## Network analysis and visualization for social media

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School of advanced sciences of Luchon, July 3rd, 2014





- Examples
- Practical Session 1: Basics of Gephi
  - Download http://gephi.org/download/
  - Example network: http://gephi.org/datasets/LesMiserables.gexf
- Practical Session 2: Create and visualize your own networks OR

Modeling the structure and evolution of online discussion cascades





# Part I: Examples for Network Analysis and Visualisation in Social Media





## **Outline Part I**

- Political User interaction on Twitter
- Political Affiliation on Wikipedia
- Emotional styles on Wikipedia
- Geographical distance and Friendship
- Sister Cities
- 6 Links between biographies on Wikipedia





## Outline

- Political User interaction on Twitter
- Political Affiliation on Wikipedia
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# Analysis of the Spanish General Elections of 2011

#### Research Questions

- Do political parties interact on Twitter?
- Do political parties use Twitter to engage in conversations or as one-way flow broadcast medium?
- Are there differences between the parties?

#### Dataset collected between Nov 4 and 24, 2011

- $\bullet \sim 3$  million tweets.
- $\bullet \sim 380.000$  users.

### Results published in



P. Aragón, K. Kappler, A. Kaltenbrunner, D. Laniado and Y. Volkovich.

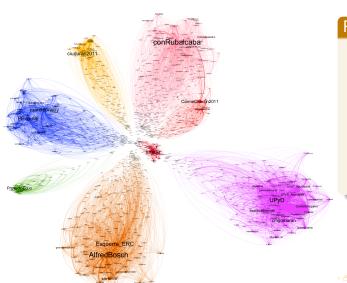
Communication Dynamics in Twitter During Political Campaigns: The Case of the 2011

Spanish National Election.

Policy & Internet, 5 (2), 2013.

#### Retweets

Users almost exclusively propagated contents from members of their own party



## Political parties

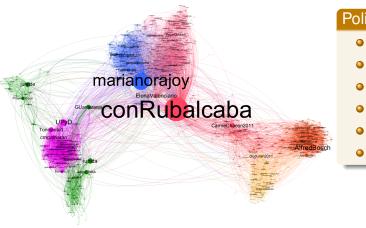
- PSOE
- PP
- EQUO
- IU
- ERC
- CiU
- UPyD





## Replies

The most intensive communication flows occur between members of the same party



## Political parties

- PSOE
- PP
- IU+EQUO
- ERC
- CiU
- UPyD

Some amount of communication also among members of

• PP - PSOE

• IU - UPyD - EQUO

• ERC - CiU

## Conclusions and Future Research

#### **Conclusions**

- Retweets: Balkanisation of Spain's (online) political sphere
- Replies: Inter-party communication happens but most of the interactions still occur within the parties.
- Political parties use Twitter as a one-way flow broadcast.
  - Low number of replies by candidate and party profiles
  - Low ratio between sent and received replies.
- New and minor parties tend to be more clustered and better connected ⇒ a more cohesive community.

#### **Future Research**

 In-depth analysis of the topological patterns of party networks to characterise the different party apparatus (centralised, decentralised, or distributed).

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## Motivation

## Does political polarisation also take place in Wikipedia?

 Obtain a deeper understanding of online interaction and collaboration among members of distinct political parties.

### Research questions

- Do political users in Wikipedia exhibit a preference for interacting with members of their same political party?
- Do we see a division in patterns of participation along party lines?

### Results published in



J. J. Neff, D. Laniado, K. E. Kappler, Y. Volkovich, P. Aragón & A. Kaltenbrunner.

Jointly They Edit: Examining the Impact of Community Identification on Political Interaction in Wikipedia.

PLoS ONE, vol. 8, no. 4, page e60584, 2013.

#### Introduction

#### Wikipedia visible side

Article Discussion



Main page

Help About Wikipedia

h Toolhov

▶ Print/export

Kiswahili

Русский

Svenska

ไทย

Norsk (bokmål)

Česky

Featured content

Donate to Wikipedia

Community portal

Contact Wikipedia

Recent changes

WIKIPEDIA The Free Encyclopedia

Presidency of Barack Obama From Wikipedia, the free encyclopedia

See also: Timeline of the Presidency of Barack Obama

The Presidency of Barack Obama began at noon EST on January 20, 2009 when he became the 44th President of the United States, Obama was a United States Senator from Illinois at the time of his victory over Arizona Senator John McCain in the 2008 presidential election. Obama became the first African-American president of the United States.

