

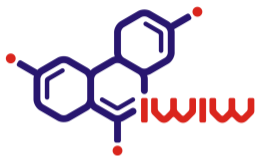
# Online social network sites and how to study them

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October 22, 2022



## Data: iWiW



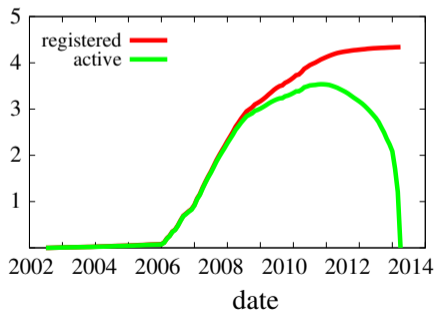
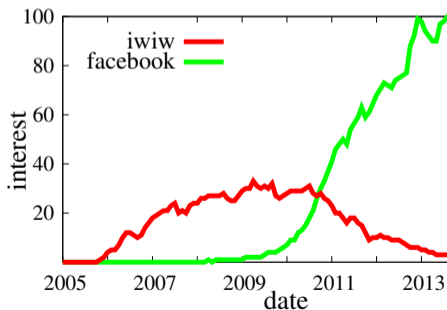
- ▶ Established in 2005 (similar to facebook, but more general in the beginning)
- ▶ Commercialized in 2006
- ▶ Growth by invitation system (linear)
- ▶ Top 2-3 site in Hungary for years till 2011
- ▶ 35% of the Hungarian population online
- ▶ 60% of those with Internet access

# Questions

- ▶ Why did it fail and how?
- ▶ How does ICT data differ from reality?
- ▶ What does the data tell us about the society?

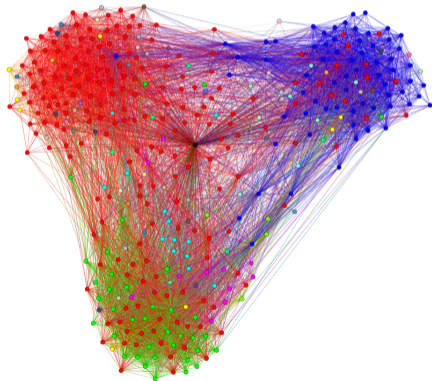
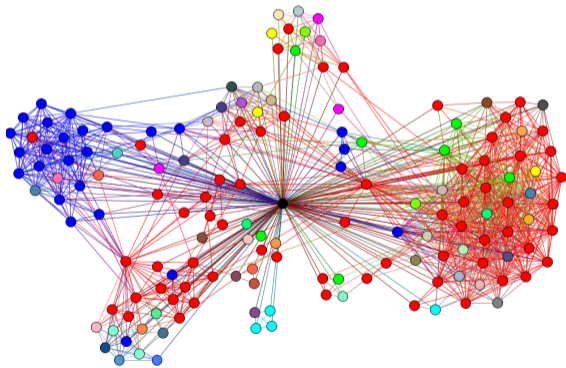
## iWiW: Life and death

- ▶ Linear increase due to limited invitations (one per person in a month or two)
- ▶ Growth till end of 2010 (3.5 million active users)
- ▶ Stagnation 2011
- ▶ Problems 2012 (more than 3 million active users till May 2012)
- ▶ Collapse 2013



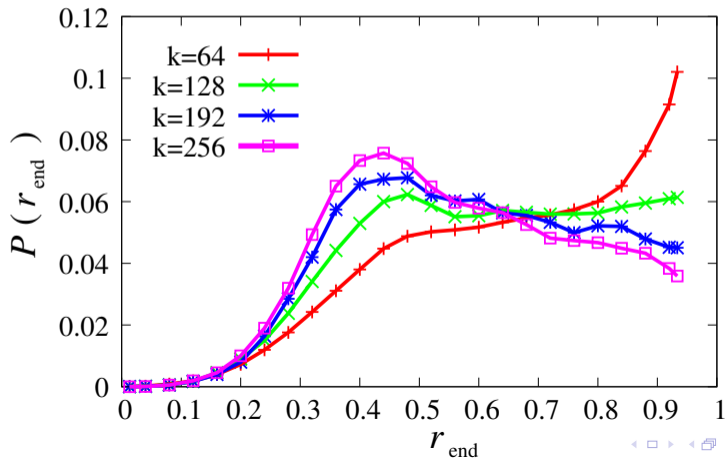
# Egocentric networks

- ▶ My egocentric network
- ▶ Egocentric network from Becsehely
- ▶ Color according to cities



