

Don't be a tattle tale: Preventing leakages through data dependencies on access control protected data

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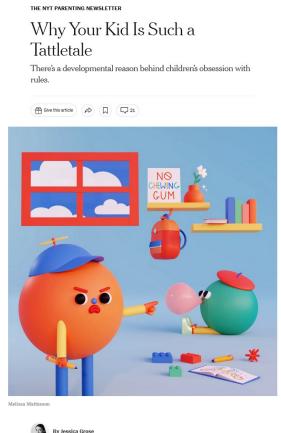








What is a Tattle-Tale?





To stop a child from tattling, you must understand why they are tattling in the first place. If they need attention, reassess how much attention you've been giving them and demonstrate an interest in them. If your child is struggling with social skills, help them



Tattle-Tale in Databases

	Eid	EName	Zip	State	Role	WorkHrs	SalPerHr
t_1	34	Tina	45678	WA	Student	20	40
t_2	56	Bobby	54321	CA	Faculty	40	200
t_3	78	Dale	53567	CA	Faculty	40	200
t ₄	12	Khan	54321	CA	Staff	30	70

[State = CA, Role] → [SalPerHr]

Tattle-Tale in Databases

	Eid	EName	Zip	State	Role	WorkHrs	SalPerHr
t_1	34	Tina	45678	WA	Student	20	40
t_2	56	Bobby	54321	CA	Faculty	40	
t ₃	78	Dale	53567	CA	Faculty	40	200
t ₄	12	Khan	54321	CA	Staff	30	70

Using [State = CA, Role] → [SalPerHr] and Dale's SalPerHr Bobby's SalPerHr can be inferred

Our Goal

Detect and prevent leakages due to data dependencies by hiding "minimal" number of cells

Hide Bobby's State for protecting his SalPerHr [State = CA, Role] → [SalPerHr]

	Eid	EName	Zip	State	Role	WorkHrs	SalPerHr
t_1	34	Tina	45678	WA	Student	20	40
t_2	56	Bobby	54321	CA	Faculty	40	200
t_3	78	Dale	53567	CA	Faculty	40	200
t_4	12	Khan	54321	CA	Staff	30	70

Our Goal

Detect and prevent leakages due to data dependencies by hiding "minimal" number of cells

Additionally hide Bobby's Zip for protecting his State

Zip→ State

	Eid	EName	Zip	State	Role	WorkHrs	SalPerHr
t_1	34	Tina	45678	WA	Student	20	40
t ₂	56	Bobby	54321	CA	Faculty	40	200
t_3	78	Dale	53567	CA	Faculty	40	180
t ₄	12	Khan	54321	CA	Staff	30	70

Extent of Leakage

- Tested on Tax dataset which contains address and tax information of individuals
- 14 attributes and 10 associated dependencies
 - E.g., if two persons live in the same state, the one earning a lower salary has a lower tax rate
- Salary attribute marked as sensitive and tested against a real-world adversary
 - Holoclean [VLDB2017] which is a state-of-the-art tool for inferring missing data.

Able to reconstruct the actual values of sensitive cells 100% of the time highlighting the importance of preventing leakages through dependencies

Prior Work

Weak security model. [Brodsky et al, TKDE'00]

Query time prevention

Not practical for query answering
[Miklau & Suciu, SIGMOD'04]

Perfect secrecy on views

Design time prevention

[Delugach&Hinke, TKDE'96], [Yip&Levitt, CSF'98], etc.

Poor data availability.

Randomized algorithms (DP/OSDP)

[Kotsogiannis et al, ICDE'20]

Suppresses too many cells.

None of the prior works have studied leakage on sensitive data due to data dependencies with strong security guarantees and practical utility.

