

Title	Big Data in Media - results from a Finnish study
Author(s)	Bäck,Asta; Ollikainen,Ville; Södergård, Caj; Vainikainen, Sari
Citation	Nxt Media Conference, 17.11.2015, Trondheim, Norway
Date	2015
Rights	This presentation may be downloaded for personal use only.

VTT
<http://www.vtt.fi>
P.O. box 1000
FI-02044 VTT
Finland

By using VTT Digital Open Access Repository you are bound by the following Terms & Conditions.

I have read and I understand the following statement:

This document is protected by copyright and other intellectual property rights, and duplication or sale of all or part of any of this document is not permitted, except duplication for research use or educational purposes in electronic or print form. You must obtain permission for any other use. Electronic or print copies may not be offered for sale.



Big Data in Media – results from a Finnish study

Nxt Media seminar 17.11.2015

Trondheim

**Asta Bäck, Ville Ollikainen, Caj Södergård,
Sari Vainikainen**

What is needed (according to media houses) ?

- Current drawbacks
 - Log data is not collected systematically
 - Legal barriers for linking usage and users
 - Separate, unconnected data from each channel (Mobile, e-paper, PC)
- To be developed
 - Better understanding of subscribers
 - Better article recommendations
 - Adding metadata
 - Ad targeting
 - Predicting ad clicks

Aim

- Help media houses to use their **own** data for
 - Content recommendation
 - Customer segmentation
 - Predicting user behavior

- Define and set up a test environment

- Run trials on case data sets

- Funded by Finnish Media Fund (Viestintäalan tutkimussäätiö)

Two user data sets

- **Data set 1:** Cookies (one month, several services)
- **Data set 2:** Click data (one month, two services)
 - 800.000 data rows
- Both sets contained:
 - Cookie / Anonym(ized) user ID
 - URL of the clicked page
 - Time of the click (1 hour/1 second accuracy)

We used several analyzing programs

- R – Data exploration and visualization
- Weka – Association rules and k-clustering
- Microsoft Azure Machine Learning – usage prediction
- UPCV (developed by VTT)

