

Inference-Proof Data Publishing by Minimally Weakening a Database Instance

Joachim Biskup Marcel Preuß

Information Systems and Security (ISSI)

Technische Universität Dortmund, Germany

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└─ Motivating this Work



Inference-Proof Data Publishing

Nowadays: Data publishing is ubiquitous

- Governments and companies provide data
- People share data about their private lifes

But: Original data often contains sensitive (personal) information

- Set up a confidentiality policy
- Release only "inference-proof views" of original data
 - No information to be protected is revealed
 - Even if an adversary tries to deduce inferences

Basics of Relational Databases



Supposed Database Setting

Relational schema $\langle R | A_R | \emptyset \rangle$

- Relational symbol R
- Attribute set $A_R = \{A_1, \ldots, A_n\}$
- No database constraints declared (for now)
- Infinite set Dom of constant symbols

Complete relational instance *r* over $\langle R | A_R | \emptyset \rangle$

- Finite number of valid database tuples over Dom
- CWA: Each constant combination not contained in r is invalid
 - Infinite number of invalid tuples
 - No constant combination is undefined

Basics of Relational Databases



First-Order Logic for Modeling Databases

Given first-order language ${\mathscr L}$ with equality

- Predicate symbol *R* with arity $|A_R| = n$
- Predicate symbol \equiv for expressing equality
- Infinite set Dom of constant symbols

Database-specific semantics: \mathcal{I} is DB-Interpretation, if

- Dom is the universe of \mathcal{I} and $\mathcal{I}(v) = v$ for each $v \in Dom$,
- *R* interpreted by finite $\mathcal{I}(R) \subset Dom^n$,
- ▶ ≡ interpreted by $\mathcal{I}(\equiv) = \{(v, v) \mid v \in Dom\}$

└─ Basics of Relational Databases



Logic-Oriented Modeling of Relational Instances

Given instance r:

+	_	R(a, b, c), R(a, c, c), R(b, a, c)
(a, b, c) (a, c, c) (b, a, c)	(a, a, a) (a, a, b) (a, a, c) \vdots	$(\forall X)(\forall Y)(\forall Z) [(X \equiv a \land Y \equiv b \land Z \equiv c) \lor (X \equiv a \land Y \equiv c \land Z \equiv c) \lor (X \equiv b \land Y \equiv a \land Z \equiv c) \lor (X \equiv b \land Y \equiv a \land Z \equiv c) \lor \neg R(X, Y, Z)]$
]

Idea of logic-oriented modeling:

- Each valid tuple as corresponding ground atom
- Infinite set of invalid tuples as completeness-sentence
 - List all tuples which are not invalid $(\rightarrow$ Finite set)
 - All other tuples are invalid $(\rightarrow \text{Infinitely many})$

Basics of Relational Databases



Confidentiality Policy

Confidentiality policy psec

- Finite set of potential secrets
- ▶ Potential secret: Ground atom R(c) with $c \in Dom^n$

Semantics of potential secret $\Psi \in psec$

- If Ψ is valid in r: Adversary **must not** get to know this
- Otherwise: Adversary may know that Ψ is invalid in r

Assume: Adversary is aware of policy



Some Thoughts about Easy Cases



Definition of Inference-Proofness

Given:

- Complete original instance r over $\langle R | A_R | \emptyset \rangle$
- Confidentiality policy psec
- Weakening algorithm weak (r, psec)

Inference-Proofness: From adversary's point of view

- For each potential secret $\Psi \in \textit{psec}$
- Existence of complete alternative instance r^{Ψ} over $\langle R | A_R | \emptyset \rangle$
 - r^{Ψ} does **not** satisfy Ψ
 - r^{Ψ} is indistinguishable from original instance r
 - \rightarrow weak (r^{Ψ} , psec) = weak (r, psec)

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└─Some Thoughts about Easy Cases



Case Study 1: Given Setting

Policy:
$$psec = \{ \Psi_1 = R(a, b, c), \Psi_2 = R(a, c, c) \}$$

Original instance r:

Obviously: $\mathcal{I}_r \models_M \Psi_1$, $\mathcal{I}_r \models_M \Psi_2$

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Inference-Proof Weakenings

Some Thoughts about Easy Cases

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Case Study 1: Weakening

Policy:
$$psec = \{ \Psi_1 = R(a, b, c), \Psi_2 = R(a, c, c) \}$$

Weakening weak(r, psec):

$$\begin{array}{c|c}
+ & - \\
\hline
(a, b, c) & (a, a, a) \\
\hline
(a, c, c) & (a, a, b) \\
(b, a, c) & (a, a, c) \\
& \vdots \\
\end{array}$$

