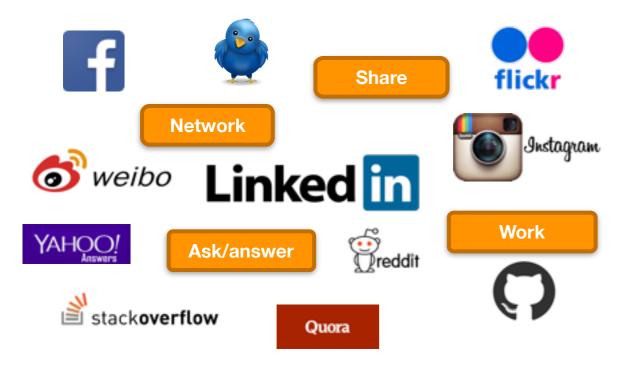


# Recommender Systems and Beyond in Web Information Systems

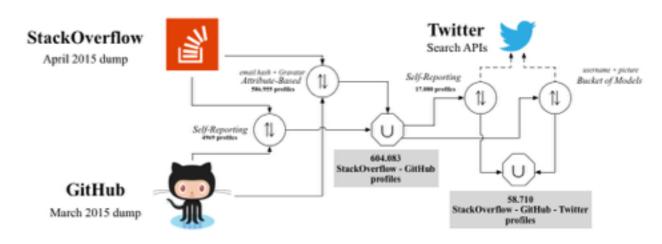
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### What do we do?

#### **Social Web Data Science**



Example 1: modeling user expertise, engagement, and topical diversity for question recommendation in community Q&A systems.



Example 2: modeling user cultural background, and city gene for POI recommendation in socio-spatial data analytics systems.

### **Topics:**

- User Modeling
- Crowdsourcing
- Urban Analytics
- Recommendation

Resident
Tourist

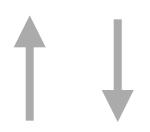
Airport

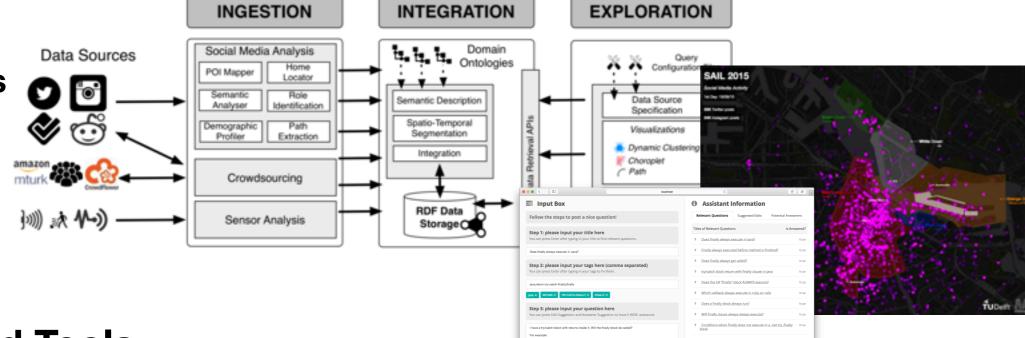
Knowledge Crowdsourcing
Acceleration

## What do we contribute?

# Systems - Social G

- Social Glass
- E-WISE





 $G_3 = |\{u_4, u_5\}|$ 

 $G_5 = \{u_1, u_2, u_3, u_4, u_5\}$ 

 $G_2 = |\{u_3\}$ 

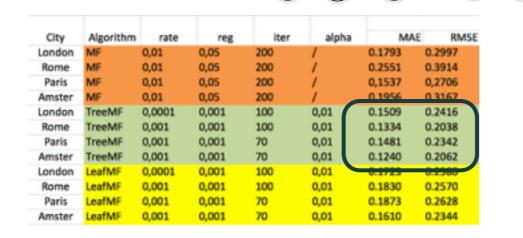
 $G_4 = [\{u_1, u_2, u_3\}]$ 

 $G_1 = \{u_1, u_2\}$ 

### **Methods and Tools**

user model based recommendation

- Matrix and tensor factorisation
- Learning to rank
- LibMTF



$\frac{\partial \mathcal{J}}{\partial \mathbf{U}} = ((\mathbf{O} \odot \mathbf{U} \mathbf{V}^T) \mathbf{V} - (\mathbf{O} \odot \mathbf{R}) \mathbf{V} + \alpha \mathcal{L} \mathbf{U}$	$+\lambda \mathbf{U}),$
$\frac{\partial \mathcal{J}}{\partial \mathbf{V}} = ((\mathbf{O}^T \odot \mathbf{V} \mathbf{U}^T) \mathbf{U} - (\mathbf{O}^T \odot \mathbf{R}^T) \mathbf{U} + \lambda$	$(\mathbf{V}).$

$$\frac{\partial \mathcal{J}}{\partial g_p} = \prod_{\forall a: G_a \in ancestors(G_p)} s_a Dis(G_p),$$

$$\partial \mathcal{J}$$

$$\frac{\partial \mathcal{J}}{\partial s_p} = \prod_{\forall a: G_a \in ancestors(G_p)} s_a(\sum_{\forall G_c \in children(G_p)} \mathbf{W}(G_c)).$$



# Opportunities?