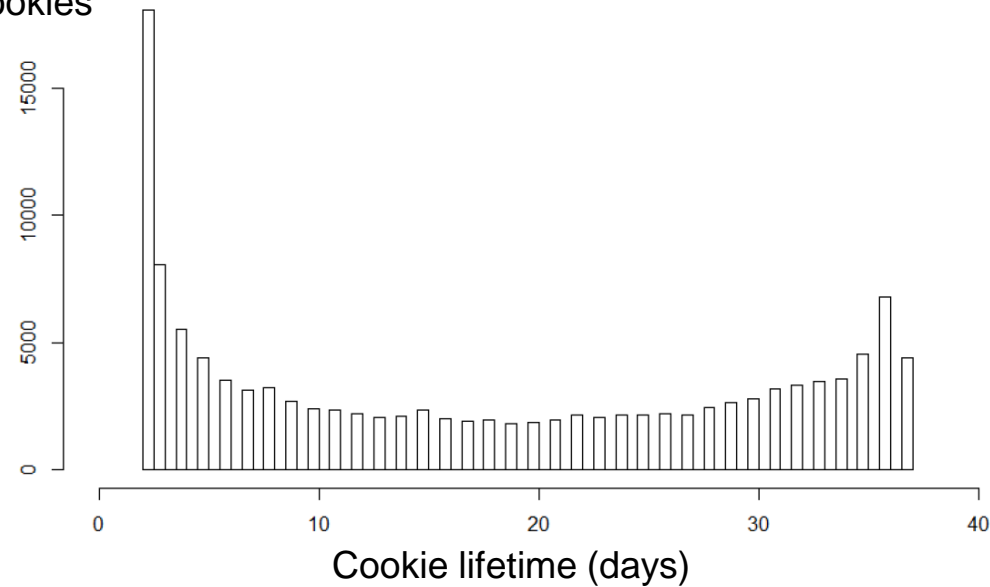
Observations

- Large variation in life span of cookies

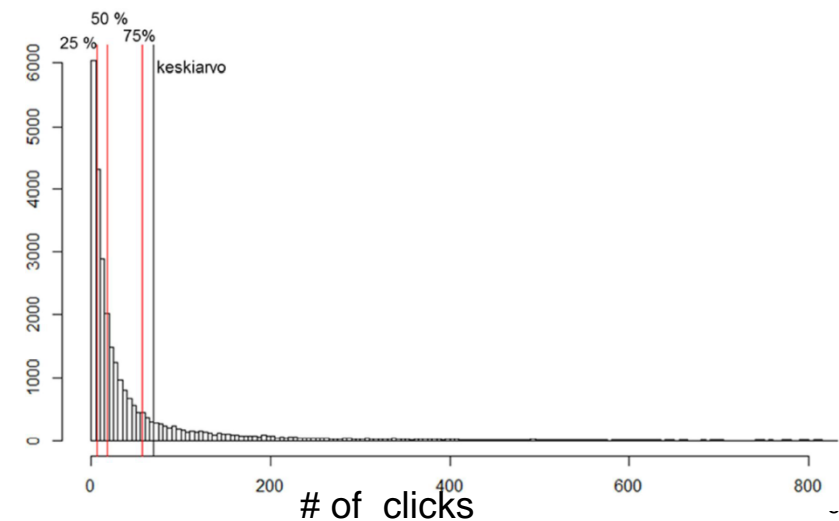
- Most users had only a small number of clicks
 - Used many content services
 - More clicks per service

- Hardcore users:

of cookies



of users



Surfing paths – evening news is the entry point

iltalehti.fi	+	telkku.com
iltalehti.fi	+	kauppalehti.fi
iltalehti.fi	+	kotikokki.net
etuovi.com	+	iltalehti.fi
aamulehti.fi	+	iltalehti.fi
kotikokki.net	+	telkku.com

