on tree tuple models. Then two new kinds of functional dependencies based on possible worlds model for probabilistic XML dataset are introduced: probabilistic XML functional dependencies (pXFDs) and probabilistic approximate XML functional dependencies (pAXFDs). pXFDs extend the concept of XFDs of deterministic XML dataset by considering the probability of each possible world of probabilistic XML dataset, and pAXFDs extend the concept of probabilistic XML functional dependencies of probabilistic XML dataset by considering the degree of truth of tree tuples in each possible world of probabilistic XML dataset.

Keywords: uncertain XML, functional dependency, inference rule, closed set

1 Introduction

XML (Extensible Markup Language) has become the de facto standard of data exchange and is widely used in many fields. With the increase in applications such as data integration, web extraction, sensor networks, etc., XML datasets may be obtained from heterogeneous data sources and are not always deterministic. In such cases, XML datasets may contain uncertain data for the same attribute or element due to different data sources, information extraction, approximate query, and data measurement. Although functional dependencies (FDs) of uncertain XML datasets are more complicated than the counterparts of traditional relational databases and deterministic XML datasets, it is possible and necessary to study them in uncertain XML datasets as shown in the paper.

Related work. Although there has been a lot of significant work in functional dependencies for relational databases and XML datasets, none of them can be directly applied to uncertain XML datasets. We analyse the related work in the following three aspects:

(1) For traditional relational databases, functional dependencies are thoroughly studied for several decades [1, 2]. Ref. [3] proposed a concept of functional dependencies in relational databases, which can deal with slight variations of data values. Ref. [4] proposed the conditional functional dependencies to detect and correct data inconsistency. It is obviously that the techniques of traditional relational databases cannot be directly applied to XML due to the significant difference in structure between XML documents and relational databases.

(2) For traditional deterministic XML datasets, functional dependencies are also thoroughly studied for some years. There are two major approaches to define functional dependencies in XML research community, i.e. path-based approach and sub-tree/sub-graph-based approach. In path-based approach [5-11], XML datasets are represented by a tree structure, and some paths of the tree with their values are used in defining XML functional dependencies. In Sub-tree/Sub-graph-based approach [12, 13], functional dependencies of XML datasets are defined by sub-graph or sub-tree in XML datasets. A sub-graph or a sub-tree is a set of paths of XML datasets. As an improvement over Sub-tree/Sub-graph-based approach functional dependencies, Refs. [14, 15] deal with XML functional dependencies with some constraint condition such that there exists a sub-tree is equal. The above XML functional dependencies cannot deal with uncertainty in XML datasets, which is the research topic of the paper. Ref. [16] proposes an approach to discover a set of minimal XML Conditional Functional Dependencies (XCFDs) from a given XML instance to improve data consistency. The XCFDs extends XML Functional Dependencies (XFDs) by incorporating conditions into XFD specifications. It is
easy to see that all the functional dependencies defined above cannot deal with uncertainty in XML datasets.

(3) For uncertain relational databases, Ref. [17] proposes the probabilistic functional dependency for probabilistic relational databases which associated with a likelihood of the traditional functional dependency is satisfied. Ref. [18] proposes some kinds of functional dependencies for probabilistic relational databases, such as Probabilistic Approximate Functional Dependencies (pAFD), Conditional Probabilistic Functional Dependencies (CpFD), and Conditional Probabilistic Approximate Functional Dependencies (CpAFD), which combine approximate, conditional, and approximate/conditional characteristics into traditional functional dependencies to defined corresponding functional dependencies for probabilistic relational databases. Ref. [19] proposes horizontal functional dependencies and vertical functional dependencies for uncertain relational databases, which extends the traditional relational functional dependencies into the uncertain relational databases. Although these work of uncertain relational databases are meaningful and significant, they can not directly applied in uncertain XML datasets, as XML are more complicated in structure then relational databases.

Contributions. In this paper, we will extend the concept of traditional deterministic XML functional dependencies (XFDs) to study the FDs of probabilistic XML datasets by considering the probability of each possible world and the degree of truth of tuples in each possible world. The main contributions of the paper are detailed as followings:

(1) We first proposed XML functional dependencies (XFDs) of deterministic XML dataset based on tree tuple models.

