

I8734: Foundations of Privacy

Course Overview

Anupam Datta
CMU
Fall 2016

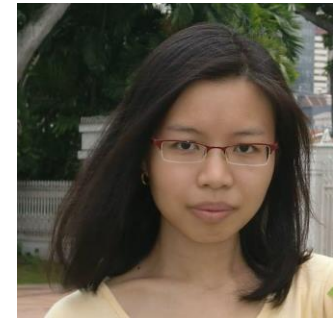
Logistics



Course Staff

- ▶ **Instructor: Anupam Datta**
 - ▶ Office: B23, 221 (SV)
 - ▶ Email: danupam@cmu.edu
 - ▶ Office hours: Mon 12-1PM Pacific at SV + Google Hangouts

- ▶ **TA: Lay Kuan Loh**
 - ▶ Office: CIC 2221E (Pittsburgh)
 - ▶ Office hours: Fri 10:30-11:30am Pacific in CIC 2206 + Skype



Extra office hours on demand

Logistics

- ▶ Lectures: Monday & Wednesday, 1:30-3:20 PM Pacific (usually 90 minutes)
- ▶ Recitation: Friday 9:30-10:20am Pacific (attend!)
- ▶ Web page: <http://www.ece.cmu.edu/~ece734/>
- ▶ Course blackboard (for grades)
- ▶ Piazza (for all other communication)
 - ▶ Please enroll; you should have received invitation
- ▶ Course work and grading:
 - ▶ Homework (60%) – 4 x 15%
 - ▶ Best 4 of 5 homeworks
 - ▶ Course project (30%)
 - ▶ Class participation (10%)

Logistics (2)

- ▶ **Course Project:**
 - ▶ Teams of 2 (form team by end of week)
 - ▶ Project proposal: 1-2 pages + in-class presentation (Sept 26)
 - ▶ Deliverable Part I + in-class presentation (Oct 31)
 - ▶ Deliverable Part II + Written report: 5-10 pages (Dec 5)
 - ▶ In-class presentation (Dec 5, 7)

Logistics (3)

Collaboration policy:

- ▶ You are allowed to discuss homework problems and approaches for their solution with other students in the class, but are required to figure out and write out detailed solutions independently and to acknowledge any collaboration or other source

[CMU Computing Policy](#)

[CMU Academic Integrity Policy](#)

Logistics (4)

Example Violations:

- ▶ Submission of work completed or edited in whole or in part by another person.
- ▶ Supplying or communicating unauthorized information or materials, including graded work and answer keys from previous course offerings, in any way to another student.
- ▶ Use of unauthorized information or materials, including graded work and answer keys from previous course offerings.
- ▶ ...not exhaustive list

If in doubt, ask me!

Logistics (6)

- ▶ Consent form for video recording lectures

Prerequisites

- ▶ An undergraduate course equivalent to 15-251 is required or permission of instructor
- ▶ An introductory course in computer security such as 18-487, 18-630, or 18-730 is recommended, but not required
- ▶ If in doubt, please talk to me after class
- ▶ Quick class poll

Personal Information is Everywhere



Google

facebook



amazon.com



flickr® from YAHOO!

