

Evolution of Privacy Loss in Wikipedia

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Motivation

- Social media and online privacy are two of today's hot topics.
- Given that we know that digital traces reveal more than users might think [**Kosinski, PNAS 2013**], we ask the next questions.
- Goals of this work:
 - Does online user's privacy degrade over time?
 - What factors contribute most to revealing private traits?
 - Can I stop leaking personal information if I stop posting online?



Content of this presentation

Presentation outline

- Case study: Wikipedia dataset
- Profiling of editing behavior
- Measuring predictability of personal traits
- Marginal utility of features over time
- Conclusion and the way ahead

Case study: Wikipedia

Why Wikipedia?

- 13 years long, public: ideal for longitudinal study;
- tens of thousands of editors, of different geographic locations, religious, educational and political backgrounds;
- *apparently harmless dataset*: a reservoir of knowledge, no focus on personal information.

Dataset dimensionality:

- 188,805,088 revisions
- 117,523 editors
- 8,679 editor badges
- 22,172,813 edited pages
- 430,410 page categories
- Time extent: January 2001 - July 2013.

Encoding editing behavior (1)

Editor activity profiles:

- *basic set*: #revisions over 6 predefined categories (Wiki namespaces);
- *extended set*: adding Wikipedia's 23 high level thematic categories (Math, Geography, History etc.)

	Feature name
Basic feature set	CONTENT
	TALK-C
	USER
	TALK-U
	WIKI
	INFRA

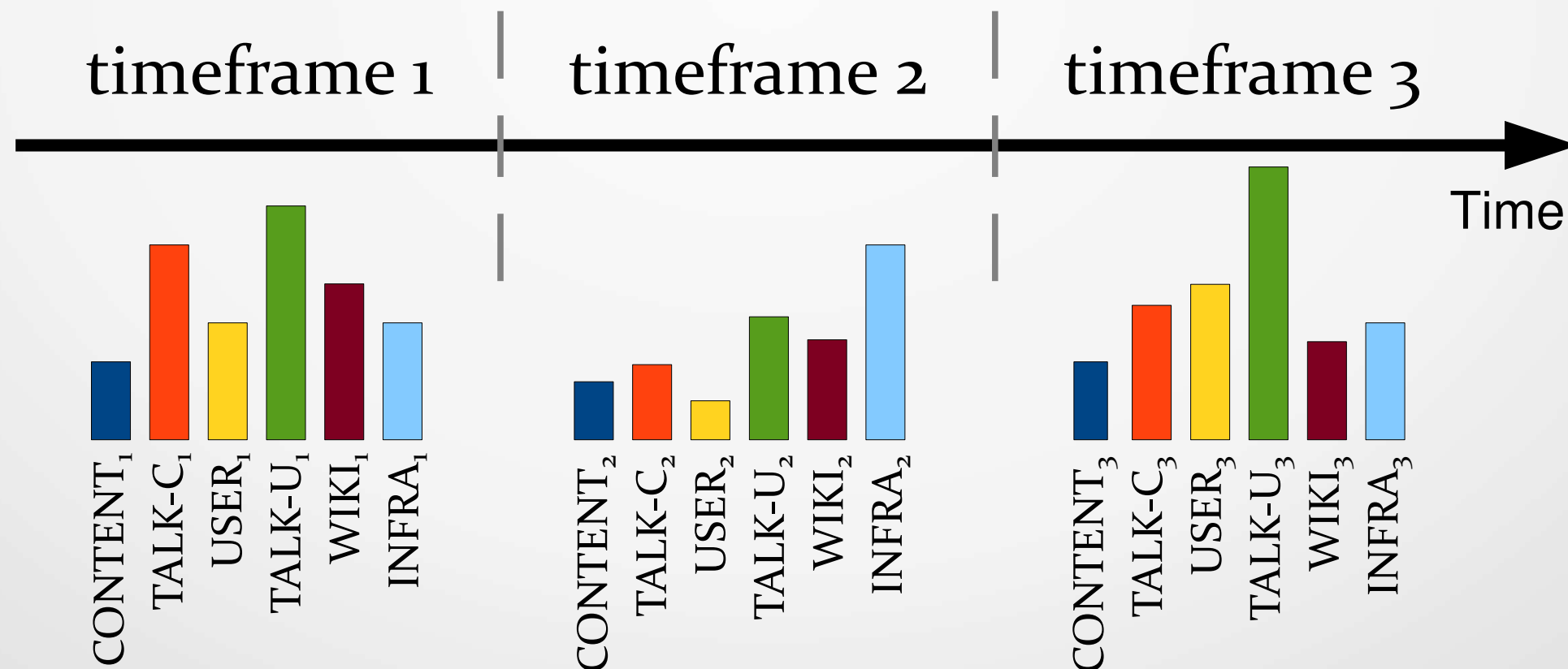
Editor personal information:

- Extracted from the badges editors put on their editor pages;
- **Gender** (6936 out of more than 117k), **ethnic origin**, **religious views** (7685), **education** (9224), sexual orientation *etc.*



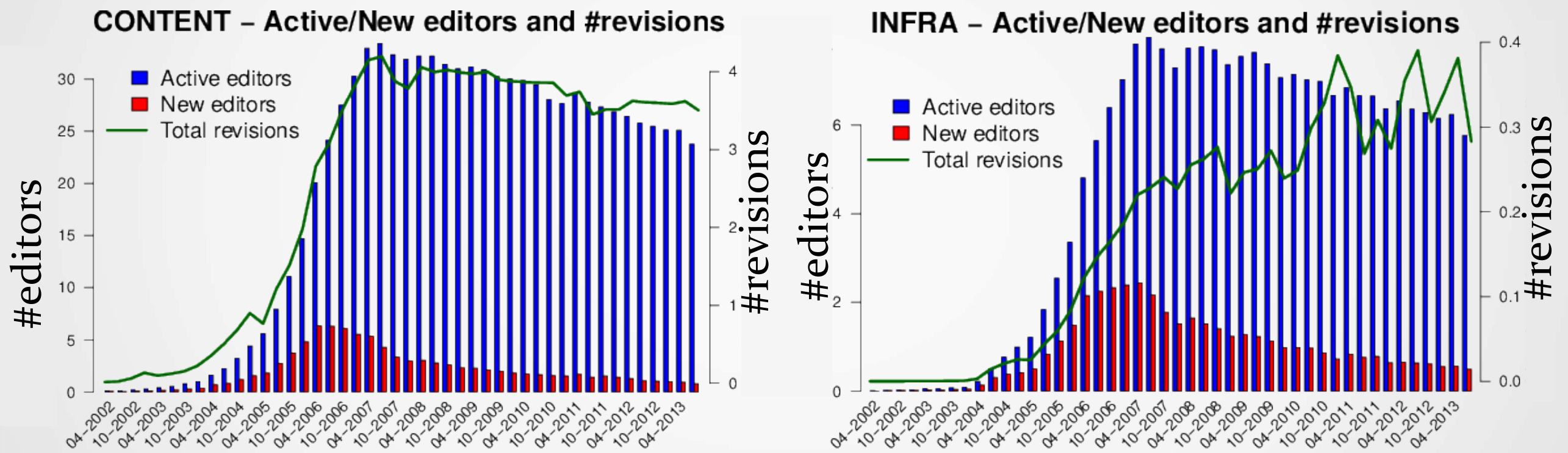
Encoding editing behavior (2)

- 3-month timeframes
- description for each editor per timeframe, each feature counts revision over categories
- feature set temporally embeds increasing amounts of information



Profiling of editing behavior (1)

