

# CONSIDERING STRUCTURAL AND VOCABULARY HETEROGENEITY IN XML QUERY: FPTPQ AND HOLISTIC EVALUATION.

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## Abstract

*The integration of XML data sources which have different schemas/DTD can originate structural and vocabular heterogeneity. In this context, it is difficult to write satisfiable queries. As a solution, many Information Systems focus on building approximate evaluation techniques for exact queries. As a project, we build flexible and preference XML query languages and associated evaluation algorithms. In this paper, we propose the Flexible Preference Tree Pattern Query (FPTPQ), a new TPQ that allows multiple items/names (resp. paths) for the same node, in order to integrate (resp. to locate) all the different instances of the database nodes. The FPTPQ enable to have preference nodes and ordering operators among label items and paths. We also provide a holistic algorithm that evaluates the FPTPQ and capitalises the preferences to determine the best available solutions. Illustrations and experimentations are realized to show the effectiveness of our solutions.*

## Keywords

*XML Query with structural preference, Structural heterogeneity, vocabular heterogeneity, flexible query, holistic matching algorithm, ranked results.*

## 1. Introduction

XML has become the standard format for information representation and data exchange among different systems. For interoperability needs, data of various sources which have been modelled independently, can be merged or exploited simultaneously. Generally, the integration of XML data that come from different sources may cause heterogeneity problems [1]. Heterogeneity in the domain of databases have different appreciations [2, 3]. Among the types of heterogeneity, we denote structural heterogeneity and vocabular heterogeneity. The first refers to the fact that the same information instances may have different paths whose name and number of the nested database tree nodes (tag) are different. The second signifies that the instances of the same object (tag, attribute, text value) are represented with different names, that can be synonyms (e.g. skill, competency and expertise), abbreviations (course, crse), or any other group of similar words (hours and times, football and soccer). This problem has serious impact on the querying process. As a matter of fact, during the exact matching of exact queries, since all the constraints are considered as filters, a single erroneous object name among numerous ones means total failure,

The consideration of only one denomination among all the object instance names or paths may result to an incomplete set of solutions. For example, Let’s consider the merging of the documents *uwm.xml* and *wsu.xml* available on [4] which contain respectively the schedule of the University of Wisconsin-Milwaukee (UWM) and the Washington State University (WSU) courses. Table 1 shows fragments of each of these documents. Even though these documents have been built for the same goal and for the same subject (courses scheduling), some information instances are contained in different tag (with similar names) or/and are located with different paths. The correspondent of the *uwm.xml* tags *course*, *course\_listing* and *hours* are respectively *crs*, *course* and *times* in *wsu.xml*. The starting time of a course is located by the path *root/course\_listing/section\_listing/start* in *uwm.xml* and by *root/course/time/start* in *wsu.xml*. The integration of these two databases creates structural heterogeneity and vocabular heterogeneity. A non-aware user who needs the title of all the courses may have as result only the ones of *WSU* courses if he uses the tag *course* without considering *course\_listing*. He will face the same consequences if all the different paths of the needed database node are not considered in the query.

Table 1. Fragment of the XML files of the two USA university course schedule, available on [4].

a) Fragment of <i>uwm.xml</i>	b) Fragment of <i>wsu.xml</i>
<pre> &lt;root&gt;   &lt;course_listing&gt;     &lt;note&gt;&lt;/note&gt;     &lt;course&gt;400-327&lt;/course&gt;     &lt;title&gt;CONTEMPORARY FRENCH ...&lt;/title&gt;     &lt;credits&gt;3&lt;/credits&gt;     &lt;level&gt;U&lt;/level&gt;     &lt;restrictions&gt;; (HU) PREREQ: FRENCH 303(215)     ...&lt;/restrictions&gt;     &lt;section_listing&gt;       &lt;section_note&gt;&lt;/section_note&gt;       &lt;section&gt;Se 001&lt;/section&gt;       &lt;days&gt;TR&lt;/days&gt;       &lt;hours&gt;         &lt;start&gt;12:30pm&lt;/start&gt;         &lt;end&gt;1:45pm &lt;/end&gt;       &lt;/hours&gt;       &lt;bldg_and_rm&gt;         &lt;bldg&gt;CRT&lt;/bldg&gt;         &lt;rm&gt;B13 &lt;/rm&gt;       &lt;/bldg_and_rm&gt;       &lt;instructor&gt;Alkhas Alkhas&lt;/instructor&gt;     &lt;/section_listing&gt;   &lt;/course_listing&gt;   ... &lt;/root&gt; </pre>	<pre> &lt;root&gt;   &lt;course&gt;     &lt;footnote&gt;&lt;/footnote&gt;     &lt;sln&gt;10637&lt;/sln&gt;     &lt;prefix&gt;ACCTG&lt;/prefix&gt;     &lt;crs&gt;230&lt;/crs&gt;     &lt;lab&gt;&lt;/lab&gt;     &lt;sect&gt;01&lt;/sect&gt;     &lt;title&gt;INT FIN ACCT&lt;/title&gt;     &lt;credit&gt;3.0&lt;/credit&gt;     &lt;days&gt;TU,TH&lt;/days&gt;     &lt;times&gt;       &lt;start&gt;7:45&lt;/start&gt;       &lt;end&gt;9&lt;/end&gt;     &lt;/times&gt;     &lt;place&gt;       &lt;bldg&gt;TODD&lt;/bldg&gt;       &lt;room&gt;230&lt;/room&gt;     &lt;/place&gt;     &lt;instructor&gt;B.     MCELDFOWNEY&lt;/instructor&gt;     &lt;limit&gt;0112&lt;/limit&gt;     &lt;enrolled&gt;0108&lt;/enrolled&gt;   &lt;/course&gt;   ... &lt;/root&gt; </pre>

The characterization of the searched information alongside with the query evaluation process most consider these issues in order to find the complete set of solutions available for the user in the database. In the literature, some authors focus on tailoring a homogeneous database from the heterogeneous XML document collection. Other authors make some mapping between data sources schemas, that help to construct from each source document (resp query), a target one which is more appropriated for the querying process [5] [6]. In [7], AlHamad et al. produce a global schema for the entire database, alongside with mapping procedures between it and the multiple sources schemas, in order to provide a homogeneous view over heterogeneous XML data

[8], [9]. Further works [10], use ontology, web semantic [11], thesaurus, passed user experiences and other artificial intelligence techniques to automatically rewrite query. The purpose is to obtain from the user query, a more suitable one, with the respect of the database structure/format [12]. The resulting queries are more likely to produce solutions than the user one, since the user may not be familiar to database schema and vocabulary. These techniques are used in mediation systems. A rewriting system can produce as much target queries as there are data sources or clusters with different schemas. Consequently, multiple evaluations are needed. The flexibility features are used after the formulation of the query by the user. The cited works do not allow the user to propose himself, some additional label items that may be useful to determine alternative solutions, or the complete set of the needed solutions. In fact, the existing query languages do not allow to have multiple paths to localized the occurrences of the same object, or multiple (similar) words as a node label, where each word can be one of the names used for the same database node. Moreover, the user may have in mind some potential related substitute words or objects which can help the evaluation process to also select alternative solutions in case the ideal query (query that is supposed to give the user a maximum satisfaction) does not have database images. But he will still be obliged to write other queries by continuously make several adjustments from the initial one, replacing some labels by others, sometime without changing the initial query organization. If this situation is more common in e-commerce context, it is a general problem for database querying system. Note that, as an XML document has a tree shape, XML queries that are written with the most popular languages XPath and XQuery can be represented as a combination of one or many trees, called Tree Pattern Queries (TPQ).