Privacy and Fairness Problems

WHAT THEY KNOW

When the Most Personal Secrets Get Outed on Facebook

By GEOFFREY A. FOWLER

Google's iPhone Tracking

Web Giant, Others Bypassed Apple Browser Settings for Guarding Privacy

By JULIA ANGWIN And JENNIFER VALENTINO-DEVRIES

February 17, 2012

WHAT THEY KNOW

Websites Vary Prices, Deals Based on Users' Information

By JENNIFER VALENTINO-DEVRIES, JEREMY SINGER-VINE and

ASHKAN SOLTANI

December 24, 2012

The screenshot shows the top navigation bar of The New York Times website. On the left, there are links for 'SECTIONS', 'HOME', and 'SEARCH'. The 'The New York Times' logo is centered. On the right, there are buttons for 'SUBSCRIBE NOW', 'LOG IN', and a settings gear icon. Below the navigation bar is the 'TheUpshot' section header. To the right of this header are social media icons for Facebook, Twitter, and RSS, with the text 'FOLLOW US: GET THE UPSHOT IN YOUR INBOX'. The main article title is 'When Algorithms Discriminate'. Below the title is the author's name 'Claire Cain Miller @clairecm' and the date 'JULY 9, 2015'. At the bottom right of the article preview are social media sharing icons for Facebook, Twitter, Email, and Print, along with a bookmark icon and a comment count of '147'.

Privacy Solutions

Wednesday, May 21, 2014

CARNEGIE MELLON, MICROSOFT RESEARCH AUTOMATE PRIVACY COMPLIANCE FOR BIG DATA SYSTEMS

Search Engine Code Is Moving Target That Eludes Manual Audits

NEWS

Google's RAPPOR aims to preserve privacy while snaring software stats

**APPLE'S 'DIFFERENTIAL
PRIVACY' IS ABOUT COLLECTING
YOUR DATA—BUT NOT YOUR
DATA**

Organizing Questions

- ▶ **What is privacy?**
 - ▶ From philosophical and legal conceptions to computer science and engineering
 - ▶ Inspiration from conceptions, but greater precision often through greater specificity

- ▶ **How can we protect privacy?**
 - ▶ Beyond creating laws and institutions
 - ▶ Computational mechanisms for privacy protection