Disjunctive knowledge: $R(a, b, c) \lor R(a, c, c)$

$$R(b, a, c)$$

$$R(a, b, c) \lor R(a, c, c)$$

$$(\forall X)(\forall Y)(\forall Z) [$$

$$(X \equiv a \land Y \equiv b \land Z \equiv c) \lor$$

$$(X \equiv a \land Y \equiv c \land Z \equiv c) \lor$$

$$(X \equiv b \land Y \equiv a \land Z \equiv c) \lor$$

$$\neg R(X, Y, Z)]$$

Achievement: weak $(r, psec) \not\models_{DB} \Psi_1$, weak $(r, psec) \not\models_{DB} \Psi_2$

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Case Study 1: Alternative Instance Protecting Ψ_1

Policy:
$$psec = \{ \Psi_1 = R(a, b, c), \Psi_2 = R(a, c, c) \}$$

Alternative instance r^{Ψ_1} from adversary's POV:

i

$$\begin{array}{c|c} + & - \\ & (a, a, a) \\ (a, c, c) & (a, a, b) \\ (b, a, c) & \vdots \\ & (a, b, c) \\ & \vdots \end{array}$$
Question: Is r^{Ψ_1} credible from adversary's POV?

Adversary's view: $\mathcal{I}_{r^{\Psi_1}} \not\models_M \Psi_1$, $\mathcal{I}_{r^{\Psi_1}} \models_M \Psi_2$

Some Thoughts about Easy Cases

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Case Study 1: Indistinguishability of Instance r^{Ψ_1} Policy: $psec = \{ \Psi_1 = R(a, b, c), \Psi_2 = R(a, c, c) \}$

Adversary's simulation of weak $(r^{\Psi_1}, psec)$:



Disjunctive knowledge: $R(a, b, c) \lor R(a, c, c)$

 r^{Ψ_1} and r are indistinguishable: $weak(r^{\Psi_1}, psec) = weak(r, psec)$

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Case Study 1: Alternative Instance Protecting Ψ_2

Policy:
$$psec = \{ \Psi_1 = R(a, b, c), \Psi_2 = R(a, c, c) \}$$

Alternative instance r^{Ψ_2} from adversary's POV:

i.

+-
$$(a, b, c)$$
 (a, a, a) Question: Is r^{Ψ_2} credible from
adversary's POV? (b, a, c) \vdots Again: Simulation of
 $weak(r^{\Psi_2}, psec)$

Adversary's view: $\mathcal{I}_{r^{\Psi_2}} \models_M \Psi_1$, $\mathcal{I}_{r^{\Psi_2}} \not\models_M \Psi_2$

└─Some Thoughts about Easy Cases



Case Study 2: Given Setting

Policy:
$$psec = \{ \Psi_1 = R(a, b, c), \Psi_2 = R(a, b, d) \}$$

Original instance r:

+	-	R(a, b, c), R(a, c, c), R(b, a, c)
(a, b, c)	(a, a, a)	$(\forall X)(\forall Y)(\forall Z)$
(a, c, c)	(a, a, b)	$(X \equiv a \land Y \equiv b \land Z \equiv c) \lor$
(b, a, c)		$(X \equiv a \land Y \equiv c \land Z \equiv c) \lor$
	(a, b, d)	$(X \equiv b \land Y \equiv a \land Z \equiv c) \lor$
	:	$\neg R(X, Y, Z)$]

Obviously: $\mathcal{I}_r \models_M \Psi_1$, $\mathcal{I}_r \not\models_M \Psi_2$

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Inference-Proof Weakenings

Some Thoughts about Easy Cases

Case Study 2: Weakening Policy: $psec = \{ \Psi_1 = R(a, b, c), \Psi_2 = R(a, b, d) \}$ Weakening weak(r, psec):

$$\begin{array}{c|c}
+ & - \\
\hline
(a, b, c) & (a, a, a) \\
(a, c, c) & (a, a, b) \\
(b, a, c) & \vdots \\
\hline
(a, b, d) \\
\vdots
\end{array}$$

Disjunctive knowledge:

