

Type Based XML Projection

VLDB 2006, Seoul, Korea

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- 1 Introduction
- 2 Notations
- 3 Algorithm
- 4 Formal results
- 5 Experiments
- 6 Conclusion

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```
<bib>  
  <book year="1994">  
    <title>TCP/IP Illustrated</title>  
    <author><last>Stevens</last><first>W.</first></author>  
    <publisher>Addison-Wesley</publisher>  
    <price>65.95</price>  
  </book>  
  <book year="1992">  
    <title>Advanced Programming in the Unix environment</title>  
    <author><last>Stevens</last><first>W.</first></author>  
    <publisher>Addison-Wesley</publisher>  
    <price>65.95</price>  
  </book>  
  ...  
</bib>
```

Projecting XML Documents, Amélie Marian and Jérôme Siméon,
VLDB 2003.

Queries + document analysis to compute the pruning :

- + Algorithm is formally specified
- + Achieve high precision in pruning
- Does not take into account backward axis.
- Performances degrade in the presence of //

Indeed, for some queries, pruning the document is more expensive than actually executing the query.

Solution : Queries + **type** based optimisations

By using a DTD and a given set of queries, we can statically infer a *projector* for the set of queries and use it to prune the document.

Main advantages :

- Efficient : no additional cost at runtime
- Take into account backward axis
- Soundness : executing the query on the projected document gives the same result as on the original.
- Precision (and even exact for a large class of DTDs and queries).

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Definition (DTD)

A DTD is a pair (X, E) where X is a distinguished name (the root) and E is a set of productions of the form

$$\{X_1 \rightarrow R_1, \dots, X_n \rightarrow R_n\}$$

where each R_i is of the form :

$$a_i[\text{Regex}] \text{ or String}$$

where a_i is a unique *tag* name and *Regex* a regular expression of X_i . $\text{Names}(E)$ is the set of names occurring in E .

Definition (Projector)

For a given DTD (X, E) a type projector for (X, E) is a set of names \mathcal{P} such that :

- 1 $\mathcal{P} \subseteq \text{Names}(E)$
- 2 $\forall X_i \in \mathcal{P}$ there is at least one derivation from X to X_i in E

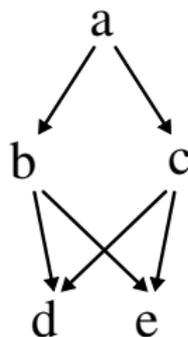
Definition (Pruning)

The pruning (or projection) of a document D valid w.r.t a DTD (X, E) with the projector \mathcal{P} is a document D' where every node not generated by a name in \mathcal{P} is erased (replaced by the empty sequence).

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```
<!ELEMENT a    (b, c)>  
<!ELEMENT b    (d, e)>  
<!ELEMENT c    (d, e)>  
<!ELEMENT d    (#PCDATA)>  
<!ELEMENT e    (#PCDATA)>
```



```
for $x in $doc/descendant-or-self::c
return $x
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```
 $X_a \rightarrow a[X_b, X_c]$   
 $X_b \rightarrow b[X_d, X_e]$      $X_c \rightarrow c[X_d, X_e]$   
 $X_d \rightarrow \text{String}$          $X_e \rightarrow \text{String}$ 
```

```
<a>  
  <b>  
    <d>LotsOfData</d>  
    <e>SuperLongString<e/>  
  </b>  
  <c>  
    <d>bar</d>  
    <e>foo<e/>  
  </c>  
</a>
```

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 $\mathcal{P} = \{X_a, X_c, X_d, X_e\}$ 
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$$\begin{array}{ll} X_a \rightarrow a[X_b, X_c] & \\ X_b \rightarrow b[X_d, X_e] & X_c \rightarrow c[X_d, X_e] \\ X_d \rightarrow \text{String} & X_e \rightarrow \text{String} \end{array}$$
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</a>
```

Definition (XPath (1.0))

Q	::=	$Axis :: Test$	$Expr$::=	Q
		$Axis :: Test[Expr]$			$Expr \text{ op } Expr$
		Q/Q			$f(Expr, \dots, Expr)$
					$Base$

$Axis$::= self | child | descendant | parent | ancestor | ...

