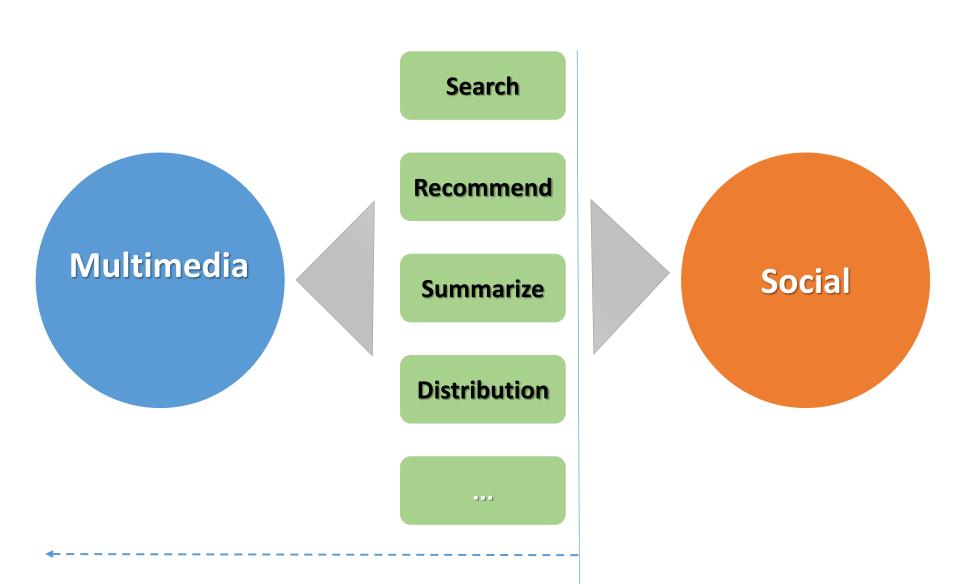
## **Social-Sensed Multimedia Computing**

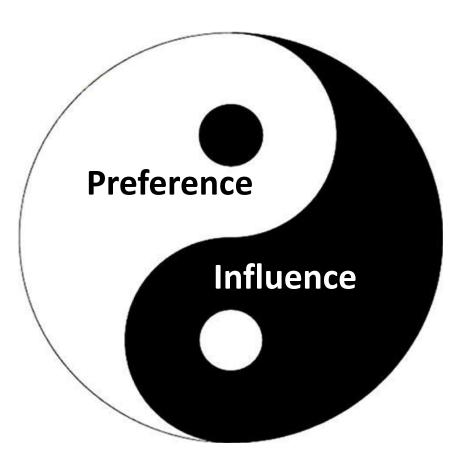
Wenwu Zhu

**Tsinghua University** 

## **Multimedia Computing**

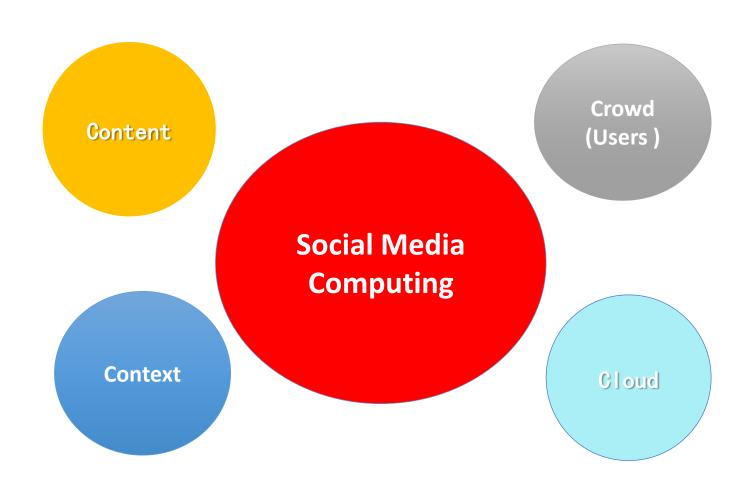


## **Sense from Social**



User behaviors
User preference
User influence
Use relations

## Social Media Computing



### **Social-sensed Multimedia Computing**

**Social-sensed Image Search** 

MMM'13 Best Paper Award

**Social-sensed Recommendation** 

SIGIR'11, TMM'13

**Social-sensed Media Summarization** 

**MM'12 Grand Challenge Award** 

**Social-sensed Media Distribution** 

**ACM MM '12 Best Paper Award** 

**Social-sensed Emotion Prediction** 

**ACM MM'13** 

## Social Sensed Image Search

## **Today's Image Search**

50% search sessions fail to provide any satisfactory results, even worse for image search.