R(a, c, c), R(b, a, c) $R(a, b, c) \lor R(a, b, d)$ $(\forall X)(\forall Y)(\forall Z) [$ $(X \equiv a \land Y \equiv b \land Z \equiv c) \lor$ $(X \equiv a \land Y \equiv b \land Z \equiv d) \lor$ $(X \equiv a \land Y \equiv c \land Z \equiv c) \lor$ $(X \equiv b \land Y \equiv a \land Z \equiv c) \lor$ $\neg R(X, Y, Z)]$

 $R(a, b, c) \lor R(a, b, d)$

Achievement: weak $(r, psec) \not\models_{DB} \Psi_1$, weak $(r, psec) \not\models_{DB} \Psi_2$

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Some Thoughts about Easy Cases



Case Study 3: The Easy Case

Policy:
$$psec = \{ \Psi_1 = R(a, a, a), \Psi_2 = R(a, a, b) \}$$

Original instance r:

$$\begin{array}{c|c} + & - \\ \hline (a, b, c) & (a, a, a) \\ (a, c, c) & (a, a, b) \\ (b, a, c) & (a, a, c) \\ & \vdots \end{array}$$

Nothing to weaken!

Neither Ψ_1 nor Ψ_2 need to be protected.

$$\rightarrow$$
 weak (r, psec) := r

Obviously: $\mathcal{I}_r \not\models_M \Psi_1$, $\mathcal{I}_r \not\models_M \Psi_2$

Treating Non-Simple Confidentiality Policies



Clustering of Non-Simple Policies (1)

How to deal with non-simple policies of an arbitrary size?

- Partition the policy into a set of disjoint clusters
- ▶ For each cluster C: Construct disjunction $\bigvee_{\Psi \in C} \Psi$

How to achieve only meaningful disjunctions?

- Declare a set of admissible clusters
 - \rightarrow Employ high level languages such as SQL
- ► Goal: Each admissible disjunction should be well-balanced
 - Provide as much useful information as possible
 - All alternatives provided should be equally probable
- Only admissible clusters allowed in final disjoint clustering

Treating Non-Simple Confidentiality Policies



Clustering of Non-Simple Policies (2)

How to balance availability and confidentiality requirements?

- ► Size of cluster C induces length of disjunction \V_{Ψ∈C}Ψ
- In the following: Goal is to maximize availability
 - Keep size of clusters as small as possible
 - ► Only one alternative instance per potential secret required → Clusters of size 2 comply with security definition

└─ Treating Non-Simple Confidentiality Policies



(Partitioning)

Preparing the Clustering Algorithm

Requirements for clustering summarized

- 1. Each cluster is of size 2 (Maximizing availability)
- 2. Each cluster is admissible (Meaningful clusters)
- 3. Different clusters are pairwise disjoint
- 4. Each policy element is in a cluster

How to implement this efficiently on the operational level?

Model all admissible clusters within simple and undirected **Indistinguishability-Graph** G = (V, E) with

- ▶ V := psec
- $E := \{ \{ \Psi_1, \Psi_2 \} \in V \times V \mid \Psi_1 \lor \Psi_2 \text{ is admissible} \}$



Treating Non-Simple Confidentiality Policies



First Idea for Clustering Algorithm

Compute maximum matching M on indistinguishability-graph G

- ► $M \subseteq E$ is a matching on G, if each pair of different $\{\Psi_1, \Psi_2\}, \{\overline{\Psi_1}, \overline{\Psi_2}\} \in M$ is disjoint
- *M* is maximum if there is no matching *M*' with |M'| > |M|
- Is a maximum matching M on G the wanted clustering?
 - 1. Each cluster is of size 2 ✓
 - 2. Each cluster is admissible √
 - 3. Different clusters are pairwise disjoint ✓
 - There may be policy elements not contained in a cluster (Although matching is maximum)



└─ Treating Non-Simple Confidentiality Policies



Improved Idea for Clustering Algorithm

How to ensure that each policy element is in a cluster?

- Compute a maximum matching M
- Compute a matching extension M* of M
 - Initially: $M^* := M$
 - For each potential secret Ψ not covered by M
 - Create a suitable additional potential secret Ψ^A for Ψ
 - Add cluster $\{\Psi, \Psi^A\}$ to M^*

How to create a **suitable** additional potential secret Ψ^A for Ψ ?

