

Checking for k-Anonymity Violation by Views

X. Sean Wang
The University of Vermont

Co-authors:

Chao Yao, Sushil Jajodia
George Mason University



Outline

- Motivation
- Problem definition
- Complexity of the general problem
- Polynomial cases
- Conservative checking methods
- Conclusion



Anonymity set and k-anonymity

- K-anonymity. Two versions:
 - Property k-anonymity: Given a set of k distinct *properties*, and a particular *person*. The adversary knows that the *person* has one of the k properties, but he does not know exactly which property the *person* has.
 - Person k-anonymity: Given a set of k distinct persons, and a particular *property*. The adversary knows that one of the k persons has the *property*, but she can't tell among the k persons who has that *property*.
- *Implicit assumption: probability of a particular association (between person and property) is rather small.*



Prior work

- Samarati and Sweeney, PODS 1998; Meyerson and Williams PODS 2005, and others.
- Problem studied: Given a *private/base table* (one tuple per person), how to “generalize” or “obfuscate” values so that adversary can only tell that each published tuple “originates” from at least k tuples in the private table.
- Example (Person 2-anonymity):

SSN	Problem
111-11-1111	P11
111-11-1112	P21
111-11-1123	P31
111-11-1124	P32

SSN	Problem
111-11-111*	P11
111-11-111*	P21
111-11-112*	P31
111-11-112*	P32



Property 2-anonymity

- Not handled by prior work (although techniques do apply).
- Hybrid solution:

SSN	Problem
111-11-1111	P11
111-11-1112	P21
111-11-1123	P31
111-11-1124	P32

SSN	Problem
111-11-1111	P?1
111-11-1112	P?1
111-11-112*	P31
111-11-112*	P32



Publishing with views

Name	Job	Salary	Problem
George	Manager	70K	Cold
John	Manager	90K	Obesity
Bill	Lawyer	110K	HIV

Private table P1

$$v_1 = \Pi_{Name, Job}(P_1)$$

Name	Job
George	Manager
John	Manager
Bill	Lawyer

$$v_2 = \Pi_{Job, Problem}(P_1)$$

Job	Problem
Manager	Cold
Manager	Obesity
Lawyer	HIV

v_1 and v_2 together: *Violation of property 2-anonymity!*



A little more complicated example

Name	Job	Salary	Problem
George	Manager	70K	Cold
John	Manager	90K	Obesity
Bill	Lawyer	110K	HIV

Private table P1

$\Pi_{Name} \sigma_{Salary > 80K} (P_1)$

Name
John
Bill

$\Pi_{Problem} \sigma_{80K < Salary < 100K} (P_1)$

Problem
Obesity

Name
George
John

$\Pi_{Name} \sigma_{Salary < 105K} (P_1)$



Functional Dependency: Name \rightarrow Problem

Name	Problem	Charge
George	Cold	20K
John	Obesity	20K
John	Obesity	30K
Bill	HIV	30K

Private table P2

$\Pi_{Name, Charge}(P_2)$

Name	Charge
George	20K
John	20K
John	30K
Bill	30K

$\Pi_{Charge, Problem}(P_2)$

Charge	Problem
20K	Cold
20K	Obesity
30K	Obesity
30K	HIV



Prior work on views

- Miklau and Suciu 2004; Dalvi, Miklau, and Suciu 2005; Deutsch and Papakonstantinou 2005; Dalvi and Suciu 2005 VLDB (“to some extent”).
- Probability models
- *Not* at the tuple level



Outline

- Motivation
- *Problem definition*
- Complexity of the general problem
- Polynomial cases
- Conservative checking methods
- Conclusion



Assumptions

- Provided to the public
 - View set v : a set of materialized views
 - View definitions (i.e., the queries)
- In addition, the “public” knows the constraints (FDs) on the private (base) table.
- Notation:
 - I^v is the set of all possible base/private table instances, each yielding (exactly) the given view set.



Assumptions (II)

- Two (fixed) attributes: ID and P (Property) on the base/private table
- The secret (to be protected) is the projection:

$$S(I) = \prod_{ID,P} (I)$$

- Looking for property k-anonymity



Definitions

- (Secret) association:
 - A binary tuple on (ID, P)
- An *association cover* A wrt a view set v is:
 - a set of associations,
 - all have the same ID value, and
 - for each I in I^v , $S(I) \cap A \neq \emptyset$.
- Intuition: If there exists A with $|A| < 2$, then there is “information leak”.
 - What if $|A| < k$?



Association cover example

Name	Job	Salary	Problem
George	Manager	70K	Cold
John	Manager	90K	Obesity
Bill	Lawyer	110K	HIV

Private table P1

$$v_1 = \Pi_{Name, Job}(P_1)$$

Name	Job
George	Manager
John	Manager
Bill	Lawyer

$$v_2 = \Pi_{Job, Problem}(P_1)$$

Job	Problem
Manager	Cold
Manager	Obesity
Lawyer	HIV

One association cover: $\{(Bill, HIV)\}$

Another: $\{(George, Cold), (George, Obesity)\}$



K-anonymity

Given a view set v and integer $k \geq 2$, we say v violates *k-anonymity* if there exists an association cover wrt v of size less than k .



