

Inference-Proof Materialized Views

Doctoral Examination

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Context of this Work

Inference-Proof Data Publishing

Nowadays: Data publishing is ubiquitous

- ▶ Governments and companies provide data
- ▶ People share data about their private lives

But: Original data often contains sensitive (personal) information

- ▶ Set up a confidentiality policy
- ▶ Release “secure views” instead of original data
 - ▶ Do not reveal any confidential information
 - ▶ Consider adversary’s abilities to infer information

Framework and Goal

Framework: Relational model relying on first-order logic

- ▶ Complete original instance r (definite knowledge: $+/-$)
- ▶ Confidentiality policy $psec$ of potential secrets
($\exists \mathbf{X}$) $R(\mathbf{X}, \mathbf{c})$ s.t. each variable X occurs only once
- ▶ Adversary is aware of policy and protection mechanism

Goal: Enforce policy **efficiently** by weakened view on r s.t.

- ▶ Weakened view $weak(r, psec)$ contains only true knowledge
- ▶ Inference-proofness from adversary's point of view:
For each $\Psi \in psec$ there is a "secure" alternative instance r^Ψ
 - ▶ r^Ψ does **not satisfy** Ψ
 - ▶ r^Ψ is **indistinguishable** from original instance r
→ $weak(r^\Psi, psec) = weak(r, psec)$

Confidentiality by Weakening

Construction of Weakened Views

Stage 1: Disjoint disjunction templates (*independent of r*)

- ▶ Partition the policy $psec$ into disjoint clusters C_1, \dots, C_q (inducing disjunction templates) of a certain minimum size
- ▶ If necessary: Construct additional potential secrets

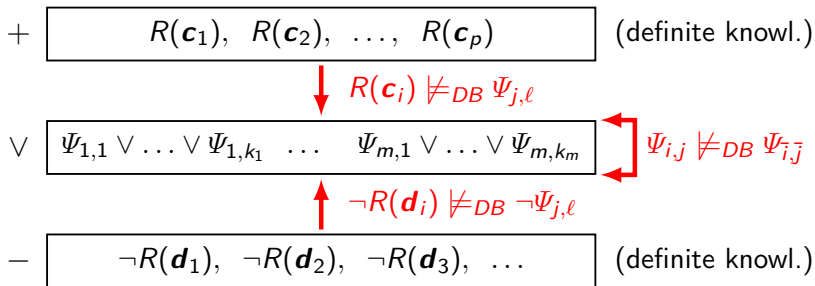
Stage 2: Weakened view $weak(r, psec)$ (*dependent on r*)

- ▶ Keep each tuple of r not satisfying any $\Psi \in C_i$
- ▶ Introduce each disjunction $\bigvee_{\Psi \in C_i} \Psi$ satisfied by r
- ▶ Knowledge not satisfying kept tuples or disjuncts is negative

→ Three classes of knowledge: +, \vee , -

Inference-Proofness by Isolation

Structure of weakened views:



Hence: For each $\Psi \in \Psi_{i,1} \vee \dots \vee \Psi_{i,k_i}$ alternative instance r^Ψ with

- ▶ $r^\Psi \not\models_M \Psi$ ✓ (but: $r^\Psi \models_M \Psi_{i,1} \vee \dots \vee \Psi_{i,k_i}$)
- ▶ $r^\Psi \models_M +, \vee, -$ \rightsquigarrow indistinguishability by construction of weakened views ✓

About the Clustering of Policy Elements

Desired properties for disjoint disjunction templates

- ▶ Credibility of all disjuncts \rightsquigarrow confidentiality
- ▶ Semantically meaningful \rightsquigarrow availability
- ▶ Certain length \rightsquigarrow level of confidentiality/availability

Desired properties for disjoint clustering of policy elements

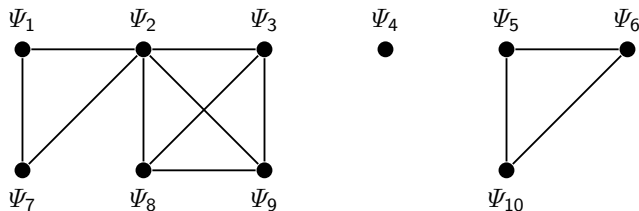
- ▶ Consider (high-level) specification of admissible clusters
→ Depends on application scenario
- ▶ Each cluster must have a certain (minimum) size k^*
- ▶ Minimize number of additional potential secrets

Clustering problem is NP-hard for $k^* \geq 3$ (Reduction of X3C)

