

## **Analyzing Temporal Dynamics in Twitter Profiles for Personalized Recommendations on the Social Web** WebScience, Koblenz, June 17, 2011

Fabian Abel, Qi Gao, Geert-Jan Houben, Ke Tao Web Information Systems, TU Delft



# The Social Web

Who is this? What are his personal demands? How can we make him happy?

### Recommend me news articles that *now* interest me!

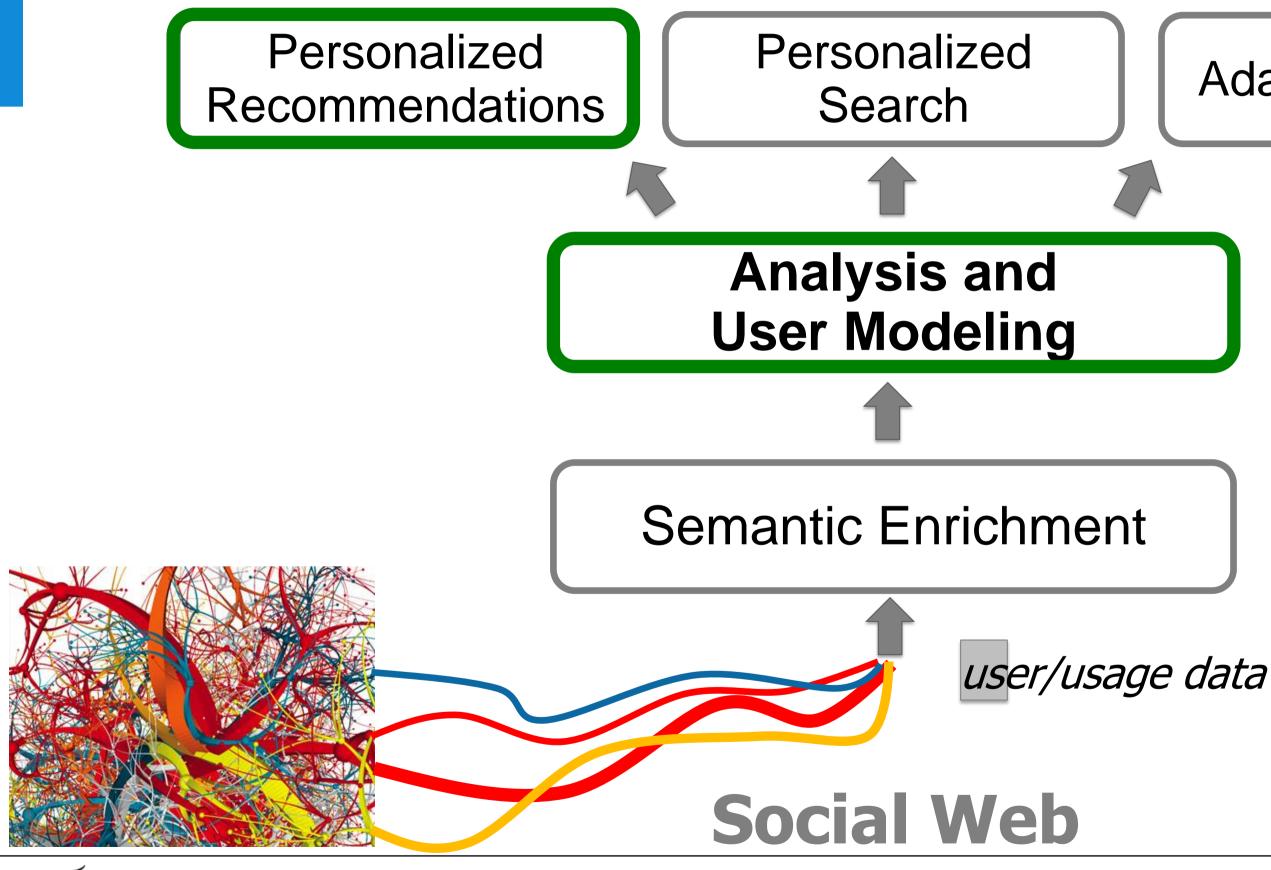
# Personalize my Web experience!

### Help me to tackle the *information* overload!

Help me to find interesting (social) media!