Privacy Problems

Module I: Privacy through Accountability

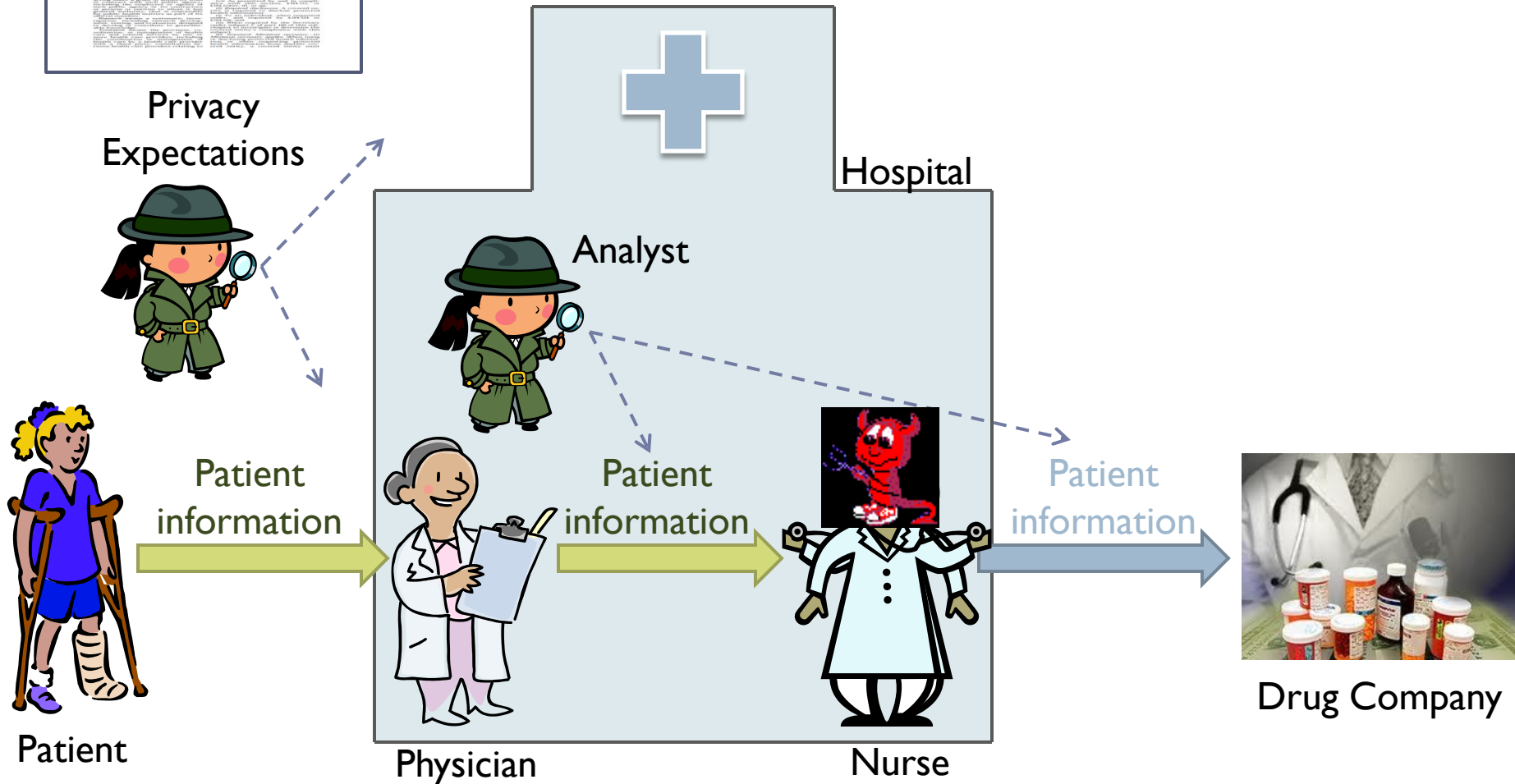
Research Challenge



Programs and People

Ensure organizations respect privacy expectations in the collection, use, and disclosure of personal information

Healthcare Privacy



Example from HIPAA Privacy Rule

A covered entity may disclose an individual's protected health information (phi) to law-enforcement officials for the purpose of