Main Contributions

*Covered in this presentation

**Refer to the full paper

Formalizing leakage attack based on two types of data dependencies

- Denial Constraints
- Function-based Constraints

Defining a security model

- Tattle-Tale Condition for Leakage Detection
- Full Deniability
- Relaxation of the assumptions in the model

Developing algorithmic solutions to implement security model

- With focus on **Utility**, **Efficiency**, and Convergence
- Optimizations to improve performance
- Evaluated on 2 different datasets
- End-to-end System implementation in MySQL

Formalizing Leakage Attacks

Access Control Policies mark cells in the database as sensitive

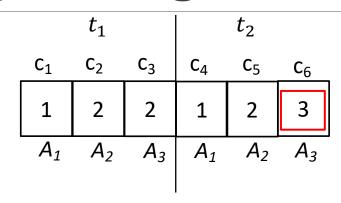
Formalizing Leakage Attacks

Data Dependencies causes the leakage

Expressed in the form of Denial Constraints (DCs)

$$\delta_1^{\sim}$$
: $\forall t_i, t_j \neg ((t_i[A] = t_j[A]) \land (t_i[B] \neq t_j[B]))$

Formalizing Leakage Attack



 $\forall A_i Dom(A_i) = \{1, 2, 3\}$

$$V_0$$
 c_1 c_2 c_3 c_4 c_5 c_6 $*$ $*$ $*$ $*$

V_1 c_1 c_2 c_3 c_4 c_5 c_6 c_6 c_7 c_8 c_9 c_9

Base view



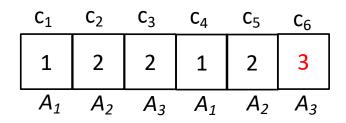
$$c_6 = \{1, 2, 3\}$$

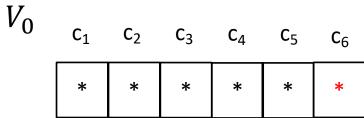


Adversary Infers

$$c_6 = \{1, 2, 3\}$$

Formalizing Leakage Attack





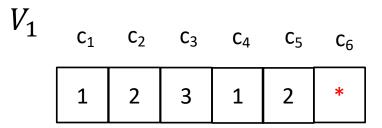
Base view

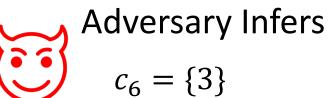
$$\delta^{\sim}: A_1 \to A_3$$

$$\delta_1: \neg(c_1 = c_4 \land c_3 \neq c_6)$$



$$c_6 = \{1, 2, 3\}$$





$$c_6 = \{3\}$$

Formalizing Leakage Attack

View V_2 achieves Full Deniability i.e., adversary is unable to infer nothing more than the base view V_0

$$V_2$$
 C_1
 C_2
 C_3
 C_4
 C_5
 C_6
 C_7
 C_8
 C_8

$$c_1$$
. $(c_1 - c_4 \land c_3 \neq c_6)$



Adversary Infers

$$c_6 = \{1, 2, 3\}$$



Adversary Infers $c_6 = \{1, 2, 3\}$

$$c_6 = \{1, 2, 3\}$$

What caused leakage?

Shared View

$$\delta_1: \neg(c_1 = c_4 \land c_3 \neq c_6)$$

Tattle-Tale is True when all the other predicates, except the one with the sensitive cell, evaluate as True

$$\neg((1 = 1) \land (3 \neq *)$$
True ?????

Truth value of the last predicate must be False in a clean database



$$c_6 = \{3\}$$

Remember that, in the base view we had c_6 = {1, 2, 3}

What prevented leakage?

Shared View

 V_2

$$\delta_1: \neg(c_1 = c_4 \land c_3 \neq c_6)$$

Tattle-Tale is False when at least 1 other predicate evaluate as False or Unknown.

