



Exchanging more than Complete Data

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Outline: First part

- ▶ The data exchange problem
 - ▶ Some fundamental results in relational data exchange
- ▶ The need for a more general data exchange framework
 - ▶ Two important scenarios: Incomplete databases (open-world databases: RDF) and knowledge bases

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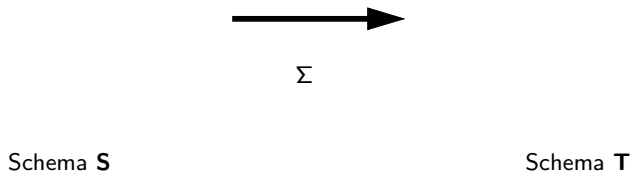
The problem of data exchange

Given: A source schema **S**, a target schema **T** and a specification Σ of the relationship between these schemas

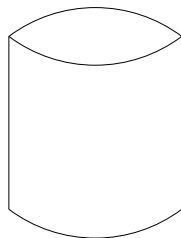
Data exchange: Problem of materializing an instance of **T** given an instance of **S**

- ▶ Target instance should reflect the source data as accurately as possible, given the constraints imposed by Σ and **T**
- ▶ It should be efficiently computable
- ▶ It should allow one to evaluate queries on the target in a way that is *semantically consistent* with the source data

Data exchange in a picture



Data exchange in a picture



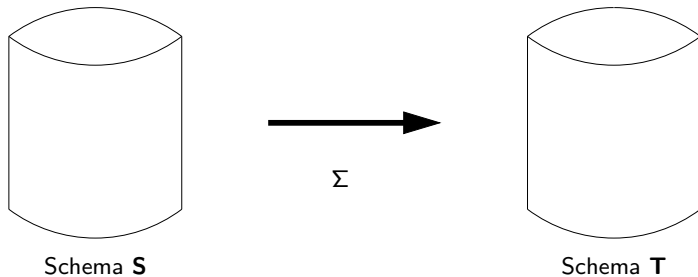
Schema **S**



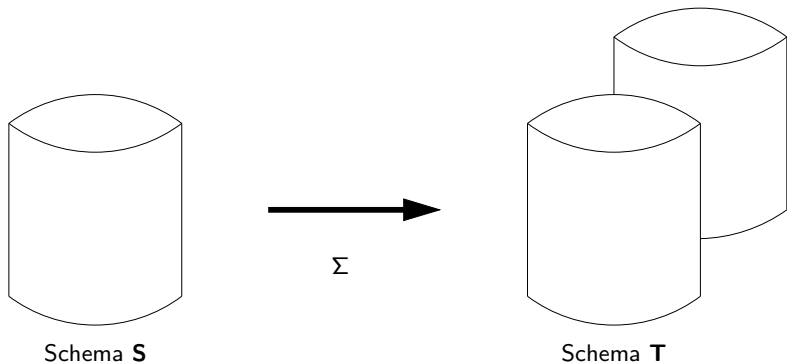
Σ

Schema **T**

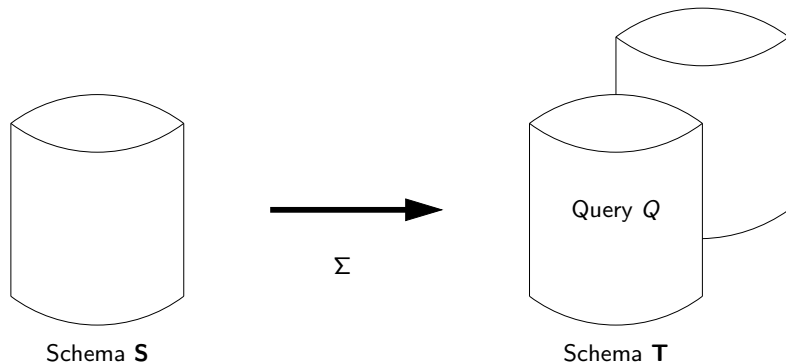
Data exchange in a picture



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Data exchange in a picture



Data exchange: Some fundamental questions

What are the challenges in the area?

- ▶ What is a good language for specifying the relationship between source and target data?
 - ▶ Expressiveness versus complexity
- ▶ What is a good instance to materialize?
- ▶ What does it mean to answer a query over target data?
- ▶ How do we answer queries over target data? Can we do this efficiently?