Efficient Clustering for $k^* = 2$ (1)

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

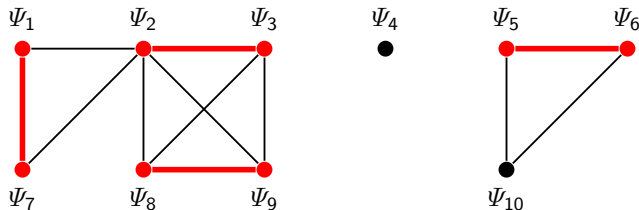
- ▶ $V := \{ \Psi \in psec \mid \Psi \text{ is to be clustered} \}$
- ▶ $E := \{ \{ \Psi, \Psi' \} \mid \Psi \vee \Psi' \text{ is admissible} \}$



Efficient Clustering for $k^* = 2$ (2)

Compute **maximum matching** on indistinguishability graph

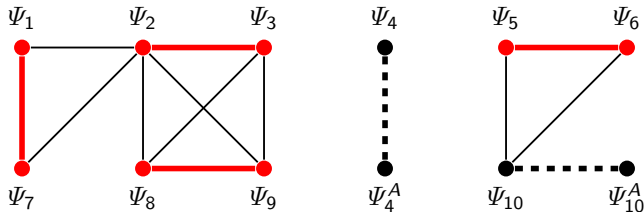
- ▶ Matching: Subset of pairwise vertex-disjoint edges
- ▶ Induces set of disjoint and admissible disjunction templates



Efficient Clustering for $k^* = 2$ (3)

How to handle policy elements not covered by the matching?

- ▶ Pair with **additional** (artificial) potential secrets
- ▶ Minimum number of these due to maximum matching



Inference-Proofness under A Priori Knowledge

Introducing A Priori Knowledge

Usually: Adversary also has some a priori knowledge *prior*

Challenge for inference-proofness: “secure” alternative instance r^Ψ

- ▶ r^Ψ does **not satisfy** Ψ
 - ▶ r^Ψ is **indistinguishable** from original r
 - ▶ r^Ψ **satisfies** *prior*
- } (already known)

Assumed *prior*: “Single Premise TGDs” of the form

$$\Gamma := (\forall \mathbf{X}) [R(\mathbf{X}, \mathbf{c}_1) \Rightarrow (\exists \mathbf{Y}) R(\mathbf{X}, \mathbf{Y}, \mathbf{c}_2)] \quad \text{s.t.}$$

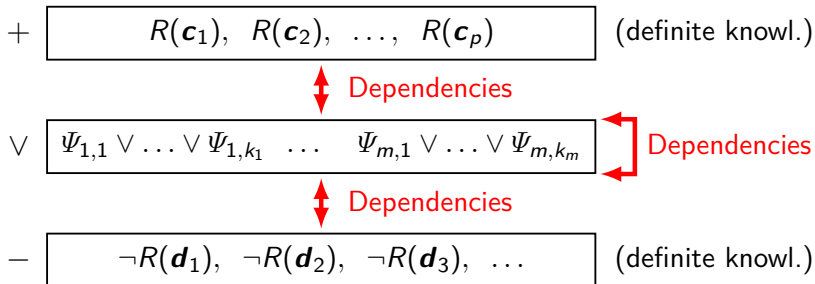
- ▶ each X occurs only once in $prem(\Gamma)$ and
- ▶ each X, Y occurs only once in $concl(\Gamma)$

Confidentiality Compromising Dependencies

Semantics of Single Premise TGDs: (also via transitive chains)

- ▶ Existent DB-Tuple \Rightarrow Existence of other DB-Tuple
- ▶ Non-Existent DB-Tuple \Rightarrow Non-Existence of other DB-Tuple

Broken isolation in weakened views:



Re-Establishing Sufficient Isolation (1)

Handling of dependency Γ interfering with policy elements

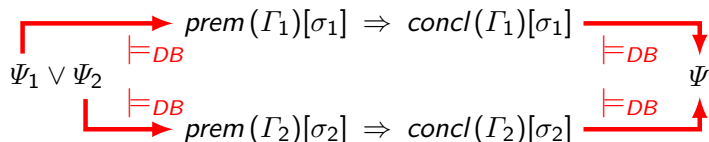
- ▶ Add policy elements protecting $prem(\Gamma)$ and $concl(\Gamma)$
→ Do not reveal satisfaction-status of premise or conclusion
- ▶ Attention: New policy elements \rightsquigarrow further interferences

Problem: Disjunctions do not always guarantee distortion
of non-satisfaction of conclusions

Only escape: Resort to distortion by complete refusal 

Re-Establishing Sufficient Isolation (2)

Inference-channel within disjunctive knowledge:



How to eliminate this kind of inference-channel?