$$\neg((*=1) \land (3 < *)$$

Either of the predicates could be False



$$c_6 = \{1, 2, 3\}$$

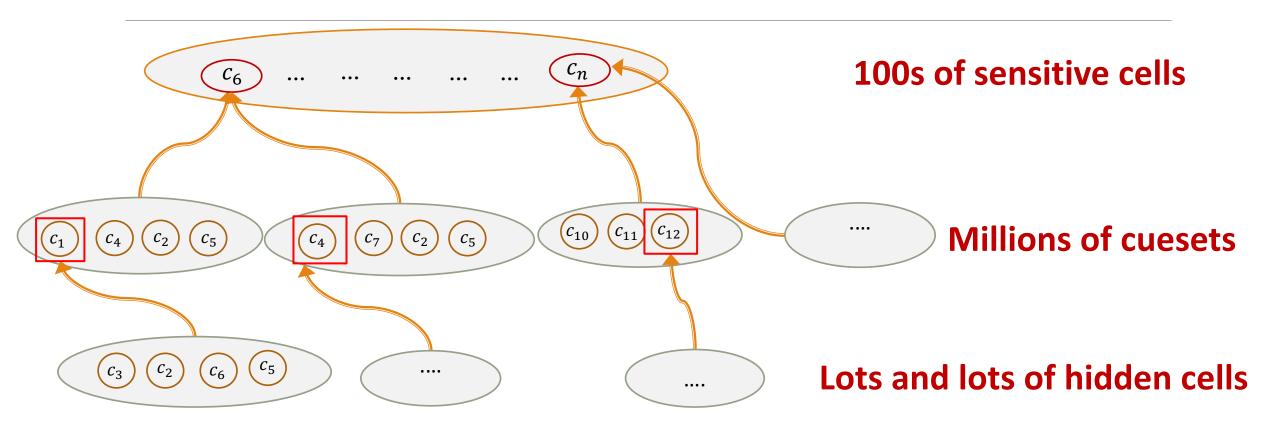
Same as in the base view



Security model

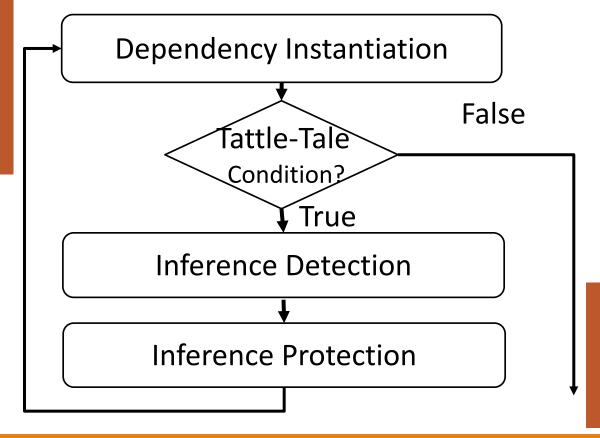
Full deniability is achieved for a shared view if for all the hidden cells in that view and their dependency instantiations, **Tattle-Tale Condition** is False.

The Tail of Tattle-Tales!



Our approach

Input: Database *D*,
Set of Sensitive/Hidden
Cells, Set of data
dependencies,



Output: View V that achieves full deniability

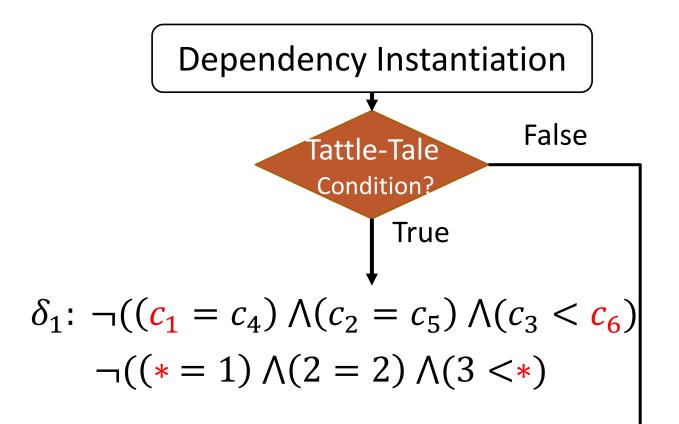
Step 1: Instantiations...!

Dependency Instantiation

$$\delta_1$$
: $\neg((c_1 = c_4) \land (c_2 = c_5) \land (c_3 < c_6)$
 δ_2 : $\neg((c_7 = c_4) \land (c_8 = c_5) \land (c_9 < c_6)$

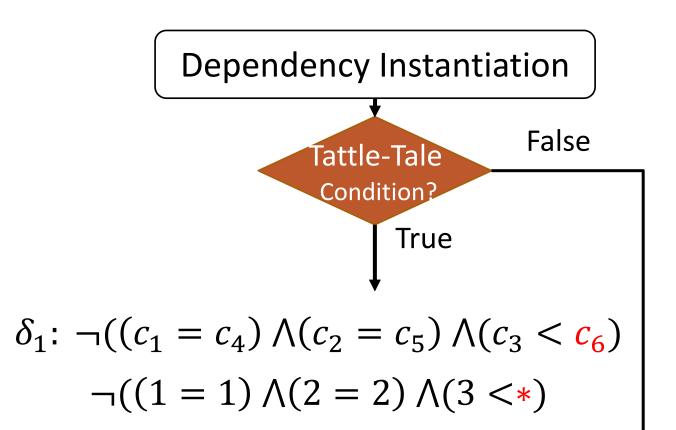
- For each hidden cell, instantiate all the dependencies
 Challenge: In the worst, there are
- Challenge: In the worst, there are $|D|^2$ instantiations for each sensitive cell
- **Solution**: We converted dependency instantiation operation into an efficient join query to reduce the complexity.

