Spatial-temporal Analysis
of International Connections
Based on Textual Social Media Data

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# Background & Motivation

### Social Media

WHY? FREE, BIG, FROM USER

WHAT? Opinions, Events, Trends

Social Media + Geoinformation

**HOW?** Geotagging

WHAT? Travel behavior, local event detection

**Problems?** small percentage, bias

### **Geoinformation in text**



To map international connections from social media data, and reveal potential spatial and temporal patterns

## **Related work**

Stefanidis et al. (2013): mapping virtual Syria through social media content

## **Problems?**

analytics 1.900gle.com, analy How many Languages should be involved?

### How to solve?

Multiple subjects, one state, one language

In order to demonstrate how the structure of a virtual polycentric Syria can be gleaned through the analysis of social media content we use for this publication tweets that we collected over a period of one week (10 July through 17 July 2012) and included mentions to Syria or hashtag equivalent #Syria. As per standard Twitter

## **Context & Research Questions**

Sina Weibo: The biggest microblog platform in China Almost all users are Chinese people and post in Chinese

"How famous are other countries in Chinese community?"

"What are they famous for?"

"Do different genders have different preference?"

RQ1: What are the possible methods to harvest weibos that contain textual geoinformation at a state level.

RQ2: How should we process the textual content of weibo?

RQ3: What information visualization techniques should be used to present the different information we harvested from actual weibos and their metadata?

# Methodology

Determine the list of states

Weibo data collection

**Data formatting** 

Visual analysis

Keyword extraction

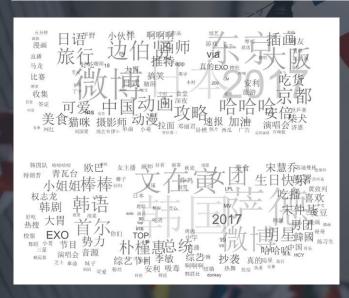
### Workflow

- Determine the list of states
  - 193 member states of UN, Vatican, Palestine, Taiwan. Transform into commonly used Chinese name
- Weibo data collection based on the state list

  No available Sina API. A web crawler was created to harvest JSON data
- Extract keywords from Weibo content with jieba NLP package 7 other attributes: State, user name, user gender, platform, follower, repost count, post time. (key: weibo ID)
- Visual analysis of the data



- 1. Closer states intend to have higher level of concern
- 2. The space-time distance has less influence (Saint Vincent And The Grenadines has 5 mentions)





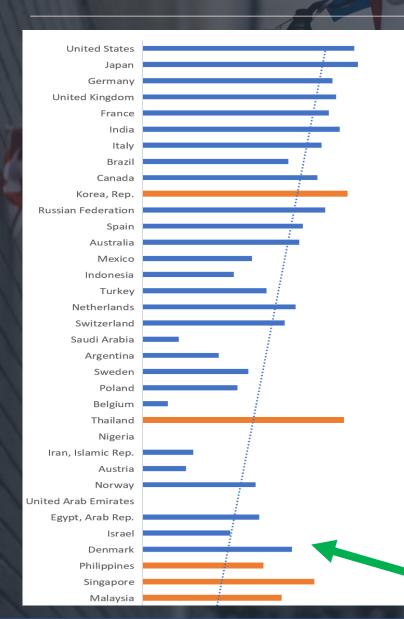
Higher culture similarity states, weibo user discussed more about popular culture

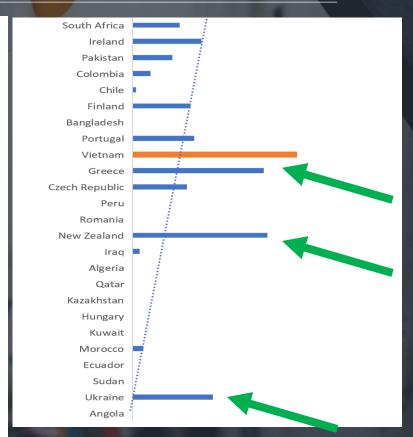
- Japanese animation
- South Korea pop music

lower culture similarity states, the discussions lean to general news or specialized interests

- Terrorism attack in UK and Brexit
- Football and political talks under Germany

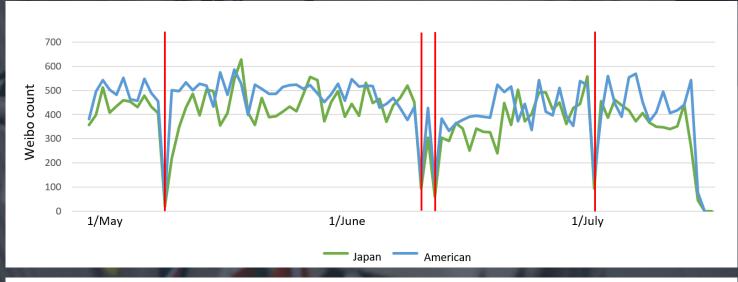
Word clouds could indicate the "stereotypes"