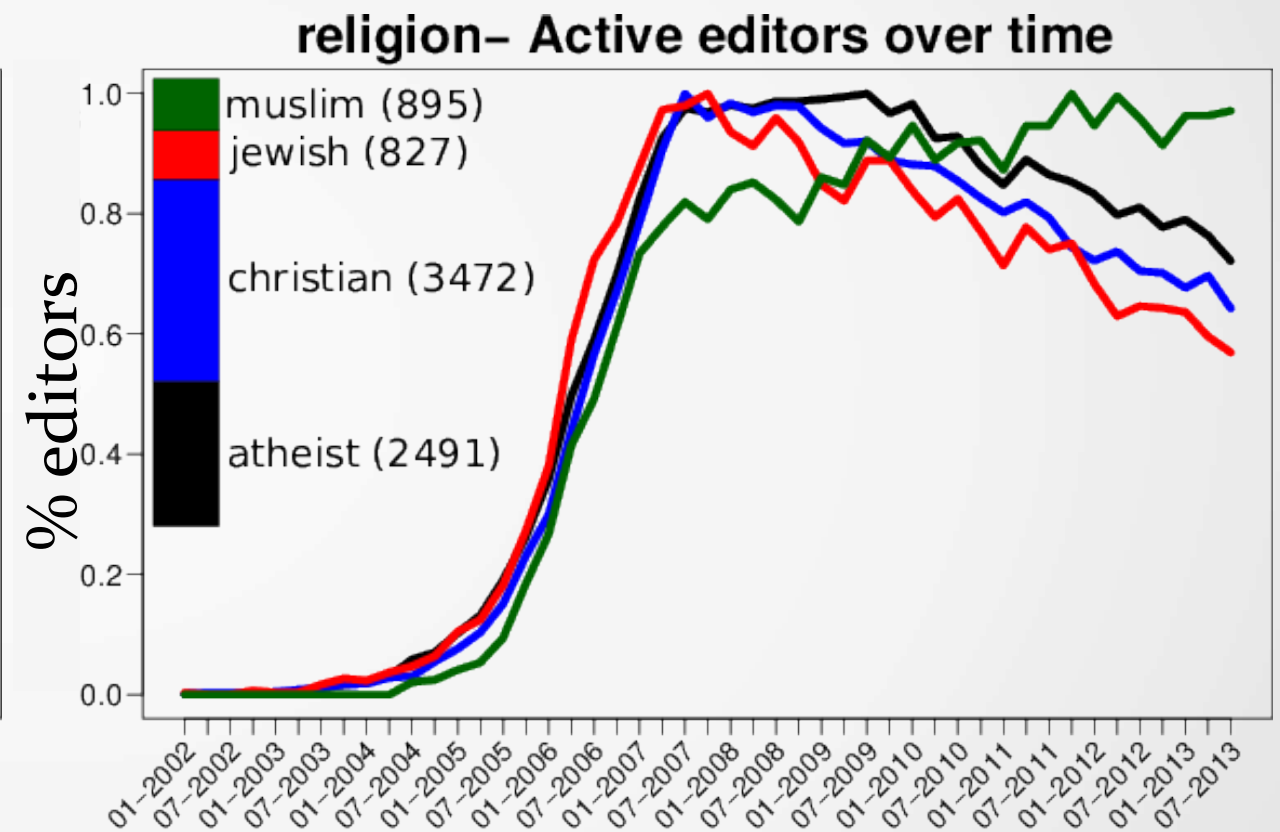
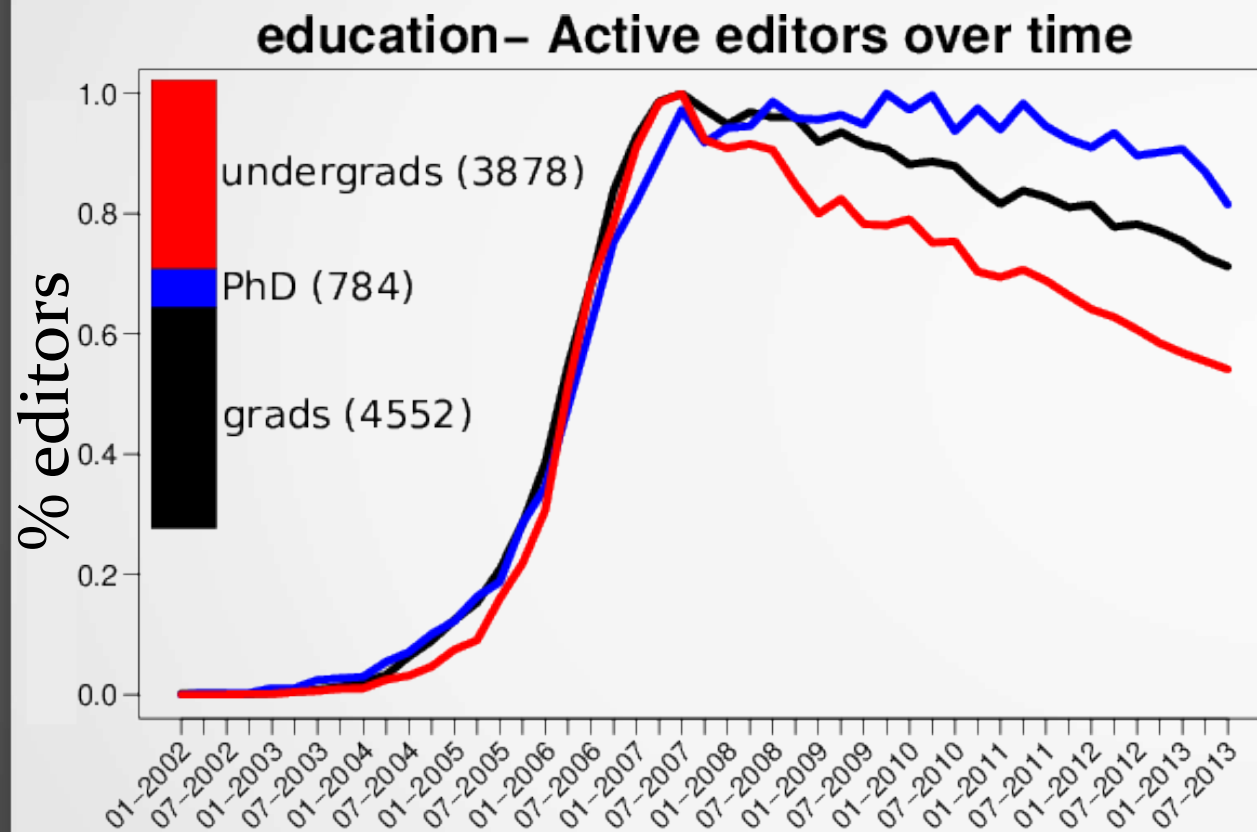
Decline of editorship and rise of maintenance



While the Wikipedia “slowdown” has been previously reported [[Suh '09](#), [Halfaker '12](#)], we break down this evolution per category and detect a *rise of maintenance effort*