Different types of tree pattern query exist in the literature [13]. The most expressive ones enable the utilization of the wildcard \* to match any single database node, or the Ancestor-Descendant (//) operators to allow the matching of the same object occurrences that have different paths. As example, If the user wants to select the titles of books and articles in the database dblp, he may use the wildcard "\*" to have the paths dblp/\*/title. To select all the start time of the courses in *merged\_wsu-uwm.xml*, he may use *root//start*. But these operators are responsible of many useless solutions and does not enable any preference order. *dblp/\*/title* will select the title of all the documents (articles, books, but also improceedings, master thesis and PhD thesis). *root//start* will select all the occurrences of the node *start* which are the descendant of the *root*, no matter what is in-between. Rather than using the *wildcard* and the *A-D* operators, a list of additional elements can be added as replacement items to query node labels/paths. This type of flexibility ensures that only the needed solutions are returned to the user. In some case, the replacement items can allow the selection of alternative solutions that are closed to user needs (example room if studio are not available). Such tree pattern is not useful only for the user. It can also be used to represent the results of a query reformulation [14] [15] [16] by a mediator system.

Flexibilities have been imported in database query process to write soft queries and allow more solutions possibilities. The concept of preference query, also none as bipolar query, is used to write query that have two parts: a first part has "must be satisfied" constraints, and a preference part is made of soft constraints. Contrary to the first part whose constraints are considered as filter (conditions whose every solution must absolutely satisfy), the satisfaction of the preference part is optional, but it enhances the correspondent solution value. i.e., the solutions which also satisfy the second part are must likely to be preferred by the user than those who satisfy only the obligatory constraints. None of the existing TPQ, even preference one, allow the integration of difference instance names for the same object, as node label. None of them allow multiples paths for the same query leaf node. In this paper, we propose (in section 3) a more general Tree Pattern Query called Flexible Preference Tree Pattern Query (FPTPQ). The FPTPQ enable replacement

items for node labels, and multiple paths to locate the multiple instances of the same database object. The items can be of equal value or classified in preference order using ordering operators, when the replacement items are the attributes of alternative solutions that have difference preference values, according to the user. For example, a user who need a room can add studio as alternative. Attributes of room will be more rated than the attributes of studios. The FPTPQ improves the preference operator of the language *prefXPath*, proposed in our preview article [17]. For the evaluation of the FPTPQ, we proposed (section 4.4), the holistic matching algorithm FlexPrefTreeMatch which is an improved version of TreeMatch [18]. The matching of the query is paired with the calculation of each solution weight, utilized to determine the best solutions. Illustrations (Section 5) and experimentations (section 6) are made, in order to show the effectiveness of the FPTPQ and the algorithm FlexPrefTreeMatch: that is to show how the complete solution set is returned, and how the useless solutions caused by \* and // are avoided.

## 2. Preliminaries

In this section, we define the concepts related to XML databases querying process, that are useful to understand this work.

### 2.1. XML and heterogeneous database.

An XML database is a collection of XML documents. An XML document [19] Consists of a set of hierarchical tags that describe the data they contain. An XML document must be well-formed, that is it must be in accordance with XML recommendations. The structure of XML documents is fundamentally tree oriented, so it can be modeled as a rooted tree  $t = (N, E)$  where N is a set of nodes labeled with the tag name for internal nodes, data or attributes for leaf nodes. E is the set of edges, each one represented as a couple  $(n_i, n_j) \in (N \times N)$  that connect a node  $n_i$ , to  $n_j$ . DTDs or XML schemas when used, impose a structure/format to the XML database documents. They define the nature and the type of the elements that may be included in a valid document of the collection, and the way these elements are nested. An XML database can therefore be represented as a forest. XML documents validated against a DTD or XML Schema are said to be "Valid", The corresponding database object instances usually have uniform structure and same tag names, and can be considered as homogeneous. In the other hand, the integration of many XML data sources with different structure/format may result to a heterogeneous database. XML database heterogeneity can be interpreted in different way. In this paper, we are interested in vocabular heterogeneity and structural heterogeneity [20] . We talk of vocabular heterogeneity when the same object instances are expressed by different tag names, synonyms, abbreviations or other languages borrowed words. Structural heterogeneity refers to the fact that the same information instances may be located with more than one paths. Heterogeneity makes it difficult to write satisfiable query, using the common XML query languages. As XML document, XML queries can be modeled as a combination of trees, called Tree Pattern Queries.

### 2.2. XML query and Tree Pattern Query (TPQ)

To extract specified data from an XML database, many query languages have been developed. The most famous are XPath [21] and XQuery [22]. A common feature of these languages is a possibility to formulate paths in the database tree or forest. Such a path is a sequence of tree nodes from the root to the searched element occurrences. Regular expressions of XML query languages provide valuable methods for paths specifications based on XPath, and some formular to join the path images. XML queries can be translated into one or many trees, called Tree Pattern Query

[13] (TPQ) with the respect of the structure and the complexity of the query. Each TPQ is used to represent a useful fragment of the principal query. A TPQ is a tree  $t_Q = (N_Q, E_Q)$  where  $N_Q$  is a set of nodes containing the root of  $t_Q$  and  $E_Q$  is the set of edges represented as couple  $(n_{Q_i}, n_{Q_j}) \in (N_Q, E_Q)$  that connect the query nodes  $n_{Q_i}$  to  $n_{Q_j}$ . Several TPQ models exist in the literature with different kind of features, some are more expressive than others, according to the operators and relations they offer. One of the first tree pattern queries is the Tree Algebra for XML Tree Pattern Query (TAXTPQ) [23]. Its main features are the ancestor-descendant (A/D) and parent-child (P/C) relations [24]. If had brought the basics features for the other TPQ. The TAXTPQ is too rigid and the user has to master the structure of the database in order to adapt his request accordingly. The absent of only one edge or node in a potential solution tree prevents it from being in the final result of the matching, even if the candidate subtree is "almost perfect" [13]. As a response, Chen et al proposed the Generalized Tree Pattern (GTP) [25], a TPQ that enable some edges or nodes to be optional, by associating a mandatory/optional status to them, in other to increase the possible matched subtrees in the database tree. More than a limitation Through the Annotated Pattern Tree, Papisos et al [26] allow the addition of a specification to an edge  $(u, v)$ , which specifies how many matches to node  $v$  are to be obtained from each match with node  $u$ . Moreover, Lu et al [18] have proposed the Extended XML tree pattern that enable more relaxation with the wildcard node "\*". The wildcard can match any single tree element. It is usually used when the associate's element is unknow or is not important. The risk with this operator is the abundancy of the corresponding matching element alongside with the query solutions. In fact, it uses to originate many useless solutions that imply costly filtering.

Sometimes, the user or the potential mediator system could have some additional elements that may reinforce the satisfiability of the query, when there is no assurance that the first (principal) searched element will be available. In this context, rather than using the wildcard and give a totally freedom to the matching algorithm to select anything, it is better to insert the additional elements as replacement items that will help to produce alternatives solutions which are near to the user initial needs. But the existing tree pattern does not allow it. The optional operator "?" is a beginning of a solution, but allow only an element to be optional [26]. With the logical operator OR and XOR, Izazi et al [27] allow the selection of one of two proposed sub path that finished at different leaf nodes. They don't integrate the fact that only some internal nodes/paths may need some replacement items. For example, if we consider the two documents of Table 1, we see that the starting *time* of the two documents courses are inside the tag *start*, while the parent tags are different, *hours* for the first document, and *times* for the second. The TPQ of Izazi et al does not allow to have *hours* and *times* as items of the same node.

### 2.3. Exact query, Flexibles and preference queries languages

Exact query is considered as query where all the constraints most obligatory be satisfied. All the constraints are considered as filter. The none satisfaction of all the conditions by a solution disqualify it from being in the final result set. Flexible queries are relaxed, allow soft matching and favour more solutions possibilities. Queries containing the operators "?" and "\*" can be consider as flexible queries, since the first accept the fact that an element of the image tree can be absent and the second can be matched by any single database node (when different database nodes can be the image of the same query node). The wildcard is responsible of big solution set. At the end of the matching, all the solutions of such queries are considered to be of equal weight and are returned to the users without any ranking, obviously with all the useless solutions.