(2) Then a new kind of FDs called probabilistic XML functional dependencies (pXFDs) based on possible worlds model for probabilistic XML dataset is introduced, which extends the concept of XFDs of deterministic XML dataset by considering the probability of each possible world of probabilistic XML dataset. pXFDs, and pAXFDs A sound and complete inference rules are given for the three types of uncertain XML functional dependencies.

(3) Finally another new kind of FDs called probabilistic approximate XML functional dependencies (pAXFDs) is introduced, which extend the concept of pXFDs of probabilistic XML dataset by considering the degree of truth of tree tuples in each possible world of probabilistic XML dataset.

(4) We also analyze the relationship among the tree kinds of FDs: XFDs, pXFDs and pAXFDs. pXFDs extend XFDs by considering the probability of each possible world of probabilistic XML dataset, and pAXFDs extend pXFDs by considering the degree of truth of tree tuples in each possible world of probabilistic XML dataset. More generally speaking, pXFDs are more general than XFDs, and pAXFDs are more general than pXFDs.

Organizations. The rest of the paper is organized as following: Section 2 gives an example as our research motivation and demonstration of the concepts throughout the paper. Three types of functional dependencies, including XFDs, pXFDs, and pAXFDs, of probabilistic XML datasets are given in Section 3. Finally, Section 4 concludes the paper and points out the future directions of the paper.

2 A Motivating example

Suppose we want to know the people impressions of relationship between a man’s salary and his diploma/height. In terms of functional dependencies, if it is the case that [diploma, height]→salary? We design a questionnaire and obtain some data as the following:

TABLE 1 Four interviewees’ impressions about the relationship between a man’s salary and his diploma/height

<table>
<thead>
<tr>
<th>Record</th>
<th>Interviewee ID</th>
<th>diploma</th>
<th>height</th>
<th>salary</th>
<th>probability</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>1</td>
<td>High</td>
<td>high</td>
<td>high</td>
<td>0.9</td>
</tr>
<tr>
<td>2</td>
<td>1</td>
<td>High</td>
<td>low</td>
<td>high</td>
<td>0.8</td>
</tr>
<tr>
<td>3</td>
<td>1</td>
<td>Low</td>
<td>high</td>
<td>low</td>
<td>0.5</td>
</tr>
<tr>
<td>4</td>
<td>1</td>
<td>Low</td>
<td>high</td>
<td>low</td>
<td>0.7</td>
</tr>
<tr>
<td>5</td>
<td>2</td>
<td>High</td>
<td>high</td>
<td>high</td>
<td>0.9</td>
</tr>
<tr>
<td>6</td>
<td>2</td>
<td>High</td>
<td>low</td>
<td>high</td>
<td>0.7</td>
</tr>
<tr>
<td>7</td>
<td>2</td>
<td>Low</td>
<td>high</td>
<td>high</td>
<td>0.5</td>
</tr>
<tr>
<td>8</td>
<td>2</td>
<td>Low</td>
<td>low</td>
<td>low</td>
<td>0.8</td>
</tr>
<tr>
<td>9</td>
<td>3</td>
<td>High</td>
<td>high</td>
<td>high</td>
<td>0.6</td>
</tr>
<tr>
<td>10</td>
<td>3</td>
<td>Low</td>
<td>low</td>
<td>low</td>
<td>0.7</td>
</tr>
<tr>
<td>11</td>
<td>4</td>
<td>High</td>
<td>low</td>
<td>high</td>
<td>0.7</td>
</tr>
<tr>
<td>12</td>
<td>4</td>
<td>Low</td>
<td>high</td>
<td>high</td>
<td>0.5</td>
</tr>
</tbody>
</table>

For Record 1 in Table 1, interviewee 1 thinks (with confidence of 0.9) that if a man’s diploma is high and height is high, then his salary is high. We can store Table 1 as an XML file (survey.xml) as following (attribute “Prob” stands for the confidence of an interviewee’s answer):

<survey>
    <interviewee>
        <ID>1</ID>
        <answer Prob="0.9">
            <diploma>high</diploma>
            <height>high</height>
            <salary>high</salary>
        </answer>
    </interviewee>
    ……
    <interviewee>
        <ID>2</ID>
</surve
We will use the above XML file as demonstration of our proposed concepts throughout the paper.