- ▶ Partitioning of *prior* s.t. Γ_1 and Γ_2 in same partition, if
 - ▶ their conclusions imply the same Ψ (under some σ_1, σ_2) or
 - ▶ they can possibly form a transitive chain
- ▶ Do not construct disjunction, if
 - all disjuncts imply a premise of the same partition

Conclusion & Future Work

Conclusion & Future Work

Main contributions:

- ▶ Confidentiality by cooperative weakening without lies
- ▶ Even if adversary employs Single Premise TGDs
- ▶ Efficient computation for disjunctions of length $k^* = 2$
- ▶ Without *prior*: Confidentiality level can provably be varied

Possible future work:

- ▶ Clustering algorithm for $k^* \geq 3$ (\rightarrow Reasonable heuristic)
- ▶ More expressive classes of a priori knowledge
- ▶ Proof for different levels of confidentiality under *prior*
- ▶ Model k -anonymity/ ℓ -diversity within weakening approach

Backup Slides

Confidentiality by Weakening: Example (1)

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

Complete original instance r :

+			
-			
(a, b, c)	(a, a, a)	\implies	$R(a, b, c), R(a, c, c), R(b, a, c)$
(a, c, c)	(a, a, b)		$(\forall X)(\forall Y)(\forall Z) [$
(b, a, c)	\vdots		$(X \equiv a \wedge Y \equiv b \wedge Z \equiv c) \vee$
(a, b, d)	\vdots		$(X \equiv a \wedge Y \equiv c \wedge Z \equiv c) \vee$
\vdots	\vdots		$(X \equiv b \wedge Y \equiv a \wedge Z \equiv c) \vee$
			$\neg R(X, Y, Z) \quad]$

Obviously: r satisfies Ψ_1 (\rightarrow to be weakened)

Confidentiality by Weakening: Example (2)

Disjunction template: $\Psi_1 \vee \Psi_2 = R(a, b, c) \vee R(a, b, d)$

Weakened view $weak(r, psec)$:

+	-	
(a, b, c)	(a, a, a)	$R(a, c, c), R(b, a, c)$
(a, c, c)	(a, a, b)	$R(a, b, c) \vee R(a, b, d)$
(b, a, c)	⋮	(∀X)(∀Y)(∀Z) [
	(a, b, d)	$(X \equiv a \wedge Y \equiv b \wedge Z \equiv c) \vee$
	⋮	$(X \equiv a \wedge Y \equiv b \wedge Z \equiv d) \vee$
		$(X \equiv a \wedge Y \equiv c \wedge Z \equiv c) \vee$
		$(X \equiv b \wedge Y \equiv a \wedge Z \equiv c) \vee$
		$\neg R(X, Y, Z)$]

Disjunctive knowledge:

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

Achievement: $weak(r, psec)$ does **neither** imply Ψ_1 **nor** Ψ_2

Isolation within Disjunctive Knowledge

Policy of only ground atoms: Isolation due to disjoint clustering

But: Existential quantification in policy can break up isolation

- ▶ Consider: $\Psi_1 \vee \Psi_2$ with $\Psi_1 \models_{DB} \Psi_2$
- ▶ Then: $\Psi_1 \vee \Psi_2 \models_{DB} \Psi_2$ reveals validity of Ψ_2 ⚡
- ▶ Also harmful, if Ψ_1 and Ψ_2 stem from different disjunctions

How to re-establish isolation?

- ▶ Only weakest sentences of *psec* may occur in disjunctions
→ No implication between disjuncts
- ▶ Stronger policy elements still implicitly protected

Experimental Evaluation for $k^* = 2$

About the prototype implementation

- ▶ Criterion for admissible disjunctions: “Interchangeability”
- ▶ “Boost”-library for maximum matchings on general graphs

Lessons learned from 5 experiment setups

- ▶ Algorithm efficiently handles input instances of realistic size
- ▶ Size and structure of *psec* and *prior* crucial for runtime
- ▶ Low number of additional potential secrets and refusals
But: Admissibility criterion should fit to application scenario
- ▶ Parallelization: Doubling threads nearly halves runtime
- ▶ Clustering is significantly faster with matching heuristic
→ Only slight loss of availability