Preference queries, more than flexible queries also include operators useful for the calculation of the best solutions. As an example we consider a tourist looking for a room in a luxury hotel with a swimming pool and beach. Even if hotels with a beach are difficult to find, they are more likely to be preferred by the users. If it is possible to consider the beach as optional (in a flexible query), in the final query solution set, all the hotels' rooms will be returned in a random order, with no consideration of whether they have a beach or not. The best formulation of that query is "I want a room in a luxury hotel, with a preference for hotels that have a swimming pool and a beach". In this case, it is better to use a preference query language [28] [29] [30] or the bipolar query languages of our previous works [17] [31] [32] which make it possible to write queries with two parts: a first part containing the obligatory constraints that must necessarily be satisfied and a second part containing the elements of preferences whose availability magnifies the corresponding solutions. At the end of the matching, the solutions that are not dominated, those which satisfy all the obligatory constraints and incorporate more elements of preferences than the others, are returned. Pareto's dominance concept through the skyline [33] [34] [35] [36] operator is commonly used to compare the solutions. But more is still to be done to make XML query languages and TPQ more flexible even though they integrate preferences. The non-existence of a database schema may induce the representation of an information instance by different words, in the same collection of documents. Moreover, in the context of a heterogeneous database, XML query language must allow the integration of all the different representations of an information in the query (for focused flexibility) while minimizing the utilization of the operators \* and // which cause a huge number of useless solutions.

### 3. Flexible Preferences Tree Pattern Queries (FPTPQ)

#### 3.1. motivations

Vocabulary heterogeneity and structural heterogeneity make it difficult to write a query that captures all the different representations of the database tags. Several issues are considered:

**Issue 1:** The database may contain some words (tags) and their synonyms, abbreviations, other language borrowed words. For example, in the database of a country's labor ministry, a *football club* can be considered as an *enterprise*, whereas in the database of the Football League the word used for the corresponding tag is "*club*"; For those who are used to American language, the collective sport where the ball is moved only with foot is called *soccer* rather than *football* in a European country. Another example is the utilization of the words *option* and *specialty* to indicate the area of study. The same object instance may have different paths in the database. As an example *root/course\_listing/section\_listing/start* and *root/course/time/start* in *merged\_wsu-uwm.xml*. In this issue, the listed elements are similar, and thus have the same values. We only need to precise the different word/path instances no matter the order. The purpose is to select all the available solutions. Here we introduce the expression **Flexible Node (FN)**. A FN is a node which has more than one label item. Its label items are separated by the operator "|". In the query S2 of **Figure 1**, the node whose label is *(country | location)*, is a flexible node. The purpose is to maximize the satisfiability of the constraints associated to that node.

**Issue 2:** This issue concerns a user preference node label which has multiple representations or paths in the database. Here, the objective is to maximize the satisfiability of the preference constraints. The list of the similar elements should be listed alongside with the preference operator. The associated node is called a **Flexible Preference Node (FPN)**. A FPN is a preference node which has more than one label. The preference operator remains "!" as proposed in our previous work [17]. In the last query of **Figure 1**, the node whose label is *(proc1|proc2)!*, is a FPN.

**Issue 3:** In some circumstances as e-commerce, the attributes (constraints) for alternative solutions can be added as replacement items alongside with preference order, in the same query. So that, if the ideal (user first choice) solution is missing, alternatives ones will be selected. The established preference order is used to calculate the best solutions among all the available ones. For example, in the last query, the node whose label is  $(Proc1 | proc2)>!$  is an Ordered Flexible Preference Node (OFPN). In an OFPN, the node label items are classified in ascending or in decreasing order, thus  $(Proc1 | proc2)>!$  Can also be written as  $(Proc2 | proc1)<!$ . In this node, the label item *proc1* is more preferred than *proc2*. The solutions whose tree has *Proc1* will have a greater preference value than those whose tree has *proc2*.

The different types of nodes can be combined inside the same query.

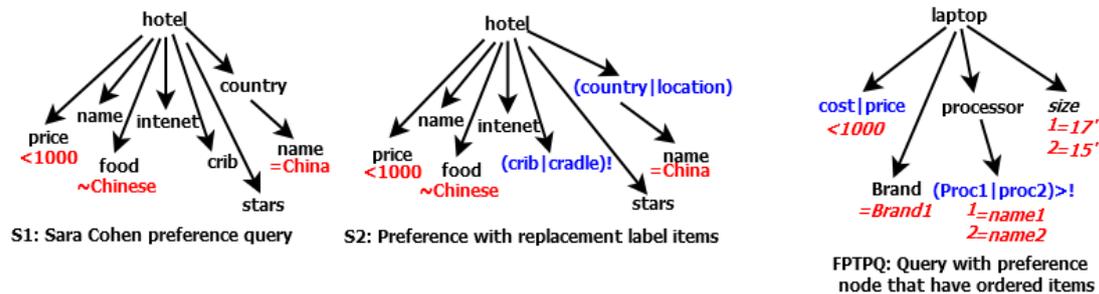


Figure 1. Sara Cohen preferences Vs flexible preferences

### 3.2. Flexibles Preferences Tree Pattern Queries (FPTPQ): language.

It is already possible to propose many preference alternatives for data values (processors name, screen size, etc.) using Sara Cohen preference language [30]. None of the existing TPQ allow to do so for internal node labels. In this section, we present the Flexible Preference Tree Pattern Query (FPTPQ), a model of Tree Pattern Query which gives the possibility of proposing many paths or/and many label items (words) for some query node, alongside with on-demand ordering operators which are useful for the calculation of solutions preference weights. This is another way to enable more flexibility at the level of the query writing module. Since the user can be the one to propose replacement items, the associated results are likely to satisfy him and save the querying systems from multiples query execution and useless filtering operations. To express preferences inside the query, we extend the *prefSXpaths* language proposed in our previews work[16]. As another example, let's consider the queries of Figure 1, from Sara Cohen et al [30]: In S1, a tourist (Sam) needs a hotel to stay at when attending a conference in China. Ideally, Sam would like a cheap (at most 1000 RMB per night) hotel in China. Since he would like to taste the local food, he would like Chinese food to be served at the hotel. Sam needs an Internet connection, to keep in touch. Finally, Sam will be bringing his wife and new baby, and so will need a crib. Despite it is represented as a preference query type, in a heterogeneous database, the *crib* can be expressed by its synonym *cradle*. In the place of the tag *country*, the database designer could have used the word *location*. In order to maximize the satisfiability, synonyms or equivalent word (*cradle* and *location*) can be added as replacement item for the label *country* and *crib* in order to have query S2. In some cases, the main label item, better contribute for user satisfaction than the other replacement items.

In a preference query, a priority/preference order can be set among the replacement items of preference nodes. In this case, the operators “<” (resp “>”) are added before “!” to indicate that the replacement item are listed in ascending (resp in decreasing) order of preference.

Figure 2-b show how to represent a FPTPQ (of Figure 2-a) that has FN and FPN, using their associated variable. Figure 2-d do the same for the FPTPQ of Figure 2-c which has both FN and OFPN. All the other nodes constraints are expressed as in the Extended Tree pattern query [18]. The preference operator still remains “!”. We can now have preferences nodes that have replacement items. The binary operator “|” is used as separator of multiple paths or label items.

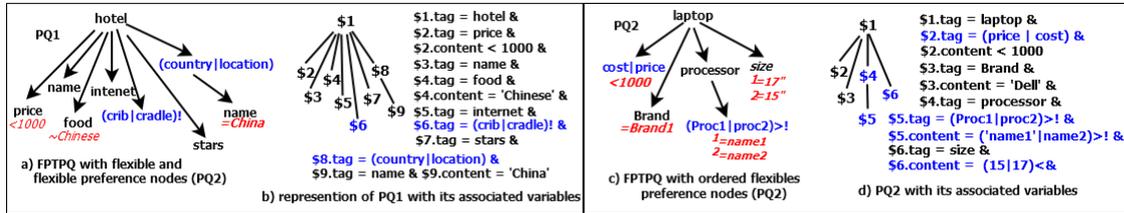


Figure 2. Expressing the FPTPQ with variables.