# What we do: Science and Engineering for the Personal Web

### domains: news social media cultural heritage public data e-learning

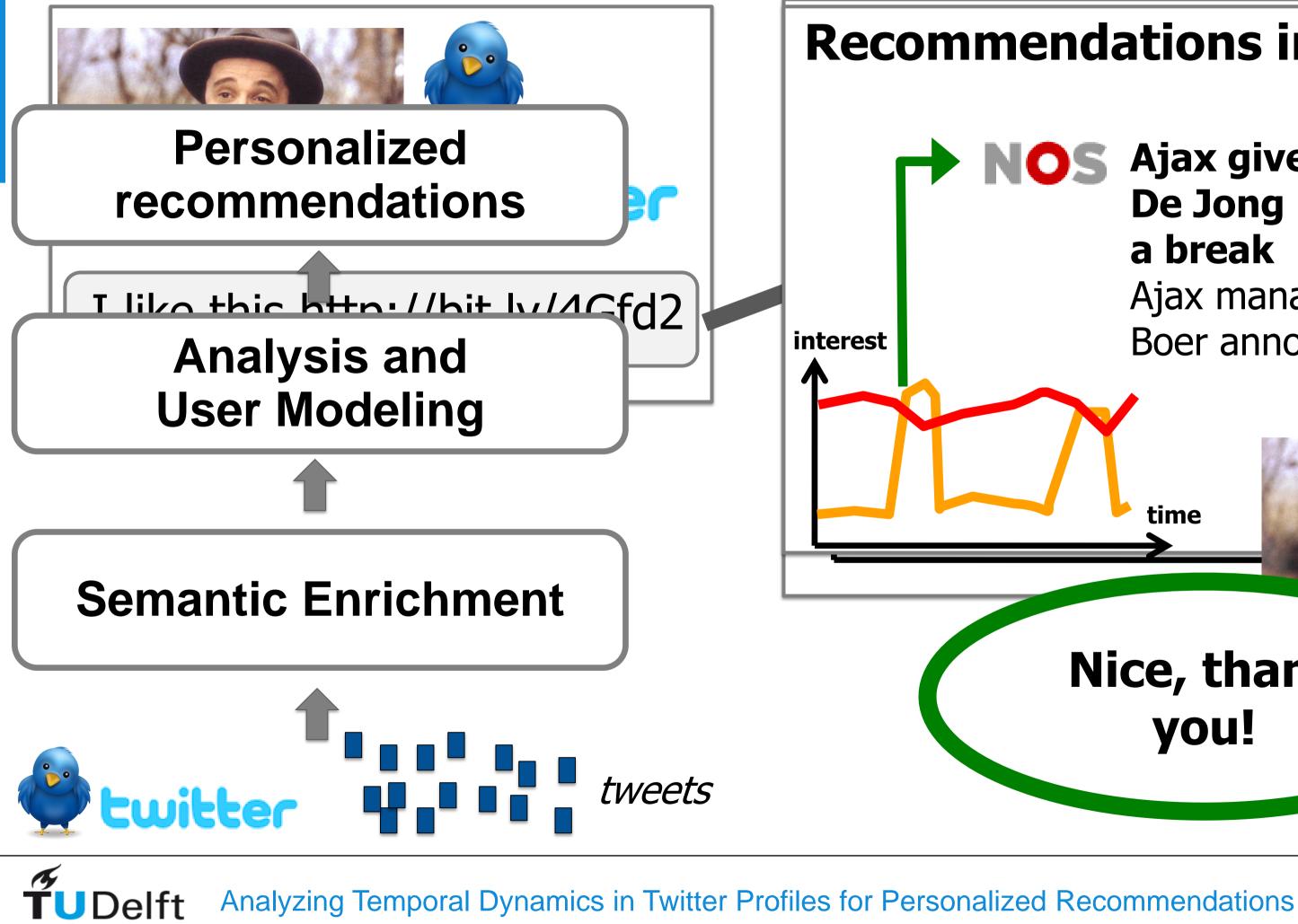


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### Adaptive Systems



# What we do in this paper



### **Recommendations in time:**

### **NOS** Ajax gives **De Jong** a break



oc:Sports

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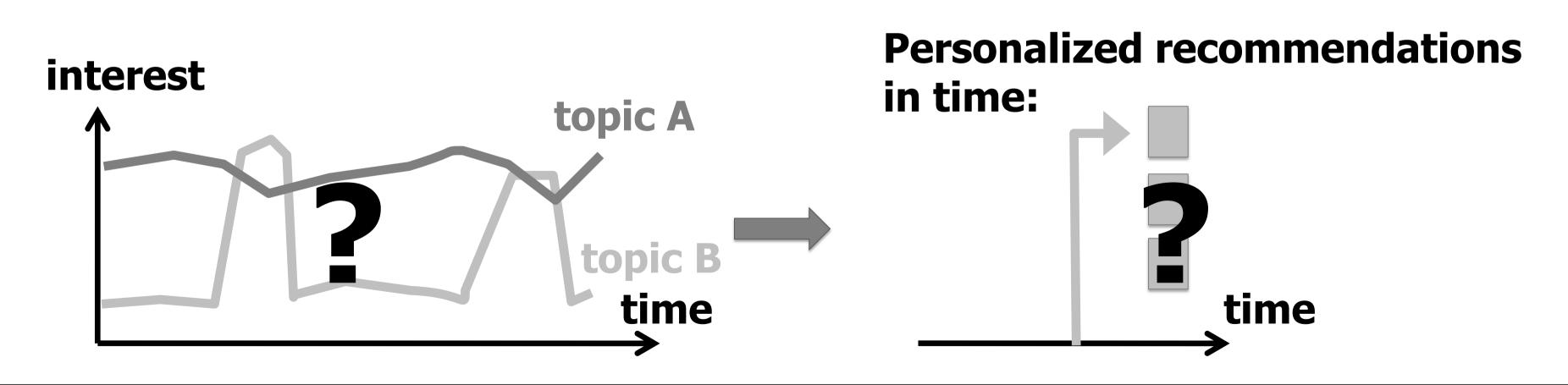
Ajax manager Frank de Boer announced that...

### Nice, thank you!

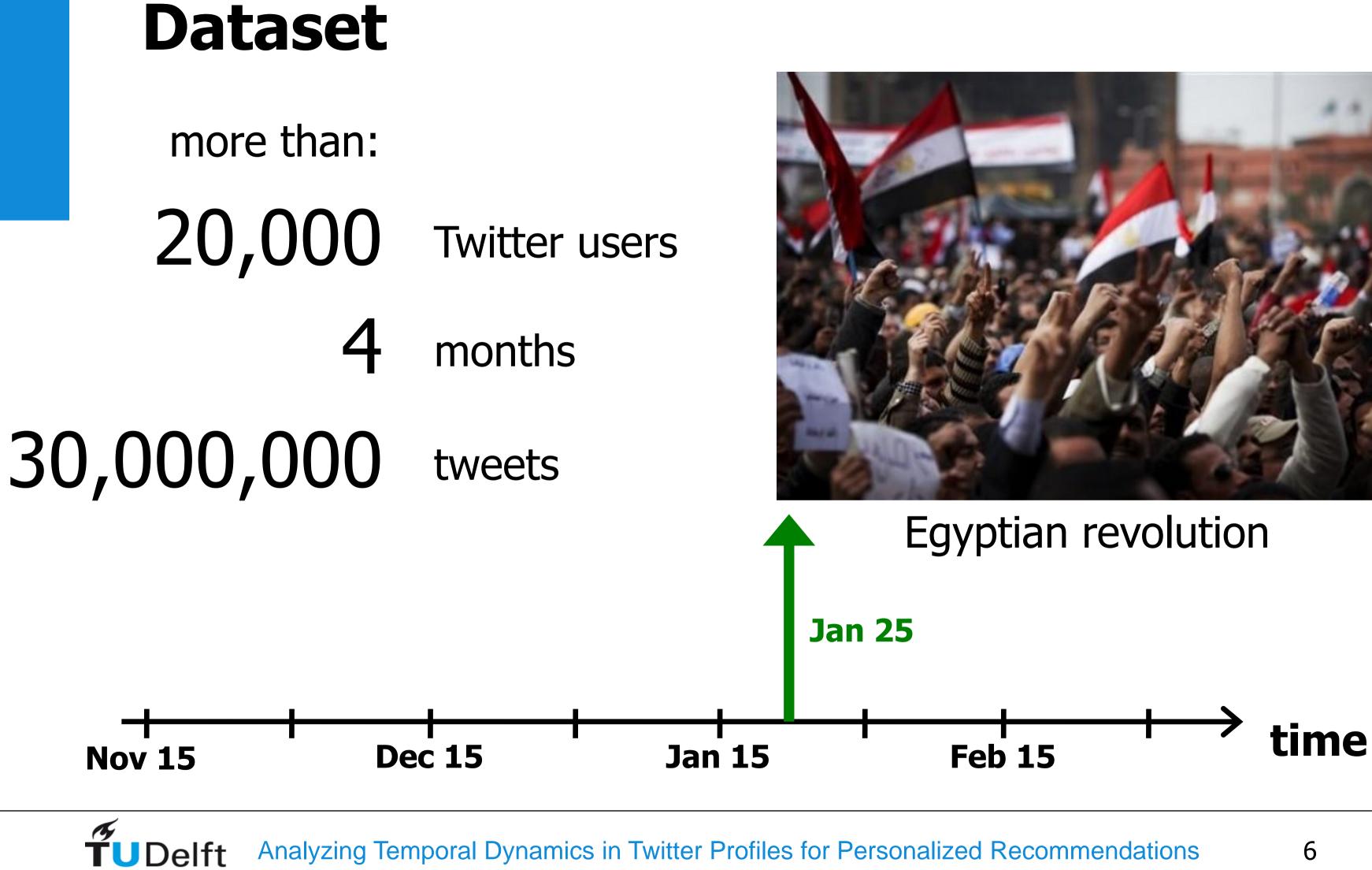
time

# **Research Questions**

- **1.** How do topics evolve over time?
- 2. How do Twitter-based user profiles evolve over time?
- 3. Can we exploit Twitter-based profiles for personalizing users' Social Web experience?



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# What are topics? How can we represent a topic? 1. How do topics evolve over time?

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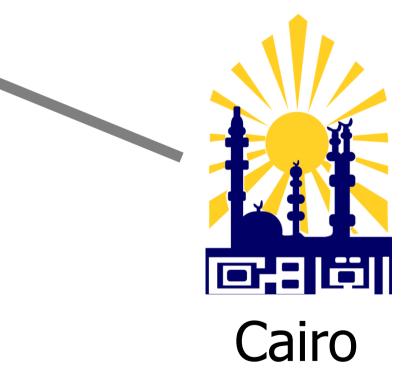
# **Representing a topic: via entities (and** hashtags)

