

# Leveraging Social Context for Modeling Topic Evolution

Janani Kalyanam<sup>\*</sup>, Amin Mantrach<sup>+</sup>, Diego Saez-Trumper<sup>+</sup>, Hossein Vahabi<sup>+</sup> and Gert Lanckriet<sup>\*</sup>

UC San Diego

<sup>\*</sup>University of California, San Diego, <sup>+</sup>Yahoo Labs



## Text corpora today

- Social Media posts every instant
- Constantly evolving corpora
- Free user vocabulary
- Volatile

Nelson Mandela's death and burial

Lindsay Lohan messes up contract with Oprah

Selina Gomez and Justin Bieber: "just friends"

South Africa's liberator dies at 95

Nelson Mandela Death (highly focused)

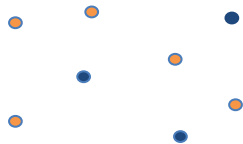
Celebrity Gossip (volatile, varied vocabulary)

## Our idea

- Traditional approaches: Bayesian (like LDA), or NMF-based
- Use textual content
- Our idea is to use side-information

## Our goal

When does side information in the form of community help in modeling topic discovery and evolution?



## Our approach

$X^t$  is documents-by-words

$U^t$  is documents-by-users

$$X^t \approx W^t H^t \quad U^t \approx W^t G^t$$

$H^t G^t$  each row is a topic/community represented as distribution over words/users

## Key assumption

$W^t$  explains each document in terms of the underlying latent topics/community.

## Temporal evolution

$$X^t \approx W^t M_T^t H^{t-1} \quad H^{t-1} \text{ Communities in the previous time step (considered known at time-}t\text{)}$$

$$U^t \approx W^t M_C^t G^{t-1} \quad M_T^t \text{ Evolution matrix}$$

## Objective function

$$\|X^t - W^t H^t\|^2 + \|X^t - W^t M_T^t H^{t-1}\|^2 + \|U^t - W^t G^t\|^2 + \|U^t - W^t M_C^t G^{t-1}\|^2$$

$L_T \qquad L_C$

Variables are  $W^t H^t G^t M_T^t M_C^t$

$$L = \mu L_T + (1 - \mu) L_C + R$$

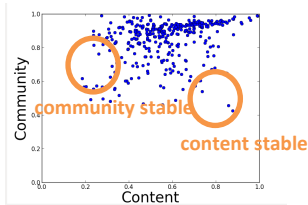
$\mu$  importance parameter

$R$  regularization

## How to validate?

- Content stable topics; Community stable topics; Mixed stable topics
- Obtain centroids of representations at time- $t$
- Cosine similarities for representation at time- $t$  and time- $(t+1)$  for all time
- Average all similarities

## Types of topics



## Experiments

### Community Stable

	K = 5	K = 10	K = 15	K = 20
NDCG	0.4081	0.4800	0.5029	0.5129
MAP	0.2653	0.3637	0.4007	0.4173

NDCG	0.3699	0.4496	0.4608	0.4138
MAP	0.2191	0.3596	0.3462	0.3420

### Content Stable

NDCG	0.6888	0.6055	0.6317	0.6623
MAP	0.5655	0.4784	0.5115	0.5559
$\mu$	1	1	0.75	0.75

NDCG	0.3699	0.4496	0.4608	0.4138
MAP	0.2191	0.3596	0.3462	0.3420

## Stability of M

$$\text{stability}(M) := 1/n \bullet \sum_i \text{abs}(\gamma_i)$$