His policy decisions have addressed a global financial crisis and have included changes in tax policies, legislation to reform the United States health care industry, foreign policy Initiatives and the phasing out of detention of prisoners at the Guantanamo Bay detention camp in Cuba. He attended the G-20 London summit and later visited U.S. troops in Iraq. On the tour of various European countries following the G-20 summit, he announced in Prague that he intended to negotiate substantial reduction in the world's nuclear arsenals, en-route to their eventual extinction. In October 2009, Obama was awarded the Nobel Peace Prize for this extraordinary efforts to strengthen international diplomacy and cooperation between peoples."

#### Contents [hide] 1 Transition period 2 Inauguration 3 First 100 days 3.1 Expectations 3.2 Legislation and executive orders 4 Approval ratings and opinion 4.1 2009 4.1.1 Handling of the economy

5 Major legislation 5.1 Legislation signed

5.1.1 2009 5.1.2 2010

5.1.3 2011 6 Personnel

6.1 Cabinet appointees

6.2 Notable non-Cabinet positions 6.3 Judicial nominees

6.3.1 Supreme Court 6.3.2 Other courts 7 Policies

7.1 Economy 7.2 Ethics

7.2.1 Lobbying reform 7.2.2 Transparency 7.3 Foreign policy

7.3.1 Guantánamo Bay detention camp 7.3.2 Overseas Contingency Operation

7.4 Gun control 7.5 Science and technology 7.5.1 Cybersecurity

7.5.3 NASA

Network analysis and visualization for social media

& Log in / create account Q



Read Edit View history Search

44th President of the United States Incumbent Assumed office

January 20, 2009

Vice President los Rideo Preceded by George W. Bush

Born August 4, 1961 (age 49)

Honolulu, Hawali, United States[1] Birth name Barack Husseln Obama

Nationality American

Political party Democratic Michelle Ohama (m.

Spouse(s)

Children Malia Ann is 1998)

Residence

Natasha (Sasha) is The White House Occidental College Columbia University



## Introduction

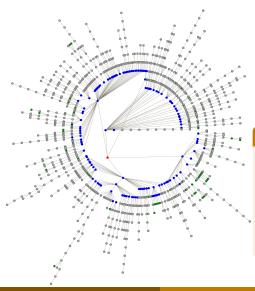
#### Article talk pages



I meant that we should put somewhere in here something about Republicans victories in general. It says something in the BIII Clinton Presidency article about the 94 Republican victories and in the George W. Blush presidency article about the Democratic victories in the 2006 elections. We need to got something in here about the 2010 Republican victories, but where Prolitics/2012 (Cash) 0.0339, 09 November 2010 (UTC) in Classarer. The information in those charge articles have been written at this the processor by a support of the property of the charge articles have been designed to an information in those charged with support of the property of the charged articles have been designed and provided and information in those charged with support of the property of the design of the property o

## **Example Structure**

Discussion tree for article "Presidency of Barack Obama"



- red → root (the article)
- blue → structural nodes
- green → anonymous comments
- grey → registered comments

#### More details in:



D. Laniado, R. Tasso, Y. Volkovich, and A. Kaltenbrunner.

When the Wikipedians talk: Network and tree structure of Wikipedia discussion pages.

In Proc. of ICWSM, 2011



## Interactions of partisan users on article talk pages

## User-boxes $\Rightarrow$ Party assign.

#### **Democrats**

dem

This user supports the U.S. **Democratic Party.** 

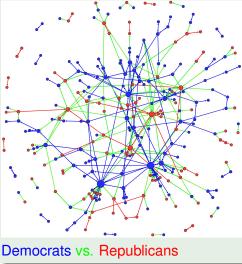
#### Republicans



### Cross-party interactions

- Shuffle test indicates neutral mixing.
- ⇒ no stat. significant preference for neither internor intra-party interaction.

## Interaction Network



## Method

Mixing coefficient r

#### Motivation

- Measures if there exists a preference for relations between users of the same or different characteristics.
- Possible characteristics:
  - Number of relations
  - Sex
  - Age
  - Race
  - Weight
  - Mother tongue
  - ...
- Examples can be found in [Newman 2003].