$Test$::= tag | node() | text()

Definition (Simple path)

Q	::=	$Axis :: Test$	$Cond$::=	$SPath$
		$Axis :: Test[Cond]$			$Cond \text{ or } Cond$
		Q/Q			

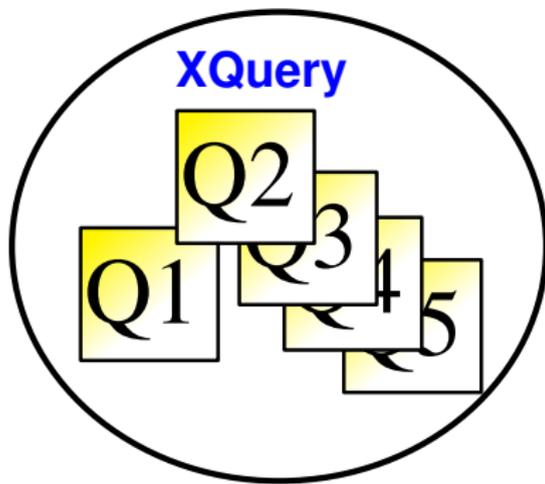
$SPath$::= $Axis :: Test$
| $Axis :: Test/SPath$

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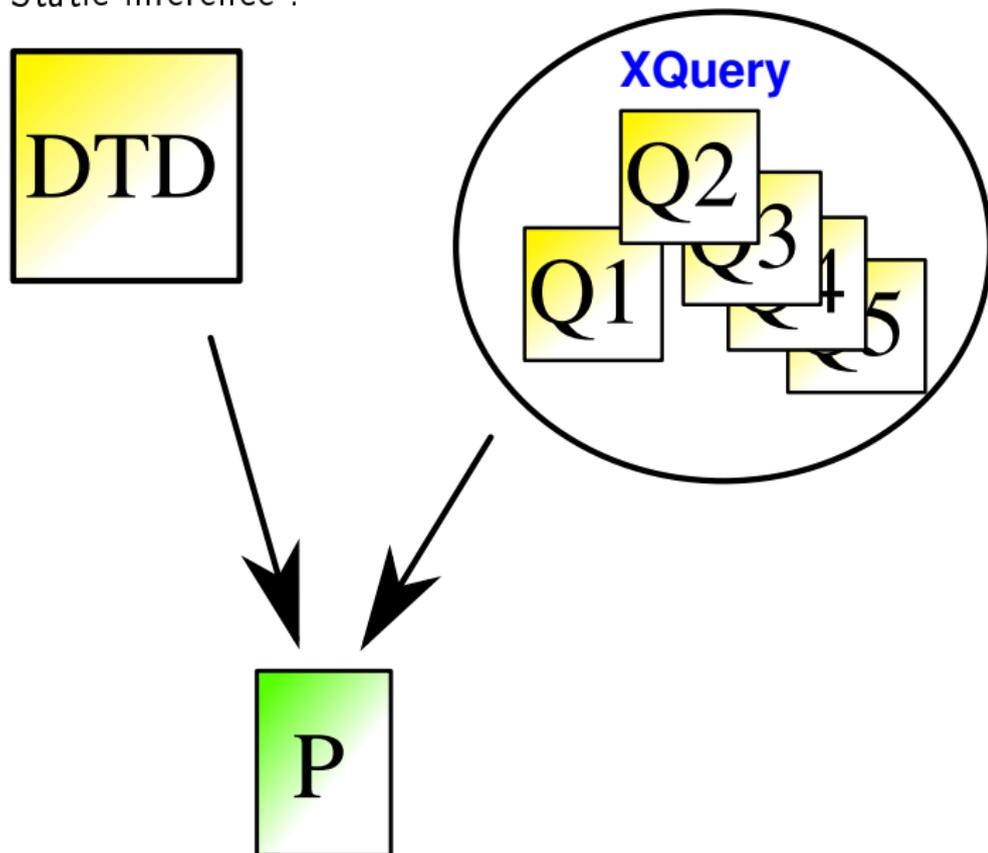
Static inference :



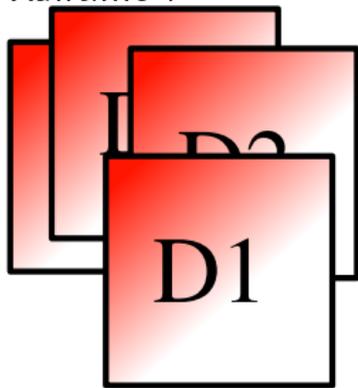
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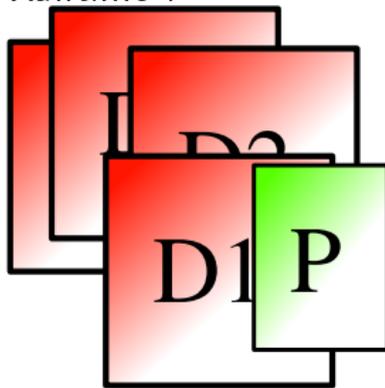