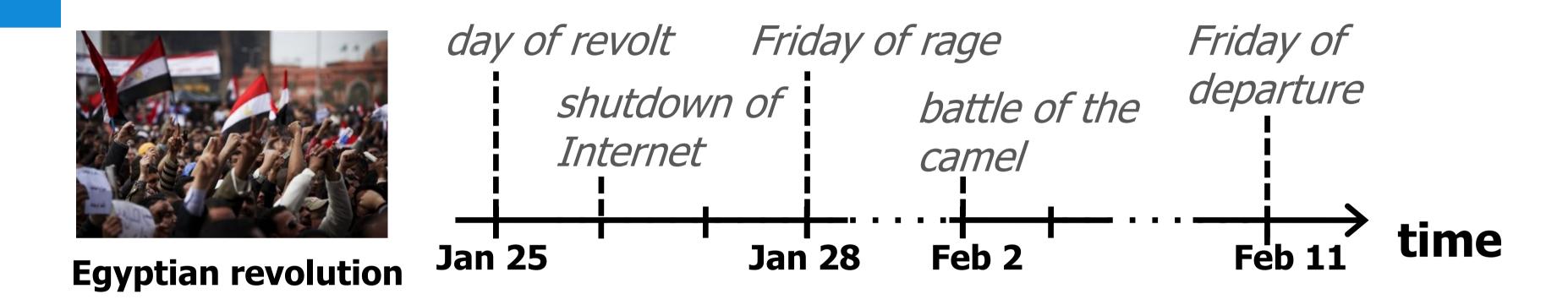


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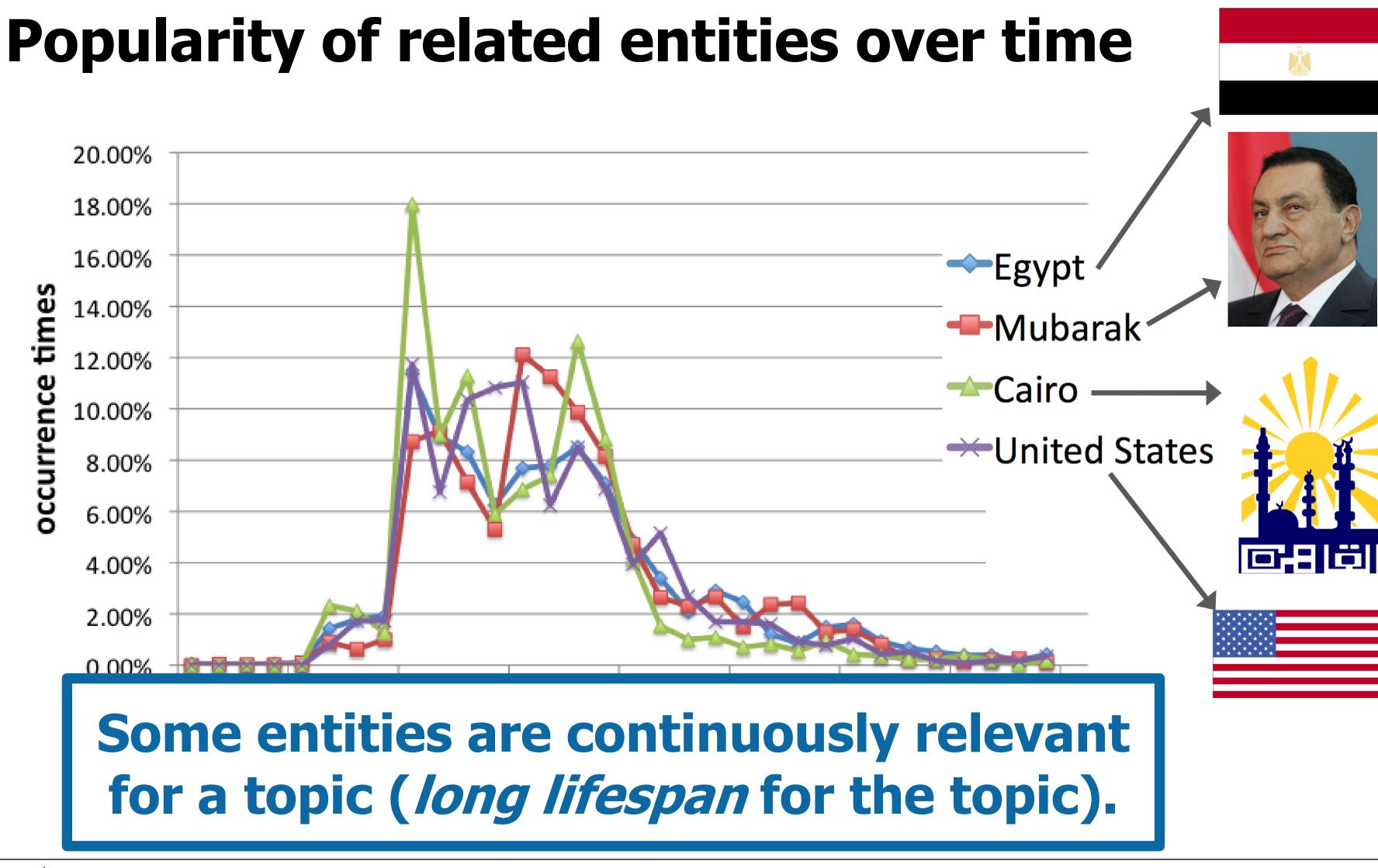
### Mubarak





# 1. How do topics evolve over time?

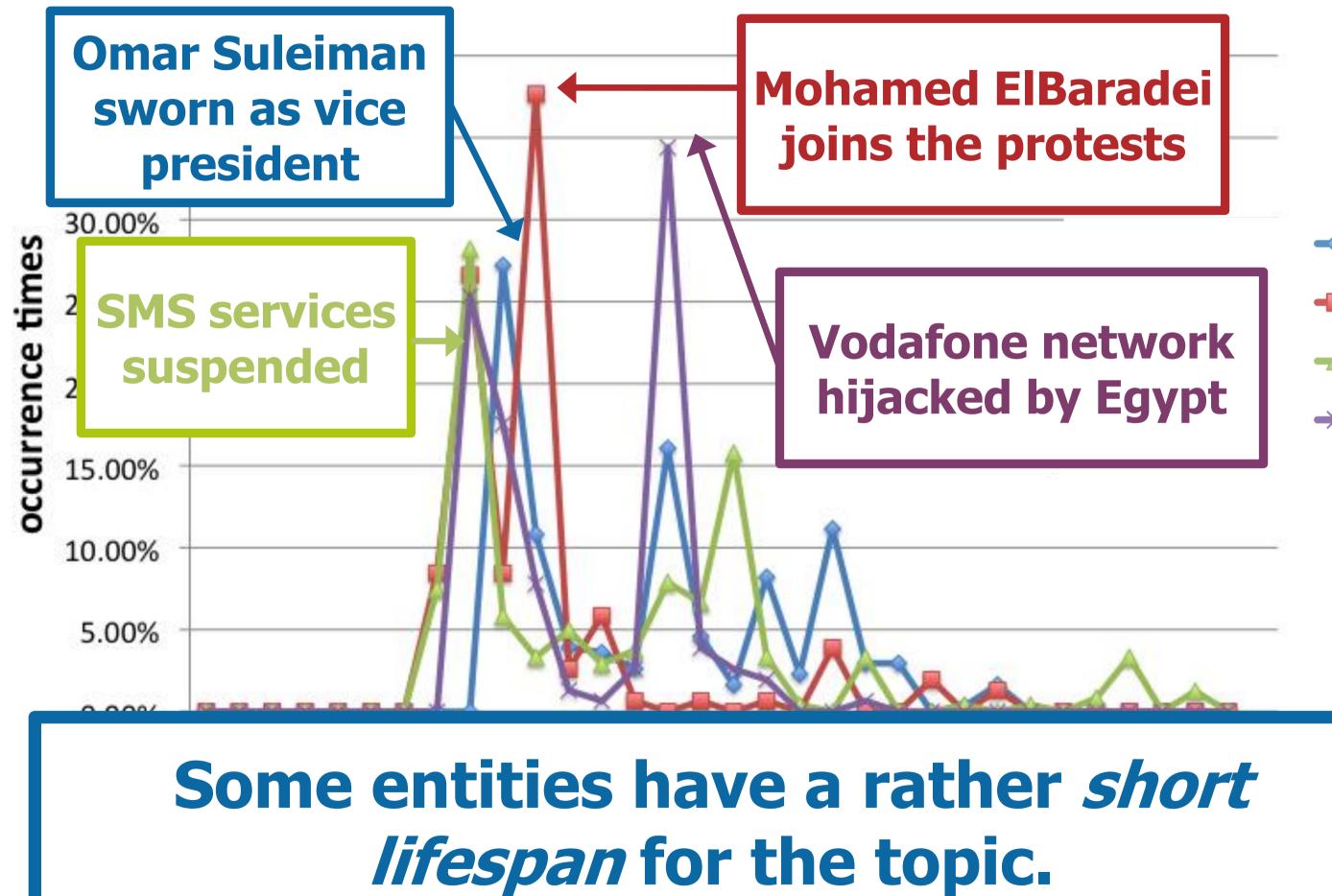




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# **Popularity of related entities over time (cont.)**





Omar Suleiman Mohamed ElBaradei -SMS Vodafone

### **1. How do topics evolve over time?** Observations

Importance of entities that represent a topic varies over time (long-term vs. short-term lifespan of entities)

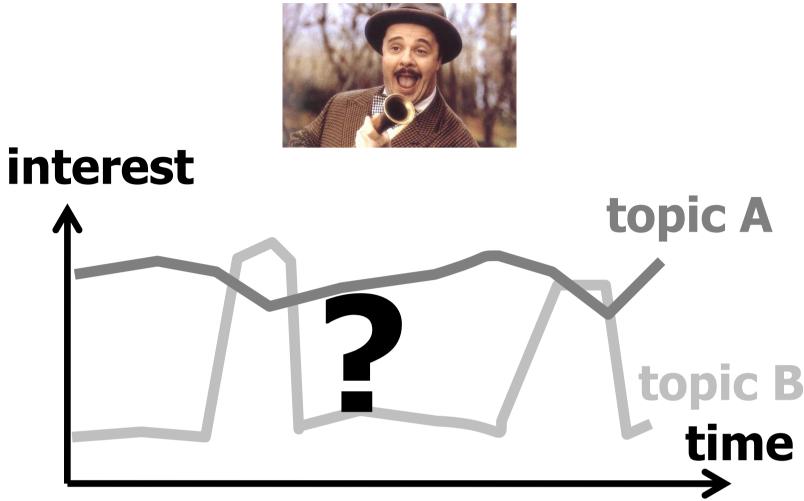
### **Representation of a topic (topic profile) depends on** the time when it is requested





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### (When) is Bob interested into the topic?



