Birds of a Feather Flock Together: How Set Bias Helps to Deanonymize You via Revealed Intersection Sizes (USENIX Security Symposium, August 2022)

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Set Membership Inference

Evaluation



- COVID-19 contact tracing
 - Alice: Health authority
 - Bob: Client
 - x_{alice} : A set Y of (tokens of) infected patients
 - x_{bob} : A set X of (tokens of) individuals in contact with
 - y_{bob} : $|X \cap Y|$



- Measurement of ad conversion revenue/lift
 - Alice: Publisher
 - Bob: Advertiser
 - ▶ x_{alice} : A table indexed by a set Y of (tokens of) individuals that click/view the ad
 - x_{bob} : A set X of (tokens of) converted individuals
 - ▶ y_{bob} : Ad conversion revenue/lift from intersecting converted individuals, and $|X \cap Y|$



From Bob's view

Intersection \Rightarrow Each token's membership regarding Alice's set \Rightarrow Each individual's relationship with Alice (*)

- (*): Linking a token to an individual is possible
 - ► COVID-19 contact tracing: Physical contact & token exchange logs¹²
 - Measurement of ad conversion revenue/lift: Sensitive tokens (e.g., email addresses, IMEI numbers)

'Yaron Gvili. Security analysis of the COVID-19 contact tracing specifications by Apple Inc. and Google Inc. (IACR). 'https://www.wired.com/story/apple-google-contact-tracing-strengths-weaknesses/. 2PC does NOT protect what can be deduced from one party's input and output!



Inference attacks in intersection-related analysis tasks

- COVID-19 contact tracing
- Measurement of ad conversion revenue
- Measurement of ad conversion lift
- Our observations
 - Existing 2PC protocols for these tasks reveal intersection size
 - These tasks need to be performed regularly
- More severe leakage if there is a non-weak set bias
 - ► Set bias: Alice's set tends to include individuals with certain features

Set Membership Inference

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SET MEMBERSHIP INFERENCE: PROBLEM DEFINITION

- Alice is the victim with a dynamic set $\{Y_0, \ldots, Y_i, \ldots\}$
- Bob is the attacker with a fixed set *X* of **target individuals**
- Bob can choose its set X_i in each protocol invocation
- Bob aims to determine whether a target individual has been in $Y = \bigcup_i Y_i$



A TOY INFERENCE ATTACK

- Each X_i contains only **one** target individual
- Learn its set membership from $|X_i \cap Y_i|$
- Require |X| protocol invocations



- Choose X_i with binary-tree-based strategy
 - Set *X* as the root, and randomly divide a node into two equal-size child nodes
 - Visit nodes via **priority-based** depth-first search (DFS), and set X_i as the current node
 - priority = intersection size (IS) / # target individuals in the node
 - ► IS in right child = IS in parent IS in left child



- ► A stronger attacker with some features regarding set bias
- Same as baseline attack, except that a node is divided using feature-based clustering
- Intuition
 - Clustering can put target individuals with similar features in the same sub-tree
 - A non-weak set bias \Rightarrow many member individuals are with similar features
- ► Implement clustering with **k-means**

- No set membership change of target individuals \Rightarrow Perfectly correct result
- Otherwise, there may be **false positives** and **false negatives**

Set Membership Inference

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Public data sources

- COVID-19 contact tracing: COVID-19 dataset of tested individuals in Israel
- Measurement of ad conversion revenue: Taobao's dataset of ad display/click records
- Measurement of ad conversion lift: Tencent's dataset of ad display records
- Frequency of protocol invocation
 - COVID-19 contact tracing: 5 / day
 - Measurement of ad conversion revenue: 1 / hour
 - Measurement of ad conversion lift: 1 / day

• Set bias (higher mutual information \Rightarrow stronger set bias regarding a feature)

Scenario	feature name (mutual information)	
COVID-19 contact tracing	fever (0.0168) , cough (0.0099) , gender (0.0004)	
Measurement of ad conversion revenue	age (0.0010), gender (0.0007), shopping_level (0.0002), work (0.0002), consumption_ability (0.0001), city_level (0.0001)	
Measurement of ad conversion lift	<pre>marriage_status (0.0013), education (0.0012), consumption_ability (0.0012), age (0.0009), work (0.0005), gender (0.0001)</pre>	

SUMMARIES OF THE VICTIM'S SET

Set size change



Simulation parameters

- Size |X| of the attacker's set
- Ratio β of # target individuals in the victim's **initial** set Y_0 to |X|

ſ	Vict	im's initial set Y_0	Public dataset	Universe
		Overlap of size $\beta \cdot X $		
	Attacker's simulated set of size $ X $			

► Feature-aware attacker can only use easy-to-collect features

Set membership leakage in COVID-19 contact tracing



Set membership leakage in measurement of ad conversion revenue



Set membership leakage in measurement of ad conversion lift



Set Membership Inference

Evaluation

- Set membership leakage does exist in the three scenarios
- COVID-19 contact tracing
 - ► Set membership regarding health authority ⇒ Whether a target token belongs to a COVID-19 patient
 - ► Can be combined with known linkage attacks in non-2PC settings ⇒ Risk of patient deanonymization
- Measurement of ad conversion revenue
 - ► Set membership regarding publisher ⇒ Whether a target individual has clicked the ad ⇒ Personal interest
- Measurement of ad conversion lift
 - ► Set membership regarding publisher ⇒ Whether a target individual has or would have seen the ad ⇒ Personal interest

- Limiting the number of 2PC protocol invocations
- Auditing intersection size
- Auditing the size of the attacker's set
- Applying differential privacy
- ▶ But there are also some challenges when using these defenses ...

- ► Set membership inference problem in intersection-size-revealing 2PC protocols
- A baseline attack and a feature-aware attack, where the latter outperforms the former given a non-weak set bias
- Evaluation in three scenarios with public datasets

Thank You

Contact xiaojie.guo@mail.nankai.edu.cn for any questions