## Method: Calculate mixing coefficient with reshuffling I

## Data: Pairs of users interacting broken by party.

article discussions	Democrats	Republicans
Democrats	193	94
Republicans	86	57
user wall	Democrats	Republicans
user wall  Democrats	Democrats 395	Republicans 243

### Definition: mixing coefficient

$$r = \frac{\text{Tr}A - ||A^2||}{1 - ||A^2||}$$

where A is a normalised matrix with elements  $a_{ij}$  and  $||A^2||$  is the sum over all  $a_{ij}^2$ 

## Calculate mixing coefficient with reshuffling II

Mixing coefficient r

## Interpretation

- r > 0: assortative mixing
  - There exists a preference for relations between similar users.
  - Users with the same characteristics relate preferentially among themselves and vice versa.
- $r \approx 0$ : neutral mixing
  - There is no preference in the relations.
- r < 0: dissortative mixing
  - There exists a preference for relations among users with different characteristics.
  - For example between users with the opposite ideological views.

## Calculate mixing coefficient with reshuffling III

### To avoid bias due to network topology

E. g. one group of users being more active than the other

## Compare with $r_{rand}$ in reshuffled networks

- keep the users fixed,
  - same party affiliations
  - same numbers of in-coming and out-going links
- randomise the links between them
- generate a sample of 100 networks
- computed the average mixing coefficient  $\hat{r}_{rand}$  of these networks and their standard deviation  $\sigma_{rand}$ .
- calculate Z-score

Z-score = 
$$(r - \hat{r}_{rand})/\sigma_{rand}$$

## Calculate mixing coefficient with reshuffling IV

### Interpretation Z-score

- High positive values of Z indicate assortative mixing
- High negative values indicate dissortative mixing.
- ullet Low absolute values (|Z|<2) correspond to neutral mixing, i.e. no statistically significant preferences [Foster 2010].

#### Results

talk page	r	$\hat{r}_{rand}$	$\sigma_{rand}$	Z-score	significant?
article	0.070	0.0028	0.0505	1.33	no
user	0.095	-0.0053	0.0301	3.33	yes

#### Conclusions

 Wikipedian identity seems to predominate over party identity in article discussions.

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# Analysis of emotions expressed in talk pages

#### Goal:

Study the emotional dimension in a large peer production community

## Research questions: How are the emotional styles of editors ...

- affected by their level of experience?
- affected by their gender and the topics they choose to work on?
- affected by interacting with others (emotional congruence)?
- related to those of the editors they interact more frequently with (emotional homophily)?

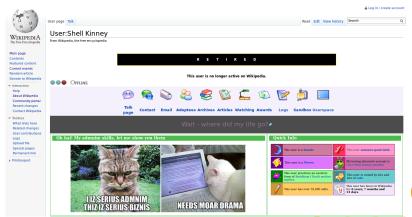
### Results are partly published in



Laniado, D., Castillo, C., Kaltenbrunner, A., and Fuster Morell, M. F. (2012) Emotions and dialogue in a peer-production community: the case of Wikipedia. 8th International Symposium on Wikis and Open Collaboration, WikiSym'12

## User gender labelling

- ullet pprox 12 000 users wrote  $\geq$  100 comments in articles talk pages
- ullet Gender identified through Wikipedia API for pprox 2~000 of them
- Out of the remaining ones, a sample of 1 385 users for manual labelling through crowd-sourcing (Crowdflower)





## User gender labelling

#### Manual labelling

- Gender could be identified only for  $\approx 50\%$  of users:
  - real name or username (50% of those identified)
  - implicitly stated gender (27% of females, 20% of males)
  - pronoun (15% of females, 10% of males)
  - other indicators: userboxes, pictures, links to personal blogs...

	Non-admins	Admins	Total
Males	1 087	1 526	2 613
Females	68	97	165
Unknown	6 850	2 603	9 453
Total	8 005	4 226	12 231

Table: Users with  $\geq 100$  comments by gender and administrator status.

- Category "unknown" includes:
  - 8 708 users that were not included in the crowd-sourced task
  - 745 users whose gender could not be identified by evaluators

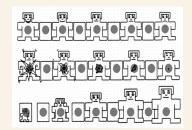


## Measuring the Emotional Content of Discussions

## Affective norms for English words (ANEW)

Rates a list of 1060 frequent words on a 9 point scale in three dimensions:

- Valence
- Arousal
- Dominance



Compare users per word frequency-weighted averages.



Bradley and Lang. (1999).

Affective norms for English words (ANEW) Technical report C-1.

The Center for Research in Psychophysiology, University of Florida, FL.





## Comparing with other word counting measures (I)

## Linguistic Inquiry and Word Count (LIWC)

- Discrete measures of emotions (anger, anxiety and sadness)
- Two scores for basic emotion (compared with ANEW valence)
  - positive valence and
  - negative valence
- by counting the proportion of positive / negative words in a comment
- ANEW assigns emotions scores to each word from the lexicon.