The visual style cannot be clear expressed by

# User Data

search is often oratory with clear goal.

**Intention Gap** between user search intent and query

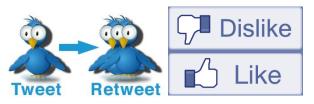
### **Social Media**



### **User Profile**

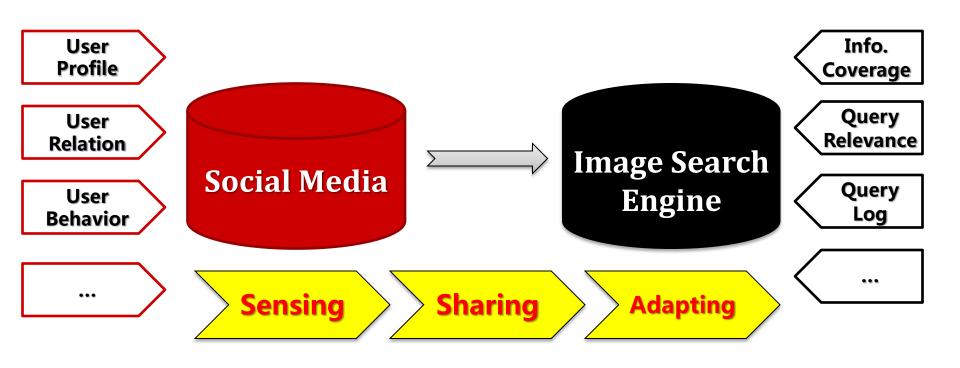


### **Behaviors**



Combining knowledge from Social Media will significantly reduce the Intention Gap in image search.