Exchanging relational data

The data exchange problem has been extensively studied in the relational world.

- ▶ It has also been commercially implemented: IBM Clio

Relational data exchange setting:

- ▶ Source and target schemas: Relational schemas
- ▶ Relationship between source and target schemas:
Source-to-target tuple-generating dependencies (st-tgds)

Semantics of data exchange has been precisely defined.

- ▶ Efficient algorithms for materializing target instances and for answering queries over the target schema have been developed

Schema mapping: The key component in relational data exchange

Schema mapping: $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$

- ▶ \mathbf{S} and \mathbf{T} are disjoint relational schemas
- ▶ Σ is a finite set of st-tgds:

$$\forall \bar{x} \forall \bar{y} (\varphi(\bar{x}, \bar{y}) \rightarrow \exists \bar{z} \psi(\bar{x}, \bar{z}))$$

$\varphi(\bar{x}, \bar{y})$: conjunction of relational atomic formulas over \mathbf{S}

$\psi(\bar{x}, \bar{z})$: conjunction of relational atomic formulas over \mathbf{T}

Relational schema mappings: An example

Example

- ▶ **S**: Employee(name)
- ▶ **T**: Dept(name, number)
- ▶ Σ :

$$\forall x \left(\text{Employee}(x) \rightarrow \exists y \text{Dept}(x, y) \right)$$

Relational schema mappings: An example

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Note

We omit universal quantifiers in st-tgds:

$$\text{Employee}(x) \rightarrow \exists y \text{Dept}(x, y)$$

Relational data exchange problem

Fixed: $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$

Problem: Given instance I of \mathbf{S} , find an instance J of \mathbf{T} such that (I, J) satisfies Σ

- ▶ (I, J) satisfies $\varphi(\bar{x}, \bar{y}) \rightarrow \exists \bar{z} \psi(\bar{x}, \bar{z})$ if whenever I satisfies $\varphi(\bar{a}, \bar{b})$, there is a tuple \bar{c} such that J satisfies $\psi(\bar{a}, \bar{c})$

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Notation

J is a **solution** for I under \mathcal{M}

- ▶ $\text{Sol}_{\mathcal{M}}(I)$: Set of solutions for I under \mathcal{M}

The notion of solution: Example

Example

- ▶ **S**: Employee(name)
- ▶ **T**: Dept(name, number)
- ▶ Σ : Employee(x) $\rightarrow \exists y$ Dept(x, y)

Solutions for $I = \{\text{Employee(Peter)}\}$:

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J_1 : {Dept(Peter,1)}

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J_2 : {Dept(Peter,1), Dept(Peter,2)}

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J_4 : {Dept(Peter, n_1)}

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J_3 : {Dept(Peter,1), Dept(John,1)}

J_4 : {Dept(Peter, n_1)}

J_5 : {Dept(Peter, n_1), Dept(Peter, n_2)}

Algorithm (Chase)

Input : $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$ and an instance I of \mathbf{S}

Output : Canonical universal solution J^* for I under \mathcal{M}

let $J^* :=$ empty instance of \mathbf{T}

for every $\varphi(\bar{x}, \bar{y}) \rightarrow \exists \bar{z} \psi(\bar{x}, \bar{z})$ in Σ **do**

for every \bar{a}, \bar{b} such that I satisfies $\varphi(\bar{a}, \bar{b})$ **do**

 create a fresh tuple \bar{n} of pairwise distinct null values

 insert $\psi(\bar{a}, \bar{n})$ into J^*

Canonical universal solution: Example

Example

Consider mapping \mathcal{M} specified by dependency:

$$\text{Employee}(x) \rightarrow \exists y \text{Dept}(x, y)$$

Canonical universal solution for

$I = \{\text{Employee}(\text{Peter}), \text{Employee}(\text{John})\}$:

- ▶ For $a = \text{Peter}$ do
 - ▶ Create a fresh null value n_1
 - ▶ Insert $\text{Dept}(\text{Peter}, n_1)$ into J^*
- ▶ For $a = \text{John}$ do
 - ▶ Create a fresh null value n_2
 - ▶ Insert $\text{Dept}(\text{John}, n_2)$ into J^*

Result: $J^* = \{\text{Dept}(\text{Peter}, n_1), \text{Dept}(\text{John}, n_2)\}$

Query answering in data exchange

Given: Mapping \mathcal{M} , source instance I and query Q over the target schema

- ▶ What does it mean to answer Q ?