Iltalehti = Evening news

Telkku = TV guide

Etuovi = Apartment advertisements

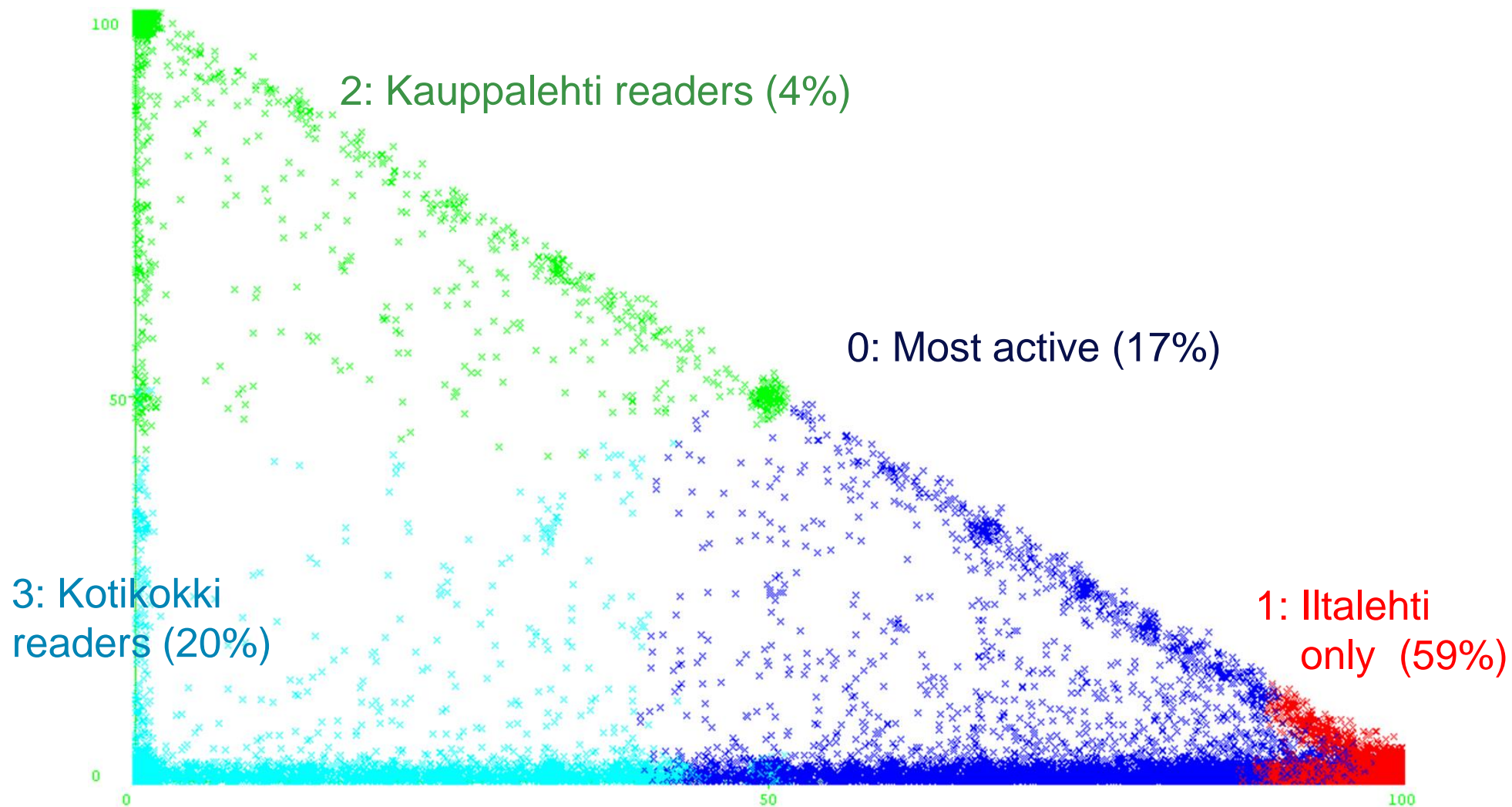
Aamulehti = Morning news

Kauppalehti = Financial news

Kotikokki = Food and recipes

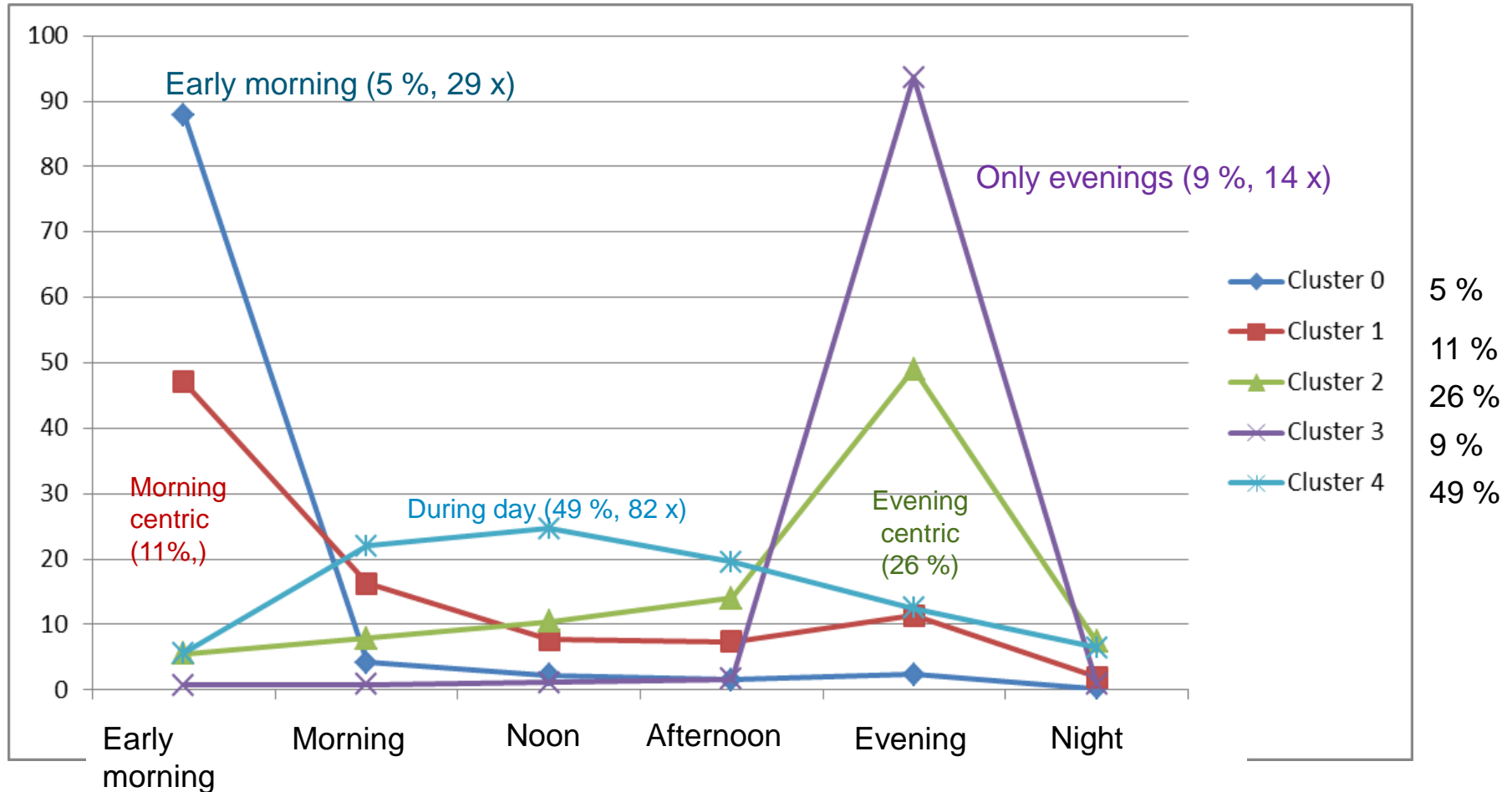
Analysis with Weka's Apriori algorithm

Clustering users based on their reading habits

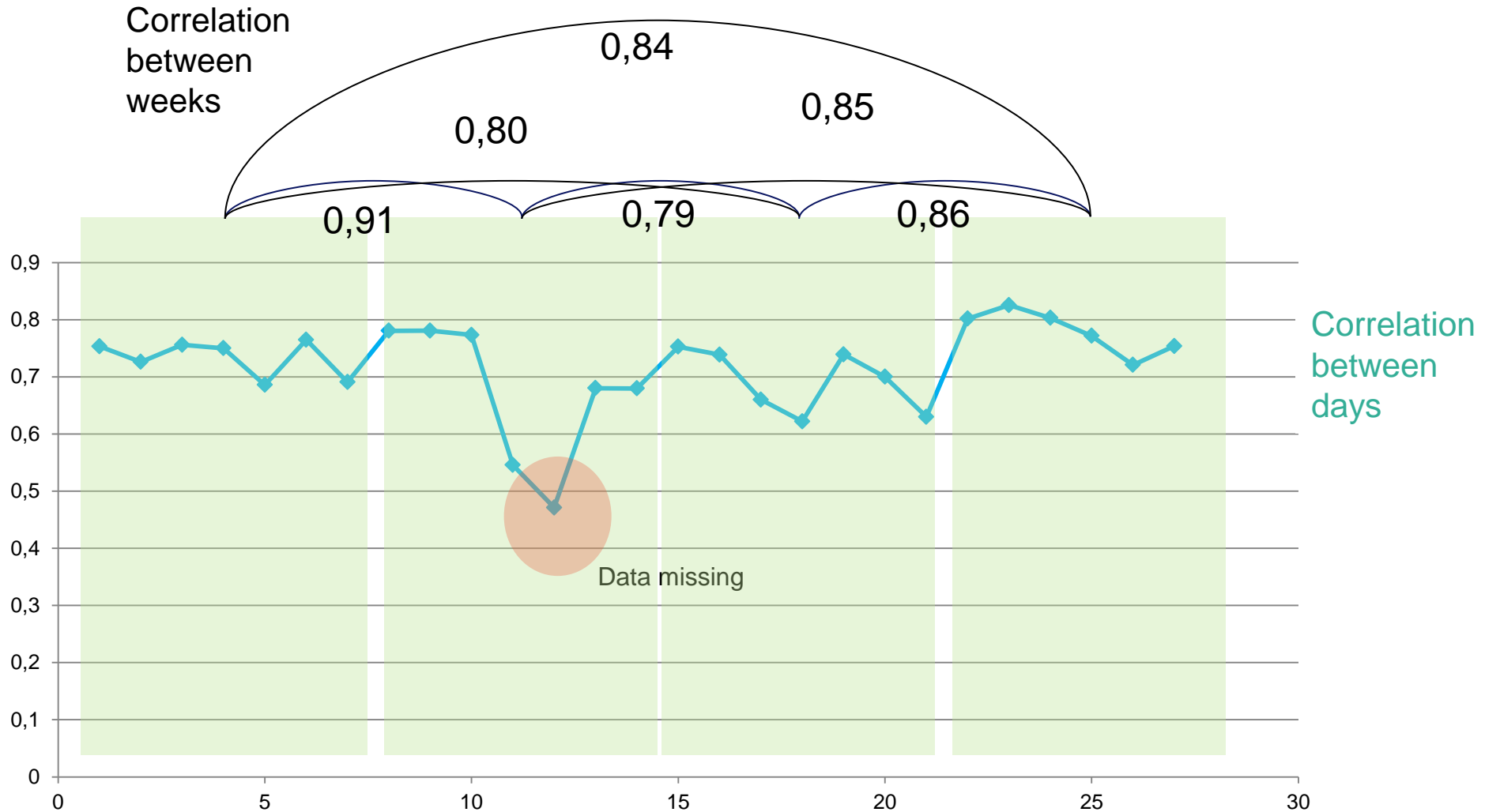


Most users read daytime

Percentage of use



Clicks between sequential days and weeks correlate



Predicting the clicks for the week 4

		Predicted Class			
		1	2	3	4
		0	1-10	11-50	51-
Actual Class	1 0	80.4%	11.8%	7.5%	0.3%
	2 1-10	18.3%	52.1%	28.4%	1.2%
	3 11-50	4.0%	20.2%	67.9%	7.9%
	4 51-	0.8%	2.0%	35.2%	61.9%

- Clicks during the week 4 were predicted based on the clicks and usage times in weeks 1...3
- 4 classes
 - 0: 1-10 clicks (2181 users)
 - 1-10 clicks (1968 users)
 - 11-50 clicks (2183 users)
 - >50 clicks (847 users)
- Machine learning method
- Fairly high overall accuracy