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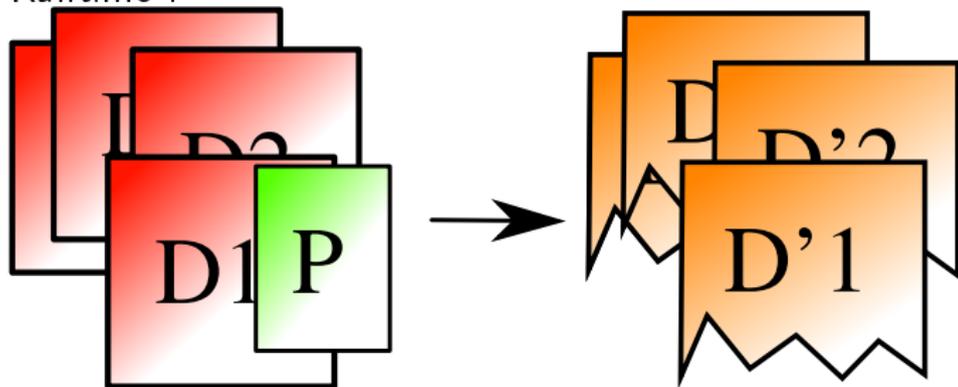
Runtime :



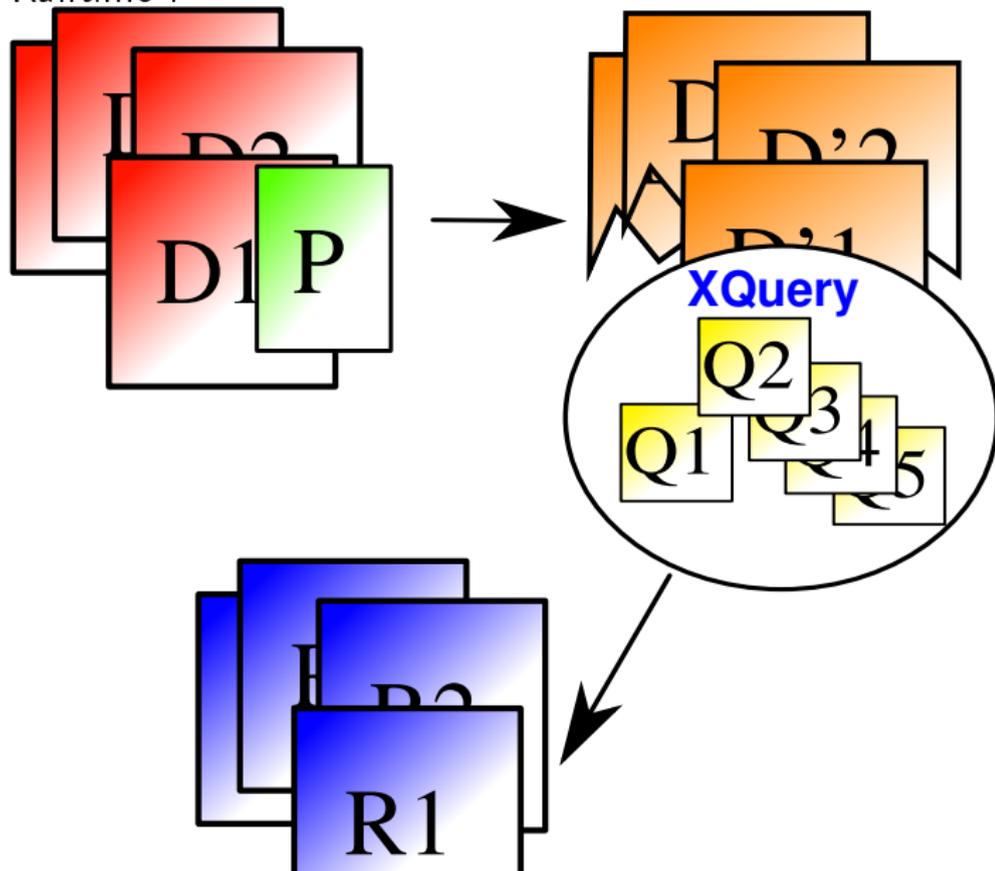
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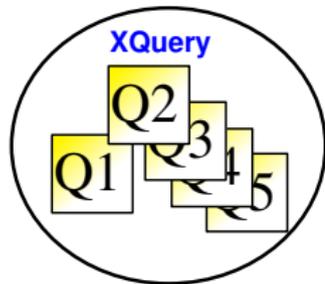


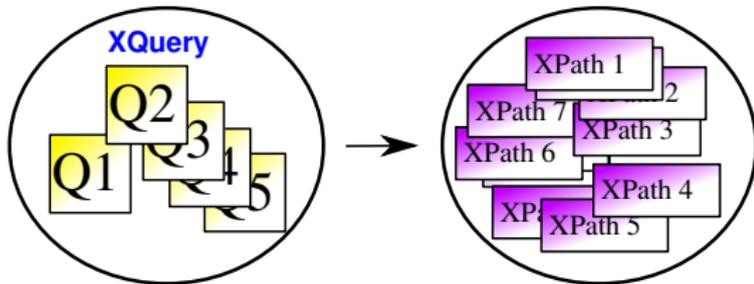
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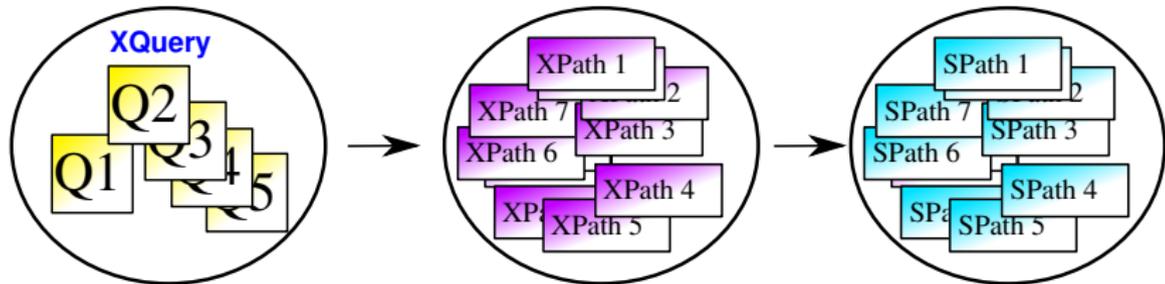


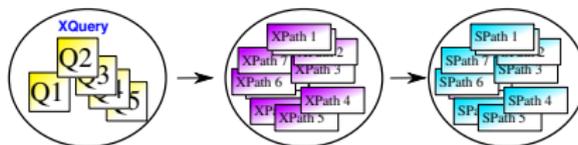