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Definition (Certain answers)

$$\text{certain}_{\mathcal{M}}(Q, I) = \bigcap_{J \text{ is a solution for } I \text{ under } \mathcal{M}} Q(J)$$

Certain answers: Example

Example

Consider mapping \mathcal{M} specified by:

$$\text{Employee}(x) \rightarrow \exists y \text{Dept}(x, y)$$

Given instance $I = \{\text{Employee}(\text{Peter})\}$:

$$\begin{aligned} \text{certain}_{\mathcal{M}}(\exists y \text{Dept}(x, y), I) &= \{\text{Peter}\} \\ \text{certain}_{\mathcal{M}}(\text{Dept}(x, y), I) &= \emptyset \end{aligned}$$

Query rewriting: An approach for answering queries

How can we compute certain answers?

- ▶ Naïve algorithm does not work: infinitely many solutions

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Approach proposed in [FKMP03]: **Query Rewriting**

Given a mapping \mathcal{M} and a target query Q , compute a query Q^* such that for every source instance I with canonical universal solution J^* :

$$\text{certain}_{\mathcal{M}}(Q, I) = Q^*(J^*)$$

Query rewriting over the canonical universal solution

Theorem (FKMP03)

Given a mapping \mathcal{M} specified by st-tgds and a union of conjunctive queries Q , there exists a query Q^ such that for every source instance I with canonical universal solution J^* :*

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Query rewriting over the canonical universal solution

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Proof idea: Assume that $\mathbf{C}(a)$ holds whenever a is a constant.

Then:

$$Q^*(x_1, \dots, x_m) = \mathbf{C}(x_1) \wedge \dots \wedge \mathbf{C}(x_m) \wedge Q(x_1, \dots, x_m)$$

Computing certain answers: Complexity

Data complexity: Data exchange setting and query are considered to be fixed.

Corollary (FKMP03)

*For mappings given by st-tgds, certain answers for **UCQ** can be computed in polynomial time (data complexity)*

Relational data exchange: Some lessons learned

Key steps in the development of the area:

- ▶ Definition of schema mappings: Precise syntax and semantics
 - ▶ Definition of the notion of solution
- ▶ Identification of good solutions
- ▶ Polynomial time algorithms for materializing good solutions
- ▶ Definition of target queries: Precise semantics
- ▶ Polynomial time algorithms for computing certain answers for **UCQ**

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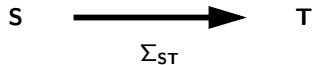
Creating schema mappings is a time consuming and expensive process

- ▶ Manual or semi-automatic process in general

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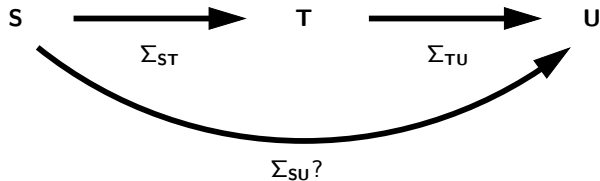
Ongoing project: Reusing schema mappings



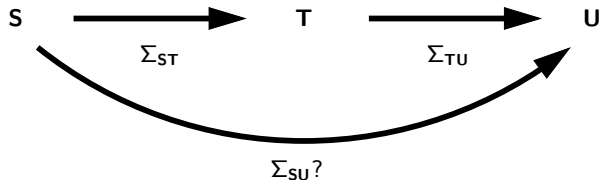
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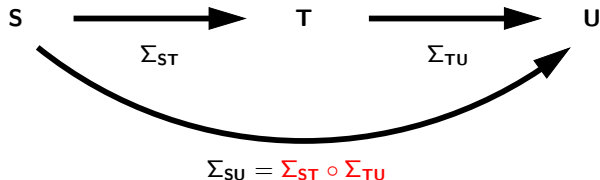


Ongoing project: Reusing schema mappings



We need some operators for schema mappings

Ongoing project: Reusing schema mappings



We need some operators for schema mappings

- ▶ **Composition** in the above case

Contributions mentioned in the previous slides are just a first step towards the development of a general framework for data exchange.

In fact, as pointed in [B03],

many information system problems involve not only the design and integration of complex application artifacts, but also their subsequent manipulation.