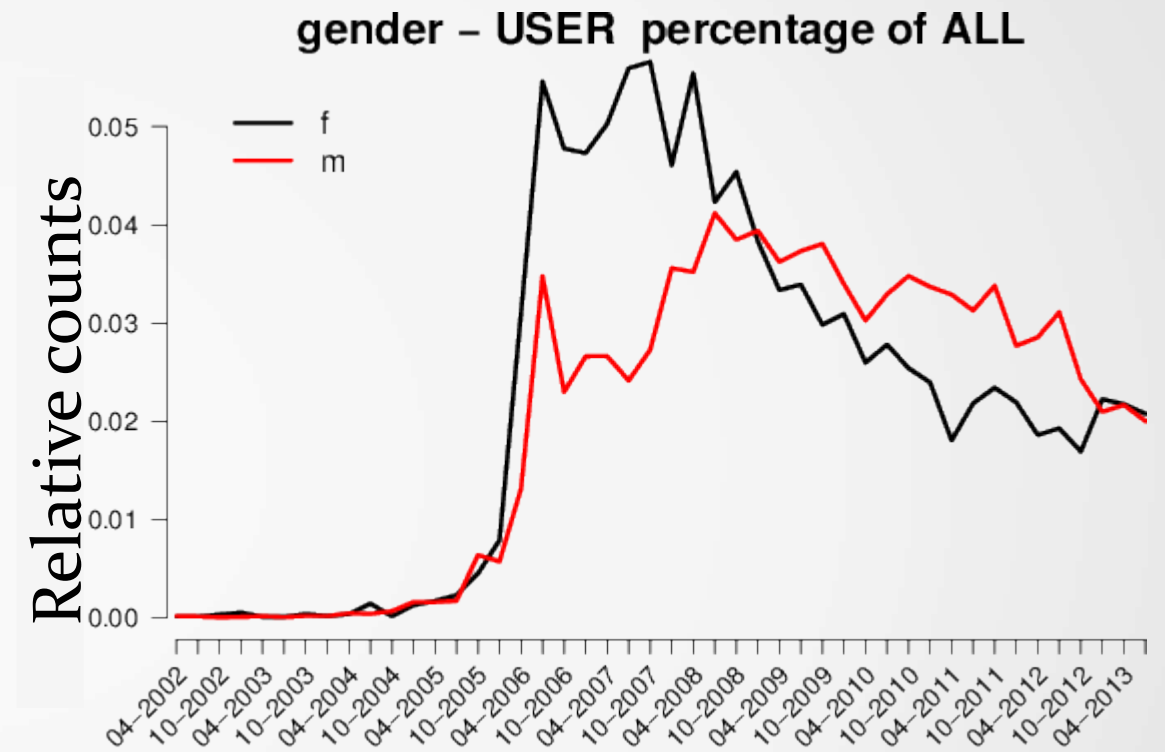
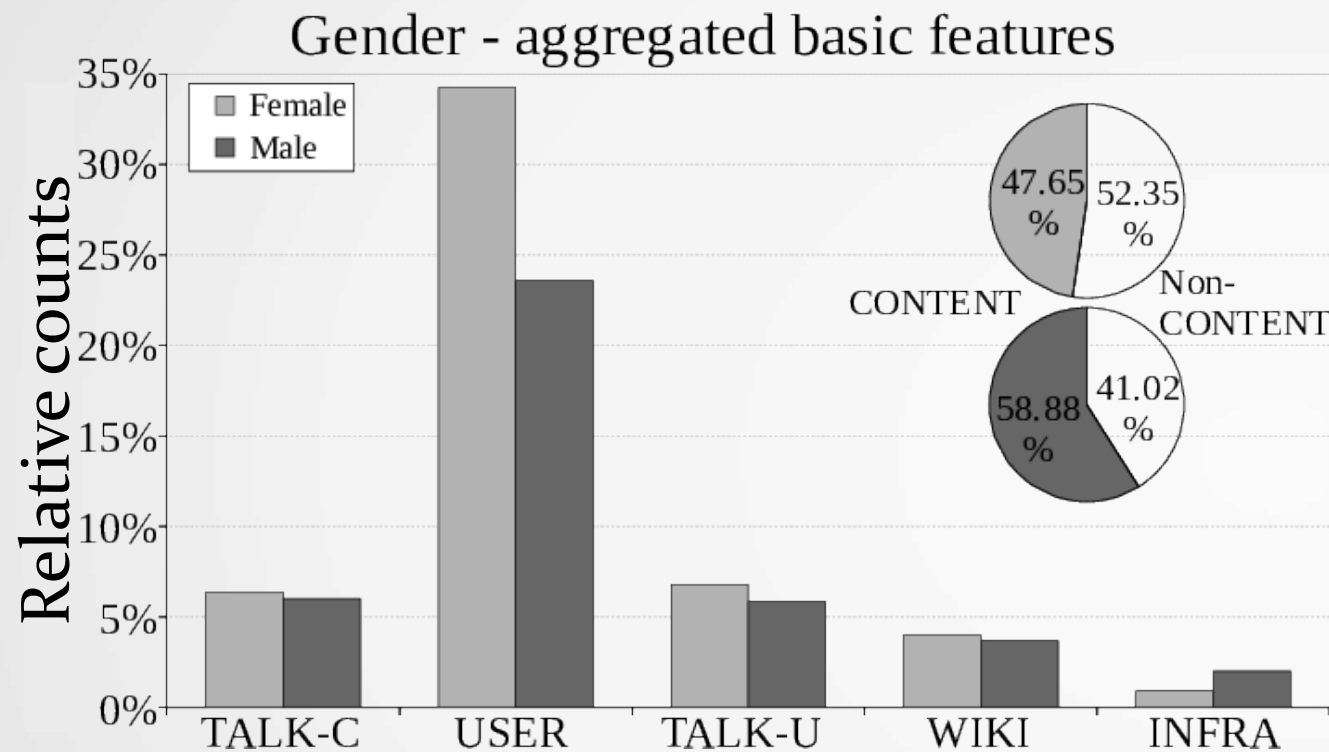
Profiling of editing behavior (2)

Different growth trends across editor demographics



Evolution trends across editor categories are unequal, providing plausible explanations for the slowdown [Gibbons '12], as well as *personal identification clues*.