3 Functional dependencies of probabilistic XML dataset

We first give some preliminary definitions such as DTD (Document Type Definition), XML tree, tree tuple, etc.:

Definition 1 (DTD). A DTD[20] is defined to be D=(E, A, P, R, r), where (1) E is a finite set of element types; (2) A is a finite set of attributes; (3) P is a mapping from E to element type definitions. For each \( \alpha \in E \), \( P(\alpha) \) is a regular expression \( \alpha \) defined as
\[
\alpha := S | \epsilon | \tau | \alpha | \alpha | \alpha
\]
where \( S \) denotes string types, \( \epsilon \) is the empty sequence, \( \tau \in E \), “\|”, “\)” and “\*” denote union, concatenation and Kleene closure respectively; (4) R is a mapping from E to the power set of A: \( P(A) \); (5) \( r \in E \) is called the element type of the root.

A path \( p \) in \( D=(E, A, P, R, r) \) is defined to be
\[
p = \alpha_0 \cdots \alpha_e \ , \ (1) \ \alpha_0 \in r \ ; \ (2) \ \alpha_i \in P(\alpha_{i-1}) , \ i \in [2,n-1] \ ; \ (3) \ \alpha_e \in P(\alpha_{e-1}) \text{ if } \alpha_e \in E \text{ and } P(\alpha_e) \neq \Phi \ , \ \text{or} \ \alpha_e = S \text{ if } \alpha_e \in E \ , \ \text{and} \ P(\alpha_e) = \Phi \ , \ \text{or} \ \alpha_e \in R(\alpha_{e-1}) \text{ if } \alpha_e \in A . \text{ Let } paths(D) = \{ p | p \in D \} .
\]

Definition 2 (XML tree). Let \( D=(E, A, P, R, r) \). An XML tree \( T \) conforming to \( D \) (denoted by \( T \models D \) ) is defined to be \( T=(V, \text{ lab }, \text{ ele }, \text{ att }, \text{ val }, \text{ root}) \), where (1) \( V \) is a finite set of nodes; (2) \( \text{ lab } \) is a mapping from \( V \) to \( E \cup A \); (3) \( \text{ ele } \) is a partial function from \( V \) to \( V' \) such that for any \( v \in V \ , \ \text{ ele } (v) = [v_1 , ... , v_e] \) if \( \text{ lab } (v_1) \cdots \text{ lab } (v_e) \) is defined in \( P(\text{ lab } (v)) \); (4) \( \text{ att } \) is a partial function from \( V \) to \( A \) such that for any \( v \in V \), \( \text{ att } (v) = R(\text{ lab } (v)) \) if \( \text{ lab } (v) \in E \) and \( R(\text{ lab } (v)) \) is defined in \( D \); (5) \( \text{ val } \) is a partial function from \( V \) to \( S \) such that for any \( v \in V \), \( \text{ val } (v) \) is defined if \( P(\text{ lab } (v)) = S \) or \( \text{ lab } (v) \in A \); (6) \( \text{ lab } (\text{ root }) = r \) is called the root of \( T \).

Given a DTD \( D \) and an XML tree \( T \models D \), a path \( p \) in \( T \) is defined to be \( p = v_1 \cdots v_e \), where (1) \( v_1 \in \text{ root } \); (2) \( v_i \in \text{ ele } (v_{i-1}) , i \in [2,n-1] \); (3) \( v_i \in \text{ ele } (v_{i-1}) \) if \( \text{ lab } (v_i) \in E \), or \( v_i \in \text{ att } (v_{i-1}) \) if \( \text{ lab } (v_i) \in A \), or \( v_i = S \) if \( P(\text{ lab } (v_{i-1})) = S \). Let \( paths(T) = \{ p | p \in T \} \).

Fig.1 is a part of an XML tree corresponding to XML file survey.xml in Section 2, in which Interview 1 thinks with confidence 0.9 that if a man’s diploma is high and height is high, then his salary is high.