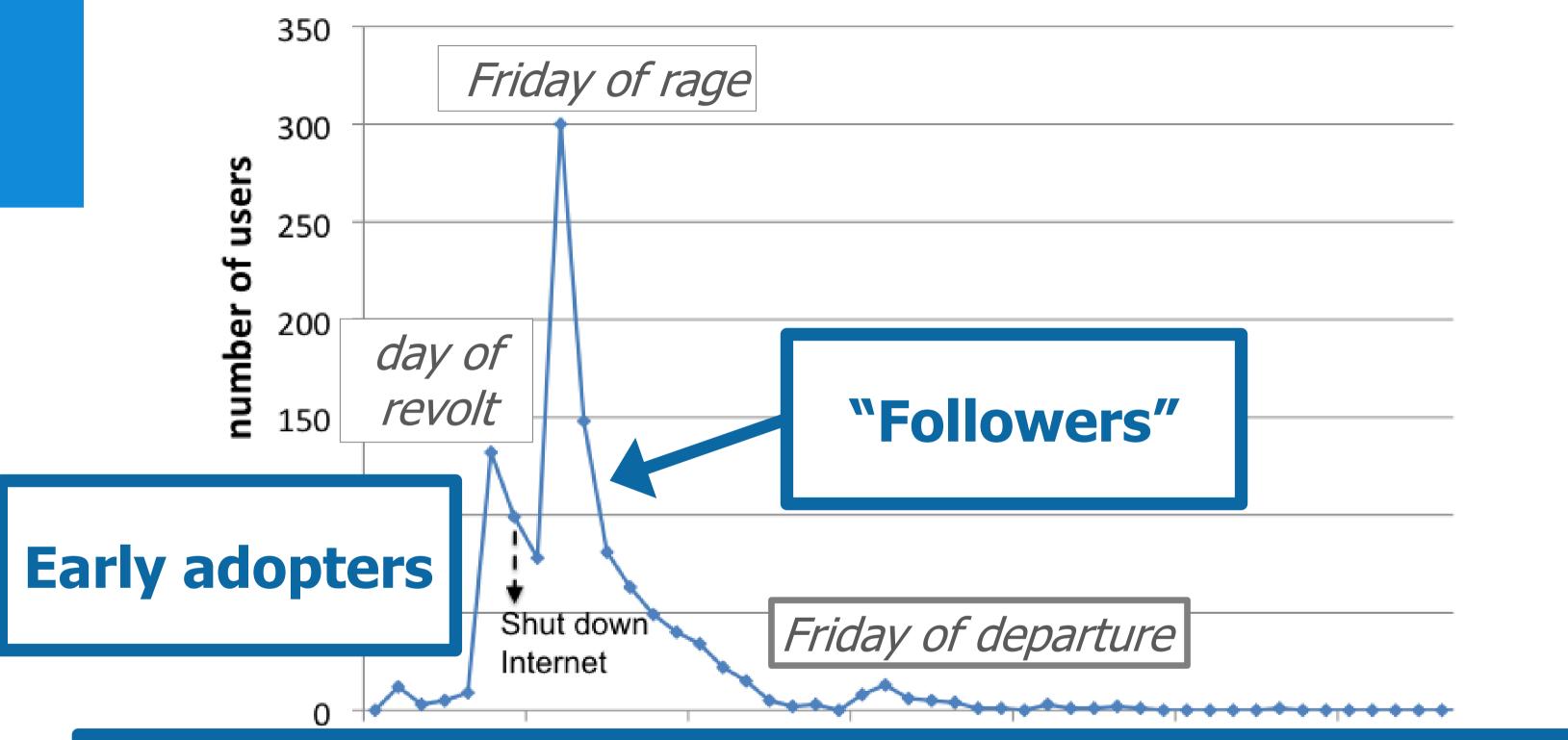
# 2. How do the interests of individual users into a topic change over time?

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time

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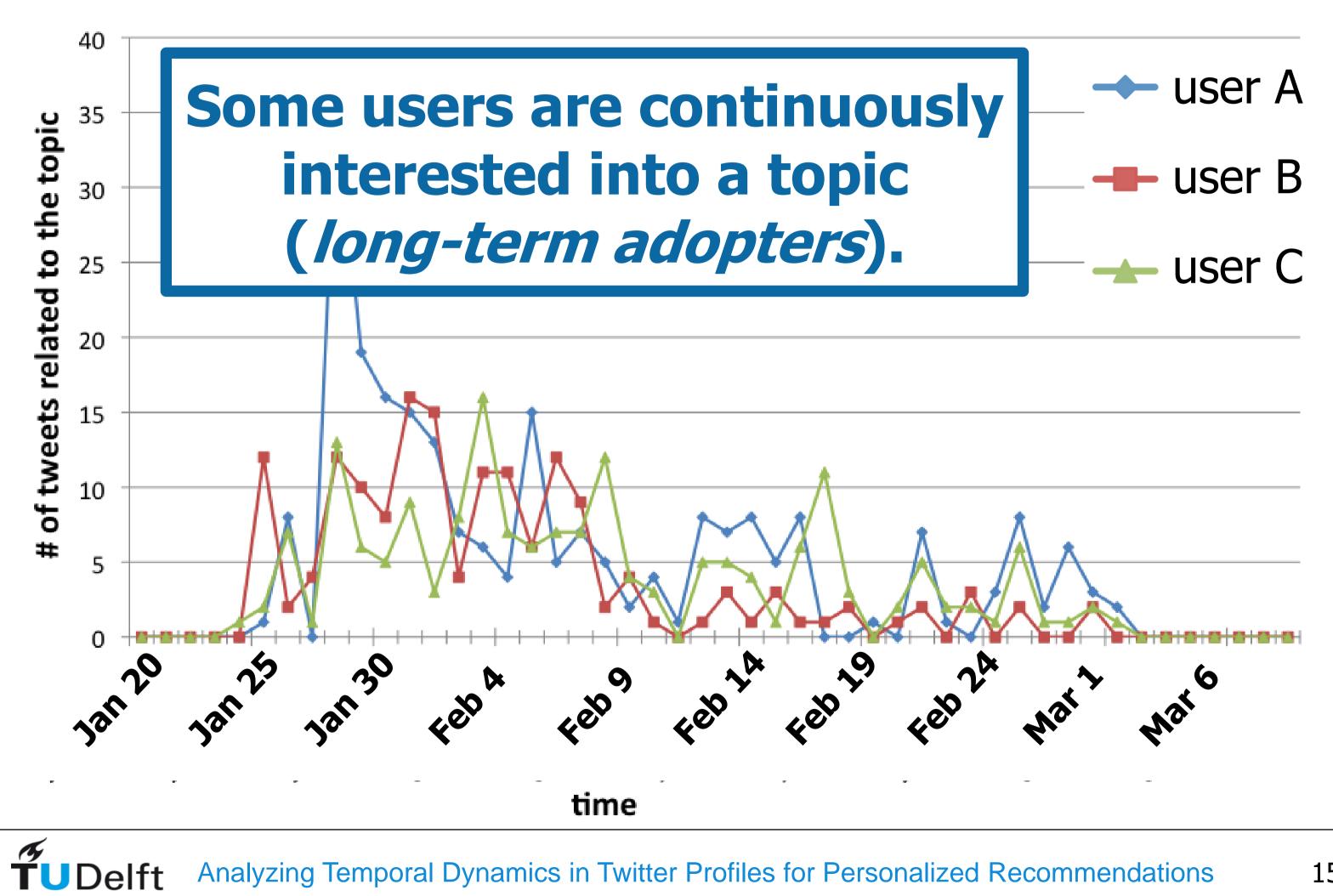
# When do users become interested?