Step 2: Who are the Tattle-Tales?



- Check for each hidden cell and their dependency instantiations
- Termination Condition: If it returns *False* for all hidden cells and their dependency instantiations, then the view has achieved Full Deniability.

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- Check for each hidden cell and their dependency instantiations
- Termination Condition: If it returns *False* for all hidden cells and their dependency instantiations, then the view has achieved Full Deniability.
- If it returns *True* for at least 1 of them, then there is leakage

Step 3: Cue them up!

Dependency Instantiation

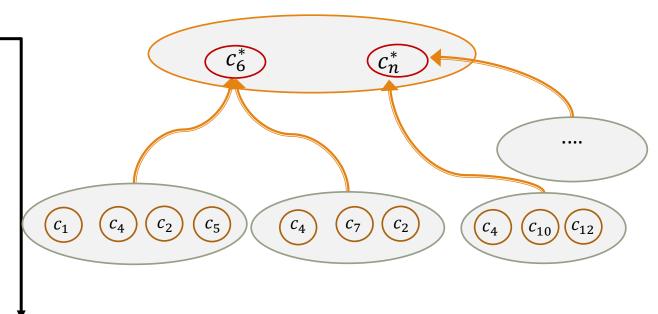
Tattle-Tale False

Condition?

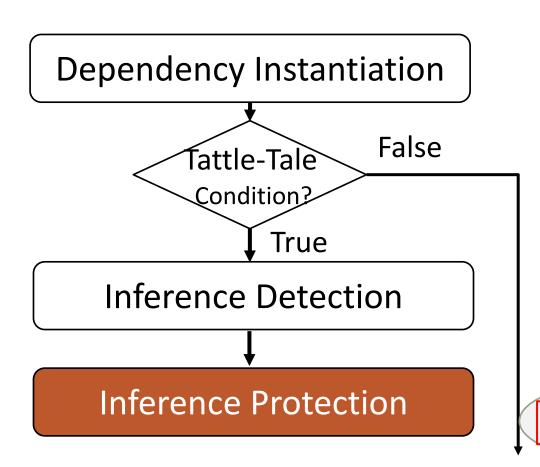
True

Inference Detection

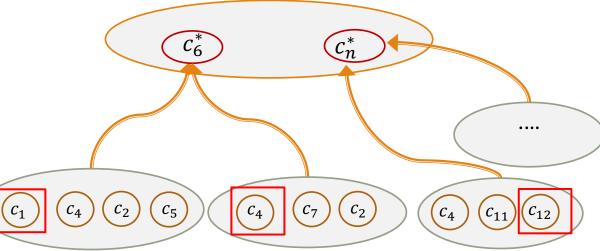
 Outputs cuesets for sensitive cells which satisfy the Tattle-Tale



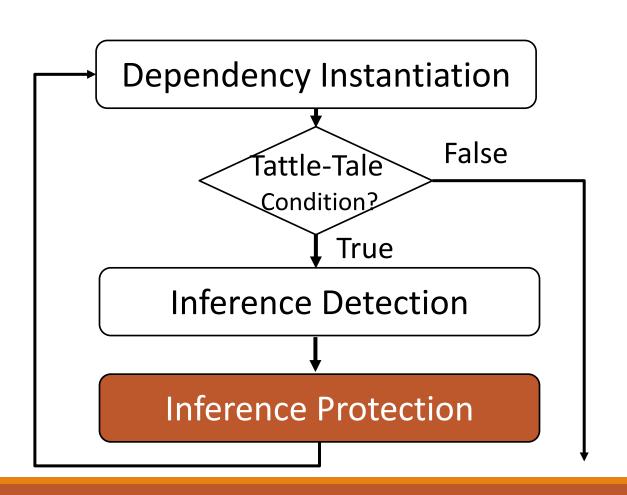
Step 4: Hide yo cells!



- Choose cells to hide from the cuesets
- Random Hiding leads to poor utility (Baseline)



Step 4: Hide yo cells, hide yo cells!



