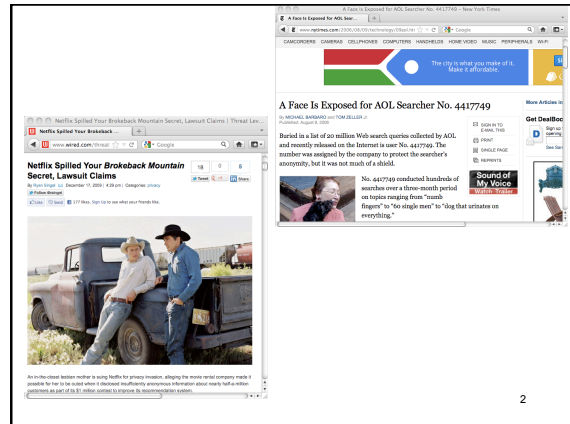
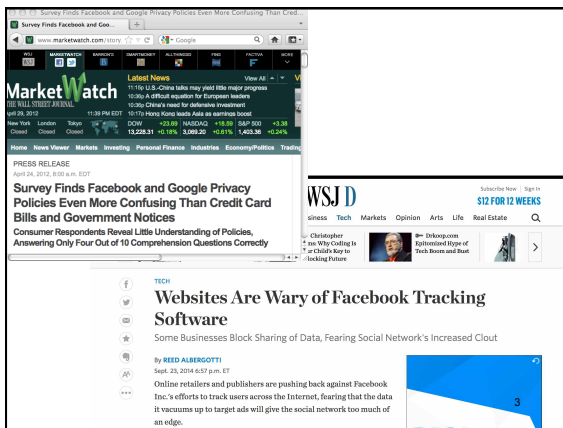


# Privacy

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## Exposing users: techniques

Look at

[You Might Also Like: Privacy Risks of Collaborative Filtering](#), Calandrino, J.A., Kilzer, A., Narayanan, A., Felten, E.W., and Shmatikov, V., *IEEE Sym. on Security and Privacy (SP)*, 2011, pp. 231 - 246.

- Various **item-to-item collaborative filtering** methods
- Practical algorithms

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## Set up

- **attacker** and **target user**
- attacker to **infer** unobservable **transaction** by target user
  - e.g. item purchased or rating given item
- attacker uses **“auxiliary information”** about some transactions of target user
- attacker **only observes**
  - does not enter ratings/ make transactions
  - no fake users

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## Sources of auxiliary information

- **provided by target system**
  - e.g. public ratings by user
- **“third-party sites”**
  - partner with target site
  - e.g. embed playlist on blog
- **other sites**
  - user places related content
  - e.g. Facebook user profile

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## “Generic Inference Attacks”

- Auxiliary information
  - target system provides **lists of related items**
  - target system provides **item-to-item covariance matrix** used by collaborative filtering
- Auxiliary information & Active attack
  - target system uses **k-nearest neighbor recommender**

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## Using related items

- system gives list of related items for each item based on user selection
- auxiliary items: attacker knows certain items associated with target user
- attacker
  - monitors related-items lists of auxiliary items
  - scores changes in lists:
    - new items appear or items move up on lists
  - if score for an item above threshold, infer item added to target user’s record

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## Using covariance matrix

- item-item covariance matrix  $M$  available
  - Hunch.com questions to users
- user record containing items interacted with
- auxiliary information: attacker knows subset  $A$  of items associated with target user  $u$ 
  - new item in record for  $u \Rightarrow$  covariances between new item and (some) items in  $A$  goes up
  - subset unique to target user?

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## Using covariance matrix, cont.

- attacker
  - monitors changes in covariance submatrix
    - columns for  $A$
    - rows  $A \cup \{\text{candidate new items}\}$
  - scores changes in submatrix
  - if score for an item above threshold, infer item added to target user’s record
- Lots of details concerning update delays in paper

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## Active attack: for kNN recommender systems

- Example target system
  - similarity measure on users
  - find  $k$  most similar users to user  $u$
  - rank items purchased by one or more of  $k$  most similar users
    - ranking by number times purchased
  - recommend items to  $u$  in rank order

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## kNN recommender systems, cont.

- auxiliary information: subset of  $m$  items target user  $U$  has purchased
  - claim  $m$  of about  $O(\log(\# \text{ users}))$  suffices
- attacker
  - creates  $k$  sibyl users
  - puts  $m$  auxiliary items on sibyls’ histories
    - “high probability” kNN of each sibyl is other  $k-1$  sibyls and  $U$
  - infer that any items recommended by system to any of sibyls and not one of  $m$  aux items is item  $U$  has purchased

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## Evaluation

- use
  - **yield**: number inferences per user per observation period
  - **accuracy**: percentage of inference that are correct
- need “ground truth”
- Several studies in paper
  - Hutch.com, LibraryThing, Last.fm

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## used on Amazon

- no ground truth
- API provides “Customers who bought x also bought y” and sales rank of items
- chose customers: top reviewers but not among top 1000 reviewers
- auxiliary info: entire set items previously reviewed by chosen customers
  - avg ~120 per customer
  - misses items purchased w/out reviewing

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## Inference for Amazon

- collected data for 6 mo
- only considered customers who reviewed in 6mo. before or during data collection
- each item, each user: retrieved top 10 most related items
- **infer**: customer purchased t if t appears or rises in related-items list associated with at least K auxiliary items for the customer
  - K parameter
- evaluate with **case studies**
  - find item later reviewed

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## Privacy issues in search, recommendations, and other information services

### In Practice:

- What is privacy?
- Kinds of problems?
- What problems are of concern?
- How address?

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