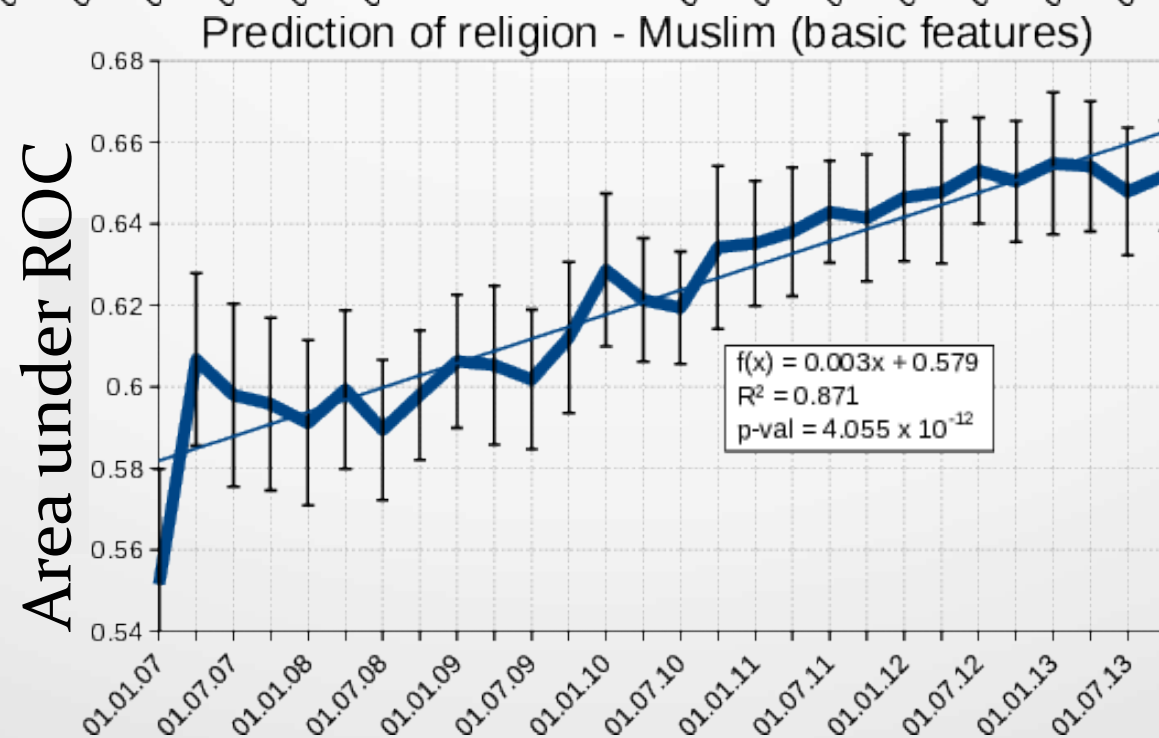
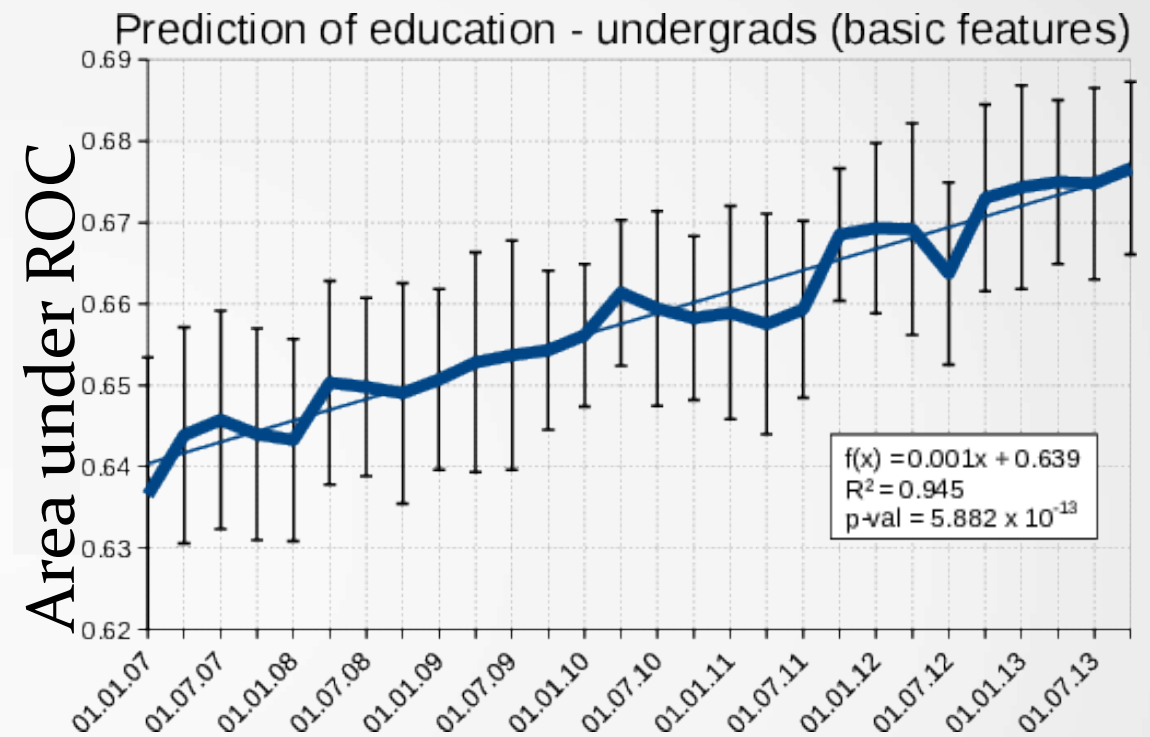
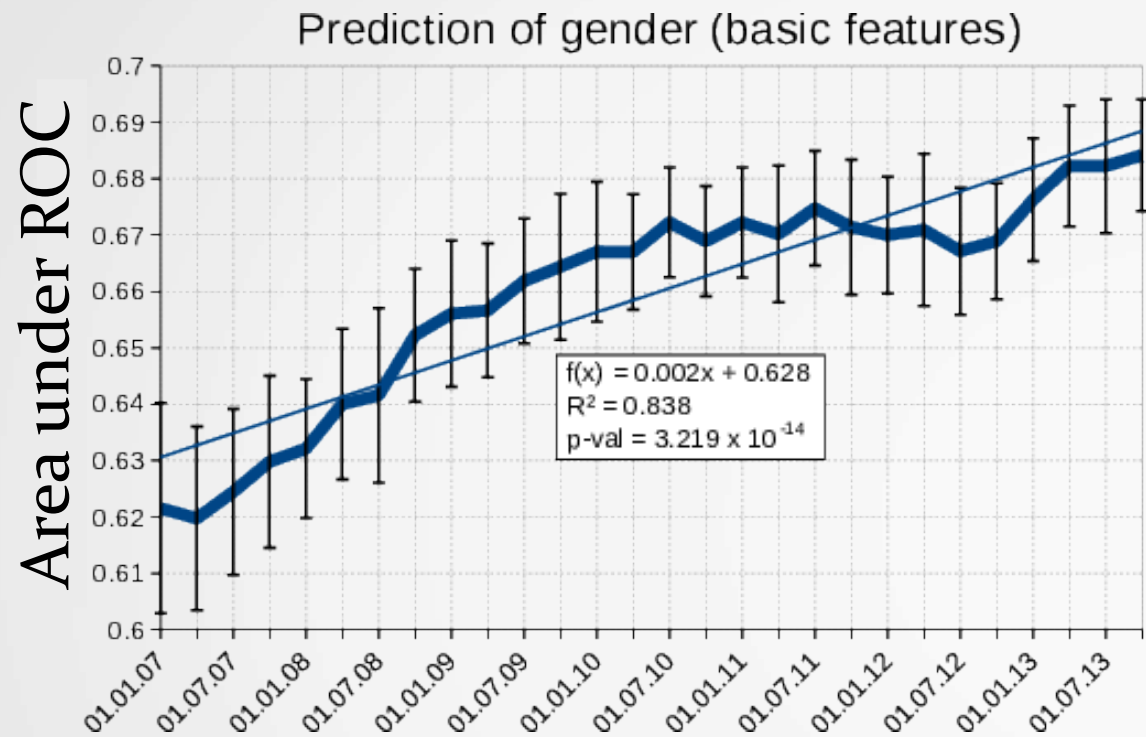
Edit behavior correlates with private traits