- Create ground atom $\Psi^A = R(\mathbf{c})$
- Ensure that Ψ^A is not in the policy and not yet in M^*
- Ensure that $\Psi \lor \Psi^A$ would be admissible if Ψ^A was in policy



The Inference-Proof Weakening Algorithm



Creation of Weakened Instance

Assume: Clustering M_r^* is given s.t. for each cluster $\{\Psi_1, \Psi_2\}$ the original instance r satisfies Ψ_1 or Ψ_2

Construction of weakened instance weak (r, psec):

- ► Positive knowledge: Ground atom R(c) for each $c \in r$ with $R(c) \not\models_{DB} \Psi$ for each $\Psi \in \bigcup_{C \in M^*} C$
- ► Disjunctive knowl.: Disjunction $\Psi_1 \lor \Psi_2$ for each cluster $\{\Psi_1, \Psi_2\} \in M_r^*$
- Negative knowledge: Each constant combination neither in positive knowledge nor in a disjunction is not valid by completeness sentence

The Inference-Proof Weakening Algorithm



The Overall Algorithmic Approach

Algorithm to compute weakenings

Inputs: original instance r, confidentiality policy psec

- **Stage 1:** Clustering of potential secrets (independent of *r*)
 - Generate indistinguishability-graph G = (V, E) from *psec*
 - Compute maximum matching $M \subseteq E$ on G
 - Construct extended matching M* based on M
- **Stage 2:** Creation of weakened instance (dependent on *r*)
 - Create set of clusters with a policy element not obeyed by r: $M_r^* := \{ \{ \Psi_1, \Psi_2 \} \in M^* \mid \mathcal{I}_r \models_M \Psi_1 \text{ or } \mathcal{I}_r \models_M \Psi_2 \}$
 - Create weakened instance weak(r, psec) based on r and M_r^*

The Inference-Proof Weakening Algorithm



Example: Stage 2 of Weakening Algorithm Clustering: { {R(a, b, b), R(a, c, b)}, {R(a, b, c), R(a, b, d)} $\{R(b, b, b), R(b, b, e)\}, \{R(b, b, d), R(b, b, f)\}$ $\{R(c, a, a), R(c, a, b)^A\}$ Instance weak (r, psec): R(a, b, a)Instance r: $R(a, b, b) \vee R(a, c, b)$ $R(c, a, a) \vee R(c, a, b)$ $(\forall X)(\forall Y)(\forall Z)$ $(a, b, a) \mid (a, a, a)$ $(X \equiv a \land Y \equiv b \land Z \equiv a) \lor$ $(a, b, b) \mid (a, a, b)$ $(X \equiv a \land Y \equiv b \land Z \equiv b) \lor$ (a, c, b): $(X \equiv a \land Y \equiv c \land Z \equiv b) \lor$ (c, a, b) $(X \equiv c \land Y \equiv a \land Z \equiv a) \lor$ $(X \equiv c \land Y \equiv a \land Z \equiv b) \lor$ $\neg R(X, Y, Z)$

└─ The Inference-Proof Weakening Algorithm



Inference-Proofness: Sketch of Proof (1)

Consider arbitrary $\tilde{\Psi} \in psec$ Suppose: $\tilde{\Psi}$ is in cluster $\{\tilde{\Psi}, \tilde{\Psi}_I\}$

Case 1:
$$\mathcal{I}_r \not\models_M \tilde{\Psi} \lor \tilde{\Psi}_l$$

- Construct alternative instance $r^{\tilde{\Psi}} := r$
- $\blacktriangleright \ r^{\tilde{\Psi}} \text{ obeys } \tilde{\Psi}: \quad \mathcal{I}_{r^{\tilde{\Psi}}} \not\models_{M} \tilde{\Psi} \lor \tilde{\Psi}_{I} \quad \text{implies} \quad \mathcal{I}_{r^{\tilde{\Psi}}} \not\models_{M} \tilde{\Psi}$
- ► Indistinguishability: $r^{\tilde{\Psi}} = r$ by construction of $r^{\tilde{\Psi}}$ $\rightarrow weak(r^{\tilde{\Psi}}, psec) = weak(r, psec)$

└─ The Inference-Proof Weakening Algorithm



Inference-Proofness: Sketch of Proof (2)