As label of a preference query node,  $(p_1|p_2|\dots|p_n)!$  means that in the absence of  $P_1$ , the items  $P_2, \dots, P_n$  will equally replace it. Figure 3-a show the tree representation of the query  $A|(D|E)!/F|B!/C$ . In this query, node 2 is a preference node,  $D$  and  $E$  are both label items of a preference node, with the same weight. In the absence of  $D$ ,  $E$  will equally be considered. In the query of Figure 3-b, the preferences nodes items are ordered. This means there are classify in increase order like  $(C|D)<!$  or in decreasing order like  $(G|H)>!$ . Here, the label item  $D$  is more valuable than  $C$ . The solutions whose tree carry  $D$  will be more valuable than those whose tree carry  $C$ . In the query of Figure 3-b, the item  $H$  can replace  $G$ , but with a lower preference value.

### 3.3. Assignment of preferences values to preference node items

We need to know that preferences values are assigned to preferences node items, according to the type of preference node there are associated to. All the items of FPN are assigned the default preference value which is “1” like in the example of Figure 3-a. For OFPN that carry a lower operator (<), the label items from the first, are assigned preference values respectively from the integer 1 to  $N$ , where  $N$  is the number of items. When the preference node has the operator “>”, preference values are assigned to its label items from “ $N$ ” to “1”. In Figure 3-b, the node 3 has the operator “<” which mean its label item are classify in ascending order. So, the label item  $C$  has “1” as preference value and the label item  $D$  has “2” as preference value. For the node 6, the label items are classified in descending order; its values then start from “2” for  $G$ , to “1” for  $H$ .

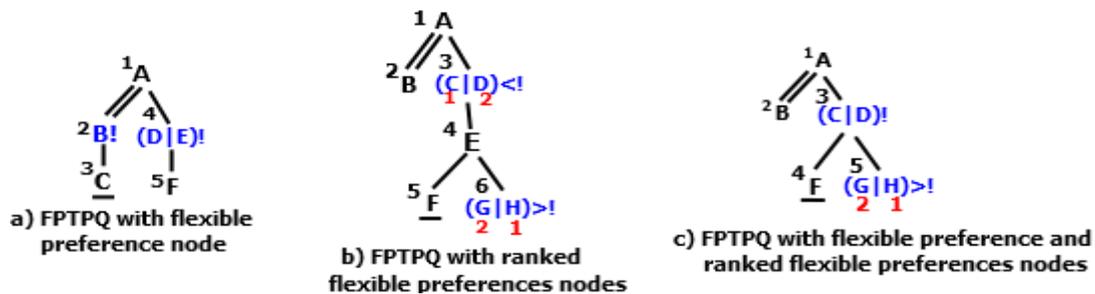


Figure 3. FPTPQ with different types of preference nodes

## 4. Evaluation of the FPTPQ: FlexPrefTreeMatch

In this section, we present an evaluation approach of the FPTPQ. The purpose is to minimize the FPTPQ by removing all the labels items that do not appear in the database, then the minimized

FPTPQ is matched with an extended Dewey based index using a holistic algorithm that is based on treeMatch [18]. During the matching, the preference value of each solution is also calculated.

#### 4.1. The algorithm TreeMatch

We used the algorithm TreeMatch as the backbone of our proposed algorithms because it is able to optimally (in term of I/O complexity) processed the Extended Tree Pattern Query, which is one of the most featured and flexible TPQ, since it allows negation function, order-based axis and wildcards. In fact, treeMatch has one of the larger optimality classes in terms of input-output. The optimality class of a tree pattern matching algorithm represent the set of TPQ it is capable of optimally evaluate. Since the FPTPQ integrate replacement items, preference operator, and ordering among items, some modifications have been made on TreeMatch for its evaluation. The evaluation process of a FPTPQ start with some precomputing which consist of the minimization of the FPTPQ label items, node's identification, attribution of preference values to each node item, the determination of all the paths associated to each query leaf node, alongside with their corresponding preference values.

#### 4.2. Precomputing: FPTPQ minimization, node identification and calculation of preference values of all paths.

Before the evaluation, the FPTPQ label item are minimized, using the database tag names list. In fact, all the label items that do not appear in the database tag list are removed from the FPTPQ. The FPTPQ query nodes are then numbered, using integer values that are used to identify them.

Because a node label can have many items, it is not appropriated to identify them with their label or with one of their label items, since each item may be manipulated separately. For example, if a node label is "A|B", A|B or A are not more appropriate identifier for it, since A|B is too long and each database index list contains the occurrences of only one object. Integer identifiers are associated to the node with the respect of their position. The Figure 3 show how the nodes of different types of queries are numbered. The couple (i, l<sub>j</sub>) is used to refers to the label item (name) l<sub>j</sub> of the node whose id is j. During the evaluation process, the preference value of each solution is progressively calculated using the preferences values associated to each query branching node that are stored in the preference locate match table. The description of this table is shown in section 4.3, alongside with the procedure that need it.

#### 4.3. Used Data structures

The inputs of the algorithm treeMatch are a *FPTPQ* and the lists  $T_q$  associated to each label item  $q$  of the minimized query.  $T_q$  contains the extended Dewey label of all the tag occurrences whose name is  $q$ .  $e_q$  is used to refer an element of a  $T_q$  list.  $Cur(T_q)$  is used to denoted the current element pointed by the cursor of  $T_q$ . The procedure  $advance(T_q)$  is used to advance the cursor of  $T_q$  to the next element. Like for the algorithm TreeMatch in [18], a set  $S_i$  is associated to each query branching node. Here, "i" no more the branching node label, but its id. Each element  $e_q$  of the set is a triplet (label, intVector, outputList) where *label* is the extended Dewey label of  $e_q$ . *intVector* is a vector of integer whose size is equal to the number of descendants of  $q$ . Compared to *bitVector* used by *TreeMatch*, *intVector* has many rules. Its first rule is to tell whether  $e_q$  has the proper children or descendant with the respect of the query (as *bitVector* for *TreeMatch*). Its second function is to save the current (partial) preference value of the potential solutions. Indeed, each integer of intVector represent the preference value of all the its associated descendant subtree. Given a child node  $q_c$  of  $q$ , let consider  $intVector(e_q)[q_c] = V: V >= 0$  if and only if there is a

database element  $e_{qc}$  such as  $e_q$  and  $e_{qc}$  satisfy the query relationship between  $q$  and  $q_c$  and  $V$  is the preference value of the subtree rooted by  $q_c$ ;  $V = -1$  if not. *OutputList* contains the elements that potentially contribute to final query answers.

#### 4.4. Algorithms flexPrefTreeMatch and associated procedures and functions.

##### 4.4.1. The principal algorithm: flexPrefTreeMatch

*Line 1* locate the first match label of each query leaf node. If a leaf node is a FN then all its label items first matches would be located and the one with the minimum match label will be first processed. Now the function *prefGetNext* select among all the query leaf nodes, the one which is going to be processed (the one that has the item with the minimum match label). *prefGetNext* return a couple  $(f_{id}, f_{act})$  where  $f_{id}$  is the *id* of the next leaf node to be processed, and  $f_{act}$  is the label item of  $f_{id}$ , that has the minimum current *matchLabel*. The purpose of line 4 and 5 is to insert the potential matching element inside the outputList of *NDB*( $f_{id}$ ). After the treatment of  $f_{id}$ , the cursor of  $T_{fact}$  is advanced to the next element. Line 7 update the set encoding and line 8 locate the next matching element to individual root to leaf path. Finally, line 9 do the appropriate update for the final solutions; for FFTPQ which has preferences, the weighted solutions table is constructed.