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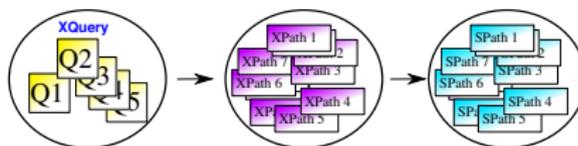






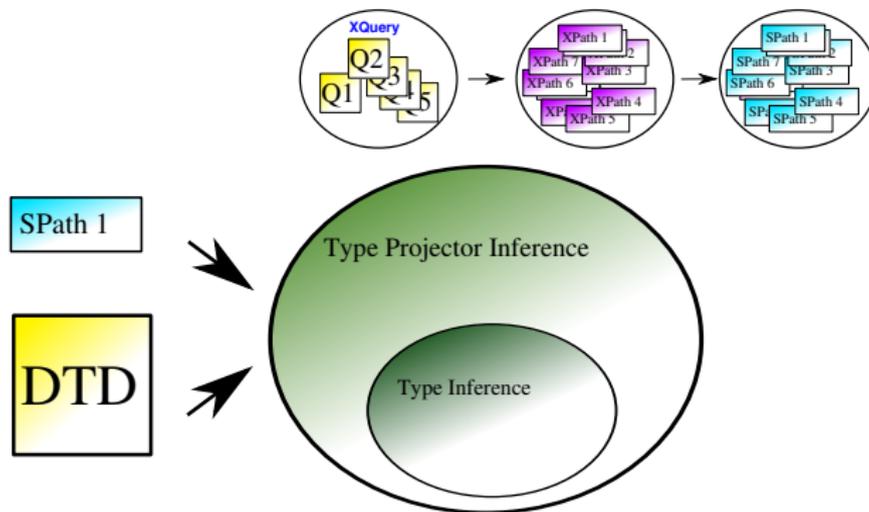


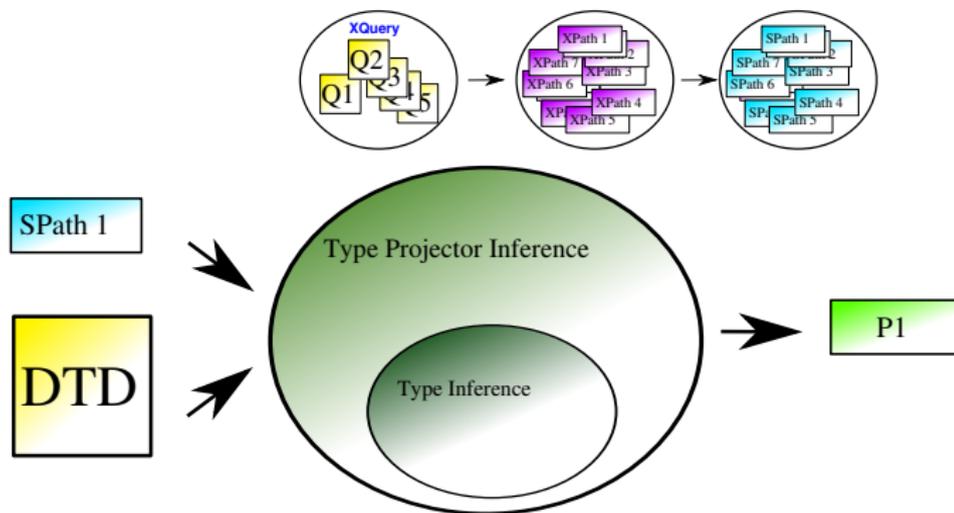
SPath 1

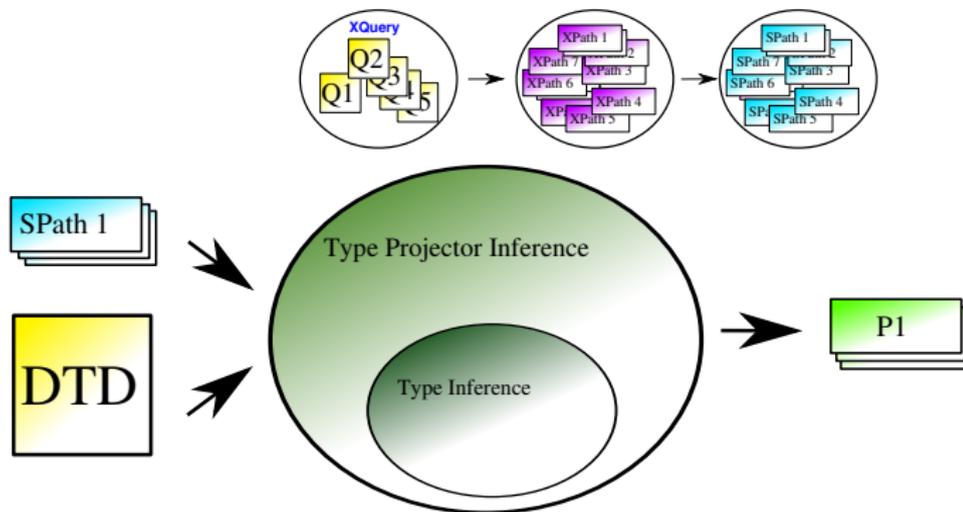


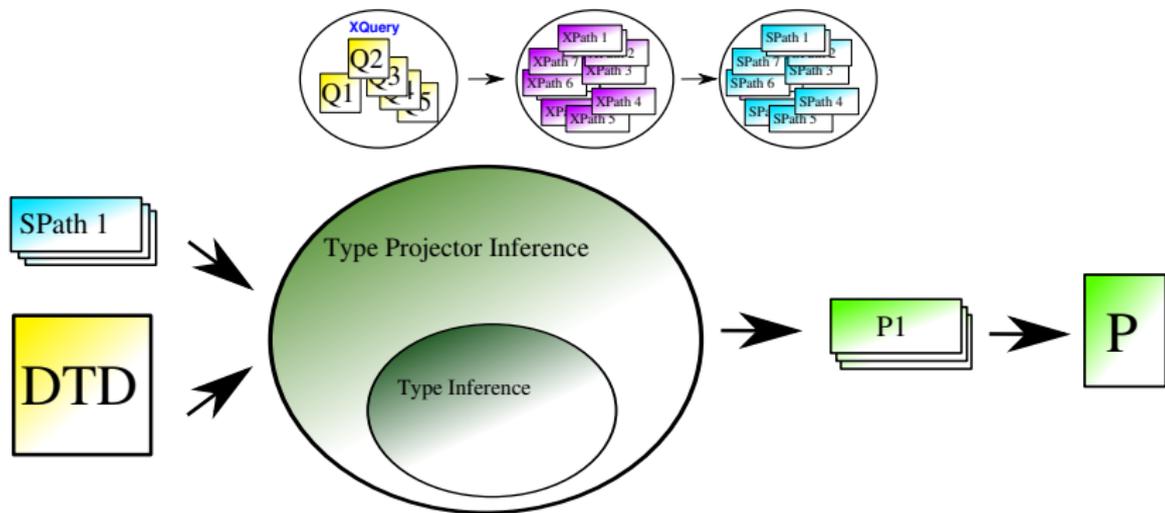
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DTD









$$\text{typeinf}(\text{DTD}, \{X\}, \text{path}) = T$$

Type T is the set of names of types of nodes in the result.