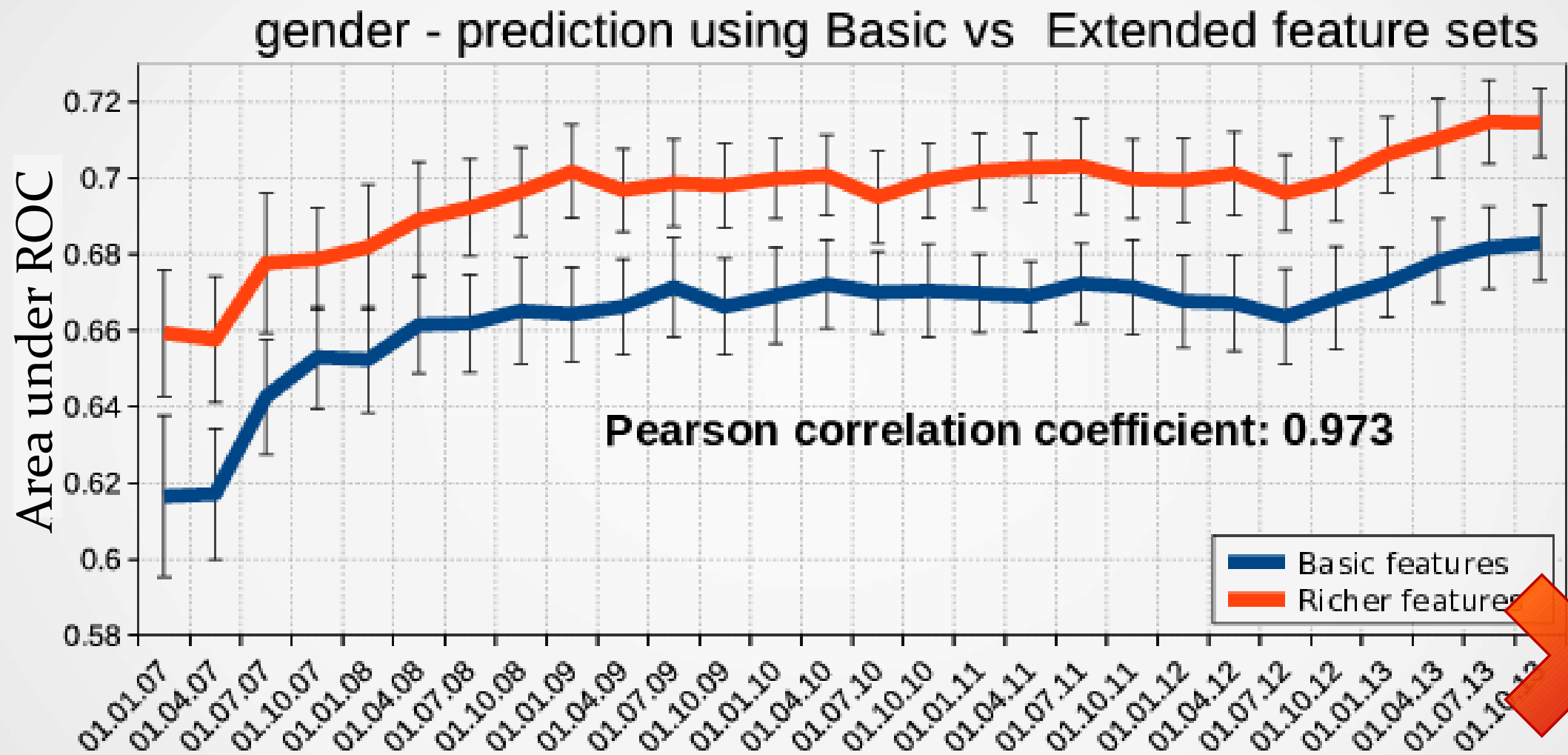
Mean editing behavior analysis shows regularities in the editing patterns for each sub-population.

Predictability improves over time

Privacy Loss as a *prediction problem*



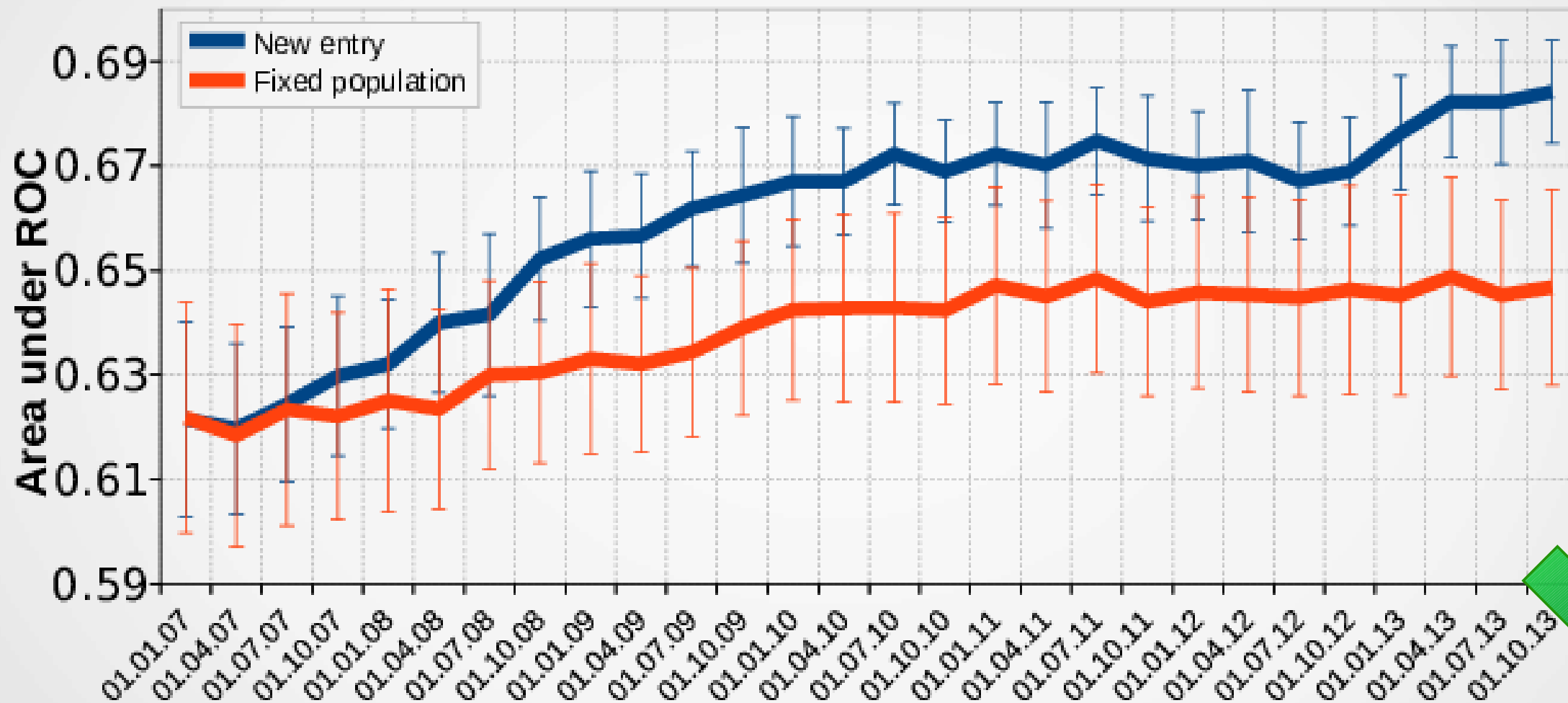
Sources of Privacy Loss (1)



**Richer features: improve prediction,
but not *Privacy Loss***

Sources of Privacy Loss (2)

gender - New Entry vs. Fixed Population (basic features)



Newcomers: information from newcomers hurts privacy

Marginal utility of features over time

Information theory measures:

- Uncertainty about private information → **entropy of target variable Y**

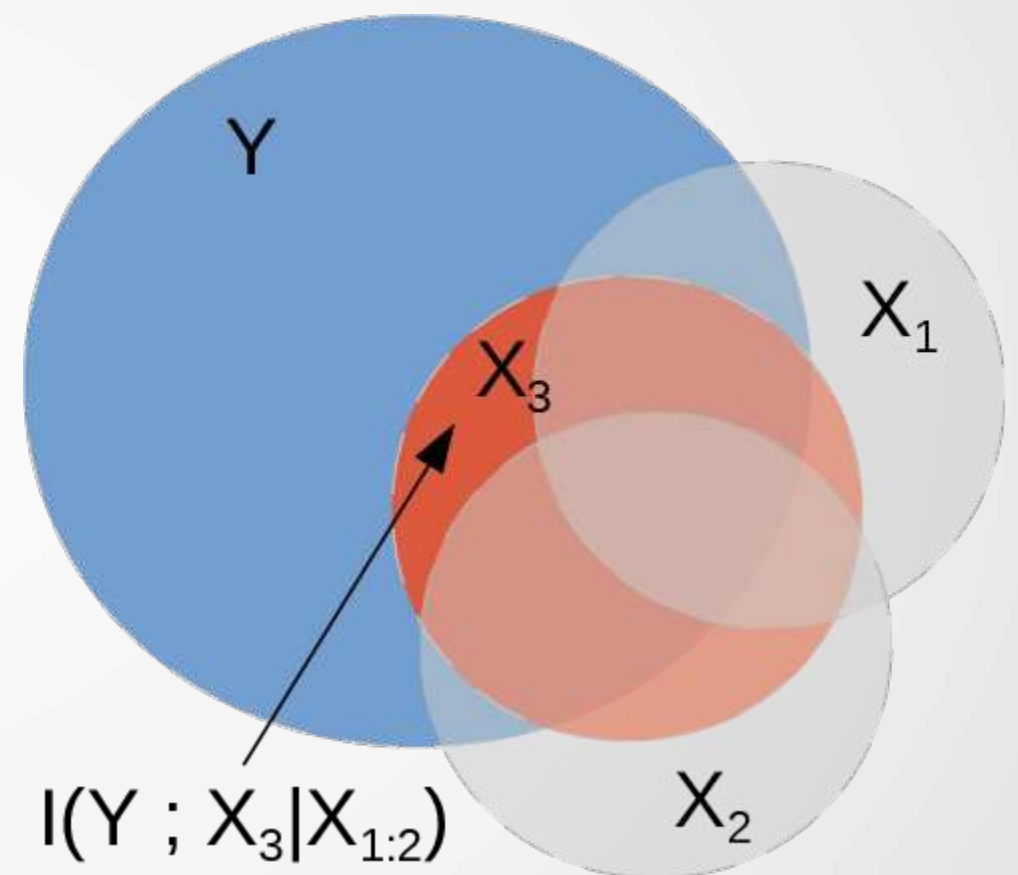
$$H(Y)$$

- Amount of information disclosed by a feature X about Y → **mutual information of X and Y**

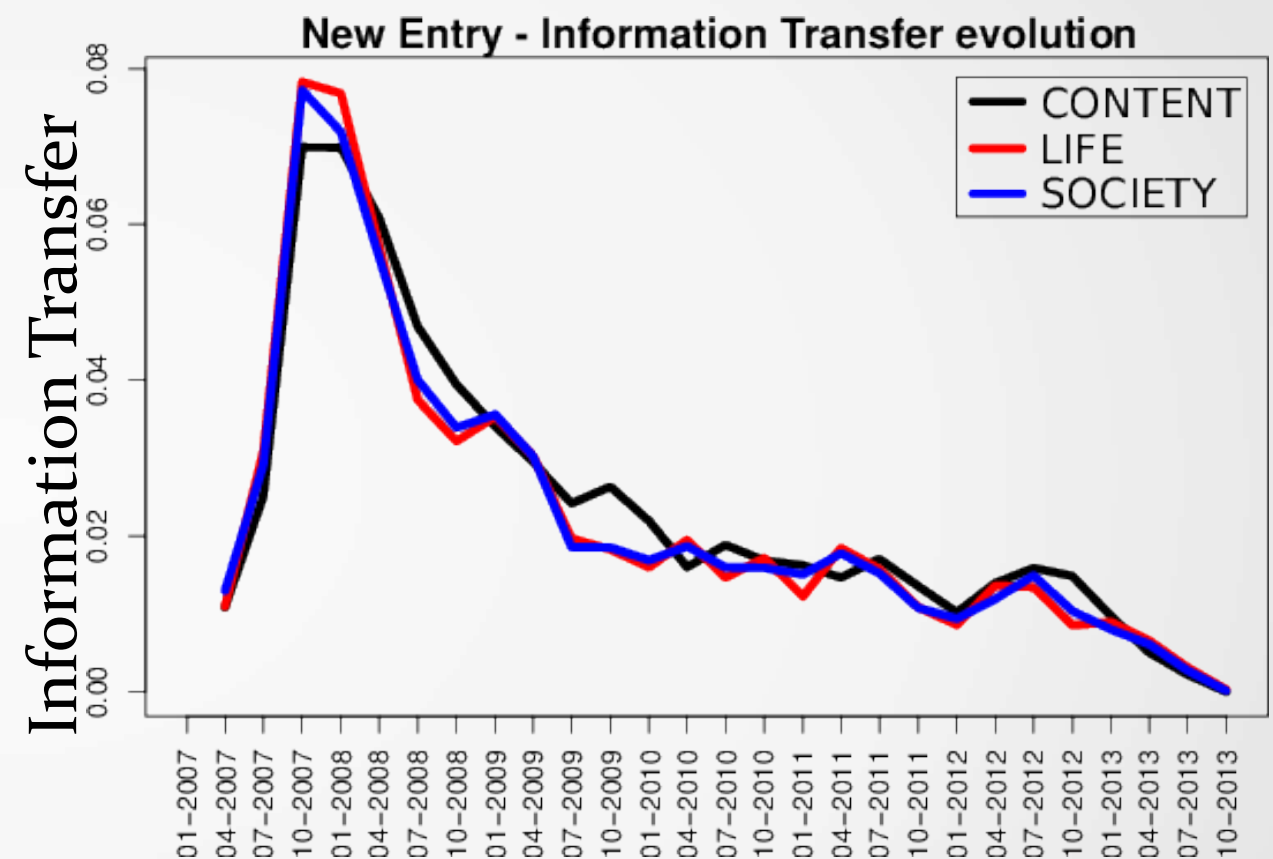
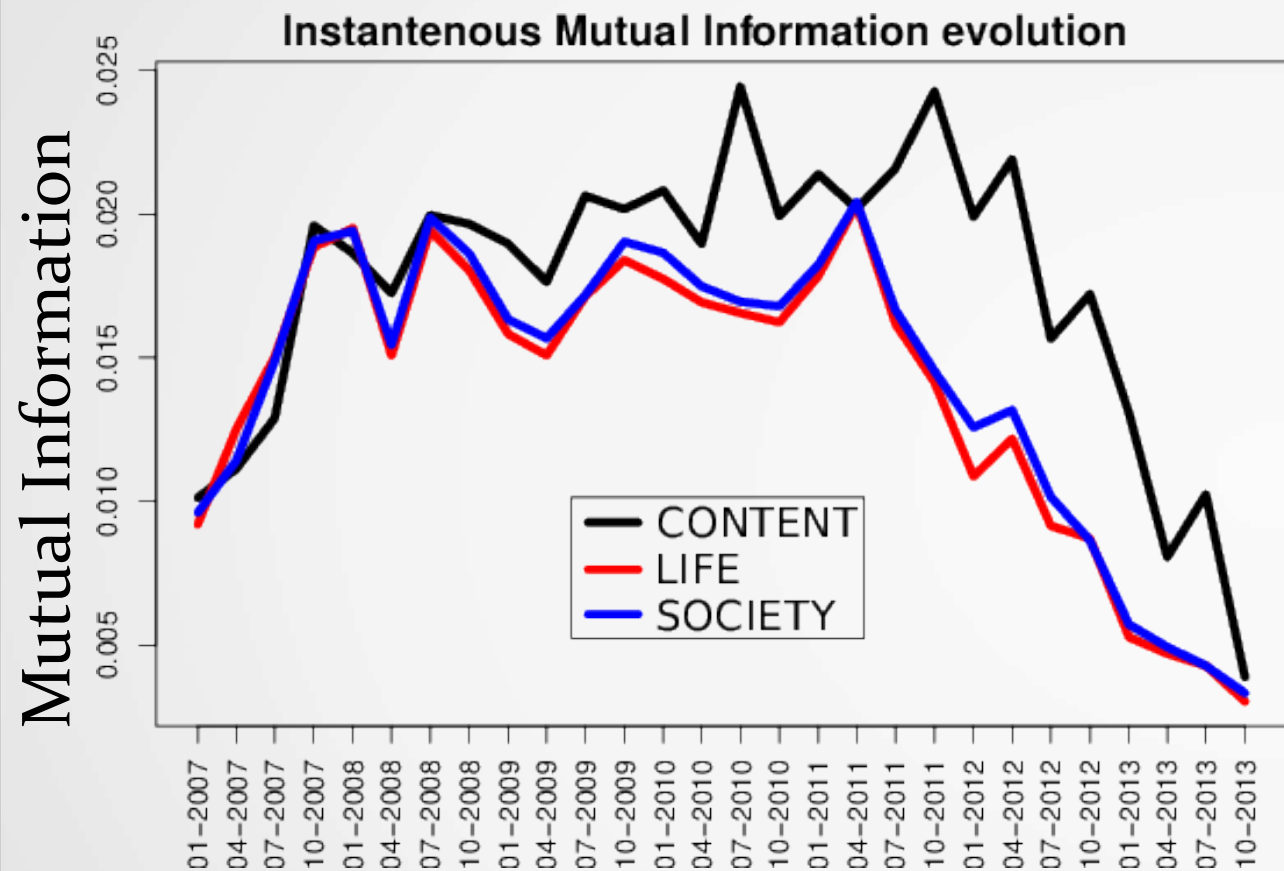
$$I(Y; X)$$