identifying an individual if the individual made a statement admitting participating in a violent crime that the covered entity believes may have caused serious physical harm to the victim

▶ Concepts in privacy policies

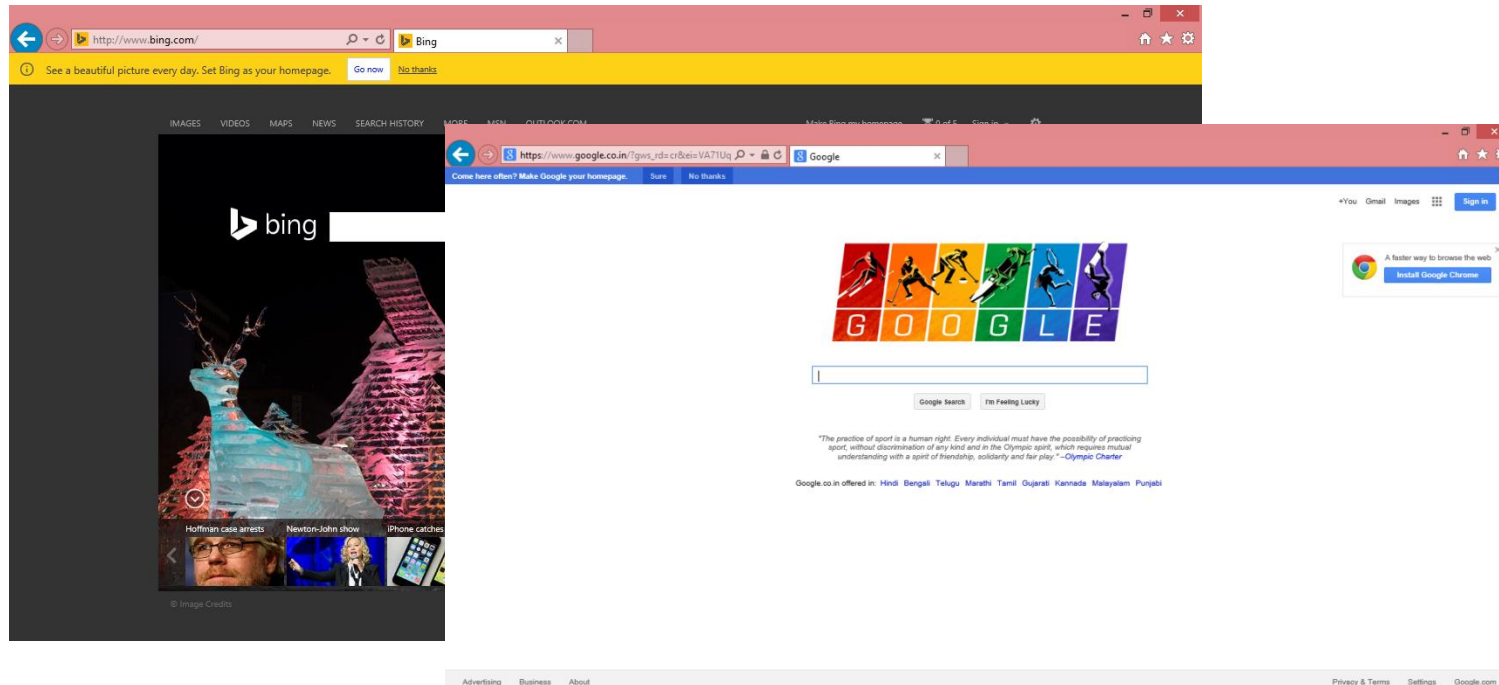
- ▶ **Actions:** send(p1, p2, m)
- ▶ **Roles:** inrole(p2, law-enforcement)
- ▶ **Data attributes:** attr_in(prescription, phi)
- ▶ **Temporal constraints:** in-the-past(state(q, m))