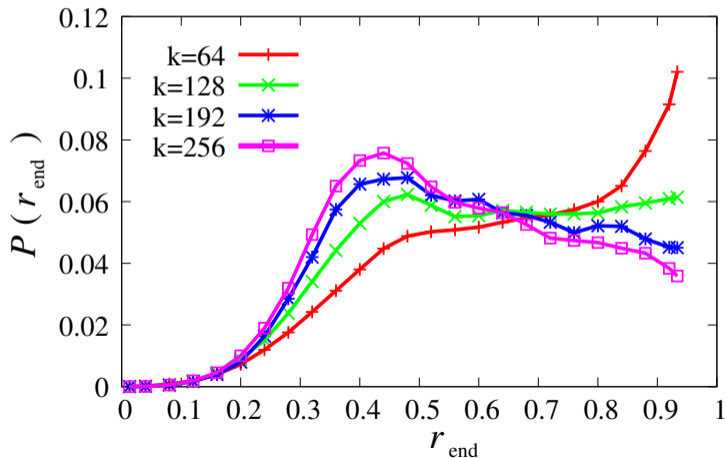
## iWiW collapse

- ▶ Fraction of active friends at the time of the ego's last login
- ▶  $k$ : degree, number of friends on iWiW
- ▶ For large  $k$  peak at  $\sim 0.4$ – $0.45$
- ▶ Two week overlap



## iWiW collapse

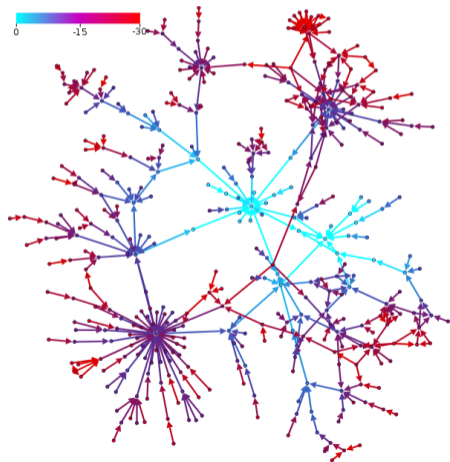
- ▶ For large  $k$  peak at  $\sim 0.4$ – $0.45$
- ▶ People with limited friends very early
- ▶ People with many friends when  $\sim 50$ – $60\%$  of their acquaintances left





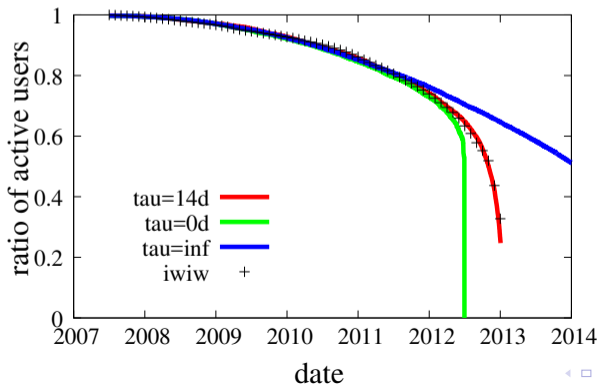
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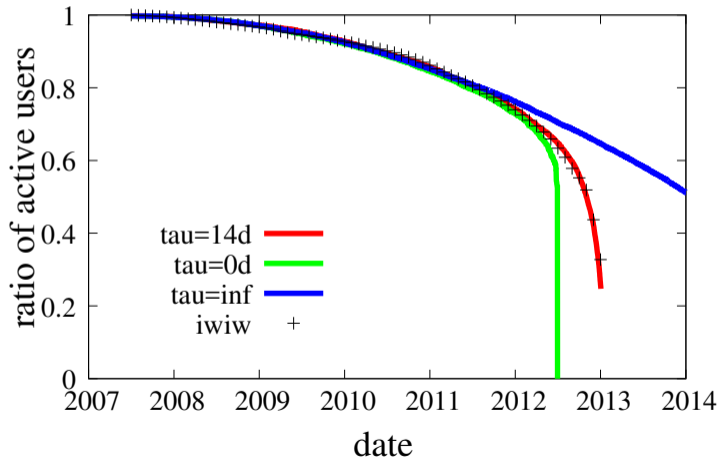
## iWiW collapse: cascade model