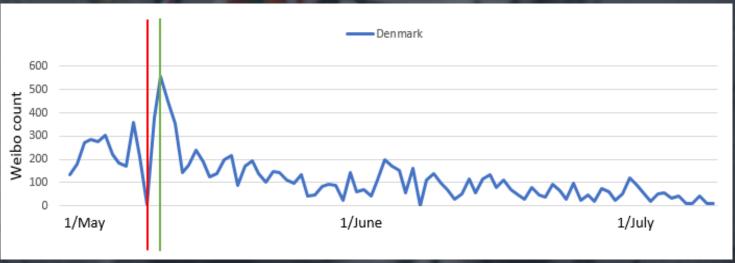


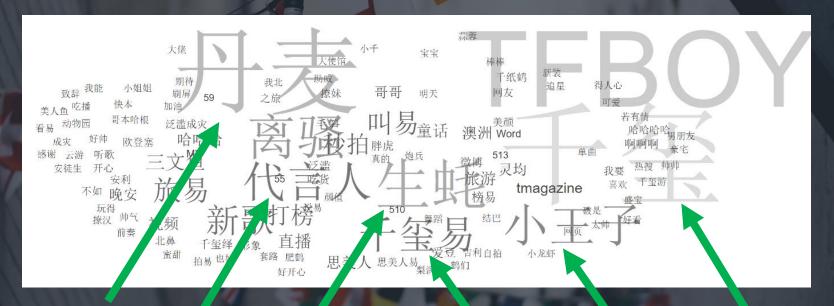


#### 2016 global GDP ranking VS data ranking

 the data amount rank generally consistent with GDP— or, "soft power"







Denmark qianxi little prince qianxi

Search for the weibos with highest repost count in May

Danish time on May 11 at 9 am, TFBOYS members of the members of the customs of the Danish tourism image spokesperson awarded the ceremony was held in Odense City, Denmark. Andersen hometown of the city of Odense and the Danish National Tourism Administration co-awarded the Commissioner Xi Danish tourism spokesperson certificate.

### Conclusion

- Textual geoinformation contained in social media could be harvested to depict international connections.
- The statistic features could be visualized. The spatial and temporal patterns could reveal interesting political and cultural events.
- With the help of nature language processing, visual result combined with human thought could lead to meaningful conclusions.

## Outlook

### Limitations

Limited choice of states

Unstable data acquisition

Based only on Chinese

Rudimentary NLP function

No user side spatial analysis

### Future work

Involve more entities. e.g. Crimea

Perform future study with API

Study other countries through Twitter

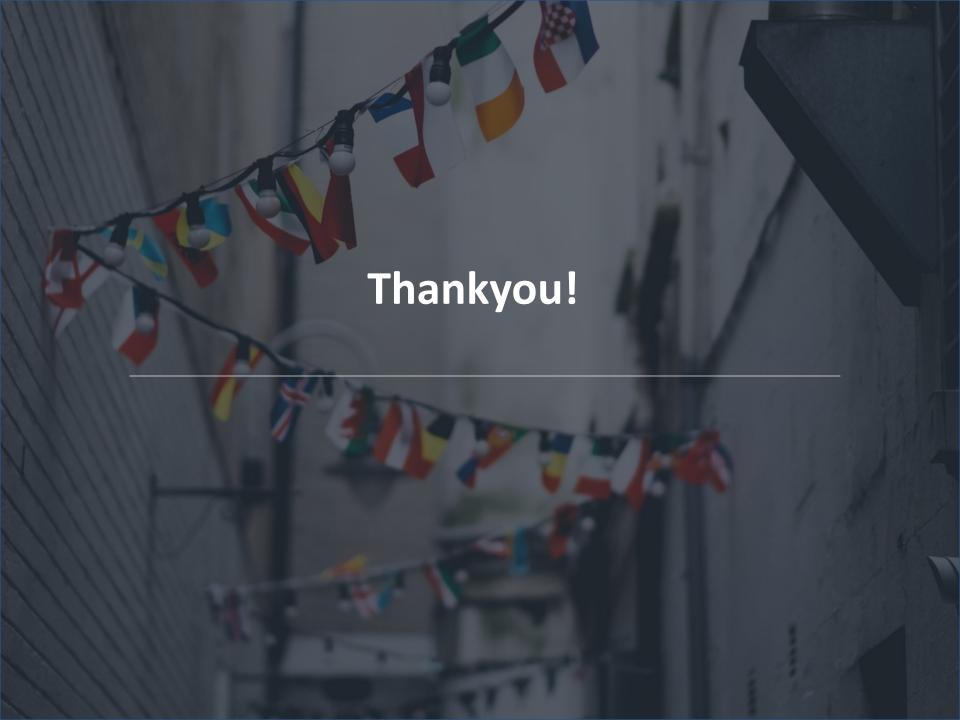
Include sentiment analysis

Consider user location study



"Our object was to highlight these emerging geographic opportunities as a result of the proliferation of social media, rather than to create a definitive map of virtual Syria."

- Stefanidis et al. (2013)



### **TF-IDF**

### frequency-inverse document frequency (TF-IDF)

- Important words tend to appear in the article more.
- Some commonly used words are frequently present in more than one articles

The importance of a word is proportional to the number of times it appears in one document, but it is inversely proportional to the frequency it appears in the corpus

$$tf_{i,j} = \frac{n_{i,j}}{\sum_{k} n_{k,j}}$$

### **Visual Data Exploration**

Initially, it is easy for the user to collect a huge amount of data from any other automated process, however

- If the data is displayed textually, the amount of data able to be presented is very limited.
- Without human participation, the value of the data cannot be interpreted.

Therefore, Visual data exploration (Visual data mining) aims at integrating human in data exploration process.