- ▶ **Purposes:** purp_in(u, id-criminal)
- ▶ **Beliefs:** believes-crime-caused-serious-harm(p, q, m)

Black-and-white concepts

Grey concepts

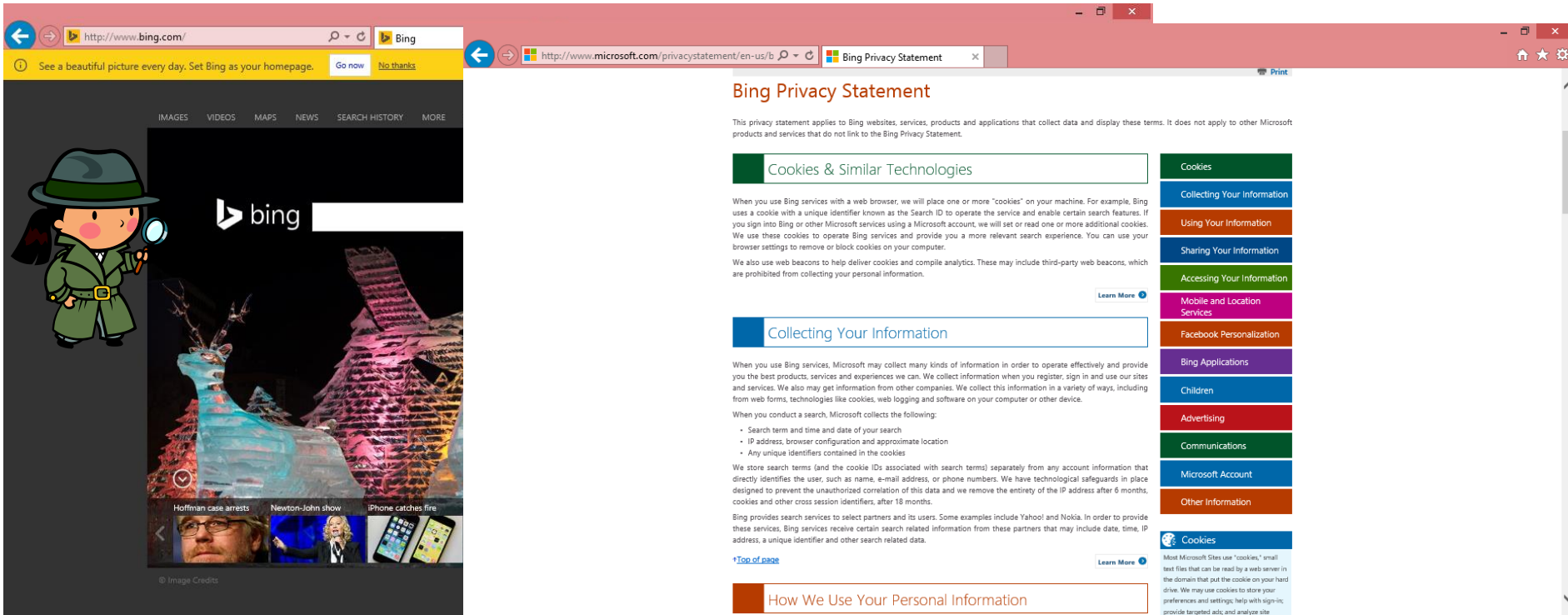
Web Advertising



Example privacy policies:

- ▶ Not use detailed location (full IP address) for advertising
- ▶ Not use health information for advertising