- ▶ Network with average degree  $\langle k \rangle$
- ▶ There are users that leave with a rate  $\gamma = \mu t / \tau$ , nodes are chosen with probability proportional to  $\sim (k + 25)^{-2}$  (exogenous effects)
- ▶ Users for which the ratio of active friends dropped below  $\lambda$  get inactive, but their friends do not realize it immediately
- ▶ Users realize the departure of their friend with rate  $\tau$



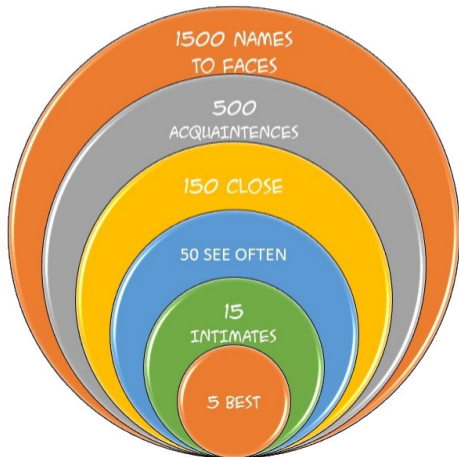
## iWiW collapse: model

- ▶ Best fit:  $\langle k \rangle = 12$ ,  $\tau = 14$  days or  $\langle k \rangle = 200$ ,  $\tau = 130$  days
- ▶ It takes two weeks to recognize inactive friends



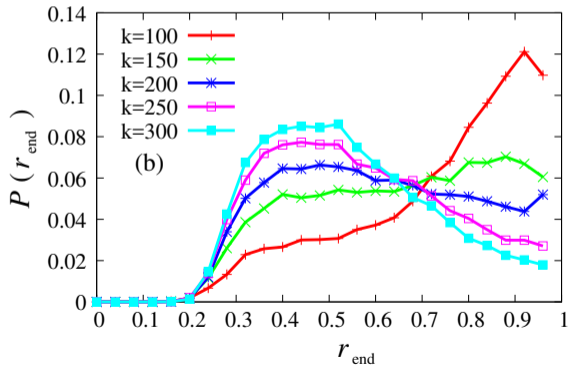
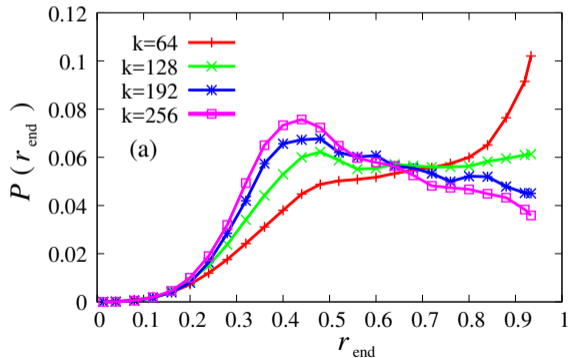
## iWiW collapse: model

- ▶ Best fit:  $\langle k \rangle = 12$ ,  $\tau = 14$  days
- ▶ It takes two weeks to recognize inactive friends
- ▶ It seems, that only the intimate circles matters (Dunbar's circles)