Metadata management

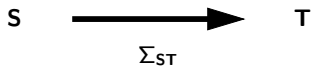
This has motivated the need for the development of a general infrastructure for managing schema mappings.

The problem of managing schema mappings is called **metadata management**.

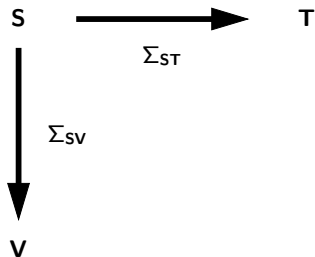
High-level algebraic operators, such as compose, are used to manipulate mappings.

- ▶ What other operators are needed?

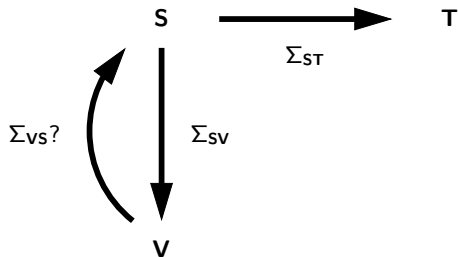
An inverse operator is also needed



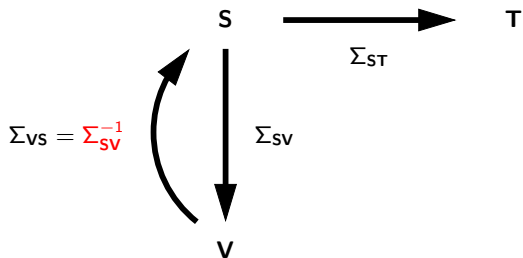
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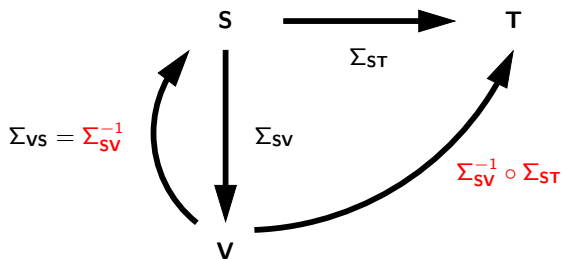
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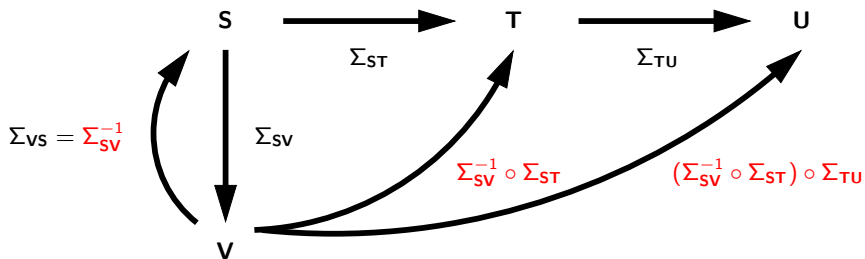


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Composition and inverse operators have to be combined

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Metadata management: A more general data exchange framework is needed

Composition and inverse operators have been extensively studied in the relational world.

- ▶ Semantics, computation, ...

Combining these operators is an open issue.

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There is a need for a data exchange framework that can handle databases with **incomplete information**.

Data exchange in the RDF world

There is an increasing interest in publishing relational data as RDF

- ▶ Resulted in the creation of the W3C RDB2RDF Working Group

The problem of translating relational data into RDF can be seen as a data exchange problem

- ▶ Schema mappings can be used to describe how the relational data is to be mapped into RDF

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But there is a mismatch here: A relational database under a **closed-world semantics** is to be translated into an RDF graph under an **open-world semantics**

- ▶ There is a need for a data exchange framework that can handle both databases with complete and incomplete information

Data exchange in the RDF world

An issue discussed at the W3C RDB2RDF Working Group: **Is a mapping information preserving?**

- ▶ In particular: For the default mapping defined by this group

How can we address this issue?

- ▶ Metadata management can help us

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Question to answer: Is a mapping invertible?

- ▶ This time an RDF graph is to be translated into a relational database!
- ▶ We want to have a **unifying** framework for all these cases

But these are not the only reasons . . .

Nowadays several applications use knowledge bases to represent data.

- ▶ A knowledge base has not only data but also **rules** that allows to infer new data
- ▶ In the Semantics Web: RDFS and OWL ontologies

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In a data exchange application over the Semantics Web:

The input is a mapping and a source specification including data and **rules**, and the output is a target specification also including data and **rules**

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There is a need for a data exchange framework that can handle **knowledge bases**.