- The group 1-10 times per week was the most difficult one

Recommending content

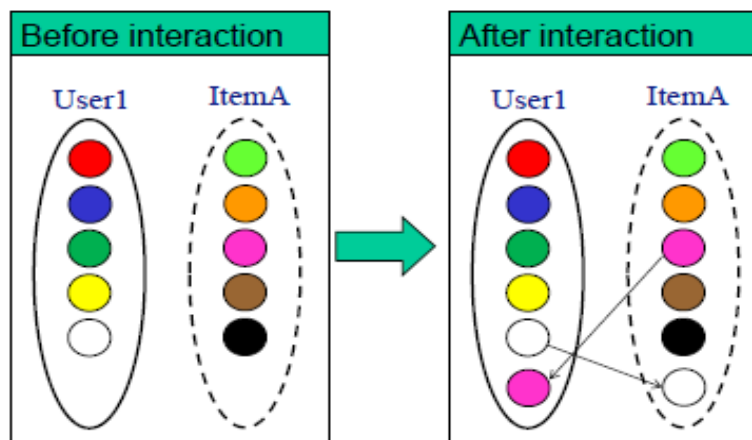
- Collaborative method, no data analysis

UPCV is a novel collaborative recommendation method preserving users' privacy and supporting distributed computing and distributed data repositories.



Figure 1. Item data is separated from User data. This allows synchronising data from several services safely.

Figure 2. In UPCV similarity will spread from users to items and from items to users making it possible to inherently provide user-item, item-item, item-user and user-user recommendations..



Results from Collaborative Recommendations

- Large data set:
 - 5,6 million users (i.e. cookies) & 308 million rows,
 - 118.000 articles
- Recommendation quality depends heavily on the user
 - Having accessed only popular articles, or only a few articles => few and rather noisy recommendations
 - Having any pattern of reading articles of a **rare topic => relevant recommendations**
 - e.g. among other readings two articles on cyber security and one article on security in general => one Data Security Officer job ad and several IT job ads, not recommended to anyone else in the evaluated subset.
 - (N.B. Pure collaborative approach: [no data analysis](#))

Summary

- Media houses want to understand user behavior (subscription, reading patterns), ad targeting and content recommendations
- User information is mostly limited in scope
 - Cookies have very varying lifetime & users have many cookies
 - Users use various accounts or do not log in
- However, also partial information of user behavior gives insights
- Clusters of users emerged -> develop services for these segments
- Usage pattern is stable from day to day and week to week
- Subscriber and demographic user data would improve the services
- Quality of collaborative recommendations depend on user behavior



TECHNOLOGY «FOR BUSINESS»

