A Study of Smartphone User Privacy from the Advertiser's Perspective



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Motivation

- "Free" apps are not entirely free: users pay the price of their privacy.
- The consequences of privacy leakages, especially when an advertiser gathers such private data across many apps, are not well studied.
- Deriving social and community information by using private data from multiple apps across users is possible, and needs more attention.

Contributions

- Modeling the *relationship inference process* in a three-layer framework by using the concept of connection, which is exemplified by two users sharing similar patterns in their leaked data.
- Conducting experiments with 10 participants and their families for over one-month time period to study the privacy leakages on using smartphones in their daily lives.
- Proposing an Activeness Based Profile as users' temporal privacy leakage profiles based on the experimental study.
- Verifying the generality of our findings from the real experiments based on trace-driven study using human mobility traces.

Concept of Connections

- The connection bridges the gap between the privacy leakage information and the users' relationship inference.
- Definition: a connection between two users exists if the same type of privacy leakage from the two users share certain spatial, temporal or content similarities.
- Contact list: share common contacts or in each other's contacts.
- Wi-Fi Access Point list: share common access points. \triangleleft
- > GPS/network-based location: leaked GPS or network-based locations are close.

Study the Consequences of Privacy Leakages (2) Trace-driven Experimental Study and Relationship study using Inference human mobility Profile Modeling traces User 1 Connection Derivation **Privacy leakage inference** Privacy framework Leakage

- Two types of social relationship:
- Fact-based Relationship: carry similar, regular and repetitive spatialtemporal connection patterns as dictated by the relationship (e.g., colleagues, classmates, roommates, and families).
- > Intelligence-based Relationship: does not necessarily carry regular patterns (e.g., friends).
- Utilizing the temporal and spatial patterns of the connections to classify the type of relationship, for example:
- > The connections of colleagues occur in work hours of weekdays.
- The connections of families in early morning and late night.
- > The connections of friends after working time and in weekends.

Real Experimental Study

- Involves 10 volunteer students and their family members over one month period including five types relationships (i.e., colleague, collaborator, classmate, friend, and family).
- Developed a tool to capture the privacy leakage information in real-time leveraging TaintDroid.
- Participants use their experimental smartphones at least three times a day with no restriction of how and when to use apps.



Activeness Based Profile

- Generating Profiles: derive the probability of each type of leakages happening in particular time windows for every participant.
- Categorize profiles with three representative user categories based on the hours that the particular user has leakages (i.e., active user, regular user, and inactive user).



- Applying the activeness based profiles to the Foursquare trace.
- Observations:
- An advertiser can achieve over 80% inference accuracy for both fact-based and intelligence-based relationship.
- The inference accuracy decreases for less active users, and longer observation windows help improve the accuracy.

Conclusions

- This work serves as the *first step* towards a comprehensive understanding of the advertiser's perspective.
- We seek to discover what an advertiser can infer about users' social relationships by combining different private data from many apps.
- We propose a privacy leakage inference framework that describes a general method for inferring users' social relationships, which can achieve high accuracy.







Aggregation