Pennebaker J. Chung C. Ireland M. Gonzales A. Booth R (2010). The development and psychometric properties of LIWC2007. Austin, TX.





## Comparing with other word counting measures (II)

## SentiStrength

- Based on LIWC and developed for short web texts
- Accounts for modes of textual expression specific to the online environment, e.g. emoticons and abbreviations.
- Provides a positive and a negative score for emotional valence.
- Emotion score is calculated at the sentence level (number of positive and negative words).
- Summarised at the comment level as strongest positive and negative emotion expressed in a comment.
- Final scores are averages over comments in a given category.



Thelwall M, Buckley K, Paltoglou G, Cai D, Kappas A (2010)
Sentiment strength detection in short informal text.

Journal of the American Society for Information Science and Technology 61: 2544 – 25



## Measuring the Emotional Content of Discussions

Example for the results of different emotional lexica

Table: Example messages with their corresponding Valence, Arousal, and Dominance(ANEW) or positive & negative scores (*LIWC*, **SentiStrength**).

	ANEW			LIWC		SentiSt.	
	V	Α	D	+	-	+	-
Sounds like a <i>good challenge</i> - to be proven or disproven. I'm <i>happy</i> if it can be shown to go further using closed cubic polynomial solutions. The <i>nice</i> thing about these are that they are <i>pretty</i> easy to test numerically	7.4	5.3	6.2	15	0	3	-2
Seems you have not yet seen female <i>lover</i> after having <u>sex</u> who do not <u>wish</u> to have <u>sex</u> with the same <i>lover</i> any more :) Once you've seen it, you understand very <i>well</i> what <u>war</u> of Venus means compared to <u>war</u> of Mars.	5.5	7.0	5.2	6.8	4.5	4	-3
-in "House (astrology)"							
What about the whirlie hazing, the alcohol <u>abuse</u> , the <u>emotional</u> <u>poverty</u> , the <u>suicide</u> in 1995/6, the biotech plans which were stopped by pitzer <u>protests</u>	1.6	5.8	3.5	4	8	1	-4
-in "Harvey Mudd College"							





## Emotions. Status and Gender

Similar results with different Lexical

#### **Emotions and Status**

- Admins express, on average, more positive emotion (p < 0.001).
- Admins also express less negative emotion (p < 0.001).
- Non-admins express more affect, in particular, more anxiety, anger and sadness (all with p < 0.001) compared to admins.

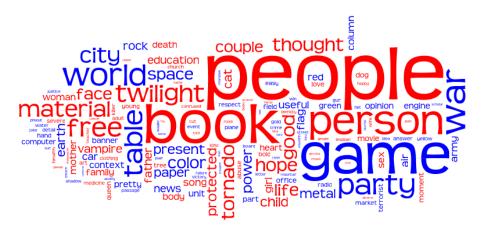
#### Emotions, Status and Gender

- Significant difference between male admins and non-admins (p < 0.001)
- No significant difference between female admins and non-admins.





## Emotions and gender



- ANEW Words more used by females and males
- Size accounts for difference in frequency





## Topics, emotions and gender

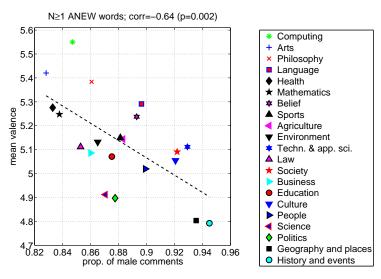


Figure: Mean valence for discussions of articles in different topic categories vs the proportion of comments written by male editors

## Emotional congruence

Replies are more positive

## On average, editors tend to reply with:

- higher valence: +0.05 (p < 0.01)
- higher dominance: +0.04 (p < 0.01)
- no statistically significant differences for arousal

 Users tend to be more positive and dominant when replying, but without recurring to words evoking stronger sentiments.

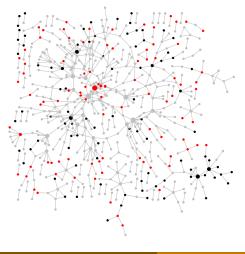




## **Emotional homophily**

Mixing patterns: do users interact preferentially with similar users?

 Assortative by emotional style: users interact more with others expressing similar emotions.