Knowledge exchange: A more general data exchange framework is needed

Example

Assume given the following source knowledge base:

Data:

Father		Mother	
Andy	Bob	Carrie	Bob
Bob	Danny		
Danny	Eddie		

Rules:

$$\begin{aligned} \text{Father}(x, y) &\rightarrow \text{Parent}(x, y) \\ \text{Mother}(x, y) &\rightarrow \text{Parent}(x, y) \\ \text{Parent}(x, y) \wedge \text{Parent}(y, z) &\rightarrow \text{Grandparent}(x, z) \end{aligned}$$

Knowledge exchange: A more general data exchange framework is needed

Example (cont'd)

Given a mapping:

$$\begin{aligned}\text{Father}(x, y) &\rightarrow \text{F}(x, y) \\ \text{Grandparent}(x, y) &\rightarrow \text{G}(x, y)\end{aligned}$$

What is a good translation of the initial knowledge base?

Knowledge exchange: A more general data exchange framework is needed

Example (cont'd)

Given a mapping:

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What is a good translation of the initial knowledge base?

Data:

F		G	
Andy	Bob	Andy	Danny
Bob	Danny	Carrie	Danny
Danny	Eddie	Bob	Eddie

Rules: \emptyset

Knowledge exchange: A more general data exchange framework is needed

Example (cont'd)

Our first alternative does not include any translation of the source rules:

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Example (cont'd)

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Bob	Danny	Carrie	Danny
Danny	Eddie	Bob	Eddie

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Is this a good translation? Why?

One can exchange more than complete data

- ▶ In data exchange one starts with a database instance (with complete information).
- ▶ What if we have an initial object that has several interpretations?
 - ▶ A representation of a set of possible instances
- ▶ We propose a new general formalism to exchange representations of possible instances
 - ▶ We apply it to the problems of exchanging instances with incomplete information and exchanging knowledge bases

Outline: Second part

- ▶ Formalism for exchanging representations systems
- ▶ Applications to incomplete instances
- ▶ Applications to knowledge bases
- ▶ Concluding remarks

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Representation systems

A representation system $\mathcal{R} = (\mathbf{W}, \text{rep})$ consists of:

- ▶ a set \mathbf{W} of *representatives*
- ▶ a function rep that assigns a set of instances to every element in \mathbf{W}

$$\text{rep}(\mathcal{V}) = \{I_1, I_2, I_3, \dots\} \text{ for every } \mathcal{V} \in \mathbf{W}$$

Uniformity assumption: For every $\mathcal{V} \in \mathbf{W}$, there exists a relational schema \mathbf{S} (the type of \mathcal{V}) such that $\text{rep}(\mathcal{V}) \subseteq \text{Inst}(\mathbf{S})$

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Incomplete instances and knowledge bases are representation systems

In classical data exchange we consider only *complete* data

Recall that given $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$, $I \in \text{Inst}(\mathbf{S})$ and $J \in \text{Inst}(\mathbf{T})$: J is a solution for I under \mathcal{M} if $(I, J) \models \Sigma$

$$J \in \text{Sol}_{\mathcal{M}}(I)$$

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This can be extended to set of instances. Given $\mathcal{X} \subseteq \text{Inst}(\mathbf{S})$:

$$\text{Sol}_{\mathcal{M}}(\mathcal{X}) = \bigcup_{I \in \mathcal{X}} \text{Sol}_{\mathcal{M}}(I)$$

Extending the definition to representation systems

Given:

- ▶ a mapping $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$
- ▶ a representation system $\mathcal{R} = (\mathbf{W}, \text{rep})$
- ▶ $\mathcal{U}, \mathcal{V} \in \mathbf{W}$ of types \mathbf{S} and \mathbf{T} , respectively

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Definition (APR11)

\mathcal{V} is an \mathcal{R} -solution of \mathcal{U} under \mathcal{M} if

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Or equivalently: \mathcal{V} is an \mathcal{R} -solution of \mathcal{U} if for every $J \in \text{rep}(\mathcal{V})$, there exists $I \in \text{rep}(\mathcal{U})$ such that $J \in \text{Sol}_{\mathcal{M}}(I)$.

What is a good solution in this framework?