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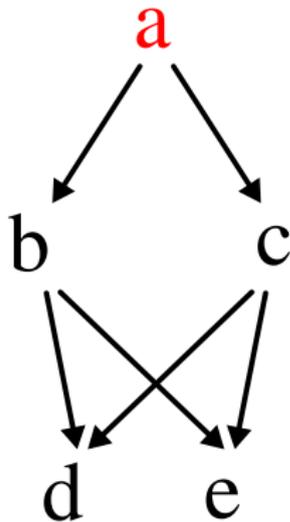
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$$\textcircled{1} \text{ typeinf}(\text{DTD}, \{X_a\}, \text{self}::\text{a}) = \{X_a\}$$



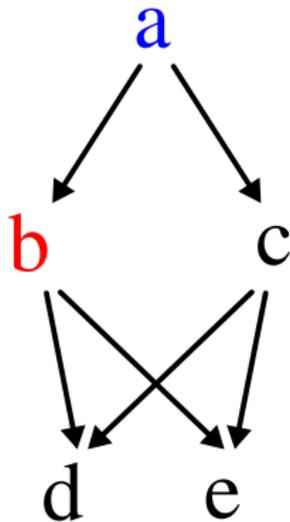
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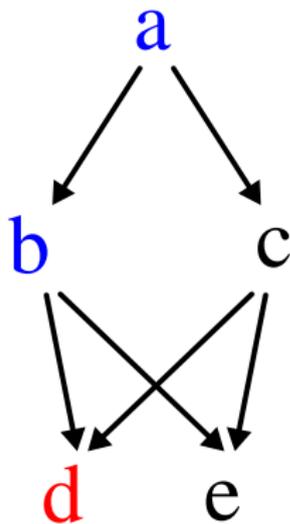
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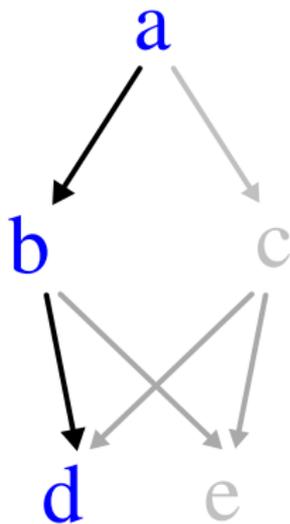
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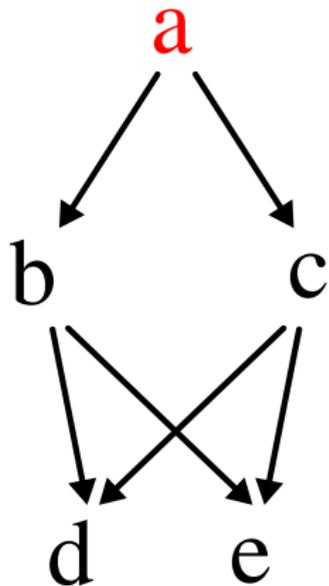
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```
/**/self::b/child::d
```

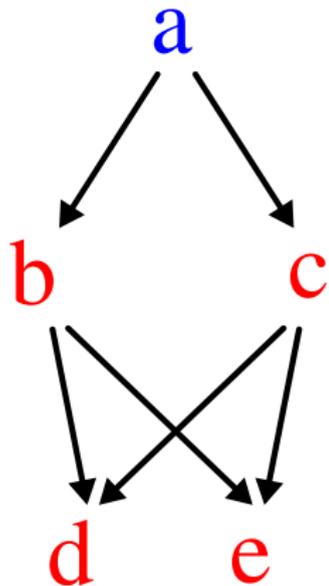
```
/**/self::b/child::d
```

❶ `typeinf(DTD, {Xa}, /**)`



```
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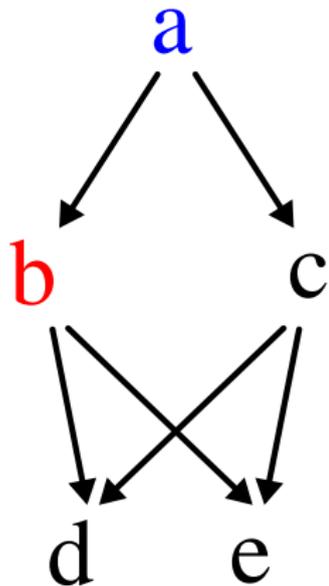
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① `typeinf(DTD, {Xa}, /**)`

① `typeinf(DTD, {Xb}, self::b/child::d) = {Xd}`

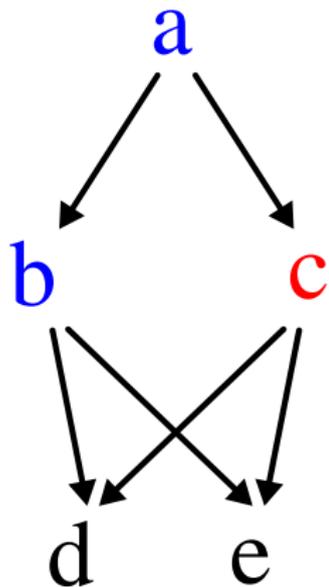


```
/**/self::b/child::d
```

① $\text{typeinf}(\text{DTD}, \{X_a\}, /**)$

① $\text{typeinf}(\text{DTD}, \{X_b\}, \text{self}::\text{b}/\text{child}::\text{d}) = \{X_d\}$

