

Standing on the Shoulders of Ants: Towards More Efficient XML-to-Relational Mapping Strategies

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Introduction

- **XML = a standard for data representation and manipulation**
 - ⇒ **A boom of implementations**
 - XML file systems, native XML databases, XML-enabled databases, ...
- **XML-enabled databases**
 - **Most practically used though less efficient than native XML databases**
 - **Exploitation of tools and functions of traditional (O)RDBMSs**
 - **Reliable and robust**
 - **Long theoretical and practical history**
 - **Major DB vendors support XML**
 - **SQL standard: new part SQL/XML**

DB-Based XML Processing Methods

(O)RDBMS

- **Key concern: Choice of the most efficient XML-to-relational mapping strategy**
 - **Various classifications:**
 - **Generic (schema-oblivious) vs. schema-driven** – omitting vs. exploiting XML schema
 - **Fixed vs. adaptive** – mapping on the basis of data model vs. target application
 - Sample data and queries
 - **User-defined vs. user-driven** – the amount of user involvement
 - User specifies target schema and required mapping vs. user locally modifies a default mapping
- ⇒ **The most promising approach: adaptive**
- **Evaluates several mappings and chooses the optimal one**

Goal of This Paper

Three improvements of adaptive strategies:

1. Improvement of searching the optimal strategy

- Ant Colony Optimization
 - Finds better suboptimal solution than simple greedy search strategies currently used in the existing papers

2. Enhancing of the adaptation process with similarity of XML data

- Classical improvement

3. Combination with user-driven techniques

- User provides information on required mapping strategies for selected data fragments

Problem Statement

- To find the optimal mapping strategy for given XML schema S_{init} into a set of relations $R = \{r_1, r_2, \dots, r_m\}$
- Cost-driven adaptive strategies exploit:
 - Set of XML-to-XML **schema transformations** $T = \{t_1, t_2, \dots, t_n\}$
 - $\forall i : t_i$ transforms schema S to schema $S_i = t_i(S)$
 - Set of **sample data** D_{sample} characterizing an application
 - XML documents $D = \{d_1, d_2, \dots, d_k\}$ valid against S_{init}
 - XML queries $Q = \{q_1, q_2, \dots, q_l\}$ over S_{init}
 - **Cost function** f_{cost}
 - Evaluates the cost of relational schema R with regard to D_{sample}
- Aim: optimal relational schema R_{opt}
 - $f_{cost}(R_{opt}, D_{sample})$ is minimal

Example

```
<!ELEMENT employee (name, address)>  
<!ELEMENT name (first, middle?, last)>  
<!ELEMENT address (city, country, zip)>  
<!ELEMENT first #PCDATA>  
...  
<!ELEMENT zip #PCDATA>
```

- $T = \{t_{in}, t_{out}\}, Q = \emptyset$
 - Inlining, outlining

```
employee_1(name_first, name_middle, name_last,  
           address_city, address_country, address_zip)
```

- $T = \{t_{in}, t_{out}, t_{shred}, t_{unshred}\}$
 - Inlining, outlining, shredding, unshredding
- $Q = \{\dots, //employee/name, \dots\}$
 - No query of the form `//employee/name/first`, `//employee/name/middle` or `//employee/name/last`

```
employee_2(name,  
           address_city, address_country, address_zip)
```

Improvement 1. Search Strategy

- **Naïve search strategy:**
 - To generate all possible transformations of S_{init} and select the optimal one
 - **Problem: Searching for R_{opt} is an NP hard problem**
 - Constraints optimization problem (COP)
- ⇒ **Current approaches use heuristics ⇒ search for suboptimum**
- Typically a kind of a greedy search strategy
 - Get stuck in local suboptimums
- **Our approach: Ant Colony Optimization (ACO)**

ACO

- **Idea: Artificial ants iteratively search a space of solutions and improve current suboptimum**
- **Ant**
 - **Searches a subspace of solutions until it “dies”**
 - **Spreads “pheromone”**
 - **Positive feedback = how good solution it has found so far**
 - **Exploits pheromones of other ants to select next step**
 - **Step = applying an XML-to-XML schema transformation**
 - **Selected randomly**
 - **Probability is given by f_{cost} and pheromones of other ants**
- **In fact: Simple application of a general heuristic on a special case**

Improvement 2. Fragments without Cost Feedback

- Aim of cost-driven methods: to find relational schema R optimal for queries in Q
- Problem: S_{init} may involve fragments which occur on no access path of queries in Q
- Motivation for solution: system **UserMap**
 - User-driven mapping
 - Schema annotations are directly applied + regarded as “hints” how to store XML patterns
 - Idea: Iteratively searches for schema fragments similar to patterns and maps them in the same way
- Modification: annotated schema fragments = schema fragments on access paths of Q

Example

```
<!ELEMENT employee (name, address)>
<!ELEMENT name (first, middle?, last)>
<!ELEMENT address (city, country, zip)>
<!ELEMENT company (co-name, address)>
<!ELEMENT co-name (title, type?)>
```

- $T = \{t_{in}, t_{out}, t_{shred}, t_{unshred}\}$
- $Q = \{//employee/name\}$
- Element company cannot be mapped adaptively \leftarrow no feedback in Q

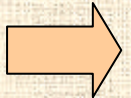
```
company_1(co-name_title, co-name_type,
          address_city, address_country, address_zip)
```

- Elements name and co-name are semantically similar \Rightarrow can be mapped in a similar way

```
company_2(co-name,
          address_city, address_country, address_zip)
```