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Definition (APR11)

\mathcal{V} is an *universal \mathcal{R} -solution* of \mathcal{U} under \mathcal{M} if

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Strong representation systems

Let \mathcal{C} be a class of mappings.

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Definition (APR11)

$\mathcal{R} = (\mathbf{W}, \text{rep})$ is a *strong representation system* for \mathcal{C} if for every $\mathcal{M} \in \mathcal{C}$ and for every $\mathcal{U} \in \mathbf{W}$, there exists a $\mathcal{V} \in \mathbf{W}$:

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$$\text{rep}(\mathcal{V}) = \text{Sol}_{\mathcal{M}}(\text{rep}(\mathcal{U}))$$

If $\mathcal{R} = (\mathbf{W}, \text{rep})$ is a strong representation system, then the universal solutions for the representatives in \mathbf{W} can be **represented** in the same system.

Outline: Second part

- ▶ Formalism for exchanging representations systems
- ▶ Applications to incomplete instances
- ▶ Applications to knowledge bases
- ▶ Concluding remarks

Motivating questions

What is a strong representation system for the class of mappings specified by st-tgds?

- ▶ Are instances including nulls enough?

Can the fundamental data exchange problems be solved in polynomial time in this setting?

- ▶ Computing (universal) solutions
- ▶ Computing certain answers

Naive instances

We have already considered **naive instances**: Instances with null values

- ▶ Example: Canonical universal solution

A naive instance \mathcal{I} has labeled nulls:

$$R(1, n_1)$$

$$R(n_1, 2)$$

$$R(1, n_2)$$

Naive instances

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A naive instance \mathcal{I} has labeled nulls:

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$R(n_1, 2)$

$R(1, n_2)$

The interpretations of \mathcal{I} are constructed by replacing nulls by constants:

$$\text{rep}(\mathcal{I}) = \{K \mid \mu(\mathcal{I}) \subseteq K \text{ for some valuation } \mu\}$$

Are naive instances expressive enough?

Naive instances have been extensively used in data exchange:

Proposition (FKMP03)

Let $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$, where Σ is a set of st-tgds. Then for every instance I of \mathbf{S} , there exists a naive instance \mathcal{J} of \mathbf{T} such that:

$$\text{rep}(\mathcal{J}) = \text{Sol}_{\mathcal{M}}(I)$$

In fact, the canonical universal solution satisfies the property mentioned above.

Are naive instances expressive enough?

But naive instances are not expressive enough to deal with incomplete information in the source instances:

Proposition (APR11)

Naive instances are not a strong representation system for the class of mappings specified by st-tgds

Are naive instances expressive enough?

Example

Consider a mapping \mathcal{M} specified by:

$$\text{Manager}(x, y) \rightarrow \text{Reports}(x, y)$$
$$\text{Manager}(x, x) \rightarrow \text{SelfManager}(x)$$

The canonical universal solution for $\mathcal{I} = \{\text{Manager}(n, \text{Peter})\}$ under \mathcal{M} :

$$\mathcal{J} = \{\text{Reports}(n, \text{Peter})\}$$

But \mathcal{J} is not a *good* solution for \mathcal{I} .

- ▶ It cannot represent the fact that if n is given value Peter, then $\text{SelfManager}(\text{Peter})$ should hold in the target.

Conditional instances

What should be added to naive instances to obtain a strong representation system?

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- ▶ Answer from database theory: Conditions on the nulls

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- ▶ Answer from database theory: Conditions on the nulls

Conditional instances: Naive instances plus *tuple conditions*

A tuple condition is a positive Boolean combinations of:

- ▶ equalities and inequalities between nulls, and between nulls and constants

Conditional instances

Example

$$\begin{array}{l|l} R(1, n_1) & n_1 = n_2 \\ R(n_1, n_2) & n_1 \neq n_2 \vee n_2 = 2 \end{array}$$

Conditional instances

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$$\begin{array}{l|l} R(1, n_1) & n_1 = n_2 \\ R(n_1, n_2) & n_1 \neq n_2 \vee n_2 = 2 \end{array}$$

Semantics:

Conditional instances

Example

$$\begin{array}{l|l} R(1, n_1) & n_1 = n_2 \\ R(n_1, n_2) & n_1 \neq n_2 \vee n_2 = 2 \end{array}$$

Semantics:

$$\underline{\mu(n_1) = \mu(n_2) = 2} \quad \underline{\mu(n_1) = \mu(n_2) = 3} \quad \underline{\mu(n_1) = 2, \mu(n_2) = 3}$$

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Interpretations of a conditional instance \mathcal{I} :

$$\text{rep}(\mathcal{I}) = \{K \mid \mu(\mathcal{I}) \subseteq K \text{ for some valuation } \mu\}$$

Positive conditional instances

Many problems are intractable over conditional instances.