![FIGURE 1 A part of an XML tree corresponding to XML file survey.xml](image)

Definition 3 (Tree tuple). Given a DTD \( D=(E, A, P, R, r) \) and an XML tree \( T=(V, \text{ lab }, \text{ ele }, \text{ att }, \text{ val }, \text{ root}) \) conforming to \( D \), a tree tuple \( I \) for a node \( v \) in \( T \) is a tree rooted on node \( v \) with all of its descendant nodes.

In XML file survey.xml, there are 4 “interviewee” tree tuples, Fig 1 shows the first “interviewee” tree tuple which is the tree rooted on the first “interviewee” node (ID=1). There are also 12 “answer” tree tuples in the XML file, and Fig 1 shows the first “answer” tree tuple, which is the tree, rooted on the first “answer” node with probability of 0.9.

Based on the above concepts, we give the definition of XML functional dependencies (XML FDs) of deterministic XML dataset, which is based on tree tuple model.
Definition 4 (XML FDs). Given a DTD $D$ and an XML tree $T \models D$, an XML Functional Dependency (XFD) has the form $\{s_i, S_i \rightarrow S \}$, where $s_i$ is the head path, $S_i = \{s_{i1}, s_{i2}, \ldots, s_{im} \}$ is the left path set (i.e. determinant path set), and $S = \{s_1, s_2, \ldots, s_n \}$ is the right path set (i.e. determined path set). For $\forall s_i \in S_i, s_k \in S_k$, $s_i, s_j, \ldots, s_k \in paths(D)$, where $i = 1, \ldots, m, j = 1, \ldots, n$. If $T$ satisfies XFD $\{s_i, S_i \rightarrow S \}$ (denoted by $T \models \{s_i, S_i \rightarrow S \}$), then it implies that for each tree tuple $t$ rooted on node $last(s_i)$, $t(s_i, s_{i1}, s_{i2}, \ldots, s_{in})$ can uniquely determines $t(s_i, s_{i1}, s_{i2}, \ldots, s_{in})$, where $last(s_i)$ denotes the last symbol of head path $s_i$.

By considering the probability of each possible world of probabilistic XML dataset, we give the definition of probabilistic XML FDs of deterministic XML dataset based on possible world model by extending the above concept of XFDs as following:

Definition 5 (Probabilistic XML FDs). Given a DTD $D$ and an uncertain XML tree $T \models D$, a Probabilistic XML FD (pXFD) has the form $\{s_i, S_i \rightarrow_{\text{prob}} S \}^T$, where $s_i$ is the head path, $S_i = \{s_{i1}, s_{i2}, \ldots, s_{in} \}$ is the left path set (i.e. determinant path set), and $S = \{s_1, s_2, \ldots, s_n \}$ is the right path set (i.e. determined path set). For $\forall s_i \in S_i, s_k \in S_k$, $s_i, s_j, \ldots, s_k \in paths(D)$, where $i = 1, \ldots, m, j = 1, \ldots, n$. $T$ satisfies pXFD $\{s_i, S_i \rightarrow_{\text{prob}} S \}$ (denoted by $T \models \{s_i, S_i \rightarrow_{\text{prob}} S \}$) with a confidence $c$ if and only if the sum of probability of the fraction of possible worlds in which the corresponding XFD $\{s_i, S_i \rightarrow S \}$ holds is $c$,

$$c = \sum p(PW) \cdot \sum p(PW) \cdot \sum p(PW)$$

where $p(PW)$ stands for the probability of possible world $PW$, $\sum p(PW)$ stands for the sum of probability of all those possible worlds which satisfy XFD $\{s_i, S_i \rightarrow S \}$, and $n$ is number of all possible worlds of $T$.