## **Bridging Image Search and Social Media**



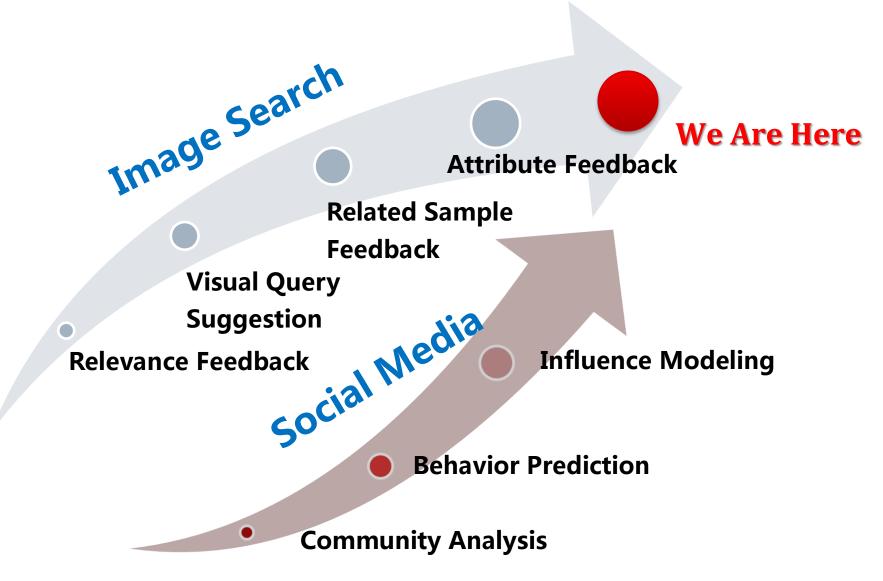
### **Social Relevance**

The degree of relevance with user interest

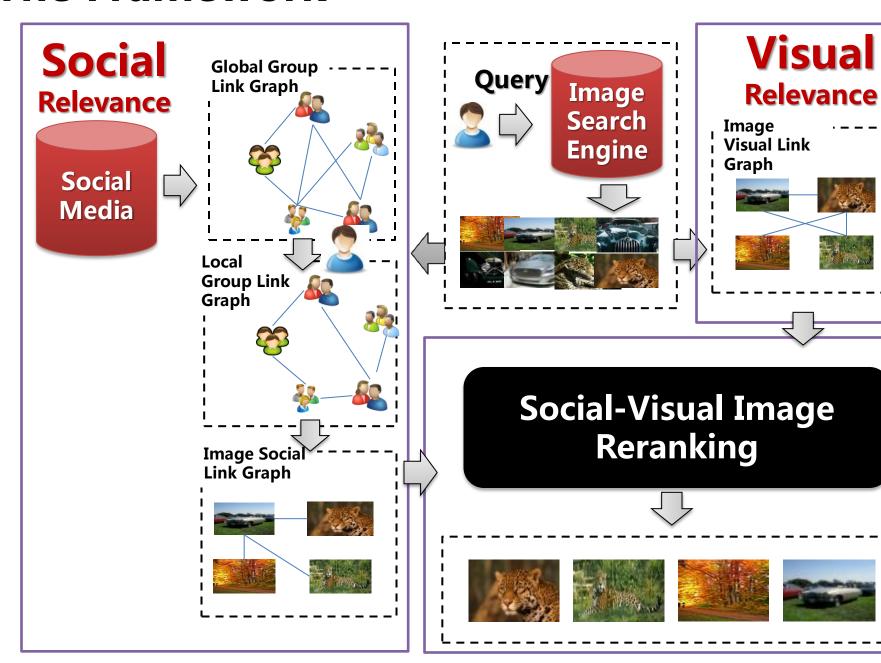
### & Visual Relevance

The degree of semantic relevance with the query

## **Related Work**



### The Framework



### **Social Sensed Recommendation**

### **Social Media Recommendation**

















#### Recommendation

- is a natural process in real life
- >drives the diffusion of new ideas, information and tools.



#### Recommendation

- should simulate the natural process
- ➤ should facilitate information seeking and sense making

### **Challenges in Social Media Recommendation**

Different Environment

Different

**Behaviors** 

Different **Multimedia** 

**New Problem** 

Two Domains → Multiple Domains

Individual → Interpersonal

**Object Label** → **Rich Semantic** 

**New Observations** 

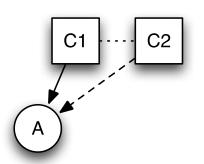
Social recommendation across multi-domains

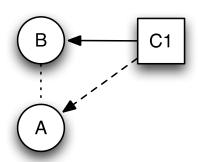
Social contextual recommendation

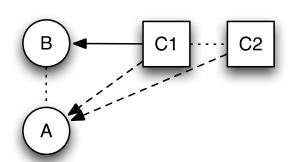
Social multimedia recommendation

**New Research** 

# Interest-oriented: Joint Social-content Recommendation







(a) Content based recommendation.

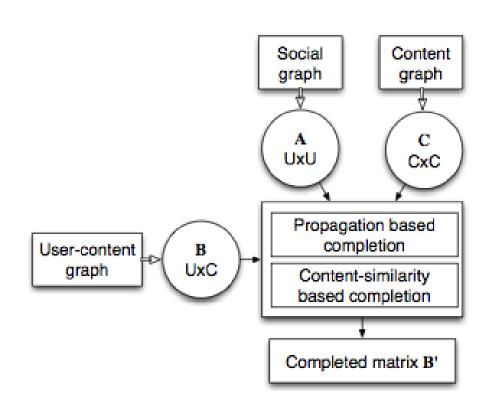
- (b) Social based recommendation.
- (c) Joint social and content recommendation.
- Joint social and content recommendation
  - Make use of user relationship, content similarity simultaneously

### **Joint Social-content Recommendation**

- Components in JSCR
  - User-content update based on social graph, content graph and user-content graph
  - User-content relative space construction
  - Social action based relative evaluation
    - Recommendation for import
    - Recommendation for re-share

### **User-content Matrix Updating**

- Propagation based updating
  - Inferring users' interests from the propagation patterns
- Content-analysis based updating
  - Inferring users' interests from content similarities