- ▶ We also consider a restricted class of conditional instances

Positive conditional instances: Conditional instances without inequalities

(Positive) conditional instances are enough

Theorem (APR11)

Both conditional instances and positive conditional instances are strong representation systems for the class of mappings specified by st-tgds.

Example

Consider again the mapping \mathcal{M} specified by:

$$\text{Manager}(x, y) \rightarrow \text{Reports}(x, y)$$
$$\text{Manager}(x, x) \rightarrow \text{SelfManager}(x)$$

The following is a universal solution for $\mathcal{I} = \{\text{Manager}(n, \text{Peter})\}$

$\text{Reports}(n, \text{Peter})$	\mid	$true$
$\text{SelfManager}(\text{Peter})$	\mid	$n = \text{Peter}$

Positive conditional instances are *exactly* the needed representation system

Positive conditional instances are *minimal*:

Theorem (APR11)

All the following are needed to obtain a strong representation system for the class of mappings specified by st-tgds:

- ▶ *equalities between nulls*
- ▶ *equalities between constant and nulls*
- ▶ *conjunctions and disjunctions*

Conditional instances are enough but not minimal.

Positive conditional instance can be used in practice!

Let $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$, where Σ is a set of st-tgds.

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Theorem (APR11)

There exists a polynomial time algorithm that, given a positive conditional instance \mathcal{I} over \mathbf{S} , computes a positive conditional instance \mathcal{J} over \mathbf{T} that is a universal solution for \mathcal{I} under \mathcal{M} .

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Let Q be a union of conjunctive queries over \mathbf{T} .

$$Q(\mathcal{J}) = \bigcap_{J \in \text{rep}(\mathcal{J})} Q(J)$$
$$\text{certain}_{\mathcal{M}}(Q, \mathcal{I}) = \bigcap_{\mathcal{J} \text{ is a solution for } \mathcal{I} \text{ under } \mathcal{M}} Q(\mathcal{J})$$

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The same result holds for the class of unions of conjunctive queries with at most one inequality per disjunct.

- ▶ The other important class of queries in the data exchange area for which certain answers can be computed in polynomial time

Outline: Second part

- ▶ Formalism for exchanging representations systems
- ▶ Applications to incomplete instances
- ▶ Applications to knowledge bases
- ▶ Concluding remarks

The semantics of *knowledge bases* is given by sets of instances

Knowledge base over \mathbf{S} : (I, Γ) such that

- ▶ $I \in \text{Inst}(\mathbf{S})$
- ▶ Γ a set of rules over \mathbf{S}

Semantics: finite models

$$\text{Mod}(I, \Gamma) = \{K \in \text{Inst}(\mathbf{S}) \mid I \subseteq K \text{ and } K \models \Gamma\}$$

We can apply our formalism to knowledge bases

(I_2, Γ_2) is a *KB-solution* for (I_1, Γ_1) under \mathcal{M} if:

$$\text{Mod}(I_2, \Gamma_2) \subseteq \text{Sol}_{\mathcal{M}}(\text{Mod}(I_1, \Gamma_1))$$

(I_2, Γ_2) is a *universal KB-solution* for (I_1, Γ_1) under \mathcal{M} if:

$$\text{Mod}(I_2, \Gamma_2) = \text{Sol}_{\mathcal{M}}(\text{Mod}(I_1, \Gamma_1))$$

Motivating questions

Same as for the case of instances with incomplete information.

- ▶ Constructing universal KB-solutions
- ▶ Answering target queries

New fundamental problem: Construct solutions including **as much implicit knowledge as possible**.

What are good knowledge-base solutions?