- Choose cells to hide from the cuesets
- Challenge: Selecting minimal cells to hide is NP-Hard.
- Use a greedy heuristic based on Minimum Subset Cover
- Run the approach again for newly hidden cells



Experimental Setup

Datasets: Tax dataset [1], (larger) Hospital dataset [2]

Dependencies: Using a data profiling tool [2]. 11 dependencies on Tax dataset and 14 dependencies on *Hospital dataset*

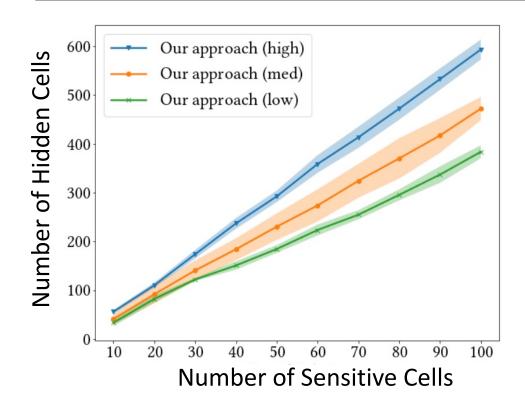
Baselines:

- Random Hiding for Inference Protection
- Oblivious of Tattle-Tale for Inference Detection

End-to-end implementation of the system with steps done at pre-processing

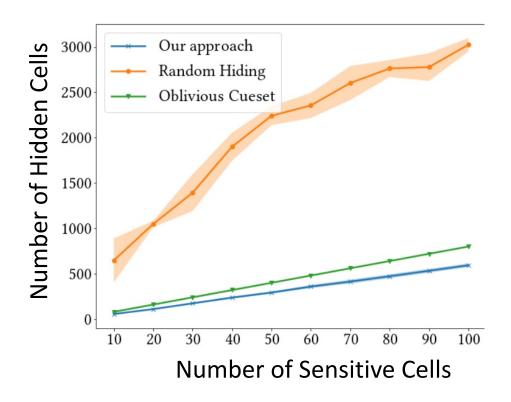
Source code available on Github

Impact of dependencies



If a sensitive cell participates in more dependencies, number of hidden cells increases!

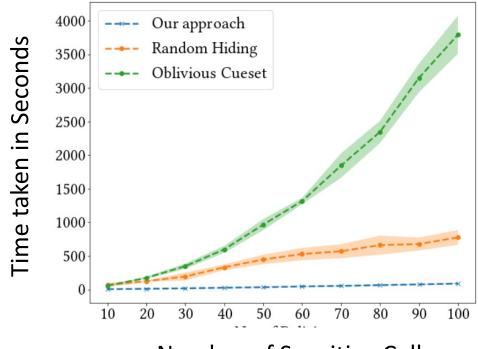
Utility Impact



Number of hidden cells increases linearly with our approach

Number of hidden cells increases exponentially when Random hiding used for Inference Protection!

Performance Impact



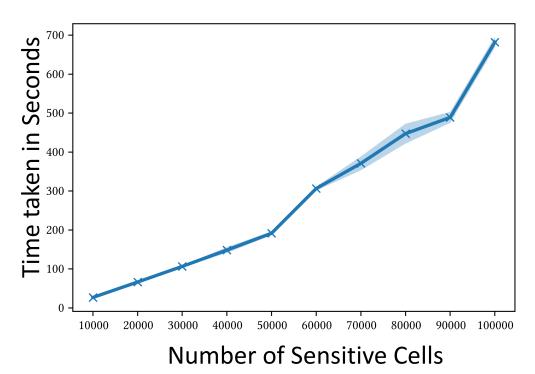
Number of Sensitive Cells

What happens if when compared against the baselines?

Overhead minimal in our approach

High overhead when Tattle-Tale condition not used for generating cuesets

Performance Impact



What happens if when size of the database is increased?

Our approach scales linearly with respect to size of the database

Takeaways

- Formalized a new type of leakage attacks based on dependencies such as Denial Constraints and Function-based Constraints
- Defined a new security model of Full Deniability (FD) and Tattle-Tale Condition for achieving FD
- Implemented algorithmic solutions for achieving FD on a given view
- Several new research directions
 - Leakage with soft dependencies
 - Combining FD with randomized response methods such as DP, OSDP to release non-sensitive data partially