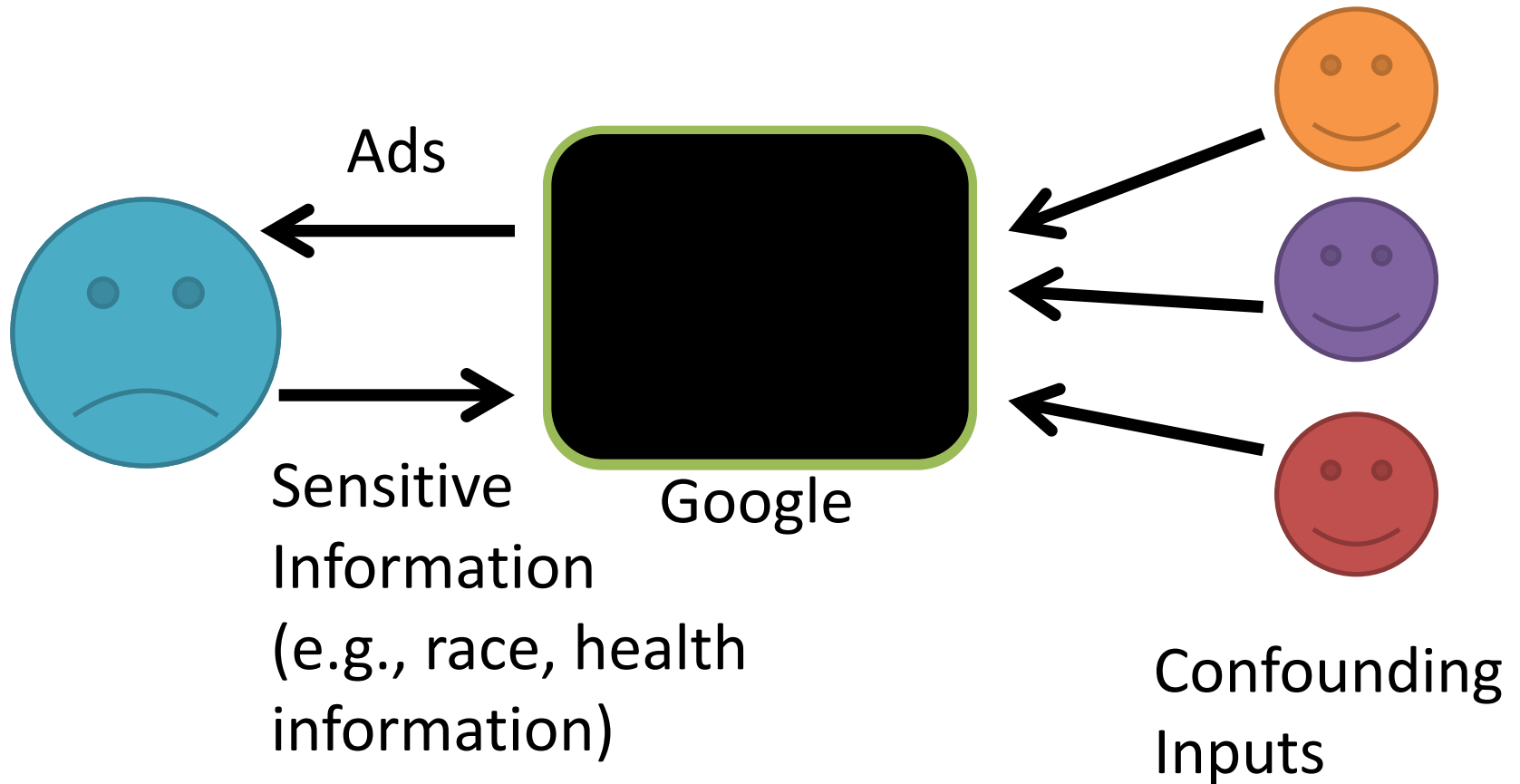
Privacy Compliance for Bing



Setting:

- ▶ Auditor has access to source code

Web Privacy: Advertising



Web Privacy: Online Tracking

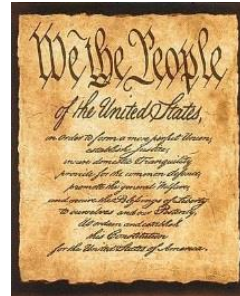


64

Independent tracking mechanisms on average on top-50 sites

Module I: Privacy through Accountability

- ▶ **Formalize Privacy Policies**
 - ▶ Precise semantics of privacy concepts
(restrictions on personal information flow)
- ▶ **Enforce Privacy Policies**
 - ▶ Accountability
 - ▶ Detect
 - ▶ Explain
 - ▶ Correct



<http://www.andrew.cmu.edu/user/danupam/privacy.html>

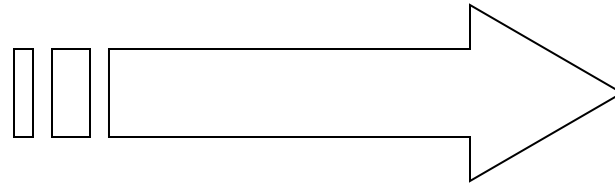
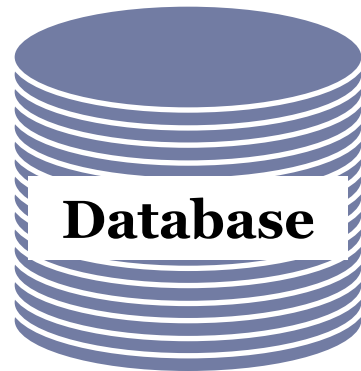
Module I: Learning Outcomes

- ▶ Understanding of real-world privacy policies and laws
- ▶ Methods for detecting privacy violations

- ▶ Experience with audit tools for healthcare privacy
- ▶ Experience with web tracking investigation tool

Module II: Protecting Privacy and Fairness in Big Data Analytics

Database Privacy Goals



Government,
marketers,
researchers, ...

- Health records
- Census data
- Web search records

Conflicting goals:

- Provide useful **information**
- Protect **individual privacy**



[CNET](#) › [News](#) › [Corporate & legal](#)

August 7, 2006 9:59 AM PDT

AOL apologizes for release of user search data

By [Dawn Kawamoto](#) and [Elinor Mills](#)

Staff Writers, CNET News

Last modified: August 7, 2006 2:30 PM PDT

Related Stories

[Should Google be forced to hand over data?](#)

March 14, 2006

[Judge to help feds against Google](#)

March 14, 2006

AOL apologized on Monday for releasing search log data on subscribers that had been intended for use with the company's newly launched research site.