First alternative: universal KB-solutions

But there exist some other KB-solutions desirable to materialize

- ▶ Minimality comes into play

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But there exist some other KB-solutions desirable to materialize

- ▶ Minimality comes into play

Given sets \mathcal{X} , \mathcal{Y} of instances:

- ▶ $\mathcal{X} \equiv_{\min} \mathcal{Y}$ if \mathcal{X} and \mathcal{Y} coincide in the minimal instances under \subseteq

Definition

(I_2, Γ_2) is a *minimal KB-solution* of (I_1, Γ_1) under \mathcal{M} if:

$$\text{Mod}(I_2, \Gamma_2) \equiv_{\min} \text{Sol}_{\mathcal{M}}(\text{Mod}(I_1, \Gamma_1))$$

Two requirements to construct minimal knowledge-base solutions

Given (I_1, Γ_1) and \mathcal{M} , when constructing a minimal KB-solution (I_2, Γ_2) we would like:

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Γ_2 is *safe* for Γ_1 and \mathcal{M}

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Definition

Γ_2 is safe for Γ_1 and \mathcal{M} , if for every I_1 there exists I_2 :

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Definition

Γ_2 is *optimal-safe* if for every other safe set Γ' :

$$\Gamma_2 \models \Gamma'$$

Computing minimal KB-solutions

To obtain algorithms for computing minimal KB-solutions, we need to specify the language used in knowledge bases.

- ▶ Full st-tgd:

$$\forall \bar{x} \forall \bar{y} (\varphi(\bar{x}, \bar{y}) \rightarrow \psi(\bar{x}))$$

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Theorem (APR11)

There exists a polynomial-time algorithm that, given $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$, where Σ is a set of full st-tgds, and given a set Γ_1 of full tgds over \mathbf{S} , computes a set Γ_2 of second-order logic sentences over \mathbf{T} that is optimal-safe for Γ_1 and \mathcal{M} .

Computing minimal KB-solutions

Unfortunately, first-order logic is not expressive enough.

Theorem (APR11)

There exist $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$, where Σ is a set of full st-tgds, and a set Γ_1 of full tgds over \mathbf{S} such that:

no FO-sentence is optimal-safe for Γ_1 and \mathcal{M} .

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no FO-sentence is optimal-safe for Γ_1 and \mathcal{M} .

How can we deal with these problems in practice?

- ▶ We need to restrict the language used to specify knowledge bases: Description logics [ABC11]

Outline: Second part

- ▶ Formalism for exchanging representations systems
- ▶ Applications to incomplete instances
- ▶ Applications to knowledge bases
- ▶ Concluding remarks

We can exchange more than complete data

We propose a general formalism to exchange *representation systems*

- ▶ Applications to incomplete instances
- ▶ Applications to knowledge bases

Next step: Apply our general setting to the Semantic Web

- ▶ Semantic Web data has *nulls* (blank nodes)
- ▶ Semantic Web specifications have rules (RDFS, OWL)

Lots of interesting problems to solve if knowledge bases are specified by means of description logics.

- ▶ Better results can be obtained

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Thank you!

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Bonus track: Computation of solutions and its associated decision problem

Decision problem: CHECK-KB-SOL

Input: $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$, where Σ is a set of st-tgds
 (I_1, Γ_1) KB over \mathbf{S} with Γ_1 a set of tgds
 (I_2, Γ_2) KB over \mathbf{T} with Γ_2 a set of tgds

Output: Is (I_2, Γ_2) a KB-solution of (I_1, Γ_1) under \mathcal{M} ?

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Output: Is (I_2, Γ_2) a KB-solution of (I_1, Γ_1) under \mathcal{M} ?

Theorem (APR11)

CHECK-KB-SOL is undecidable (even for a fixed \mathcal{M}).

Bonus track: Computation of solutions and its associated decision problem

Undecidability is a consequence of using \exists in knowledge bases.

- ▶ We need to restrict the input

CHECK-FULL-KB-SOL: Γ_1, Γ_2 are assumed to be sets of full tgds

Bonus track: Computation of solutions and its associated decision problem

Theorem (APR11)

CHECK-FULL-KB-SOL is *EXPTIME-complete*.

Bonus track: Computation of solutions and its associated decision problem

Theorem (APR11)

CHECK-FULL-KB-SOL is *EXPTIME*-complete.

Theorem (APR11)

If $\mathcal{M} = (\mathbf{S}, \mathbf{T}, \Sigma)$ is fixed:

CHECK-FULL-KB-SOL is $\Delta_2^P[O(\log n)]$ -complete.

$\Delta_2^P[O(\log n)]$: P^{NP} with a logarithmic number of calls to the *NP* oracle.