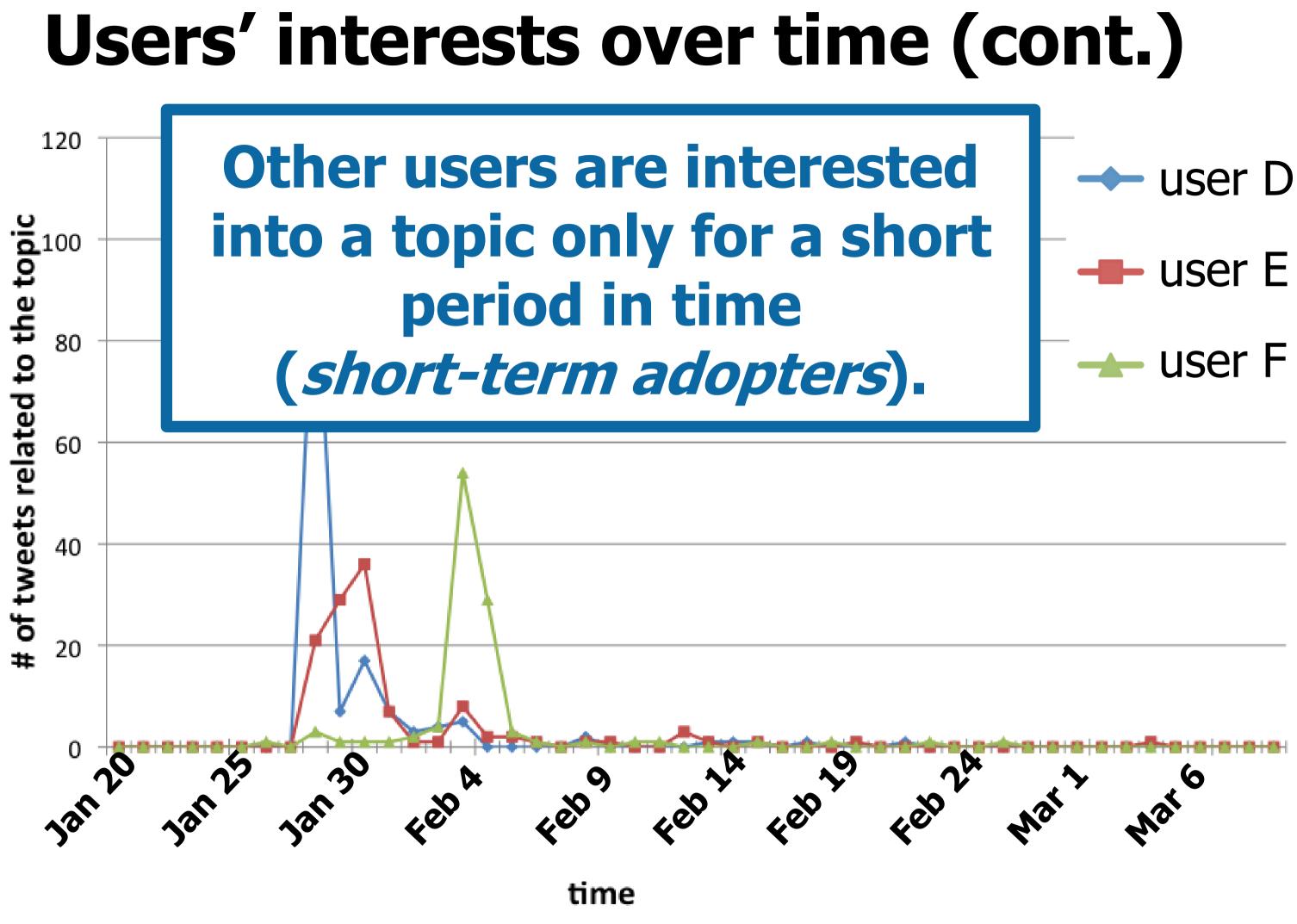
### If a user becomes interested into the topic then she become interested within a few days

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# **Users' interests over time**

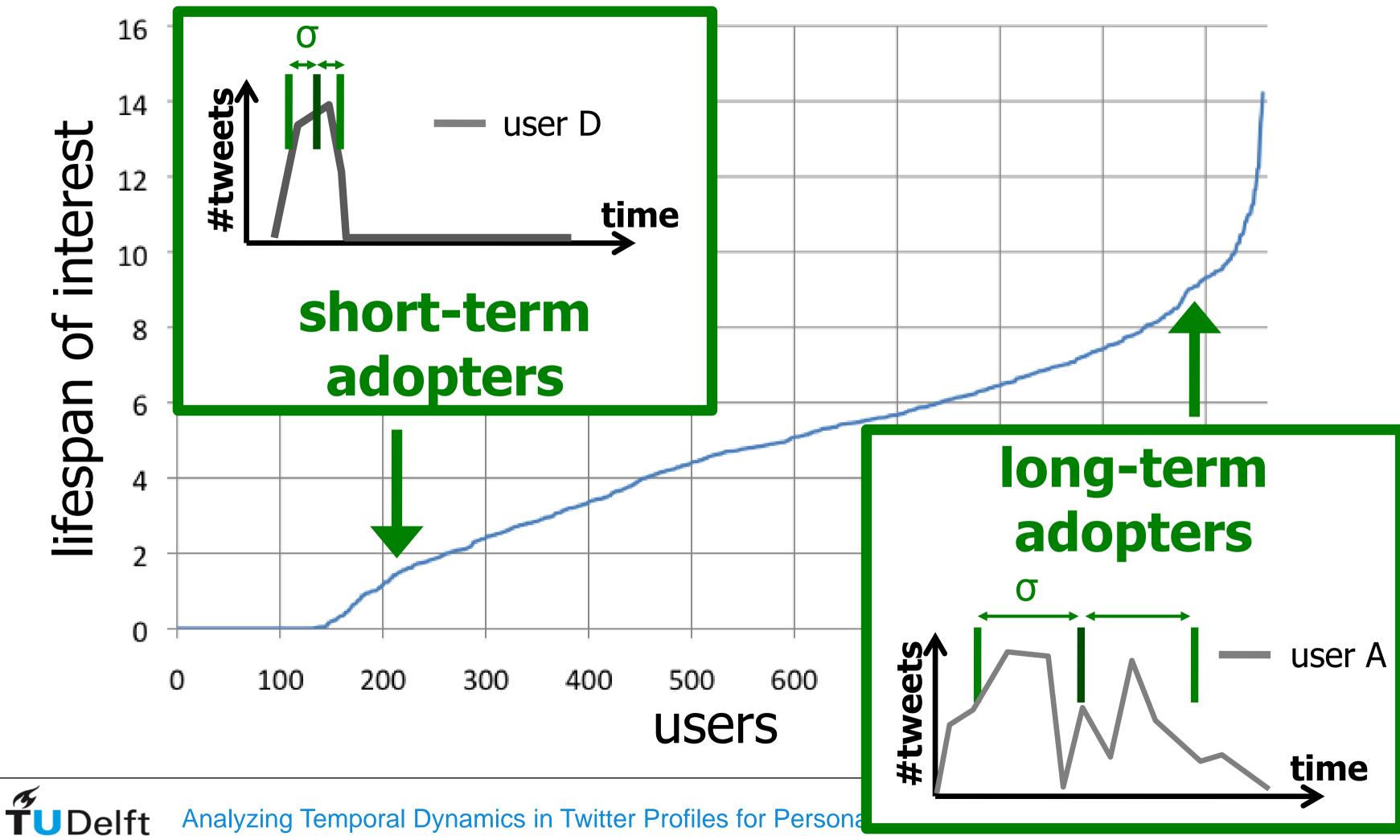


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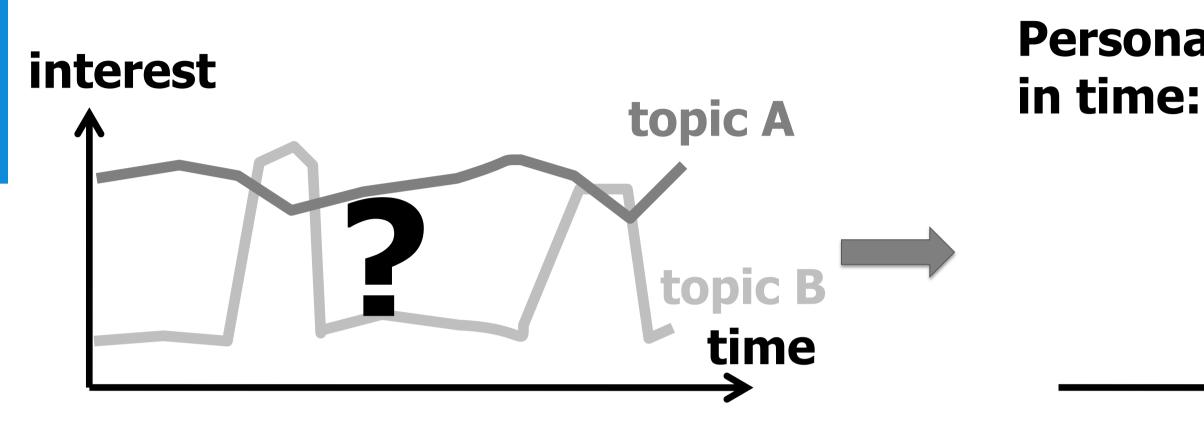
# Users' interests over time (cont.)