- Amount of *new information* disclosed by a feature at time t X_t → **Information Transfer**

$$I(Y; X_t | X_{1:t-1})$$

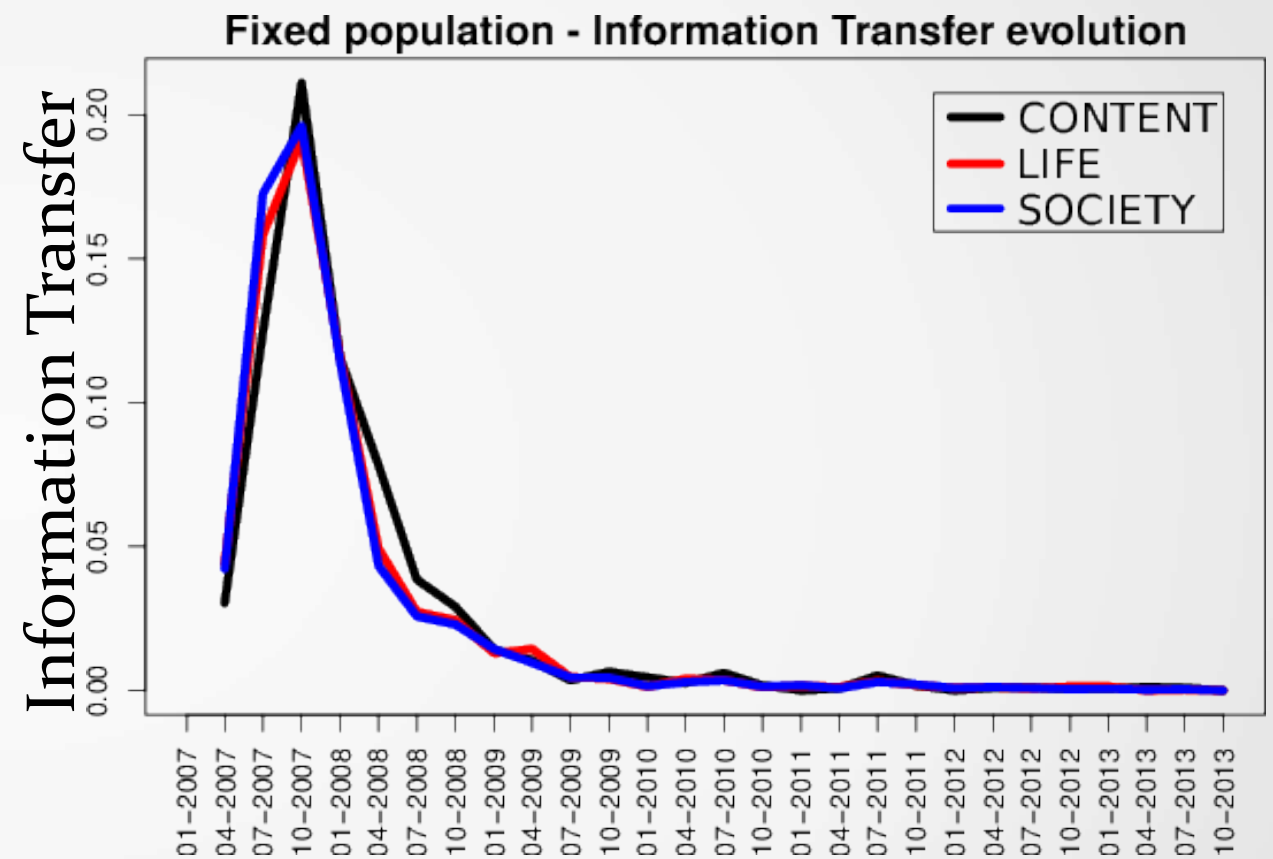
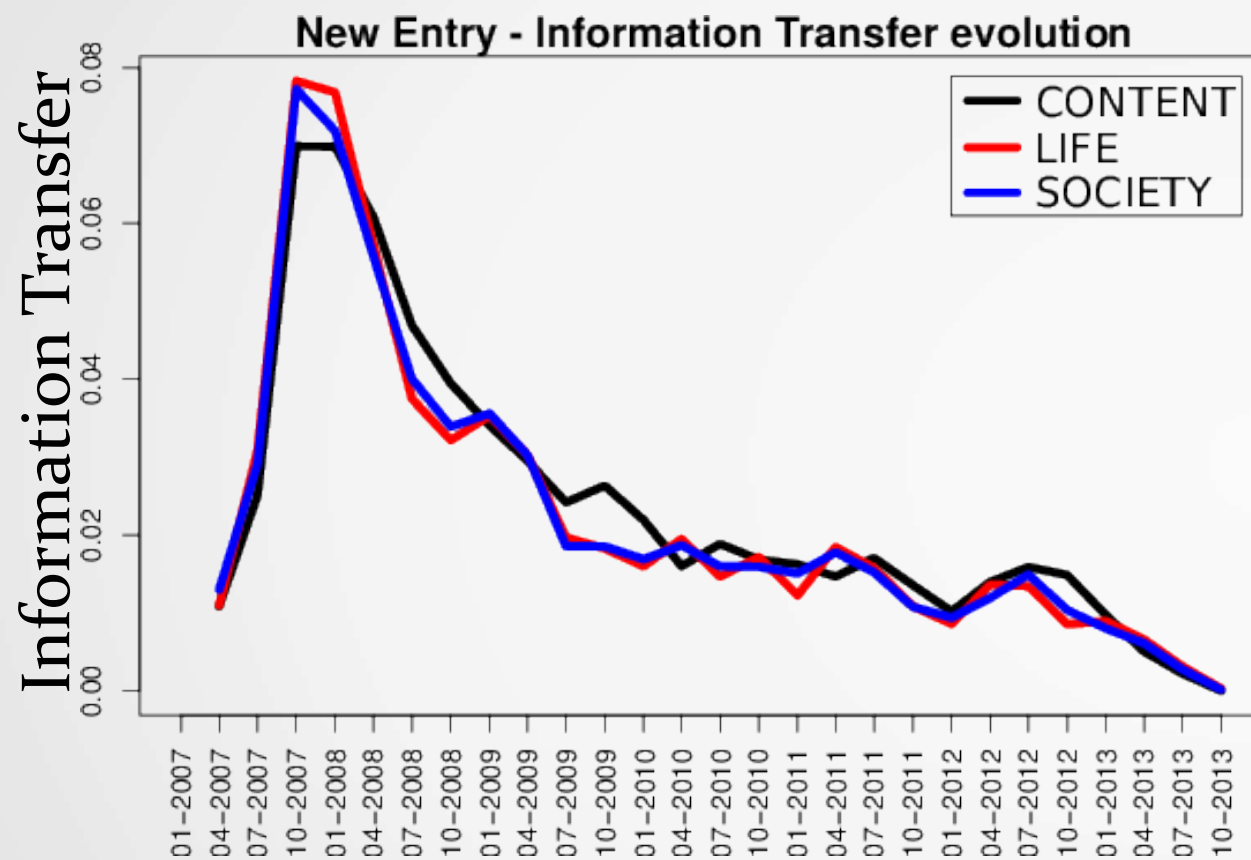