The randomly selected data, which focused on 658,000 subscribers and posted 10 days ago, was among the tools intended for use on the recently launched AOL Research site. But the Internet giant has since removed the search logs from public view.

stories

submissions

popular

blog

all stories

ask slashdot

book reviews

games

idle

yro

Anonymity of Netflix Prize Dataset Broken

Posted by **Zonk** on Tuesday November 27, 2007 @10:23AM
from the there-are-degrees-of-anonymity dept.



KentuckyFC writes

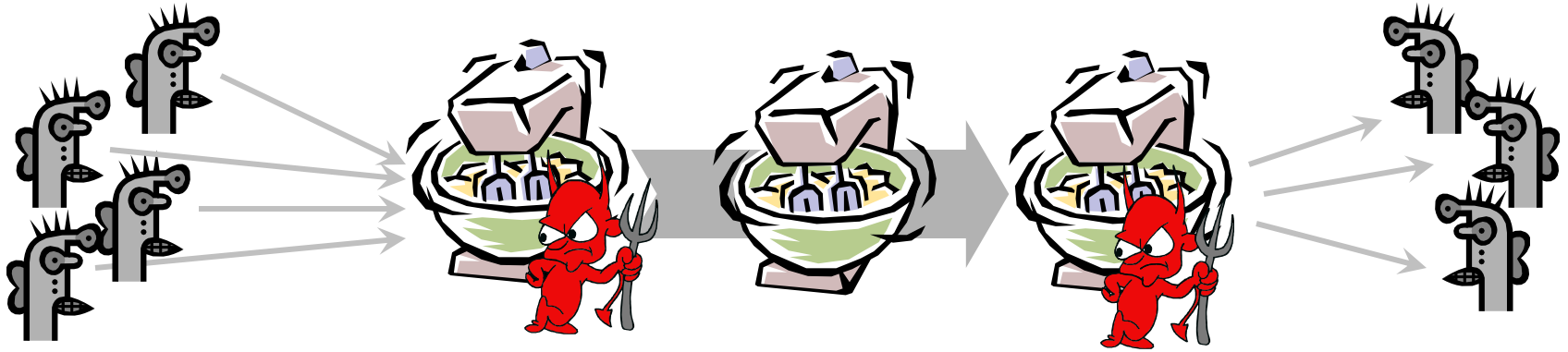
"The [anonymity of the Netflix Prize dataset has been broken](#) by a pair of computer scientists from the University of Texas, according to a report from the physics arXivblog. It turns out that an individual's set of ratings and the dates on which they were made are pretty unique, particularly if the ratings involve films outside the most popular 100 movies. So it's straightforward to find a match by comparing the anonymized data against publicly available ratings on the Internet Movie Database (IMDb) ([abstract on the physics arxiv](#)). The researchers used this method to find how individuals on the IMDb privately rated films on Netflix, in the process

Module II: Learning Outcomes

- ▶ Understanding of pitfalls in anonymizing databases
- ▶ Understanding of methods for releasing privacy-preserving statistics and their limitations
- ▶ Understanding bias in machine learning and corrective measures
- ▶ Understanding transparency (explanations) for decisions of machine learning systems

Module III: Cryptographic Mechanisms for Privacy Protection

Anonymous Communication



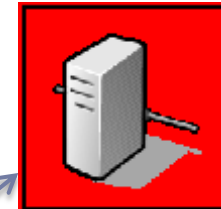
Anonymous Credentials



Organization



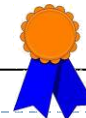
Alice



Service

“I have a cred from
Org saying
WA resident
Age >18”

Cred from Org
Name Alice
Address
Birthdate
Birthplace
Citizenship
...