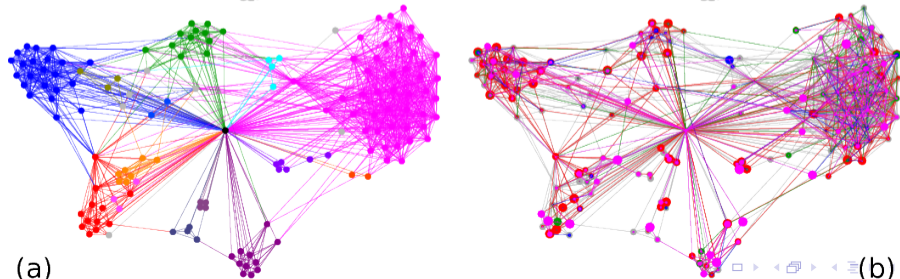
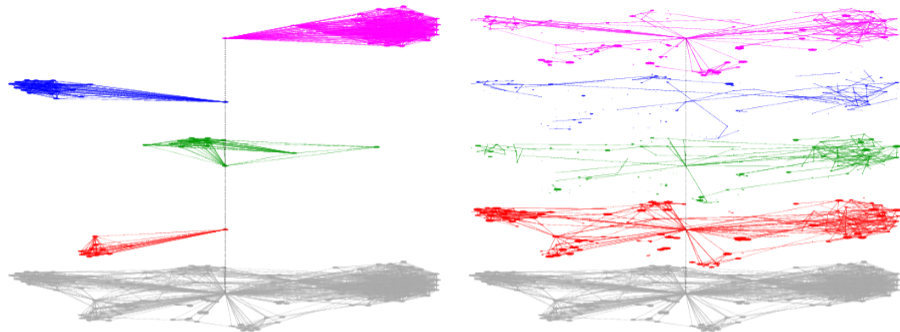


## iWiW collapse: model

- ▶ Artificial social network
- ▶ Cascade model with waiting time
- ▶ Degree dependent exogeneous leave



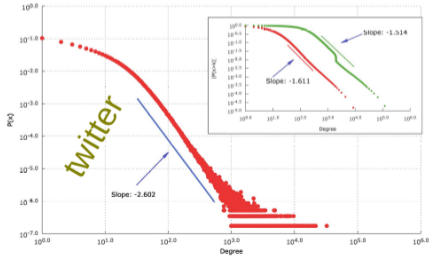
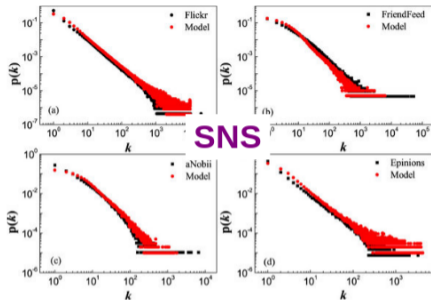
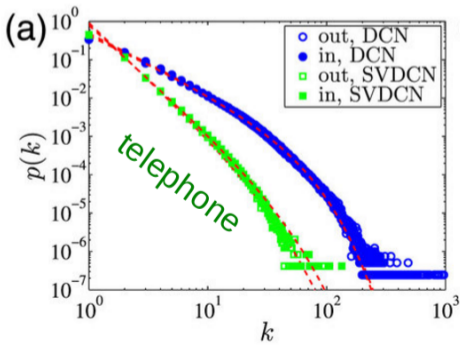
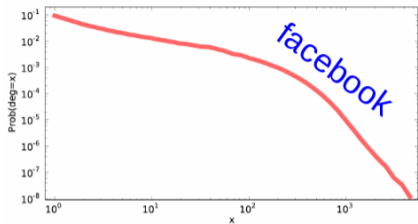
What does ICT data see? Egocentric network:



(a)

(b)

# Degree distributions



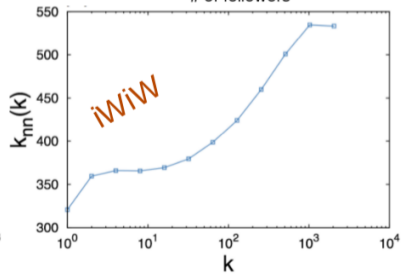
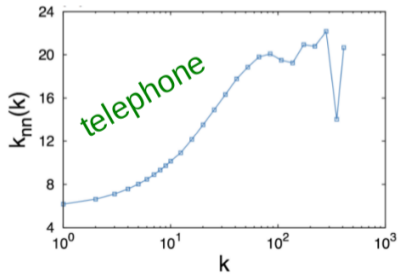
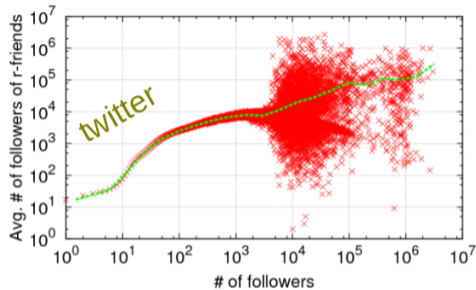
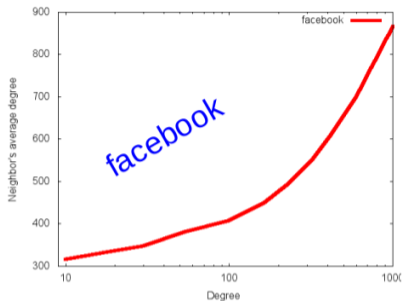
## Assortativity