##### Algorithm1: flexPrefTreeMatch

```

1: PrefLocateMatchLabel(Q);
2: while (¬end(root)) do
3:    $(f_{id}, f_{act}) = \text{prefGetNext}(\text{topBranchingNode})$ ;
4:   if  $(f_{id}$  is a return node)
5:     addToOutputList(NAB( $f_{id}$ ), cur( $T_{fact}$ ));
6:   Advance ( $T_{fact}$ ); // read the next element in  $T_{fact}$ 
7:   prefUpdateSet( $f_{id}, f_{act}$ ); // update set encoding
8:   prefLocateMatchLabel(Q); // locate the next element with matching path
9: emptyAllSets(root);

```

##### 4.4.2. PrefGetNext and other procedures and functions used by flexPrefTreeMatch

The index used by flexPrefTreeMatch remains the  $T_q$  lists, where each  $T_q$  list contains the Extended Dewey label of all the database occurrences of  $q$ . A list  $T_q$  is visited only if  $q$  is the label item of a query leaf node. Initially a pointer is positioned at the first element of  $T_q$ . *PrefLocateMatchLabel* is a very important procedure, its purpose is to locate the first elements whose path match one of the individual root-leaf query paths with the respect. In spite, for FFTPQ, many paths can be associated to a query leaf node. The preference locate match table is used to associate to each query leaf id, all its corresponding root to leaf paths in other to facilitate the matching. Table 2 associate to each leaf node item of the query of Figure 5-b all its corresponding root to leaf paths. For leaf node which are flexibles, all its label replacement items have to be considered during the matching. Let consider a leaf node whose label items are  $q_1..q_n$ ; during the matching, if the current matchLabels of these items are respectively  $e_1..e_n$ , then, the label with the minimum (by lexicographical order) will be first selected to make sure that the evaluation is being made by lexicographical order.

The function *minMatchLabel*( $n$ ) return the minimum of the current matchLabels of all the label items of the node  $n$ . The function *minItem* ( $n$ ) return amount the label item of  $n$ , the one that has the minimum current match label. That is  $n_{imin} \mid \text{cur}(T_{n_{imin}}) = \text{minMatchLabel}(n)$ .

Give the current matchLabel  $e$  of the node label  $n$ ,  $flexMB(n, b)$  return all the matchLabel of  $n$  that cover (are ancestors or parent of)  $e$ .  $flexMB$  help to make sure that if many occurrences of the branching node ( $b$ ) carry  $e$ , we always start to process the deepest one, since the evaluation is bottom-up. The function  $minValue(v)$  return the minimal value of the  $intVector$   $v$ . The function  $prefValue(e_q, e_q)$ , return the preference value of the path that link  $e_{q'}$  to  $e_q$ . It is equal to the preference value of  $e_q$  added to the sum of all preference node value between  $e_{q'}$  and  $e_q$ . This function is used by the procedure  $updateAncestorSet$  to set the integer value of the nearest ancestor branching node set. Assume that  $q$  is a branching node and  $q_i$  is it children,  $intVector(e_q, e_{q_i}) = prefValue(e_q, e_{q_i})$ . Like the matching process, the calculation of the final solution preference value (weight) of a node is bottom-up. This mean that, during the matching, the preference value of a branching node matchLabel is equal to the sum of all the preference values of its children subtree images. The function  $flexSatisfyTreePattern(e_{q_i}, e_q)$  test whether the document element  $e_{q_i}$  is covered by the branching node matchLabel  $e_q$ ; it return true if  $intVector(e_q)[e_{q_i}] >= 0$ ;

### Function $prefGetNext(n)$

```

1: if (isLeaf(n)) then
2:   return (n,  $n_{min}$ ) |  $n_{min} = minItem(n)$ 
3: else
4:   for each  $n_i \in NDB(n)$  do
5:     ( $m_i, f_i$ ) =  $prefGetNext(n_i)$ 
6:     if ( isBranching( $n_i$ ) and  $\neg empty(S_{n_i})$  )
7:       return ( $m_i, f_i$ )
8:     else  $e_i = \max \{p \mid p \in flexMB(n_i, n)\}$ ;
9:   end for
10:   $max = \maxArg_i \{e_i\}$ ;
11:  for each  $n_i \in NDB(n)$  do
12:    if ( $\forall e \in flexMB(n_i, n): e \notin ancestors(e_{max})$ )
13:      return ( $m_i, f_i$ );
14:    endif
15:  end for
16:   $min = \minarg_i \{f_i \mid f_i \text{ is not a return node}\}$ ;
17:  for each  $e \in flexMB(n_{min}, n)$ 
18:    if ( $e \in ancestors(e_{max})$ )  $updateSet(S_n; e)$ ;
19:  end for
20:  return ( $m_i, f_{min}$ );
21: end if

```

### Function $flexMB(n, b)$

```

1: if (isBranching(n)) then
2:   Let  $e$  be the maximal element in set  $S_n$ 
3: else Let  $e = minMatchLabel(n)$ ;
4: Return a set of elements  $a$  that is an ancestor of  $e$  such that  $a$  can match node  $b$  item in the path solution of  $e$  to path pattern  $p_n$ 

```

**Function prefSatisfyTreePattern( $e_q, e_{qi}$ )**

- 1: **if** ( $\text{intVector}(e_q, e_{qi}) \geq 0$ ) **return true**;
- 2: **else return false**;

**Procedure prefUpdateSet( $q; e$ )**

- 1:  $\text{prefCleanSet}(q, e)$ ;
- 2: **add**  $e$  to set  $S_q$ ;     //set the proper  $\text{intVector}(e)$  using the locate match table.
- 3: **if** ( $\neg \text{isRoot}(q) \wedge (\text{minValue}(\text{intVector}(e)) \geq 0)$ ) **then**  
     $\text{prefUpdateAncestorSet}(q)$ ;

**Procedure addToOutputList( $q, e_{qi}$ )**

- 1: **for each**  $e_q \in S_q$  **do**
- 2:     **if** ( $\text{prefSatisfyTreePattern}(e_{qi}, e_q)$ ) **then**  $\text{outputList}(e_q)$ .  $\text{add}(e_{qi})$ ;

**Procedure prefCleanSet ( $q; e$ )**

- 1: **for each** element  $e_q \in S_q$  **do**
- 2:     **if** ( $\text{prefSatisfyTreePattern}(e_q, e)$ )
- 3:         **if** ( $q$  is a return node)
- 4:              $\text{addToOutputList}(\text{NAB}(q), e)$ ;
- 5:         **if** ( $\text{isTopBranching}(q)$ )
- 6:             **if** (there is only one element in  $S_q$ )
- 7:                 output all elements in  $\text{outputList}(e_q)$ ;
- 8:             **else** from the set  $S_q$  construct the weighted solution table
- 9: delete  $e_q$  from set  $S_q$ ;

**Procedure prefUpdateAncestorSet( $q$ )**

- 1: /\*assume that  $q' = \text{NAB}(q)$ \*/
- 2: **for each**  $e \in S_q$  **do**
- 3:     **if** ( $\text{intVector}(e, q) = -1$ ) **then**
- 4:          $\text{intVector}(e, q) = \text{prefValue}(e, q)$ ;
- 5:     **if** ( $\neg \text{isRoot}(q) \wedge (\text{minValue}(\text{intVector}(e)) \geq 0)$ )
- 6:  $\text{prefUpdateAncestorSet}(q')$ ;

**4.4.3. Computation of the best results**

After the matching by the function *flexPrefTreeMatch*, the *weighted solution table* is constructed using the set of the top branching node. The *weighted solution table* contains only the solutions that integrate at least the obligatory constraints. It associates to each solution its corresponding preference weight. The solutions are sorted by increase order of weight. In fact, the weight of every solution of the final *outputList* is the sum of all the integer (weight of the children subtree) of its corresponding *intVector*. The solutions are inserted in the table with the respect of their weight values in order to avoid a sorting operation, so that at the end of the insertion, the table is already sorted. The top-K best solutions (with the highest preference values) are returned to the user. The next section shows some illustrations of different FFTPQ execution process.