Analyzing Temporal Dynamics in Twitter Profiles for Persona

### 2. How do the interests of individual users into a topic change over time? -> Observations

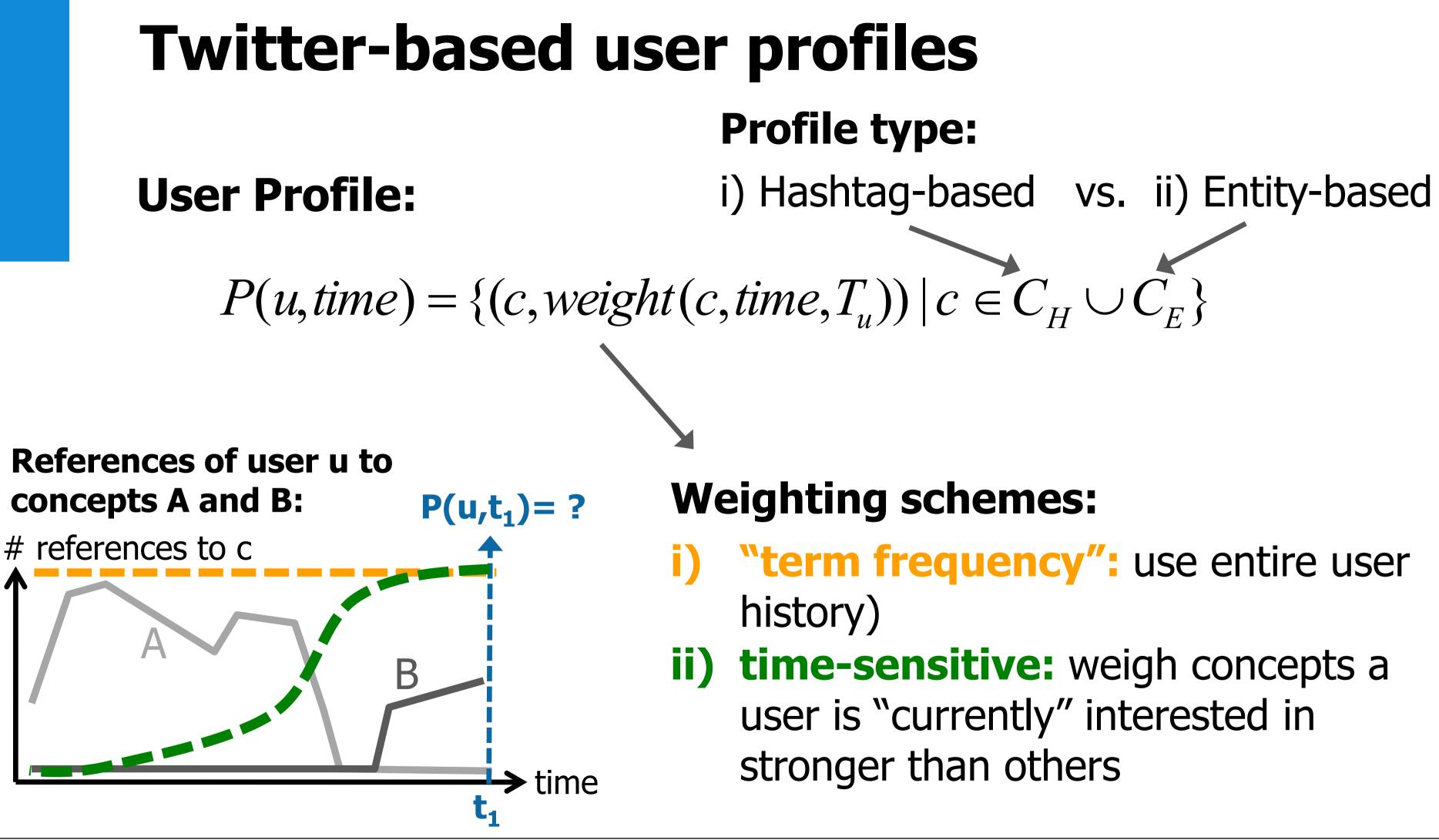
- Most users, who are interested into the topic, become interested within a few days
- Lifespan of users' interest:
  - Long-term adopters
  - Short-term adopters
- High overlap between early adopters and long-term adopters



# 3. Can we exploit Twitter-based profiles for personalizing users' Social Web experience?

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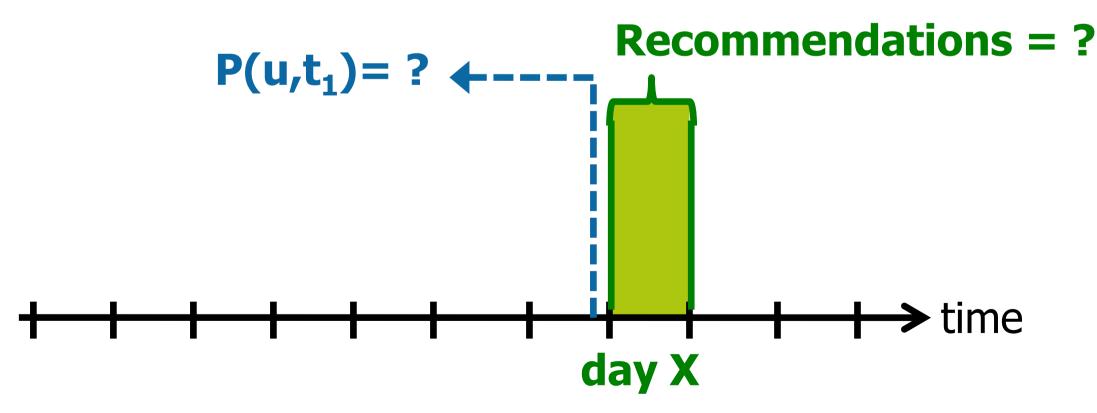
# Personalized recommendations in time:



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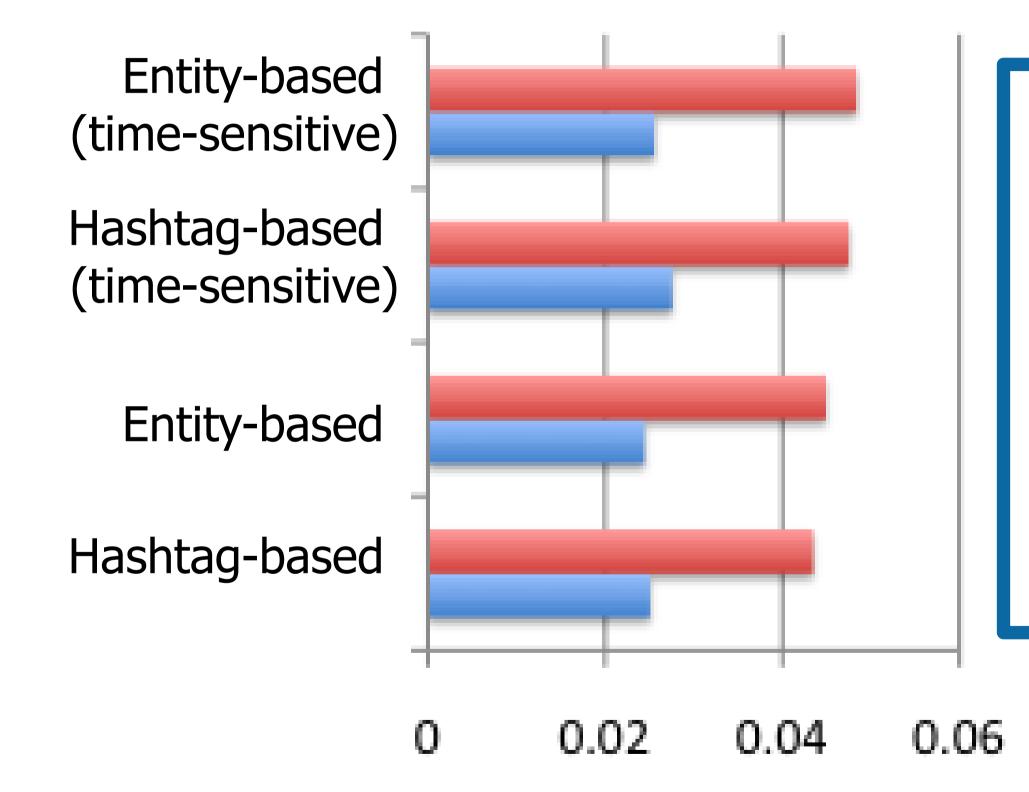
# **Twitter-based Profiles for Personalization** • **Task:** Recommending Web sites (= tweets with URLs) Recommender algorithm: cosine similarity between