Case 2: $\mathcal{I}_r \models_M \tilde{\Psi} \lor \tilde{\Psi}_l$

- ▶ Construct alternative instance $r^{ ilde{\Psi}} := (r \setminus {\{ ilde{\Psi}\}}) \cup {\{ ilde{\Psi}_l\}}$
- $r^{ ilde{\Psi}}$ obeys $ilde{\Psi}$: $\mathcal{I}_{r^{ ilde{\Psi}}}
 eq _{\mathcal{M}} ilde{\Psi}$ by construction of $r^{ ilde{\Psi}}$
- ► Indistinguishability: For each cluster $\{\Psi, \Psi_I\}$: $\mathcal{I}_{r^{\tilde{\Psi}}} \models_M \Psi \lor \Psi_I$ iff $\mathcal{I}_r \models_M \Psi \lor \Psi_I$
 - $\blacktriangleright \ \, \text{For cluster} \ \{\tilde{\Psi},\tilde{\Psi}_I\}: \quad \mathcal{I}_{r^{\tilde{\Psi}}}\models_M \tilde{\Psi} \vee \tilde{\Psi}_I \quad \text{by construction of} \ r^{\tilde{\Psi}}$
 - ► For each other $\{\Psi, \Psi_I\}$: $\mathcal{I}_{r^{\bar{\Psi}}} \models_M \Psi \lor \Psi_I$ iff $\mathcal{I}_r \models_M \Psi \lor \Psi_I$ by construction of $r^{\bar{\Psi}}$ and by disjoint clusters $\rightarrow weak(r^{\bar{\Psi}}, psec) = weak(r, psec)$

└─ The Inference-Proof Weakening Algorithm



Experimental Evaluation of Approach

About the prototype implementation

- Sample indistinguishability criterion based on local distortion
- Graph constructed with a flavor of merge-join algorithm
- Boost-Library employed for maximum matching computation

Lessons learned from evaluation of prototype

- Algorithm can handle instances and policies of realistic size
- Runtime of Stage 2 is negligible
- Runtime of Stage 1 is dominated by matching computation
- Stage 1 is significantly faster with matching heuristic → Slight loss of availability (→ more unmatched vertices)



Extending the Approach

Extending the Approach

A More Expressive Confidentiality Policy



Existentially-Quantified Atoms as Potential Secrets

Now: Improve expressiveness of potential secrets

Existentially quantified atoms like $(\exists \mathbf{X}) R(t_1, \ldots, t_n)$ in policy

- Each t_i is either a constant of Dom or a variable of X
- Each variable is existentially quantified
- Each variable occurs only once in t_1, \ldots, t_n

New difficulty arising: Too strong formulas

- Consider: $R(a, b, c) \lor (\exists X) R(a, b, X)$
- Adversary must believe R(a, b, c) to protect $(\exists X) R(a, b, X)$
- ▶ But: R(a, b, c) directly implies $(\exists X) R(a, b, X)$ *4*

- Extending the Approach

A More Expressive Confidentiality Policy



Cleaned Confidentiality Policy

Avoid too strong formulas by cleaning the policy

- Identify a maximum subset of logically weakest sentences (Without semantically equivalent sentences)
- Remove all other sentences from policy

Properties of cleaned confidentiality policy

- ► All alternatives provided by disjunctions are weakest sentences of policy → Do not imply other sentences of (original) policy
- Knowledge protected by removed stronger sentences is still protected by remaining weaker sentences

Extending the Approach

Introducing A Priori Knowledge



A Basic Kind of A Priori Knowledge

Usually: Adversary also has some a priori knowledge

- Set of sentences prior (containing database constraints)
- Original instance r must satisfy prior
- prior must not imply a sentence of the confidentiality policy

New difficulty arising: Each alternative instance must also satisfy *prior* to be credible

So far: Inference-proofness under *prior* of ground atoms R(c)

- ▶ R(c) satisfied by original instance ▶ R(c) does not imply a $\Psi \in psec$ $R(\mathbf{c})$ as **atom** in weakening
- Atoms of (positive part of) weakening in alternative instances



Conclusion & Future Work



Conclusion & Future Work

Our contribution:

- Approach creating inference-proof materialized views
- ► Therefore: Replace some definite information by disjunctions
- Limited expressiveness \rightarrow Efficient computation

Possible future work:

- Commonly used database constraints as a priori knowledge
 → Equality/Tuple Generating Dependencies
- Guarantee a certain number of k > 2 different "secure" alternative instances for each potential secret
- Elaborate connection to k-anonymity/ ℓ -diversity