Effect of *online breadcrumbs*



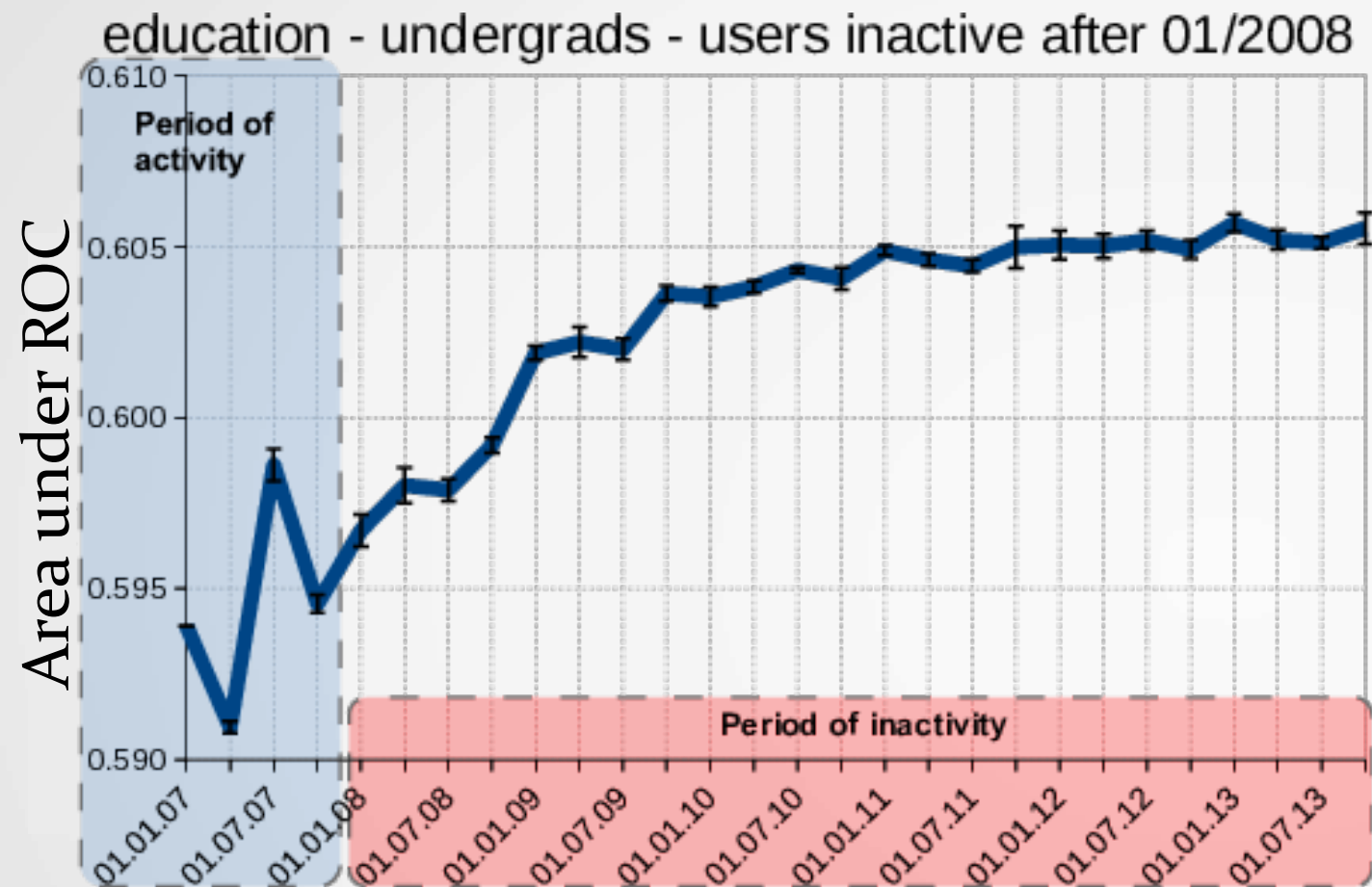
While later edits contain just as much information about a user's privacy as the earlier edits, they tend to be less harmful since most of the information they bring has already been learned.

Effect of *newcomers*

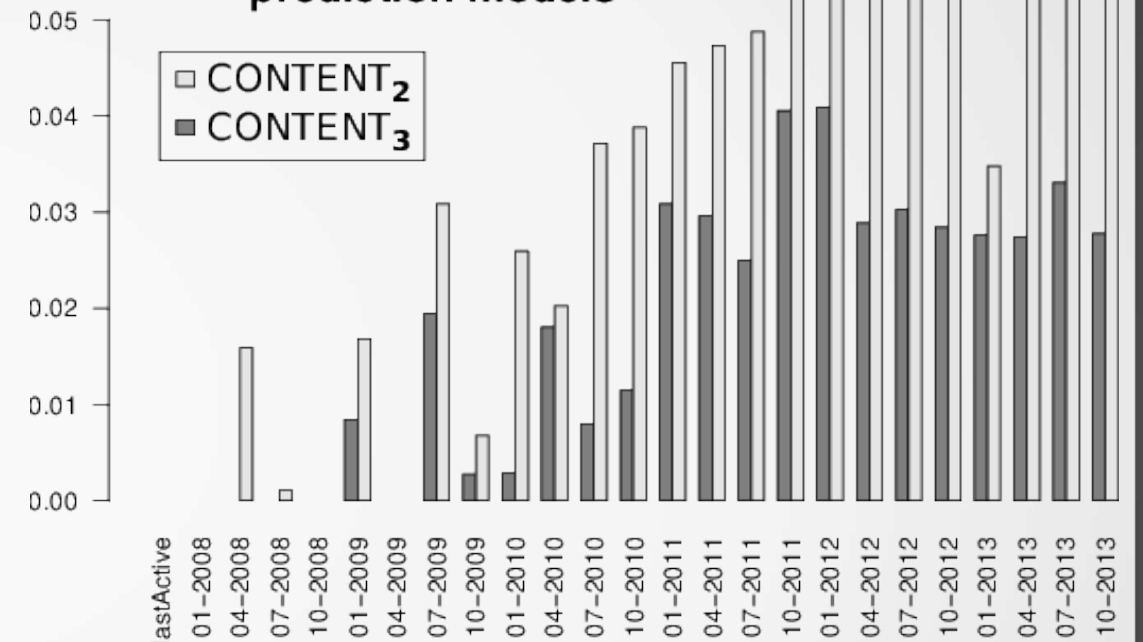


The information inferred from newcomers seems to be moderate, but consistent over time.

Privacy erodes even for *retired editors*



Learned coeff. of CONTENT feature in prediction models



Plausible explanation: observed prediction improvement originates with currently active editors, whose activity overlaps with exited editors

Conclusion

3 main conclusions:

- Time has an adverse effect on privacy
- Factors influencing Privacy Loss:
 - *online breadcrumbs* (i.e. editor's own activity)
 - activity of other editors and newcomers
- Privacy erodes even for *retired* editors

Users don't have complete control over the consequences of the information they release

The way ahead

Issues to address / Perspectives:

- **not specific to Wikipedia:** additional online platforms, richer in social data;
- editor disclosure bias;
- effective conditions for preserving privacy.

Thank you!

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