- edges connecting users who have exchanged at least 10 replies
- red nodes → 15% users expressing higher valence in article discussions
- black nodes → 15% users expressing lower valence in article discussions
- size → proportional to the number of connections





## Homophily: detailed results

Norma	alised	r	$r_{rand}$	$\sigma_{rand}$	Z
valence	(sent)	0.0269	-0.0003	0.0011	23.8
	(received)	0.0109	-0.0004	0.0010	10.8
arousal	(sent)	0.0253	-0.0004	0.0009	28.2
	(received)	0.0187	0.0013	0.0012	14.8
dominance	(sent)	0.0380	-0.0001	0.0015	26.2
	(received)	0.0121	9.8e-08	0.0011	10.8

	r	$r_{rand}$	$\sigma_{rand}$	Z
gender	0.0443	-0.0008	0.0059	7.63
#comments written	-0.0177	-0.0014	0.0017	-9.51
#replies received	-0.0060	-0.0013	0.0014	-3.50
#replied users	-0.0340	-0.0023	0.0020	-16.23
#replying users	-0.0237	-0.0014	0.0015	-14.35
#discussed articles	-0.0009	-0.0011	0.0014	0.12

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## Motivation

Distance and friendship

## online tools and long-distance travel ⇒ death of distance?

- individuals try to minimise the efforts to maintain a friendship by interacting more with their spatial neighbours.
- probability of a social interaction quickly decays as an inverse power of the relative geographic distance [Stewart 1941].
- probability of connections between two individuals on online social networking services still decreases with their geographic distance [Backstrom 2010, Liben-Nowell 2005].

### Results published in



A. Kaltenbrunner, S. Scellato, Y. Volkovich, D. Laniado, D. Currie, E. J. Jutemar & C. Mascolo.

Far from the eyes, close on the Web: impact of geographic distance on online social interactions.

In Proceedings of ACM SIGCOMM Workshop on Online Social Networks (WOSN '12). ACM, 201:



Y. Volkovich, S. Scellato, D. Laniado, C. Mascolo & A. Kaltenbrunner.

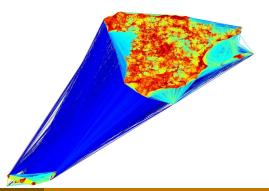
The length of bridge ties: structural and geographic properties of online social interactions.

In ICWSM-12 - 6th International AAAI Conference on Weblogs and Social Media. The AAAI Press, 2012

#### **Dataset from Tuenti**

## "Spanish Facebook", a Spain-based social networking website

- 1174 million friendship links
- ~500 directed messages exchanges during 3 months;

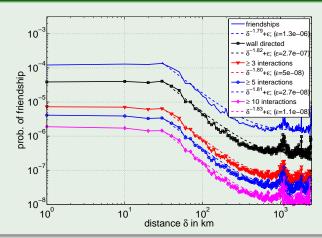




# Geographic properties

The effect of distance on friendship

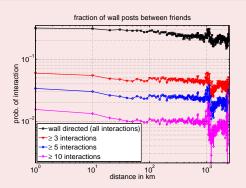
## Probability of connection as function of geographic distance



# Interaction Analysis

Interactions and distance

## Probability of message exchange between friends



 high-intensity communication takes place on social connections regardless of their geographic distance

## Conclusions

### The effect of geographic distance on online social interactions

- Spatial proximity greatly affects how users establish their connections on online social platforms.
- Social interactions are only weakly affected by distance.
- Geography affects whom we interact with, however it does not influence how much we interact.

## **Applications**

- link prediction,
- tie strength modelling,
- user profiling.





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# Analysis of institutional (sister city) relations







#### Introduction

Analysis of institutional (sister city) relations

#### Sister cities

- Institutional partnership between two cities or towns with the aim of cultural and economical exchange.
- These relations had never been analysed before.

#### We want to understand ...

- social
- geographical
- economic mechanisms

of city pairings.



#### Results published in



Andreas Kaltenbrunner, Pablo Aragón, David Laniado & Yana Volkovich.

Not All Paths Lead to Rome: Analysing the Network of Sister Cities.

In Self-Organizing Systems, Lecture Notes in Computer Science, vol. 82

## Example for Wikipedia article used for data extraction

#### List of twin towns and sister cities in Spain

From Wikipedia, the free encyclopedia

This is a list of places in Spain having standing links to local communities in other countries. In most cases, the association, especially when formalised by local government, is known as "town twinning" (though other terms, such as "partner towns" or "sister cities" are sometimes used instead), and while most of the places included are towns, the list also comprises villages, cities, districts, counties, etc. with similar links.