**5. Illustration of the evaluation of FFTPQ by treeMatch****5.1. Illustration of flexPrefTreeMatch for queries with flexible nodes.**

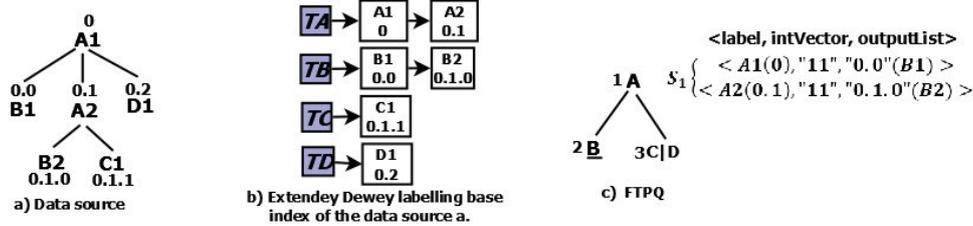


Figure 4: Illustration: evaluation of a FFTPQ that has a flexible node

Initially, the procedure *PrefLocateMatchLabel* locate *B1* for the node 2 and for the *node 3*, *C1* and *D1* are located, but *C1* is selected first because it has the minimum matchLabel. The function *prefGetNext* return (3, C) because C1 is deeper than B1. Then A2 is added in  $S_I$  with the intVector “-10” to indicate that at this moment, the second child of A2 has been found. The cursor of  $T_A$  is advanced. At the next stage, B1 and D1 are read by *PreflocateMatchLabel* and (2, B) is returned by *PrefGetNext*, A1 is added in the set with the bit vector “0-1” and outputList <0.0> (since B is a return node); the cursor of  $T_B$  is advanced. Early in stage 3, B2 and D1 are read and *PrefGetNext* return (2, B) because B2 is the deepest and has the minimum matchLabel. B is also a return node, so B2 is added in the outputList of its corresponding matchLabel ancestor element (A2). The intVector of A2 is updated from “-10” to “00”. The last element to be read is D1, and the set  $S_I$  is updated. That is the intVector of A is update from “0-1” to “00”. The query does not have any preference node, as the label items of the same flexible node, C and D are of equal values. A node (of id -1) is created to merges all the outputList of the set  $S_I$ :  $S_I = \{<-1, "0", (0.0, 0.1.0)>\}$ . The set {B1(0.0), B2(0.1.0)} is returned. Notice that B1 is also returned only because of a replacement label item (D) has been added to the FFTPQ for the node 3. Using the Extended Tree Pattern, the user may have been partially satisfied or would have been obliged to write and execute almost the same query, replacing C with D as the label of node 3.

**5.2. Illustration of flexPrefTreeMatch: evaluation of queries with ordered preference node label items.**

The Figure 5-a shows an example of a FFTPQ which has two ordered flexible preference nodes, lets describe its evaluation process using the algorithms *treeMatch*. First of void the query nodes are numbered (Figure 5-d), the labels Id are minimized (since the item C does not exist in the database, it is deleted) and the preference locate match table of Table 2 is constructed from it. The sets  $S_I$  and  $S_d$  are associated respectively to the branching nodes 1 and 4. The intVectors in blue shows the preference weight values of each path, calculated from the leaf matchLabel to the associated branching node matchLabel. In the first stage, B1, F1 and H1 (H1 is selected before G1 because it has the minimum match Label by lexicographical order) are read by *prefLocateMatchLabel*. A1 is added to the set  $S_I$  with “0.0.0” (matchLabel of B1) in his outputList, since B is the return node of the query. At this moment A1, only the first child of A1 has been read, Its corresponding intVector is “0,-1”. E1 is added to the set  $S_d$  with the intVector “0,1” whose guarantee that E1 matches all its corresponding subtree, since E1 carries two children F1 which is not a preference node item, and H1 whose preference value is 1. Therefore, the intVector A1 (the corresponding NAB matchLabel) is updated to “0,3”. The integer 3 is the preference value of the subtree rooted by D1. Later, F2 is read and it corresponding NAB (E2) is inserted in  $S_d$  with the intVector “0,-1”. When F2 is read, intVector(E2) become “0,1” since E2 has his two children and the second is the item of the preference node 6, whose preference weight is equal to “1”. Later when B2 is read, A2 is added to the set  $S_I$  with the intVector “0,1” and B2 in its outputList. Afterwards F3, G1, B3 are read in this order, followed by the insertion of

“0.2.2.0” (matchLabel of E3) in  $S_4$  and the insertion of “0.2” (matchLabel of A3) in  $S_1$  respectively with the intVectors ‘0,2’ and ‘0,4’. Then, F4 and B4 are read, E4 is added in  $S_1$  with the intVector ‘0,0’, even if it does not have a child G or H. The reason is that node 6 is a preference node, its satisfaction is not compulsory in a solution tree. A4 is added to  $S_1$  with the intVector “0,0” since it does not carry any preference node item. Finally, B5 is read, and its *NAB* A5 is added to  $S_1$  with the intVector “0,-1”. The execution is stopped, since all the elements of the input list associated to the query leaf label items have been read.

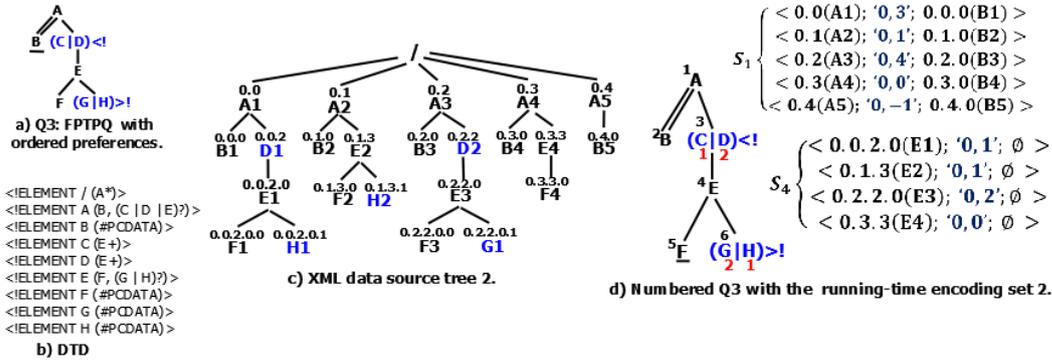


Figure 5. Illustration: evaluation of a FPTPQ with ordered preference node label items.

Table 2: Preference locates match table for the FPTPQ of Fig.6.

Leaf node id	Label items	Root to leaf paths	preference at the level of the NAB “1”	preference at the level of the NAB “4”
2	B	A/B	0	-
5	F	A/D/E/F	2	0
		A/E/F	0	0
	G	A/D/E/G	2 + 2	2
		A/E/G	0 + 2	2
6	H	A/D/E/H	2 + 1	1
		A/E/H	0 + 1	1
	E	A/D/E	2 + 0	0
		A/E	0 + 0	0

Table 3: Weighted solutions table for the example of Figure 5.

Solution id	name	weight	position
0.2.0.	B3	4	1
0.0.0	B1	3	2
0.1.0	B2	1	3
0.3.0	B4	0	4

At the end of the matching process, the weighted solutions table of Table 3. is constructed from the top branching node set  $S_1$ . We can see that the solution 0.2.0. (B3) has the greatest preference value and can then be considered as the best (most preferred) solutions. The solutions are printed per order of user preference. The possibility can be given to the user to precise the number (K) so that the top-K solutions will be returned.