- profile and tweets
- **Ground truth:** re-tweets of users
- Candidate items: URLs posted on day X
- Evaluation period: 12 days (Jan 20<sup>th</sup> Jan 30<sup>th</sup> 2011)



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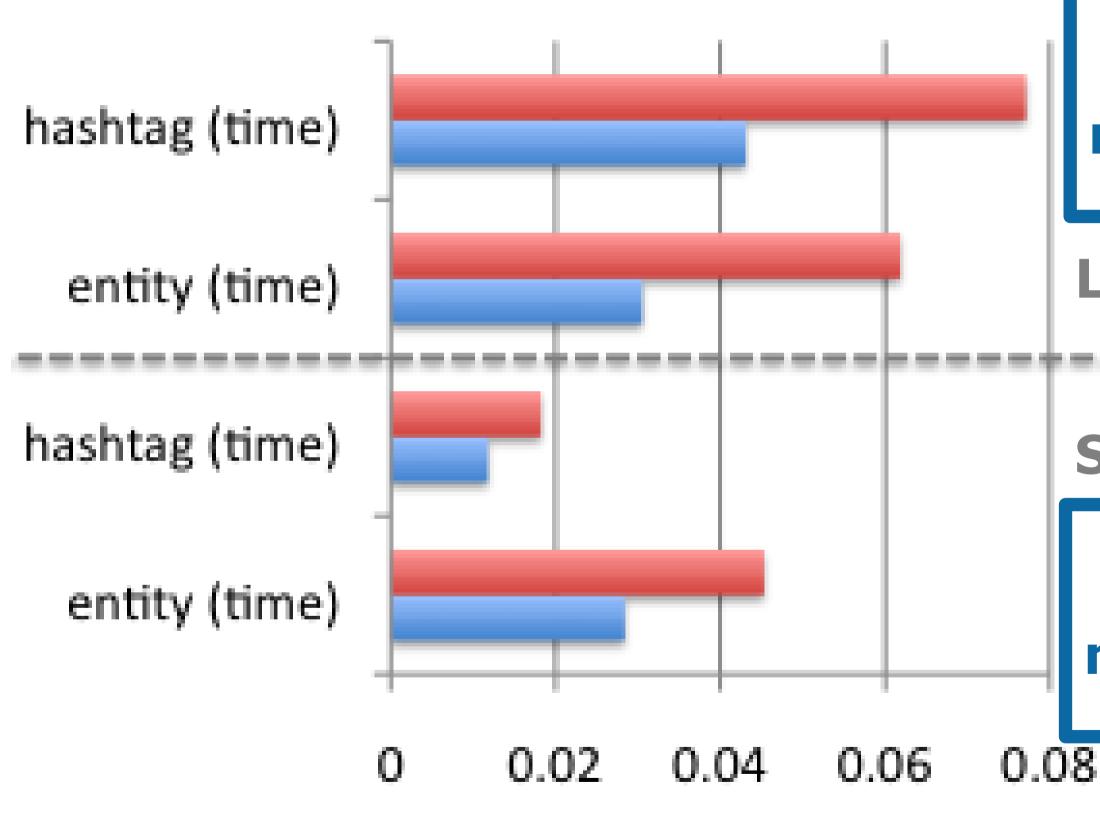
# Which user modeling strategy is best for computing personalized recommendations?



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### Time-sensitive profiles improve recommendation quality

# "Best" user modeling strategy varies for different types of users



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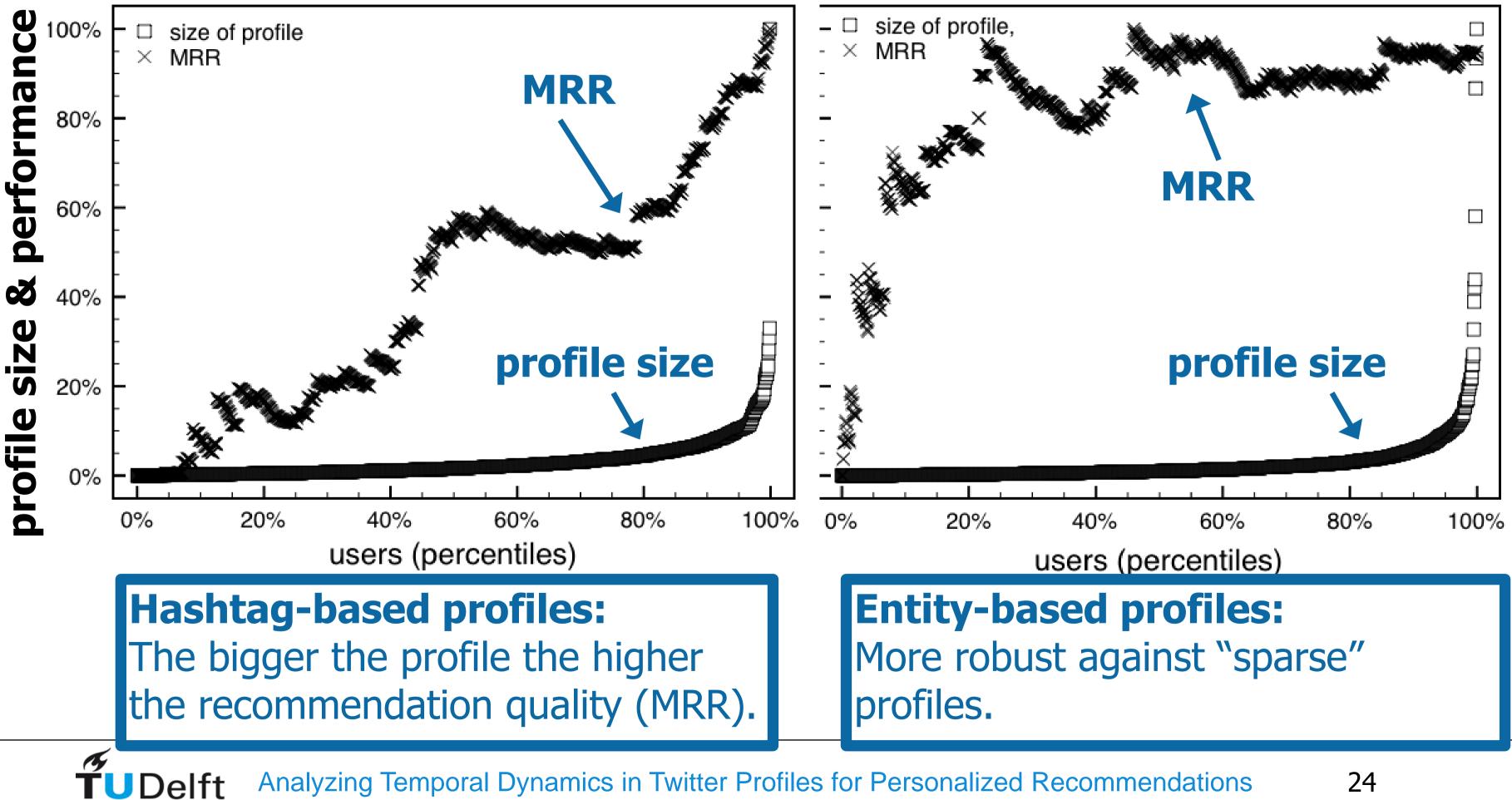
### Hashtag-based user modeling performs best

### **Long-term adopters**

### **Short-term adopters**

### **Entity-based user** modeling performs best

# Impact of profile size on recommendation quality



# **Conclusions and Future Work**

### **1.** Topics on Twitter:

 Importance of entities for a topic varies over time (long-term vs. shortterm entities)

### 2. User interests over time:

- Majority of users becomes quickly (few days) interested in a topic Long-term adopters vs. Short-term adopters

### 3. Twitter-based profiles for personalization:

- Time-sensitive user modeling improves recommendation quality
- Selection of user modeling strategy should take the type of user into account:
  - Long-term adopters: hashtag-based
  - Short-term adopters: entity-based

### **Future work:** for what type of personalization tasks can we exploit what type of Twitter profiles? http://wis.ewi.tudelft.nl/tweetum/

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### Thank you!