#### A Coruña

- Brest, Brittany, France
- 🚳 🞮 Recife, Pernambuco, Brazil

#### Antequera

Agde, Hérault, France

#### Almansa

. III Lymington, United Kingdom

#### Alcalá de Henares

Alba Iulia, Romania

• III Peterborough, United Kingdom

Altea

Altea is a founding member of the Douzelage, a town twinning association of 23 towns across the European Union. This active town twinning began in 1991 and there are regular events, such as a produce market from each of the other countries and festivals.[1][2]

Karkkila, Finland<sup>[2]</sup>

Kőszeg, Hungary<sup>[2]</sup>

Marsaskala, Malta<sup>[2]</sup>

Sweden<sup>[2]</sup>

Prienai, Lithuania<sup>[2]</sup>

Meerssen, the Netherlands<sup>[2]</sup>

Niederanven, Luxembourg<sup>[2]</sup>

- Bad Kötzting, Germany<sup>[2]</sup>
- Bellagio, Italy<sup>[2]</sup> • Bundoran, Republic of Ireland[2]
- Choina, Poland<sup>[2]</sup>
- Granville, France<sup>[2]</sup> Holstebro, Denmark<sup>[2]</sup>
- Houffalize, Belgium<sup>[2]</sup> Judenburg, Austria<sup>[2]</sup>
- Ames

Sahara Desert

Archena [edit]

Ell Chesham, United Kingdom (1995) [3]



[edit]

[edit]



- E Preveza, Greece<sup>[2]</sup>
- Sesimbra, Portugal<sup>(2)</sup>
- Sherborne, United Kingdom<sup>[2]</sup> Sigulda, Latvia<sup>[2]</sup>
- Sušice, Czech Republic<sup>[2]</sup> Türi, Estonia<sup>[2]</sup>
- Zvolen, Slovakia<sup>[2]</sup>



# Dataset extracted from the English Wikipedia

### Data extraction process

- automated parser and a manual cleaning process.
- Google Maps API to geo-locate cities.

#### Size of the dataset

network	N	K	$\langle C \rangle$	% GC	$\langle d \rangle$
city network	11 618	15 225	0.11	61.35%	6.74
country network	207	2933	0.43	100%	2.12

#### Disclaimer

- No central register.
- User generated data (only 30% of reciprocal connections).
- No guarantee that the dataset is complete.

# Top 20 cities and countries ranked by degree

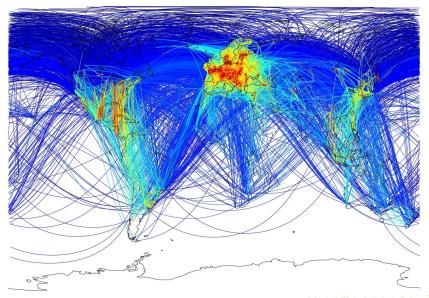
rank by betweenness centrality in parenthesis

No	city	deg.	betw.	country	w. deg.	betw.
1	Saint Petersburg	78	(1)	USA	4520	(1)
2	Shanghai	75	(4)	France	3313	(3)
3	Istanbul	69	(12)	Germany	2778	(6)
4	Kiev	63	(5)	UK	2318	(2)
5	Caracas	59	(23)	Russia	1487	(9)
6	Buenos Aires	58	(36)	Poland	1144	(33)
7	Beijing	57	(124)	Japan	1131	(20)
8	São Paulo	55	(24)	Italy	1126	(7)
9	Suzhou	54	(6)	China	1076	(4)
10	Taipei	53	(20)	Ukraine	946	(27)
11	Izmir	52	(3)	Sweden	684	(14)
12	Bethlehem	50	(2)	Norway	608	(22)
13	Moscow	49	(16)	Spain	587	(11)
14	Odessa	46	(8)	Finland	584	(35)
15	Malchow	46	(17)	Brazil	523	(13)
16	Guadalajara	44	(9)	Mexico	492	(21)
17	Vilnius	44	(14)	Canada	476	(28)
18	Rio de Janeiro	44	(29)	Romania	472	(32)
19	Madrid	40	(203)	Belgium	464	(23)
20	Barcelona	39	(60)	the Netherlands	461	(16)



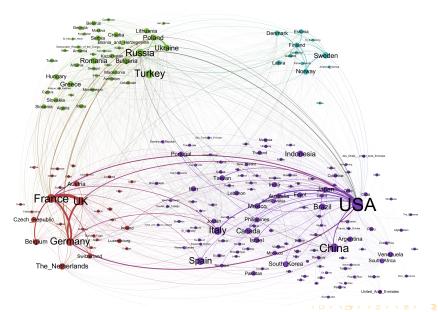


# Sister city relations





# Clustering of relations aggregated by country





# **Assortativity**

#### Method

- Compare sister city network and 100 randomised equivalents.
- Calculate assortativity measure based on the Z-score

### Degree assortativity by city

 Cities with many connections tend to be connected with cities with many connections and vice-versa.