With these illustration examples, we can show mayor differences between The FPTPQ and the other tree pattern queries. The replacement items enable more flexibility and more satisfiable results for query, than the existing most famous and featured TPQ like GTP [25] and the extended tree pattern query [18]. In fact, since they do not allow replacement items, the user would have written a query with only the first item of each flexible node. The consequence is that no solution would have been returned for the example of figure 3, since the database does not have any node (tag) named “C”. Without the replacement items H and D, only the solution B4 which has the lowest solution weight would have been returned. Moreover, if node 3 was not a preference node,

B4 would not have been selected. Unless all the flexible nodes were replaced by the wildcard “\*” which is responsible of a many useless solutions. This shows how the combination of replacement items and preference nodes contribute to ensure user satisfiability via the proposition of numerous solutions that remain close to the user needs.

Since none of the existing tree pattern query allow *flexibles nodes*, *flexible preference nodes* and *flexible ordered preference node* like describe in section 3.2. If We can assert that a new class of query have been created, the class of query which contain these three types of nodes, we cannot yet assert that flexPrefTreeMatch has a new optimality class. To do so, further analysis need to be done. Our main purpose was to propose the FFTPQ alongside with evaluation algorithms.

## 6. Experimentations: evaluation of FFTPQ with FlexPrefTreeMatch

In this section, we compare the evaluation results of the FFTPQ against the evaluation results of a TPQ written for the same need. We used the XML datasets DBLP (regular structure imposed by a DTD), Treebank (irregular, with no DTD), *uwm.xml* and *wsu.xml* which are two databases with different structures that was merged to produce *merged\_wsu-uwm.xml* which suffer of structural and vocabular heterogeneity. These datasets are available on the University of Washington XML Data repository [4]. In fact, the FFTPQ help to express all the desired information set through replacement items, contrary to the other TPQ like the GTP [25] or the extended tree pattern query [18] whose execution may result to an empty or incomplete solution set. Moreover, these TPQ may originate many useless solutions through the utilization of the wildcard “\*” and the A-D (“/”) operator. This experiment is used none to compare algorithms execution times, but to show how the FFTPQ and FlexPrefTreeMatch may be effectively used to express queries which produce only or the complete set of desired solution, when the replacement items are well inserted.

The percentage (P) of useless solution is calculated based on the exact number of available solutions. Thus, if the number of returned useless solution is greater, P will be greater than 100%.

### 6.1. FFTPQ contribution for the searching of all the needed solutions.

Table 4. Comparison of FFTPQ (with appropriate replacement items) evaluation results with the evaluation results of classic TPQ.

N°	XML data (.xml)	Evaluation results of TPQ with only one label		Evaluation results of FFTPQ with appropriate replacement items		Number of lacking solutions	% of lacking solutions
		Tree Pattern Query (TPQ)	Number of solutions	FFTPQ with appropriate replacements items	Number of solutions		
1	merged_wsu-uwm	TQ1: root//credit	3924	FQ1: root//((credit credits)	6036	2112	34,99%
2	merged_wsu-uwm	TQ2: root//sect	3924	FQ2: root//((sect section)	8499	4575	53,83%
3	merged_wsu-uwm	TQ3: root/course_listing [//hours/start//instructor	45 75	FQ3: root/(course_listing course) [//hours/start//instructor	8499	3924	46,17%
4	merged_wsu-uwm	TQ4: root/course/place/bldg	3924	FQ4: root/((course/place) ((course_listing//bldg_and_rm))/bldg	8499	4575	53,83%
5	dblp	TQ5: dblp/book/title	845	FQ5: dblp/(book article)/title	112454	111609	99,25%
6	dblp	TQ6: dblp/book/author	1153	FQ6: dblp/(mastersthesis book)/author	1158	5	0,43%
7	dblp	TQ7: dblp/mastersthesis/author	5	FQ6: dblp/(mastersthesis book)/author	1158	1153	99,57%

The FPTPQ through replacement items can help to completely express the user needs inside one query and therefore, reduce multiple query adjustment and execution. As query example, let’s consider a user who wants to print the credit of all the courses present in the document merged\_wsu\_uwm.xml (the merged result of wsu.xml and uwm.xml). Because the occurrences of credit are represented inside two differences tags (credit and credits), the query TQ1 of Table 4 will produce only 3924 titles over the needed 6036 that are returned by the FPTPQ *FQ1*. The same issue is caused by the query TQ2 that print 3924 section nodes over the available 8499 that are completely returned with the execution of the FPTPQ *FQ2*.

Let consider now a user who want the titles of all the books and articles of dblp.xml. With the TPQ TQ5, only the titles of books (only 845 solutions over 112454 needed) will be returned, and the user will be obliged to write another query, replacing “book” by “article” in order to obtain the other 112454 titles. Only one execution is need with the query FQ5 to produce the complete set of solutions. The last column of Table 4 show that an appropriate utilisation of replacement items in a FPTPQ has helped to select more than 99% of needed solutions that have not been selected with the utilization of the classic TPQ.

**6.2. The FPTPQ help to avoid useless solutions caused by the wildcard “\*”**

During the evaluation of a TPQ, the wildcard can be matched by any single node. Consequently, it causes many useless intermediate results and unsatisfiable solutions. When the user is aware of all the words used to express all the instances of an object (document tag), it better to use them rather than “\*”. The Table 5 shows FPTPQ (with appropriate multiple label items) evaluations results compare to TPQ where “\*” have been used in the place of the multiple items. In the document merged\_wsu\_uwm.xml, courses are now described inside two tags: *course* (from uwm.xml file) and *course\_listing* (from wsu.xml file).

Table 5: Avoiding useless solutions with FPTPQ over TPQ with wildcard “\*”.

N°	XML database (.xml)	Execution of TPQ with the wildcard (“*”)		Corresponding FPTPQ with appropriate replacement items.		Useless solutions caused by “*”	% of useless solutions caused by “*”
		TPQ with “*”	Number of solutions	FPTPQ with replacement items.	Number of solutions		
1	merged_wsu_uwm	SQ1: root/*[//hours/start]//instructor	8499	FQ3: root/(course_listing course)[//hours/start]//instructor	8499	0	0%
2	merged_wsu_uwm	SQ2: root/*[//bldg]	8499	FQ4: root/((course/place)(course_listing bldg_and_rm))/bldg	8499	0	0%
3	merged_wsu_uwm	SQ3: dblp/*title	328859	FQ5: dblp/(book article)/title	112454	216405	192%
4	dblp	SQ4: dblp/*author	716488	FQ6: dblp/(mastersthesis book)/author	1158	715330	61773%
5	Treebank_e	SQ5: //PP[//VP/IN]*/VBN	676	FQ7: //PP[//VP/IN]/(ADJP VP)/VBN	96	580	604%
6	Treebank_e	SQ6: //*[//VP/IN]*/VBN	28314	FQ8: //(PP SBARQ)[//VP/IN]/(ADJP VP)/VBN	89	28 225	31713%
7	Treebank_e	SQ7: //*[//VP/IN]/NP	435689	FQ9: //(ADJP NP)[//VP/IN]/NP	98352	337 337	343%

Replacing the wildcard “\*” of the query *SQ1* by these two tags names to obtain *FQ3* will originate only the needed solutions, since no other tag can be matched as a *course*. But the other tree pattern queries of Table 5. show how “\*” caused overabundant useless solutions, even for simple queries. In the merged document, not only the courses can be scheduled in a building (bldg). Different sections of the same course can be scheduled in different building. With the operator “\*”, a user who need only the buildings of courses may be obliged to find between other useless ones (the building of section). Moreover, preference operator (*course\_listing|course*)<!,

*course|course\_listing*>!) can be added to give a privilege to the buildings which are carried by the item (tag) “*course*”. i.e., the building of the UWM courses. In *dblp* dataset, the cited documents can be articles, improceedings, master thesis, PhD thesis and books. The query *SQ4* is written using the extended tree pattern query for a user who wants only the titles of books and articles. “\*” causes the selection 216405 (328859 - 112454) unneeded ones.