Outline

- Motivation
- Problem definition
- *Complexity of the general problem*
- Polynomial cases
- Conservative checking methods
- Conclusion



Computationally hard

- With FD present, it is \sum_2^P -complete to test if a view set violates k-anonymity
- Data complexity
 - Complexity is in terms of the number of tuples



Outline

- Motivation
- Problem definition
- Complexity of the general problem
- *Polynomial cases*
- Conservative checking methods
- Conclusion



Polynomial case

- No FDs
- Selection and projection queries
 - Conjunctive selection conditions



Basic definitions

- *Tuple cover* for a view set v :
 - A set of tuples T such that for each I in I^v , $I \cap T \neq \emptyset$
- A_{min} : the set of all minimal association covers
- T_{min} : the set of all minimal tuple covers



Basic mechanism

- Given a view set v ,

$$A_{min} \subseteq \{\Pi_{ID,P}(T) \mid T \text{ in } T_{min}\}$$

- Why useful?

- if $|A| < k$ for A in A_{min} , then $|\Pi_{ID,P}(T)| < k$ for a T in T_{min}
- if $|\Pi_{ID,P}(T)| < k$ for T in T_{min} , then $|A| < k$ for an A in A_{min} .
 - $\Pi_{ID,P}(T)$ is an association cover by definition
 - minimality of A_{min}



Basic mechanism

- A *projection fact* (PF) is a tuple in a view (q_i, r_i) in the view set v
- *Tuple Set* for a PF p in a view (q_i, r_i) in v is the set of all the tuples t in I^v such that $q_i(t) = p$.
- $u(p)$: the tuple set for PF p



Basic mechanism

$$T_{min} \subseteq \{u(p) \mid p \text{ is a PF}\}$$



One more... we are there

- Given a tuple p in a view (q_i, r_i) (of a view set v with n views)
- $u(p)$ can be computed as the intersection of the following n sets
 - All the tuples t that returns p with q_i
 - All the tuples that returns a FP in the first remaining view, and tuples that do not satisfy the selection condition of that first remaining view,
 - ...
 - All the tuples that returns a FP in the last remaining view, and tuples that do not satisfy the selection condition of that last remaining view.



Going back to an example

$$v_1 = \Pi_{Name, Job}(P_1)$$

Name	Job
George	Manager
John	Manager
Bill	Lawyer

$$v_2 = \Pi_{Job, Problem}(P_1)$$

Job	Problem
Manager	Cold
Manager	Obesity
Lawyer	HIV

Let $p=(\text{George}, \text{Manager})$, then $u(p)$ consists of all the tuples that project to p , and project to a tuple in v_2 (*note there is no selection condition*).

Therefore, $u(p) = \{ (\text{George}, \text{Manager}, \text{Cold}),$
 $(\text{George}, \text{Manager}, \text{Obesity}) \}$



Going back to example 2

$\Pi_{Name} \sigma_{Salary > 80K} (P_1)$

Name
John
Bill

$\Pi_{Problem} \sigma_{80K < Salary < 100K} (P_1)$

Problem
Obesity

Name
George
John

$\Pi_{Name} \sigma_{Salary < 105K} (P_1)$

Let $p=(Obesity)$ in the right view, then $u(p)$ consists of all the tuples that satisfy:

- $80K < salary < 105K$,
- name=John or name=George (due to middle view; note selection condition must be satisfied).
- name = John or name=Bill (due to the left view)

Hence: $u(p) = \{(John, s, Obesity)\}$ where $80K < s < 100K$.



The algorithm

- Represent tuple sets for each projection fact as a formula (from selection condition, or it's complement)
- Perform all the intersections as indicated earlier
- Count the number of possible tuples in each intersection.
- Complexity: basically $|v|^n$, where n is the number of views and $|v|$ is the number of tuples in each view (data complexity).



With FDs

- Some special cases based on observations on FDs
- Consider two views in the view set
 - If an FD does not contain attributes from *both* views, then we can safely ignore this FD.
 - If the two view do not have common attributes and there is a single FD $ID \rightarrow P$, then checking is easy.
 - If the single FD is not $ID \rightarrow P$, checking is NP-complete.



Outline

- Motivation
- Problem definition
- Complexity of the general problem
- Polynomial cases
- *Conservative checking methods*
- Conclusion



Conservative algorithms

- Let $S_a(I) = \sigma_{ID=a}(\Pi_{ID,P}(I))$
- b_1 and b_2 are symmetric for a : Given (a, b_1) and (a, b_2) in $ID^D \times P^D$, if exactly one of the two is in $S_a(I)$, where I is in I^v , then there is I' in I^v such that $S_a(I')$ differs from $S_a(I)$ only in having the other association (among (a, b_1) and (a, b_2)).



K-anonymity

- Given a view set ν and a value a in ID^D , ν does not violate k-anonymity for a , if there exists I in I^ν , such that the following condition is satisfied: For each association (a, b) in $S(I)$, there exists a set of $k - 1$ distinct values b_i such that b_i is symmetric to b for a and (a, b_i) is not in $S(I)$.



Going back to example

$$v_1 = \Pi_{Name, Job}(P_1)$$

Name	Job
George	Manager
John	Manager
Bill	Lawyer

$$v_2 = \Pi_{Job, Problem}(P_1)$$

Job	Problem
Manager	Cold
Manager	Obesity
Lawyer	HIV

Given ID value John, Cold and Obesity are symmetric. Then for John, 2-anonymity is NOT violated.



Conclusion & future work

- Introduced k -anonymity violation for views
- Showed computational hardness of the problem
- Gave a polynomial algorithm for a no-FD case
- Provided a general approach for conservative algorithms
- Future work
 - Value obfuscation with views?
 - Experiments?
 - Duplicate preserving projection?
 - More complex views?