#### Relations are assortative by country

- Gross Domestic Product per capita
- Human Development Index
- Political Stability Index





# Assortativity II

#### **Details**

property	r	$r_{rand}$	$\sigma_{rand}$	Z
city degree	0.3407	-0.0037	0.0076	45.52
Gross Domestic Product (GDP) <sup>a</sup>	0.0126	-0.0005	0.0087	1.51
GDP per capita <sup>b</sup>	0.0777	0.0005	0.0078	9.86
Human Development Index (HDI) <sup>c</sup>	0.0630	-0.0004	0.0075	8.46
Political Stability Index <sup>d</sup>	0.0626	0.0004	0.0090	6.94

<sup>&</sup>lt;sup>a</sup>Source

http://en.wikipedia.org/wiki/List\_of\_countries\_by\_GDP\_(nominal)

bSource: http://en.wikipedia.org/wiki/List\_of\_countries\_by\_GDP\_ (nominal)\_per\_capita

cSource: http://en.wikipedia.org/wiki/List\_of\_countries\_by\_ Human\_Development\_Index

<sup>d</sup>Source: http:

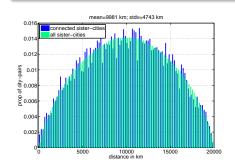
//viewswire.eiu.com/site\_info.asp?info\_name=social\_unrest\_table

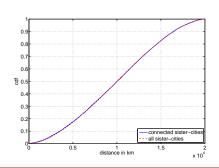


### Distances between sister cities

### Comparison of distances between two pairs of ...

- connected sister cities
- random (not necessarily connected) cities





#### First evidence for the Death of Distance

Nearly no differences between the two distributions.

### Conclusions & Future work

#### Conclusions

- Assortative mixing with respect to degree, economic and political country indexes.
- Sister city relationships reflect country predilections in and between cultural clusters.
- Geographic distance between cities does not influence city pairing.

#### Future work

- Combined analysis with networks of air traffic or good exchange.
- Analysis of network evolution (needs other data-sources).





## Outline

- Political User interaction on Twitter
- Political Affiliation on Wikipedia
- Emotional styles on Wikipedia
- Geographical distance and Friendship
- Sister Cities
- 6 Links between biographies on Wikipedia





#### Motivation

## Wikipedia as global collective memory place allows ...

- to extract from biographies how social links are recorded ...
- to generate networks of links between biographical articles.

### Research questions

- Who are the most central characters in these networks?
- Do culture related peculiarities exist?
- Which cultures are more similar?
- What is the shared knowledge about connections between persons across cultures?

### Results published in



P. Aragón, A. Kaltenbrunner, D. Laniado & Y. Volkovich.

Biographical Social Networks on Wikipedia - A cross-cultural study of links that made

history.

In Proc. of the 8th Int. Symp. on Wikis and Open Collaboration (WikiSym'12), ACM, 201

## Data extraction

Building biographical networks for 15 language editions of Wikipedia

- Selected the 15 largest language editions of Wikipedias
- Starting point: 296 511 biographies from the English Wikipedia (from DBpedia)
- Identified the corresponding articles (when existing) on the remaining 14 languages
- Generated a directed network for each language version:
  - nodes → persons
  - $\bullet$  edges  $\to$  links between the articles of the corresponding persons
- Manage alternative titles of articles: track redirects
- Data collected through Wikipedia APIs between September 8th and 13th, 2011





## Most central persons in the English Wikipedia

sorted by in-degree. Ranks for out-degree, betweenness and PageRank in parenthesis