Improvement 3. Schema Annotations

- Idea: Let us go even further...
 - Motivation:
 - Simple schema fragments: D and Q
 - e.g. the previous examples
 - Complex ones: mapping strategy
 - e.g. unshredding for XHTML fragments
 - Observation: Sequence of transformations form T = complex storage strategy = annotation
 - Combination of cost-driven and user-driven approaches
- ⇒ The set of possible steps of ants: application of T + **composite transformations** = user-specified annotations
- Can be applied on schema fragments similar to the original annotated ones
- ⇒ Speeds-up the search process



Example

```
<!ELEMENT employee (name, address)>  
<!ELEMENT name (first, middle?, last)>  
<!ELEMENT address (city, country, zip)>
```

- Relation `employee_2` can be derived using various sequences of transformations

```
employee_2(name,  
            address_city, address_country, address_zip)
```

$$s_1 = [t_{in}(city), t_{in}(country), t_{in}(zip), t_{in}(address), t_{in}(first), t_{in}(middle), t_{in}(last), t_{in}(name), t_{un}(name)]$$
$$s_2 = [t_{un}(name), t_{in}(name), t_{in}(city), t_{in}(country), t_{in}(zip), t_{in}(address)]$$

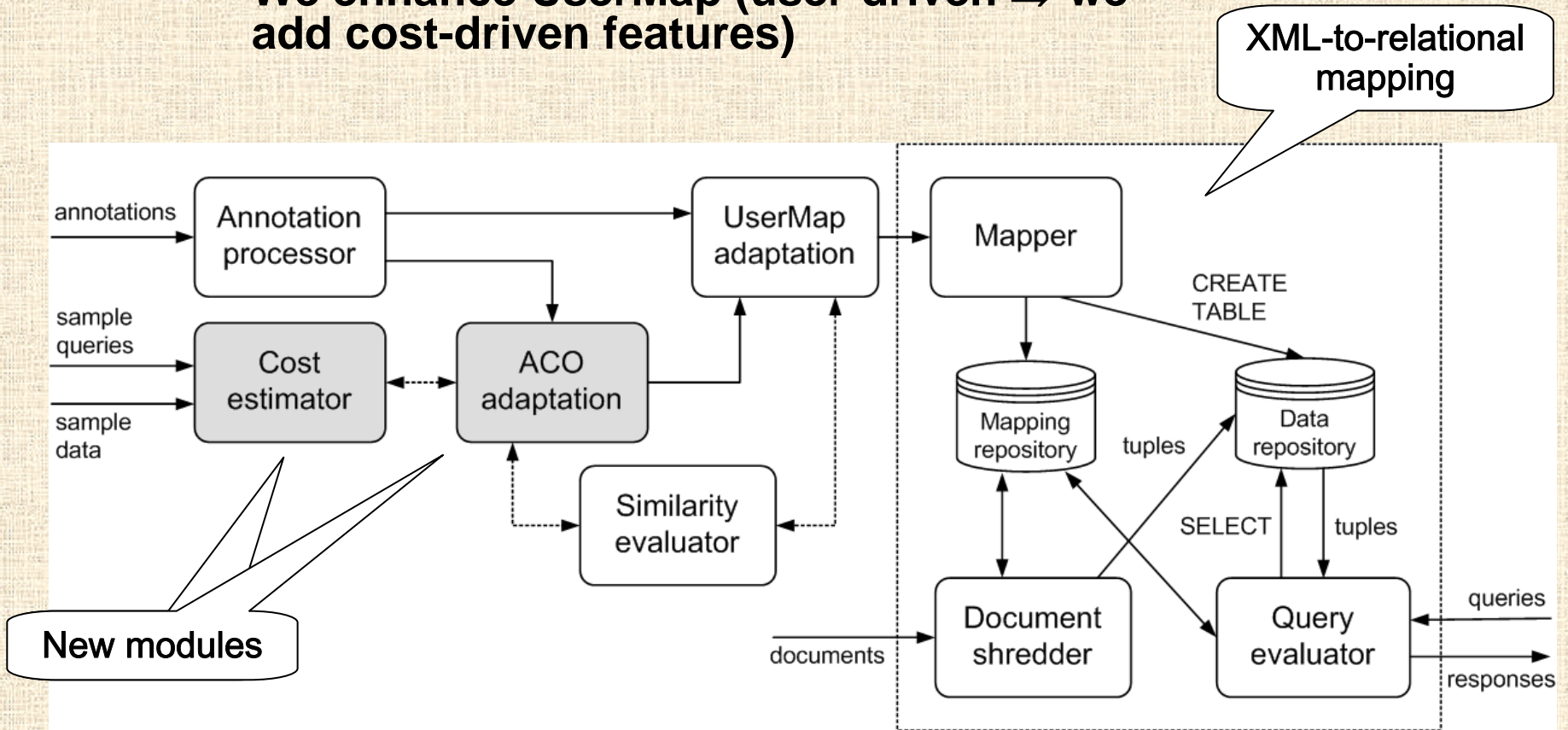
- If a user annotates element `employee` with fixed hybrid mapping (= `employee_1`)

$$s_3 = [t_{un}(name)]$$

```
employee_1(name_first, name_middle, name_last,  
            address_city, address_country, address_zip)
```

System Architecture

- Improvements could be applied on any user-driven/cost-driven system
 - We enhance UserMap (user-driven \Rightarrow we add cost-driven features)



Conclusion

- **Current work:**
 - Throughout implementation of the proposals
 - Enhancing of UserMap
 - Key aspect: cost estimator (evaluation of f_{cost})
 - Application on real-world data

⇒ What is the impact of the improvements?
- **Main advantages of improvements:**
 - Avoid getting stuck in local suboptimal solutions
 - Find an optimal mapping for greater subset of source schema
 - Speed up the search process
- **Future work: Persisting disadvantages of adaptive approaches**
 - Plenty of information on application
 - User-unfriendly
 - Efficiency can worsen with minor changes in the application

Thank you