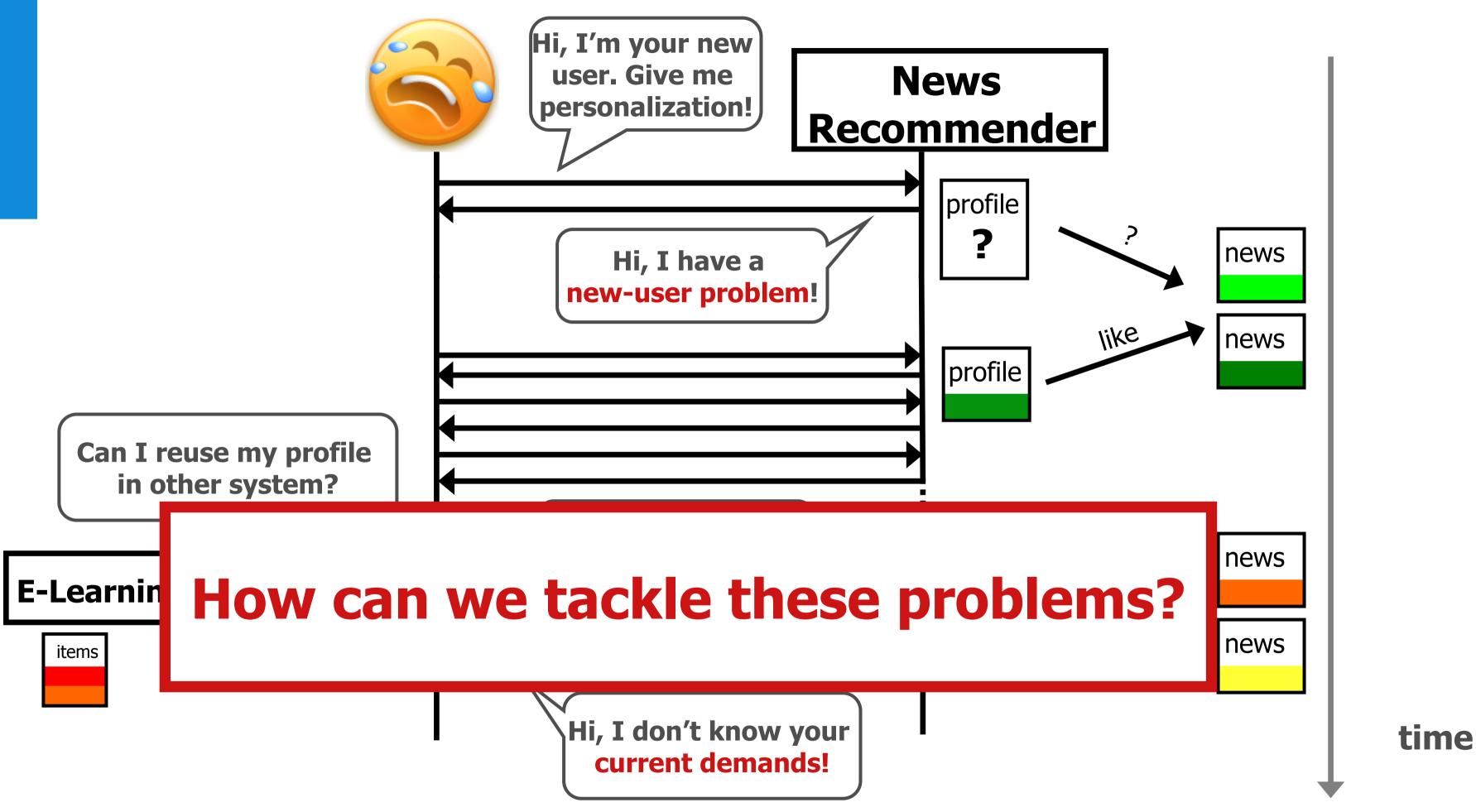
### Fabian Abel, Qi Gao, Geert-Jan Houben, Ke Tao



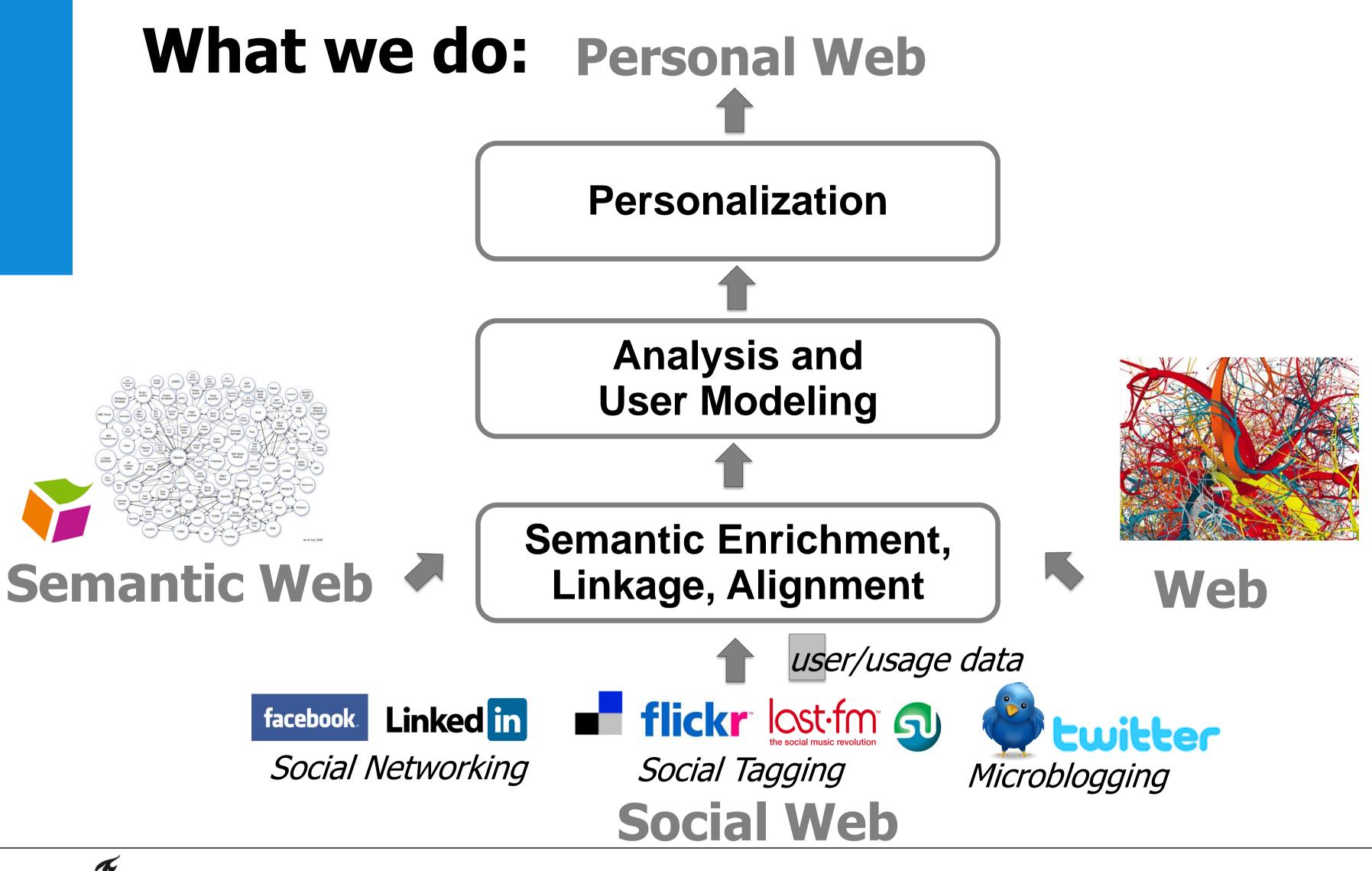
### Twitter: @persweb http://wis.ewi.tudelft.nl/tweetum/

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### **Problems of today's Web Systems**



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### Twitter

# 200,000,000 Twitter users 60,000,000 tweets per day 85% of the tweets related to "news" 2 questions...

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