② $\text{typeinf}(\text{DTD}, \{X_c\}, \text{self}::\text{b}/\text{child}::\text{d}) = \emptyset$



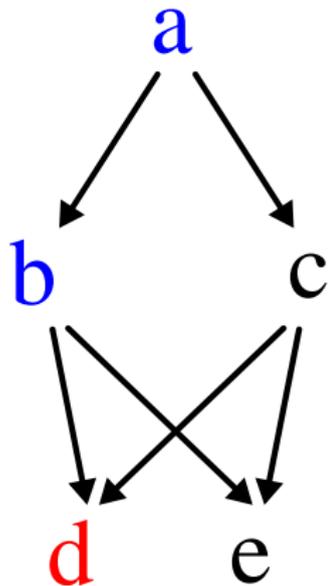
```
/**/self::b/child::d
```

① `typeinf(DTD, {Xa}, /**)`

① `typeinf(DTD, {Xb}, self::b/child::d) = {Xd}`

② `typeinf(DTD, {Xc}, self::b/child::d) = ∅`

③ `typeinf(DTD, {Xd}, self::b/child::d) = ∅`



```
/**/self::b/child::d
```

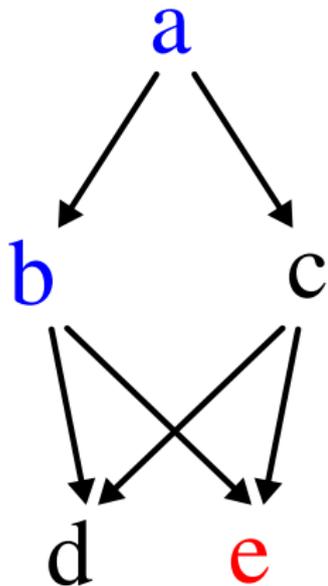
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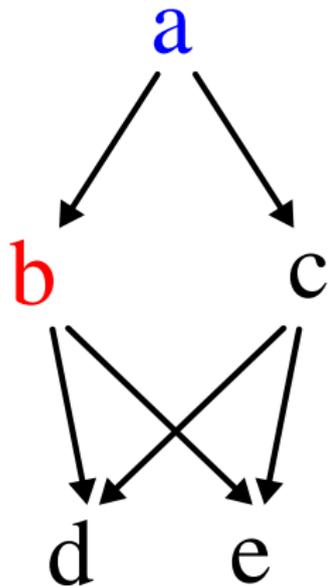
③ `typeinf(DTD, {Xd}, self::b/child::d) = ∅`

④ `typeinf(DTD, {Xe}, self::b/child::d) = ∅`



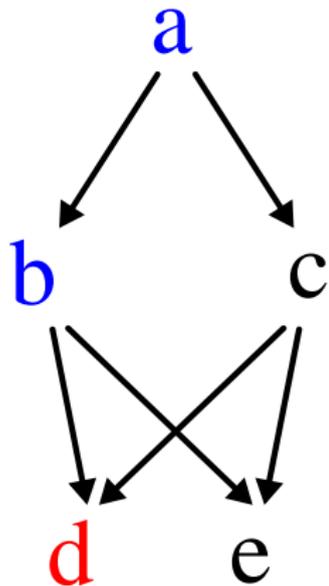
```
/**/self::b/child::d
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- 1 `typeinf(DTD, { X_a }, /**)`
- 2 `typeinf(DTD, { X_b }, self::b) = { X_b }`



```
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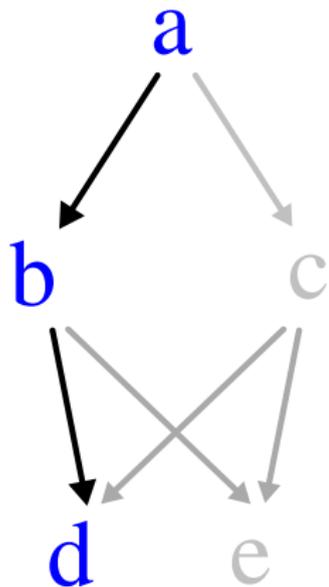
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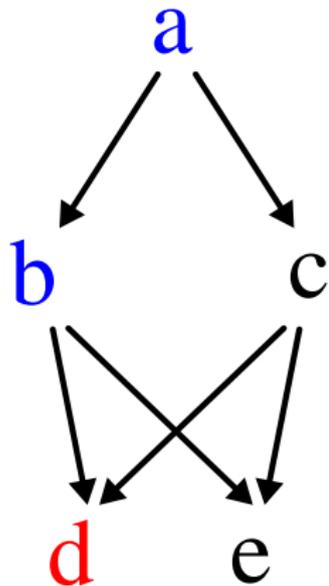
$$\mathcal{P} = \{X_a, X_b, X_d\}$$



```
/self::a/child::b/child::d/parent::node()/child::d
```

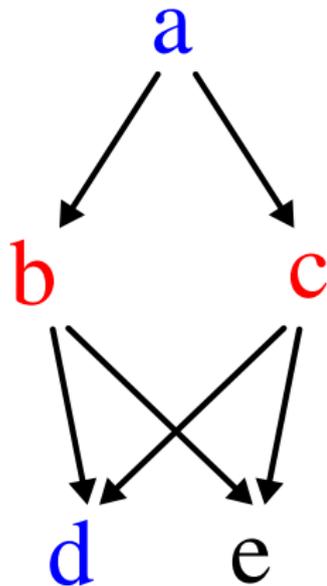
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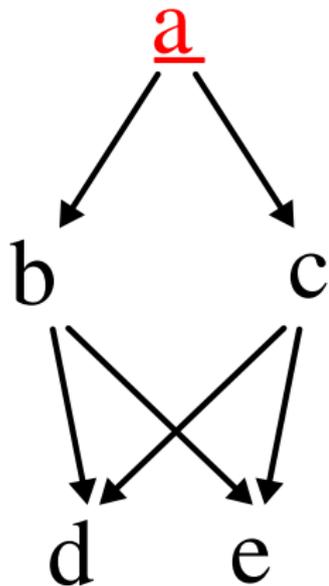
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- 3 `typeinf(DTD, {Xb}, child::d) = {Xd}`