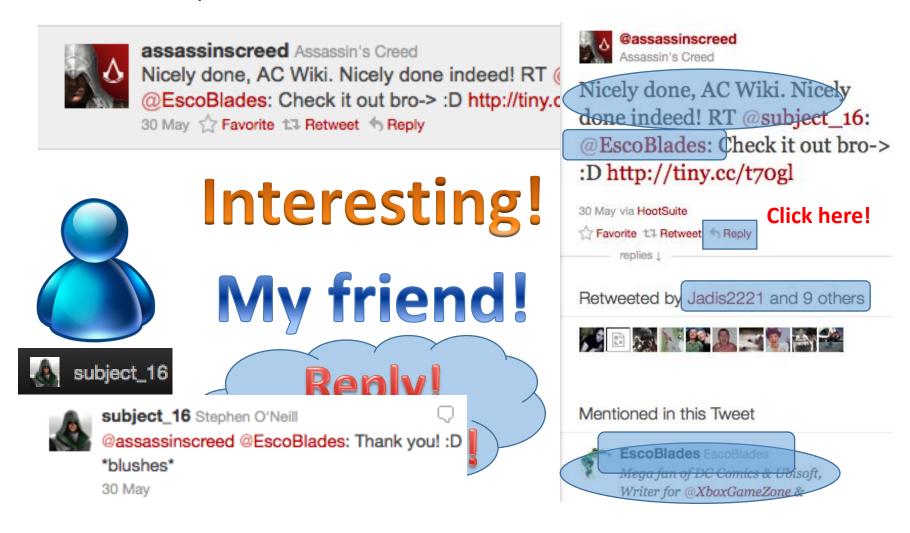


### Social Context Recommendation

- Social context information give us new thinking
  - Rich context for user to make information adoption decision
  - Context can help us to understand and then predict user behaviors
  - Intention of Information adoption: personal issue or social issue?
  - => Social context recommendation

## Information Adoption Mechanism

• In Twitter, a user receives a tweet



## Whether to Adopt the Item

 Read the content and its comments to see whether the item is interesting

## User Preference

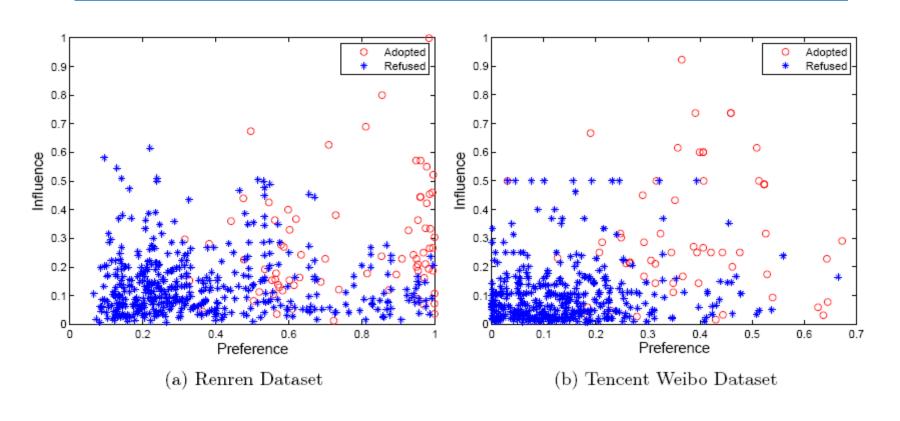
 Care about whom the sender is, whether the sender is a close friend or authoritative

## Social Influence

## Social Contextual Information

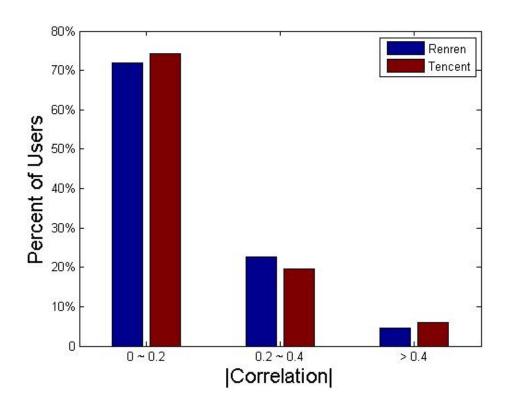
## Data-Driven Study

Accepted cases and refused cases have different distributions in the preference-influence space