Example 1. XML file survey.xml has 64 possible worlds (PW) w.r.t. element “interviewee”, the first possible world is PW1 with the following 4 tree tuples:

\[ T_1^1 = (\text{survey.interviewee.ID=1}, \text{survey.interviewee.answer.@Prob=0.9}, \text{survey.interviewee.answer.diploma=high}, \text{survey.interviewee.answer.height=high}, \text{survey.interviewee.answer.salary=high}), \]

\[ T_1^2 = (\text{survey.interviewee.ID=2}, \text{survey.interviewee.answer.@Prob=0.9}, \text{survey.interviewee.answer.diploma=high}, \text{survey.interviewee.answer.height=high}, \text{survey.interviewee.answer.salary=high}), \]

\[ T_1^3 = (\text{survey.interviewee.ID=3}, \text{survey.interviewee.answer.@Prob=0.6}, \text{survey.interviewee.answer.diploma=high}, \text{survey.interviewee.answer.height=high}, \text{survey.interviewee.answer.salary=high}), \]

\[ T_1^4 = (\text{survey.interviewee.ID=4}, \text{survey.interviewee.answer.@Prob=0.7}, \text{survey.interviewee.answer.diploma=high}, \text{survey.interviewee.answer.height=low}, \text{survey.interviewee.answer.salary=high}). \]
survey.interviewee.answer.height=high, survey.interviewee.answer.salary=low),

\[ T_{i4} = (\text{survey.interviewee.ID}=2, \]

survey.interviewee.answer.@Prob=0.9, survey.interviewee.answer.diploma=high, survey.interviewee.answer.salary=high, survey.interviewee.answer.height=high, survey.interviewee.answer.salary=high),

\[ T_{i4} = (\text{survey.interviewee.ID}=3, \]

survey.interviewee.answer.@Prob=0.6, survey.interviewee.answer.diploma=high, survey.interviewee.answer.height=high, survey.interviewee.answer.salary=high, survey.interviewee.answer.salary=high),

\[ T_{i4} = (\text{survey.interviewee.ID}=4, \]

survey.interviewee.answer.@Prob=0.5, survey.interviewee.answer.diploma=low, survey.interviewee.answer.height=high, survey.interviewee.answer.salary=high, survey.interviewee.answer.salary=high).

It is easy to see that PW_{34} does not satisfy pXFD {survey.interviewee.answer, [diploma, height]⇒_{\text{pXFD}} salary} as tree tuple T_{i4} and tree tuple T_{i4} are contradict to each other.

To solve the above problem, we propose the concept of probabilistic approximate XML FDs as following:

**Definition 6 (Probabilistic Approximate XML FDs).** Given a DTD D and an uncertain XML tree \( T \models D \), a Probabilistic Approximate XML FD (pAXFD) has the form \( \{s_i, S_j \rightarrow_{\text{pAXFD}} S_k \} \), where \( S_h \) is the head path, \( S_e = \{s_{i1}, s_{i2}, \ldots, s_{i}\} \) is the left path set (determinant path set), and \( S_h = \{s_{1}, s_{2}, \ldots, s_{n}\} \) is the right path set (determined path set). For \( \forall s_{i}, s_{e} \subseteq S_{h} \), \( s_{i}, s_{i1}, s_{i2}, \ldots, s_{in} \in \text{paths}(D) \) . \( T \) satisfies pAXFD \( \{s_i, S_j \rightarrow_{\text{pAXFD}} S_k \} \) (denoted by \( T \models (s_i, S_j \rightarrow_{\text{pAXFD}} S_k) \)) with a confidence \( c \) if and only if the sum of probability of each possible world multiplied by the maximal fraction of the tree tuples of the possible world, in which the corresponding pXFD \( \{s_i, S_j \rightarrow S_k \} \) holds is \( c = \frac{\sum (p(PW) \times \alpha_i)}{\sum p(PW)} \), where \( p(PW) \) stands for the probability of possible world \( PW \), and \( \alpha_i \) stands for the maximal tree tuples fraction in possible world \( PW \) , which satisfies the corresponding pXFD \( \{s_i, S_j \rightarrow_{\text{pAXFD}} S_k \} \), and \( n \) is the number of all possible worlds of \( T \).