- 4 `typeinf(DTD, {Xd}, parent::node()) = {Xb, Xc}`



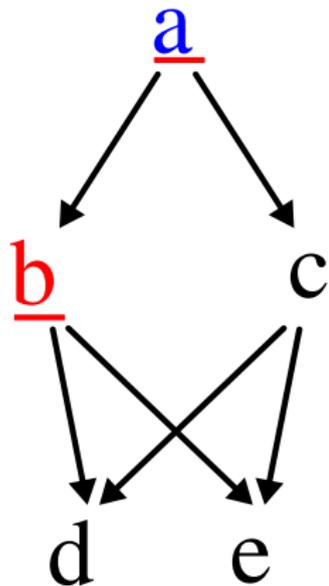
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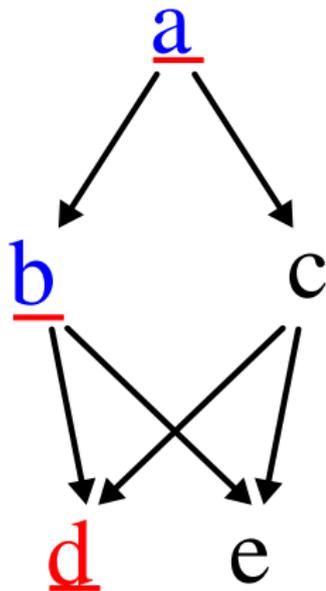
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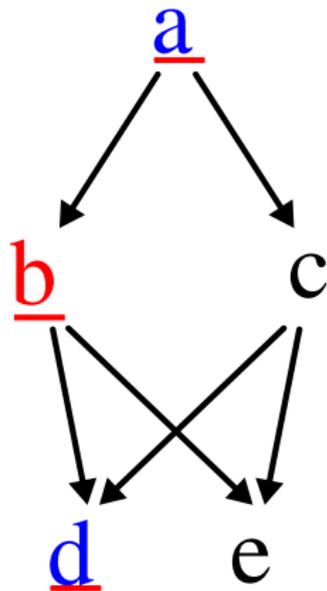
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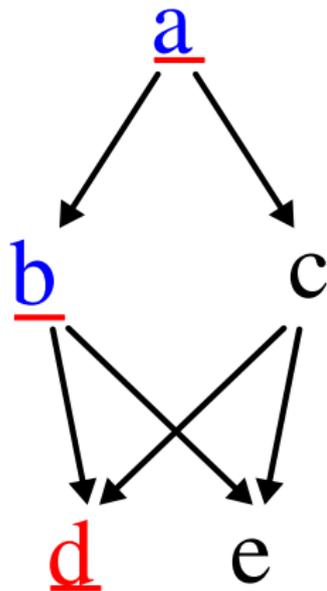
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- 3 `typeinf(DTD, {Xb}, {Xa, Xb}, child::d) = {Xd}`
- 4 `typeinf(DTD, {Xd}, {Xa, Xb, Xd}, parent::node()) = {Xb}`



`/self::a/child::b/child::d/parent::node()/child::d`

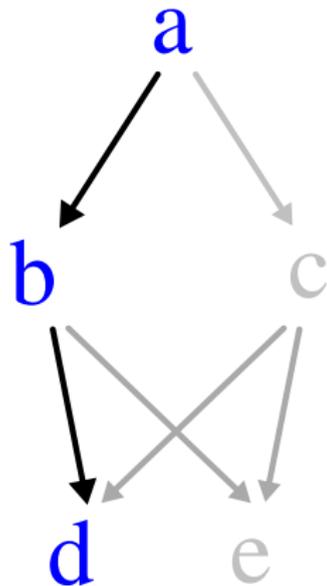
- 1 `typeinf(DTD, { X_a }, { X_a }, self::a) = { X_a }`
- 2 `typeinf(DTD, { X_a }, { X_a }, child::b) = { X_b }`
- 3 `typeinf(DTD, { X_b }, { X_a, X_b }, child::d) = { X_d }`
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$$\mathcal{P} = \{X_a, X_b, X_d\}$$



- 1 Introduction
- 2 Notations
- 3 Algorithm
- 4 Formal results**
- 5 Experiments
- 6 Conclusion

Theorem (Soundness)

Let D be a document valid w.r.t a DTD (X, E) and p a path. Let \mathcal{P} the type projector deduced from p . Let D' be the projection of D with \mathcal{P} .