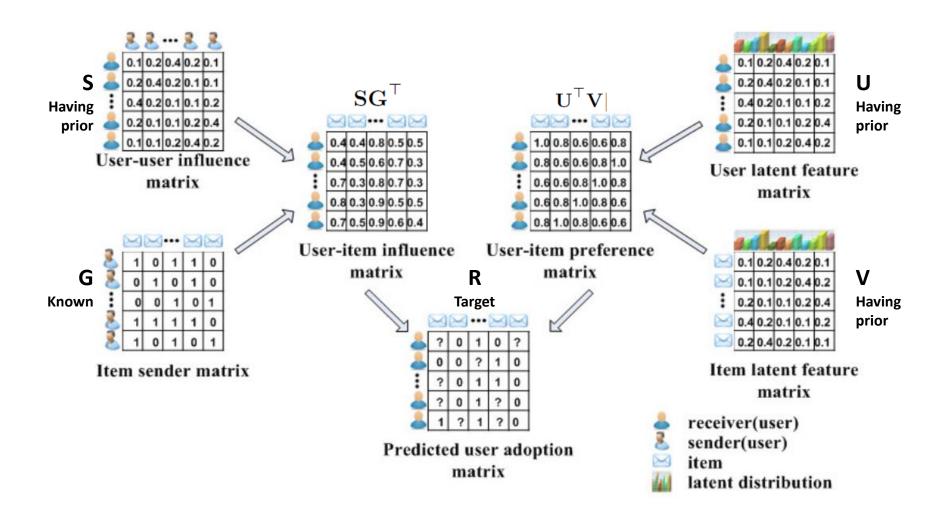


## Data-Driven Study

# Preferences and influences are weakly correlated for most users



### Social Contextual Recommendation



## **Social Sensed Video Replication**

# Social Network Changes How People Consume Multimedia



Over **700** YouTube video links are imported to Twitter every minute

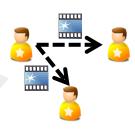
Over **500** years of YouTube videos are watched every day on Facebook

Over **30%** users select videos using social network service in China

#### **Related Works**



**Content correlation for UGC distribution** [INFOCOM'09]

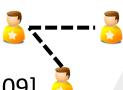


Propagationbased socialaware replication [MM'12]





Cache, CDN, and P2P [TMM'04, TMM'07, MM'09]



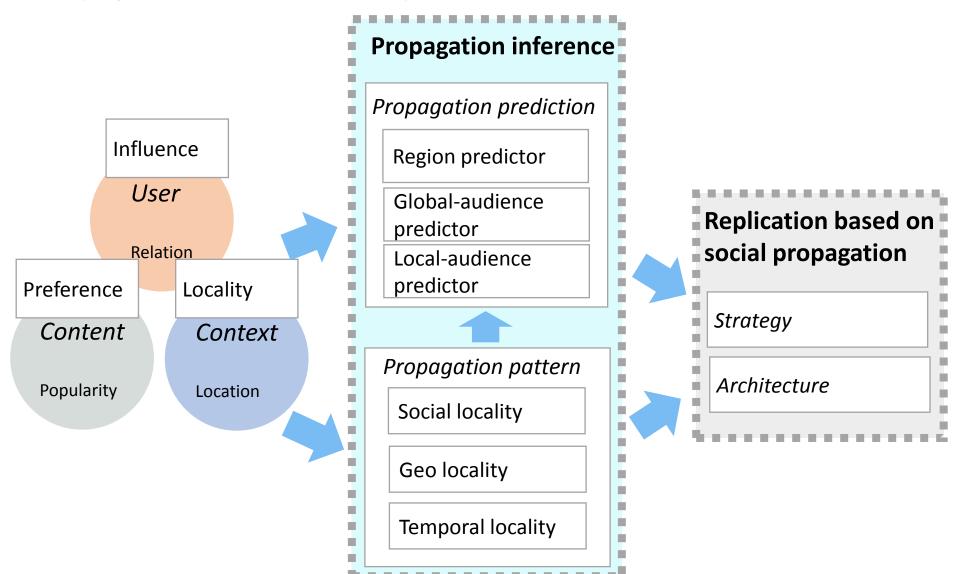
User relation/influence for social media distribution [SIGCOMM'10, INFOCOM'12]



2005

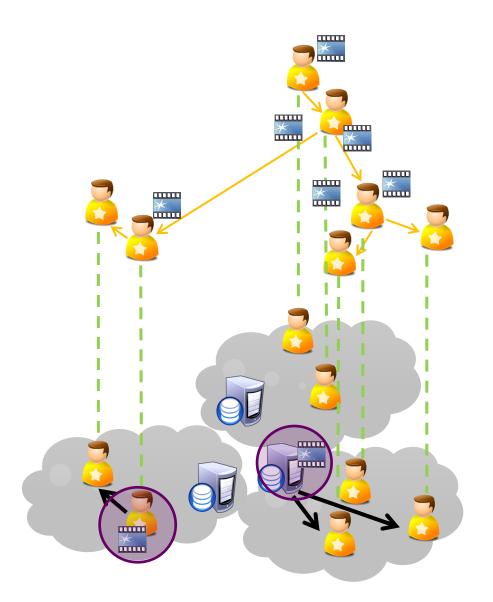
### The Framework

**Propagation Inference**: Jointly consider user, content, and context



### **Architecture Design of PSAR**

Propagation pattern: social, geographical and temporal localities



Content propagation predication



Propagation-based social-aware replication



Edge-cloud and peerassisted architecture

## Thanks!

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