**Example 2.** As noted before, possible world PW_{34} in XML file survey.xml does not satisfy pXFD {survey.interviewee.answer, [diploma, height]⇒_{\text{pXFD}} salary} as tree tuple T_{i4} and tree tuple T_{i4} are contradict to each other. If we just remove any one of them, then PW_{34} satisfies the pXFD {survey.interviewee.answer, [diploma, height]⇒_{\text{pXFD}} salary}. It is easy to see that for pXFD {survey.interviewee.answer, [diploma, height]⇒_{\text{pXFD}} salary} to be satisfied by PW_{34}, the removed minimal fraction of tree tuples is 1/4, so PW_{34} satisfies the pAXFD {survey.interviewee.answer, [diploma, height]⇒_{\text{pAXFD}} salary} with confidence \( p(PW_{34}) \times 1/4 = 0.9 \times 0.6 \times 0.5 \times 3/4 = 0.10125 \). For the entire XML file survey.xml, it satisfies the pAXFD {survey.interviewee.answer, [diploma, height]⇒_{\text{pAXFD}} salary} with the following confidence

\[ c = \frac{\sum (p(PW) \times \alpha_i)}{\sum p(PW)} = \frac{12.8191}{13.1116} \approx 0.98 \] , where \( p(PW) \)

stands for the probability of possible world \( PW \), and \( \alpha_i \) stands for the maximal tree tuples fraction in possible world \( PW \), which satisfies the corresponding pXFD {survey.interviewee.answer, [diploma, height]⇒_{\text{pXFD}} salary}, and \( n \) is the number of all possible worlds. The meaning of the pAXFD {survey.interviewee.answer, [diploma, height]⇒_{\text{pAXFD}} salary} is that the four interviewees in the survey approximately think (with confidence \( c=0.98 \)) that a man’s diploma and height can determine his salary. We can see that the pAXFD’s confidence is generally higher that that of pXFD (the former is 0.98, and the latter is 0.91).

The relationship between pXFDs and pAXFDs are given in the following theorem:

**Theorem 2 (Relationship between pXFDs and pAXFDs).** If all tree tuples of each possible world either satisfy a pXFD or not at all, then the entire probabilistic XML dataset satisfies the corresponding pAXFD with the same confidence. In general, the confidence of a pAXFD is always greater than the confidence of the corresponding pXFD.

**Proof.** (1) From the definitions of pXFDs and pAXFDs, we can see that a pAXFD is a natural extension of the corresponding pXFD by considering the degree of truth of tree tuples in each possible world of a probabilistic XML dataset. For a pAXFD, if each tree tuple of each possible world either satisfies the corresponding pXFD or not at all, then the degree of truth of tree tuples in each possible world is either 1 or 0 w.r.t. the pAXFD. So the pAXFD is equal to the corresponding pXFD with the same confidence. (2) In each possible world that the pXFD holds, the degree of the truth ( \( \alpha_i \) in Definition 6) of the corresponding pAXFD is 1. In each possible world that the pXFD does not hold, the degree of the truth of the corresponding pAXFD is always greater than 0. So the final confidence of the pAXFD, which is the weighted sum of the confidences of all possible worlds multiplied by the corresponding degree of truth, is always greater than the confidence of the corresponding pXFDs.

**4 Conclusions**

This paper studies the functional dependencies (FDs) of probabilistic XML datasets, which extends the notions of
functional dependencies of uncertain relational databases and traditional functional dependencies of XML datasets based on tree-tuple model. Three kinds of functional dependencies such as XFDs, pXFDs, and pAXFDs are given in the paper: XFDs are traditional FDs to capture the relationship between XML element in a deterministic XML dataset; pXFDs are natural extensions of XFDs by considering the probability of each possible world of probabilistic XML dataset; and pAXFDs are natural extensions of pXFDs by considering the degree of truth of each possible world of probabilistic XML dataset, which is very useful in the case that there are noisy or dirty data in probabilistic XML dataset. The relationship among XFDs, pXFDs, and pAXFDs are also studied in the paper. Generally speaking, pXFDs are more general than XFDs, and pAXFDs are more general that pXFDs.

An interesting work in the future is to assess the confidence of pXFDs and pAXFDs. As we know that both pXFDs and pAXFDs are defined on possible world model and the number of possible world is very large generally, it is a not trivial work to access the confidence of pXFDs and pAXFDs. Another interesting work is to mine the pXFDs and pAXFDs in a given probabilistic XML dataset. Specifically speaking, given a probabilistic XML dataset, find a minimal set of pXFDs or pAXFDs that are equivalent to or more general than any set of pXFDs or pAXFDs that hold over the XML dataset with a confidence higher than a user specified threshold.

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