$$\text{eval}(p, D) = \text{eval}(p, D')$$

Pruning is precise . . .

Pruning is precise ...

Theorem (Completeness)

$\mathcal{P} = \{X_1, \dots, X_n\}$ the type projector associated with a path p and a DTD (X, E) . Let $\mathcal{P}' = \mathcal{P} \setminus \{X_i\}$. There exists D a document and its projection D' such that :

$$\text{eval}(p, D) \neq \text{eval}(p, D')$$

Completeness holds with :

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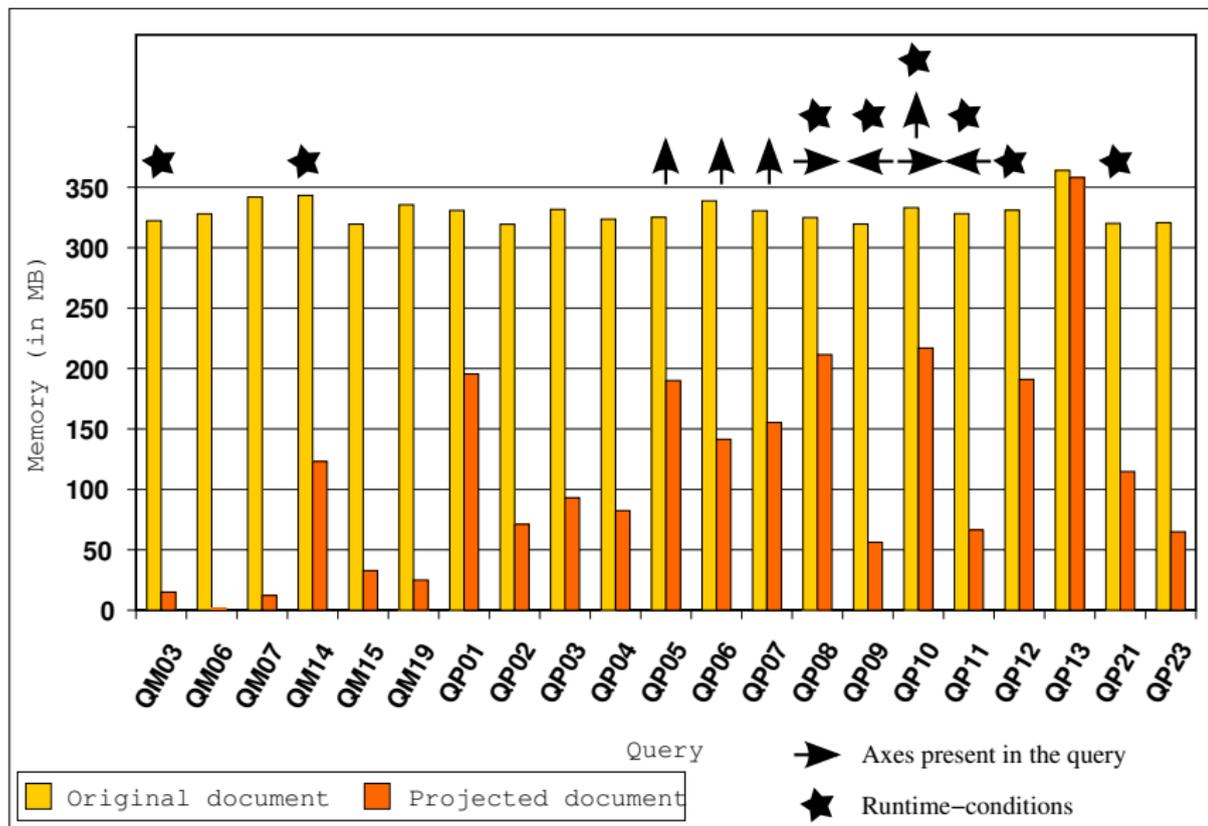
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- 2 Notations
- 3 Algorithm
- 4 Formal results
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Protocol :

- Linux desktop, 512MB Ram, 3Ghz x86 CPU (and no swap).
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 - Implementation in OCaml.
 - Validity of the result checked with the Galax query engine.
 - Pruning tested on XMark and XPathMark benchmarks.
- Pruning is **one pass bufferless** traversal of the document.
 - In practice, computing the type projector is fast.



Memory used to process a 56 MB document with Galax.

	QM03	QM06	QM07	QM14	QM15	QM19
	★			★		
Original Size (MB)	930	2048	1100	202	2048	964
Pruned Size(MB)	25	5,3	42	139	24	24
Memory Usage (MB)	374	90	380	512	245	512
% of original size	2.5	0.3	3.4	69.6	1.15	2.5
Gain in Speed (\times faster)	17.8	110.1	28.2	3.9	62.6	7.5

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Features : handles backward as well as following/preceding axes.

- 1 Introduction
- 2 Notations
- 3 Algorithm
- 4 Formal results
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- Extend the formalism to handle XML-Schema rather than DTDs.
- Integration with classical databases techniques.
- Integration with a query engine.