The utilisation of FPTPQ to write query *FQ5* (by replacing “\*” with *book|article*) help the FlexPrefTreeMatch to return only the needed solutions (only the titles of books and articles). Queries *SQ5*, *SQ6*, *SQ7* produce a huge number of useless solutions. This is because of the heterogeneity of the dataset *Treebank*.

Table 6. Advantages of replacement items over the A-D operator

N°	XML database	Result of TPQ with A-D (/)		Results of FPTPQ with replacement items in the place of “/”		Number of useless solutions caused by “/”	% of useless solutions caused by “/”
		Tree pattern query with “/”	Number of solutions	FPTPQ with needed replacements items	Number of solutions		
1	merged_wsu-uwm	<b>AQ1:</b> root[//hours/start]//instructor	8499	<b>FQ10:</b> root/(course_listing course) [//hours/start]//instructor	8499	0	0%
2	merged_wsu-uwm	<b>AQ2:</b> root//bldg	8499	<b>FQ4:</b> root(((course/place) (course_listing//bldg_and_rm))//bldg	8499	0	0%
3	dblp	<b>AQ3:</b> dblp//title	328859	<b>FQ5:</b> dblp/(book article)/title	112454	216 405	192%
4	dblp	<b>AQ4:</b> dblp//author	716488	<b>FQ6:</b> dblp/(masterthesis book)/author	1158	715 330	61773%
5	Treebank_e	<b>AQ5:</b> //S/VP//NP/NNP	55288	<b>FQ11:</b> //S/VP/(VP PP)/NP/NNP	6518	48 770	748%
6	Treebank_e	<b>AQ6:</b> //PP[//VP/IN]//VBN	6262	<b>FQ7:</b> //PP[//VP/IN]/(ADJP VP)/VBN	96	6 166	6423%

The chart of Figure 6 show the comparison result of the number of solutions returned by TPQ that have “\*” and FPTPQ where “\*” have been substituted by the appropriate replacement items. The FPTPQ contains only the needed and appropriate items. Line 1 and 2 of the table show that the number of solutions returned by the TPQ and the FPTPQ are equals. This is because the two needed nodes label “*course\_listing*” and “*course*” are the only possible ones which can be the image of “\*”. The query *FQ11* shows how a FPTPQ can avoid more than 61773% of useless solutions caused by “\*” (if the TPQ *SQ4* is used) in highly irregular databases.

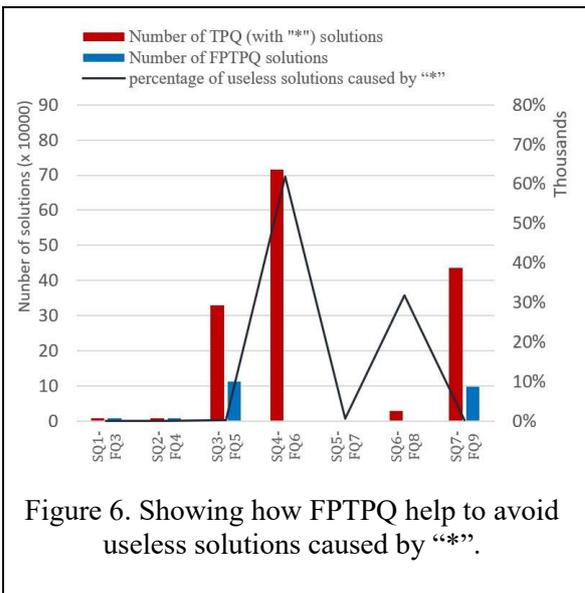


Figure 6. Showing how FPTPQ help to avoid useless solutions caused by “\*”.

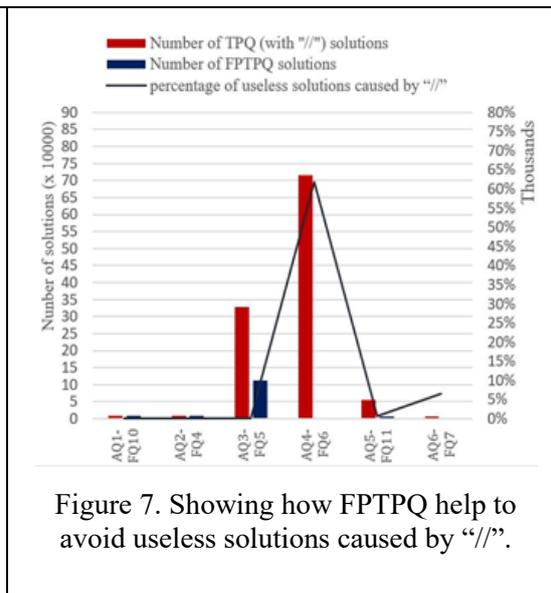


Figure 7. Showing how FPTPQ help to avoid useless solutions caused by “/”.

### 6.3. Advantages of a proper utilization of replacement items in a FPTPQ rather than A-D operators.

When the schema is absent, a distance between XML nodes (number of nodes that are in-between) may not be known, then the utilization of the ancestor-descendant operator (*//*) is justified. But an inappropriate utilization of this operator causes the selection of solutions whose paths have different lengths and different nodes labels. The execution of a query like *A//B* implies the selection of *A/B*, *A/\*B*, *A/\*/\*B*, ..., where *\** can be anything. The execution of this operator produces many useless results linked to unnecessary paths. The Table 6 shows how useless intermediate solutions can be totally avoided when the replacement items are correctly inserted inside the FPTPQ. The execution of the queries AQ3, AQ4, AQ5 and AQ6 produce a huge quantity of useless solutions. AQ4 induces 61773% (calculated based on the number of needed solutions) of useless solutions. This huge quantity of useless solutions is due to the fact that all the database instances of the return nodes are read with not enough information to filter them. In fact, there are 328859 documents in *dblp* and the operator *//* of the query AQ3 (resp AQ4) allows the selection of their title (resp of all the authors), no matter the type of document. When only the titles (resp authors) of books and articles are needed, it is preferable to use query FQ5 (resp FQ6). The execution of AQ1 and AQ2, do not generate useless solutions because all the instances of the database are needed (the number of instances needed is equal to the number of instances replaced by *//*). The chart Figure 7 shows the higher percentage of useless solutions engendered by *//*. We can see that four of the six queries produce over 150% of useless solutions. Query AQ4 engendered more than 60000% of useless solutions. The consequence of such enormous quantity of useless solutions is that it may confuse the user and make him abandon his searching.

To conclude this section, we can assert that a proper utilization of replacement items in a FPTPQ helps to express all the needed solutions, and avoid incomplete results that are caused by TPQ, since they do not allow multiple items (terms) to represent all the names of the difference tag instances. Even if the operators *\*\** and *//* remain useful, mainly when the database schema is unknown, they cause a huge number of unsatisfiable solutions. When the replacement items are known and are added in the correct places of the FPTPQ, unsatisfiable solutions are totally avoided, and only the useful solution set is returned.

## 7. Conclusion

Obtaining satisfiable queries for XML databases that have structural and vocabular heterogeneity remain an important challenge. We proposed the Flexible Preference Tree Pattern Query (FPTPQ), a TPQ that allows to have multiple items as node labels and multiple paths to locate the same query node, ordering and preference operators. The FPTPQ can be used in any XML database, to characterize in the same query the user both first choice solution and alternative ones. The FPTPQ enhances the satisfiability of both preference and non-preference nodes. For the evaluations of FPTPQ queries, we proposed the holistic algorithm FlexPrefTreeMatch which matches the FPTPQ with the database index based on extended Dewey labelling scheme, while calculating each solution's preference weight. Illustrations and experimentations verify the effectiveness of the FPTPQ and the correctness of the algorithm flexPrefTreeMatch. More types of flexibilities and preferences are being integrated in our project, to improve XML query languages.

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