person	in-degree	out-	degree	betw.	PageR	ank
George W. Bush	2123	89	(107)	(1)	0.00209	(1)
Barack Obama	1677	51	(710)	(8)	0.00162	(2)
Bill Clinton	1660	74	(205)	(4)	0.00156	(4)
Ronald Reagan	1652	90	(103)	(2)	0.00156	(3)
Adolf Hitler	1407	119	(26)	(3)	0.00149	(5)
Richard Nixon	1299	86	(127)	(7)	0.00136	(6)
William Shakespeare	1229	25	(4203)	(63)	0.00113	(9)
John F. Kennedy	1208	104	(53)	(5)	0.00123	(8)
Franklin D. Roosevelt	1052	71	(237)	(15)	0.00131	(7)
Lyndon B. Johnson	1000	106	(50)	(12)	0.00108	(11)
Jimmy Carter	953	80	(158)	(9)	0.00113	(10)
Elvis Presley	948	82	(142)	(27)	0.00063	(24)
Pope John Paul II	941	59	(444)	(11)	0.00083	(18)
Dwight D. Eisenhower	891	55	(564)	(22)	0.00095	(14)
Frank Sinatra	882	108	(47)	(18)	0.00056	(28)
George H. W. Bush	878	87	(118)	(19)	0.00096	(13)
Abraham Lincoln	846	54	(593)	(40)	0.00089	(16)
Bob Dylan	835	151	(11)	(14)	0.00055	(30)
Winston Churchill	748	84	(136)	(10)	0.00092	(15)
Harry S. Truman	743	81	(145)	(24)	0.00099	(12)
Joseph Stalin	723	69	(265)	(43)	0.00089	(17)
Michael Jackson	663	71	(237)	(34)	0.00042	(51)
Elizabeth II	653	52	(665)	(6)	0.00074	(19)
Jesus	572	38	(1595)	(51)	0.00068	(20)
Hillary Rodham Clinton	554	87	(118)	(32)	0.00063	(25)





# Most central persons in different language Wikipedias

Top 5 most central persons for each language by betweenness

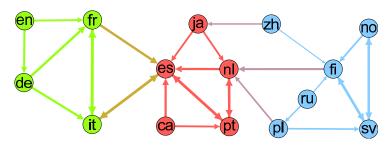
lang	#1	#2	#3	#4	#5
en	George W. Bush	Ronald Reagan	Adolf Hitler	Bill Clinton	John F. Kennedy
de	Adolf Hitler	George W. Bush	Martin Luther King, Jr	Barack Obama	Frank Sinatra
fr	Adolf Hitler	George W. Bush	William Shakespeare	Barack Obama	Jacques Chirac
it	Frank Sinatra	George W. Bush	Pope John Paul II	Michael Jackson	Elton John
es	Michael Jackson	Fidel Castro	William Shakespeare	Che Guevara	Adolf Hitler
ja	Adolf Hitler	Michael Jackson	Ronald Reagan	Yukio Mishima	Barack Obama
nl	Elvis Presley	Adolf Hitler	Bill Clinton	Joseph Stalin	William Shakespeare
pt	Michael Jackson	Richard Wagner	Adolf Hitler	Ronald Reagan	David Bowie
sv	George W. Bush	Winston Churchill	Elizabeth II	Michael Jackson	Adolf Hitler
pl	Elizabeth II	Pope John Paul II	Margaret Thatcher	George W. Bush	Ronald Reagan
fi	Barack Obama	Adolf Hitler	Michael Jackson	George W. Bush	Benito Mussolini
no	Marilyn Monroe	Adolf Hitler	John F. Kennedy	Bob Dylan	Bill Clinton
ru	William Shakespeare	Napoleon II	Kenneth Branagh	Elton John	Joseph Stalin
zh	Chiang Kai-Shek	William Shakespeare	Barack Obama	Deng Xiaoping	Adolf Hitler
ca	Adolf Hitler	Che Guevara	Juan Carlos I	Michael Schumacher	Juan Manuel Fangio

## Most are known to be (or have been) highly influential

- We find political leaders, revolutionaries, famous musicians, writers and actors.
- Hitler, Bush, Obama dominate in almost all top rankings.
- Top ranked in many languages reflect country peculiarities.

# Languages similarity network

Every language links to the two most similar ones according to Jaccard coefficient



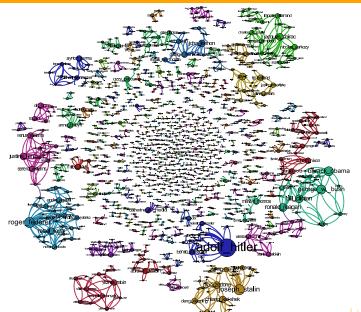
#### Definition of Jaccard coefficient J

Given the set of links A and B of two networks

$$J = \frac{|A \cap B|}{|A \cup B|}$$

• *J* is the ratio between the number of links present in both networks (their intersection) and the number of links existing in their union.

# Intersection of networks in different languages





## Conclusions and future work

#### Conclusions

- Global social network measures are largely similar for all networks.
- Most central persons unveil interesting peculiarities about the language communities.
- Networks are more similar for geographically or linguistically closer communities.
- Many connections which can be found in most of the analysed language Wikipedias.

#### Future work

- Application of the methodology to generate subnetworks of other kinds of article categories
- Consider all biographies for each language.
- Analyse links missing only in a few language Wikipedias.

## Questions?







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