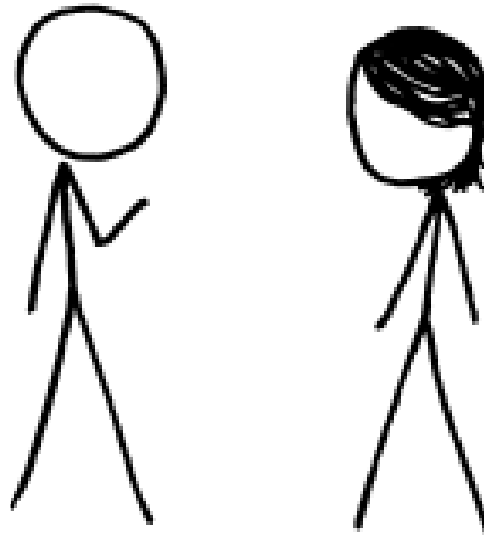


- Cannot
 - Identify Alice
 - Learn anything beyond the info she gives
 - Distinguish two users with the same attributes
 - Link multiple uses of the same credentials

Secure Two-Party Computation

Bob's Genome: ACTG...
Markers (~1000): [0,1, ..., 0]

Bob



Alice's Genome: ACTG...
Markers (~1000): [0, 0, ..., 1]

Alice



$$x = f(g_A, g_B)$$

Can Alice and Bob compute a function of their private data, without exposing anything about their data besides the result?

Module III: Learning Outcomes

- ▶ **Understanding of cryptography behind**
 - ▶ Anonymous communication
 - ▶ Anonymous credentials (zero-knowledge)
 - ▶ Biometric identification (secure computation)

Fall 2014 Course Projects

- ▶ Studies of personal information usage by Web services
 - ▶ Study on Facebook ads
 - ▶ Price Discrimination
 - ▶ Recommendations for news articles
 - ▶ Effect of cookies on Google ads
- ▶ Analytics to discover information usage by Web services
 - ▶ Abstaining Machine Learning
 - ▶ Ensemble Machine Learning
- ▶ Privacy Protecting the New York Taxicab Dataset
- ▶ Defense against Canvas Fingerprinting on the Web
- ▶ Privacy and Security issues of Android ads
- ▶ ML (Lasso Regression) over Encrypted Big Data



Fall 2015 Course Projects

- ▶ Secure Modular Embedding: Comparing Signals without revealing them
- ▶ Robust Ad Collection
- ▶ Inversion Attack on Machine Learning Models
- ▶ Privacy in Election Campaigns
- ▶ Improving Usability of Private Browsing Mode
- ▶ Investigating gender discrimination in popular employment websites
- ▶ Comparing Privacy Tools
- ▶ Google Advertising Platform Case study
- ▶ The Unexpected Danger of Multiple Social Media Accounts: Instagram and Twitter Reveal More than You Think
- ▶ Effects of Browser-Type on Internet Results

An Organizing Viewpoint

Privacy as a right to *restrictions on personal information flow*

Student Introductions

- ▶ Who are you?
- ▶ Why are you here?

Homework for Next Class

- ▶ Read the Fair Information Practices Principles

<http://www.oecd.org/internet/ieconomy/oecdguidelinesontheprotectionofprivacyandtransborderflowsofpersonaldata.htm>

- ▶ Critically read the entire privacy policy of a Web services company of your choice
 - ▶ Examine pairs of services owned by the same company (e.g., Facebook-Whatsapp)

Homework Continued

Discussion questions:

- ▶ Try to find one example of a piece of the policy that maps to each principle.
- ▶ Can you find examples of principles that are not reflected in the policy?
- ▶ Can you find examples of policy clauses that reflect a principle that is not included in these principles?
- ▶ Are there policy clauses that could be more restrictive or less restrictive with respect to information use in order to better adhere to the principles?
- ▶ Are there parts of the policy that are too vague? If so, suggest alternatives.
- ▶ Are there conflicts in policies of service pairs owned by the same company?

Thanks! Questions?