- ▶ Average degree of the friends with degree  $k$
- ▶ For random networks it is a constant
- ▶ For scale free networks it is usually decreasing (disassortative), meaning, that small degree nodes connect to hubs.
- ▶ For humans it is believed to be increasing (assortative) extrovert people have extrovert friends.



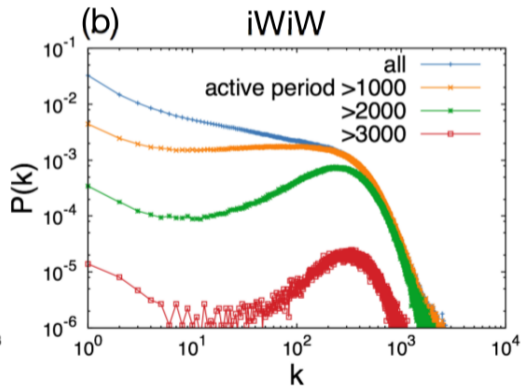
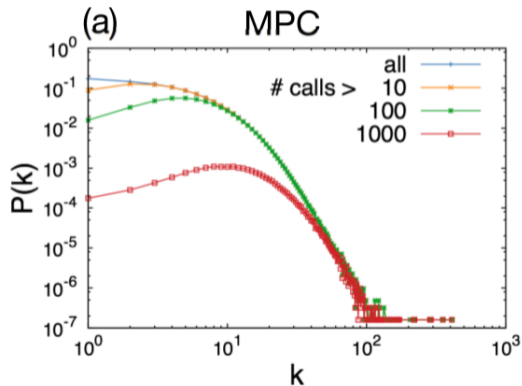


# Assortativity



## Reality?

- ▶ Service usage varies a lot
- ▶ Those who use it a lot have more friends on the service
- ▶ Maximum at iWiW for experienced users is close to the Dunbar number of acquaintances



# Model

- ▶ Model: How to choose a communication channel?
  - ▶ Which channel do we use to reach a friend?
  - ▶ Who's favourite?
  - ▶ Least uncomfortable for both!
- ▶ A kind of sampling of the underlying social network
- ▶ Start from an arbitrary network

# Model

- ▶ We start from a network (underlying, ground truth)
- ▶ Affinity: How much a user likes the channel
- ▶ Affinity from exponential distribution

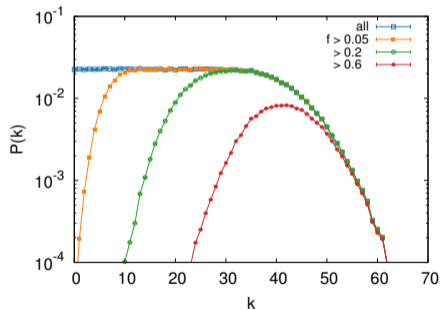
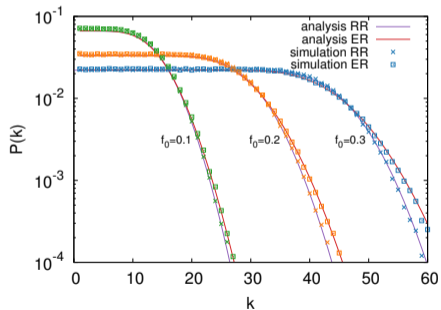
$$P(f) = \frac{1}{f_0} e^{-y/y_0}$$

- ▶ Assign affinities randomly
- ▶ Keep links from the underlying network with probability  $p_{ij}$
- ▶ Probability is the minimum of the two

$$p_{ij} = \min(f_i, f_j)$$

## Model: Degree distribution

- ▶ Model can be solved analytically for Erdős-Rényi and Random Regular network.
- ▶ Peak in the degree distribution disappears
- ▶ Node with higher affinity have nice peaked degree distribution close to the original



# Model: Assortativity

