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HEALTH RESEARCH INSTITUTE

eHealth 2011

21-23 November 2011 - Málaga, Spain

4th ICST International Conference on eHealth

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Social Media: A systematic review to understand the evidence and application in infodemiology

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OVERVIEW . . .

1. Background
2. Methods
3. Results
4. Discussion/Conclusions
5. Question period



BACKGROUND

The research institute of London Health Sciences Centre
and St. Joseph's Health Care, London.



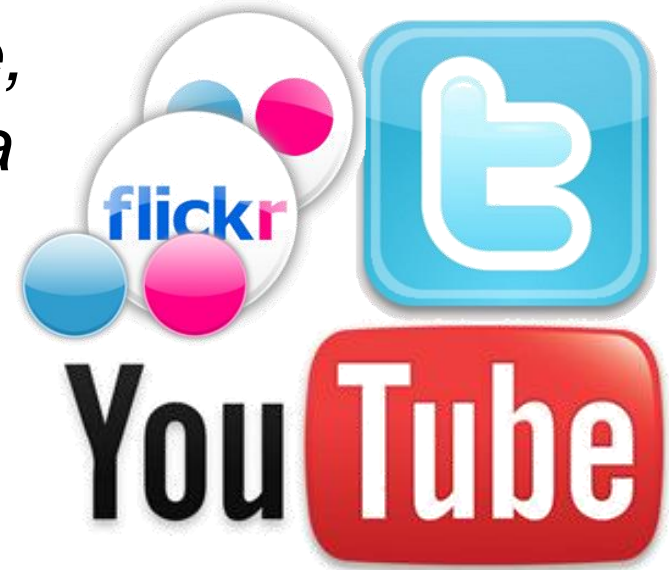
Social Media:

- # of users skyrocketing
 - **Why?**
 - Sense of community
 - Accessible (time/place)
 - Cost-effective
 - Interoperable
- *Open-source social media presents excellent opportunities in health research*



Infoveillance:

- *Systematically mine, aggregate, & analyze real-time, online data*
- Enables instant access to information around the world
- Provides insight into abnormal disease patterns



Relevance:

- ***Acute:***
 - Detect potential outbreaks
 - Allow for appropriate resource allocation
- ***Chronic:***
 - Improving health care delivery planning
 - Drawing awareness: public and HCPs
 - Address gaps: treatment, care

Objective:

To illustrate how data generated through social media can be used to inform planning and implementation of strategies to predict and address disease outbreaks.

METHODS

Data Sources

Databases:



embase[®]
BIOMEDICAL ANSWERS



PubMed

Limits:

- English
- Journal articles and conference paper/proceedings
- Published from 1999-2011

Keywords and MeSH Headings:

	Disease	Medium	Methodology
Keywords	<ul style="list-style-type: none"> • Early detect\$ • Pandemic\$ • Epidemic\$ • Communicable disease\$ • Early diagnosis 	<ul style="list-style-type: none"> • Information technol\$ • Internet • Mass medium • Medical computing • Social media • Social network\$ • Geolocation 	<ul style="list-style-type: none"> • Disease surveillance • Monitor\$ • Disease control • Algorithm\$ • Data min\$ • Query process\$ • Information retriev\$
MeSH	<ul style="list-style-type: none"> • Pandemics • Communicable diseases • Disease outbreaks • Early diagnosis 	<ul style="list-style-type: none"> • Communications media • Databases • Factual • Internet* • User-computer interface* 	<ul style="list-style-type: none"> • Population surveillance • Information storage and retrieval • Forecasting • Data mining • Sentinel surveillance

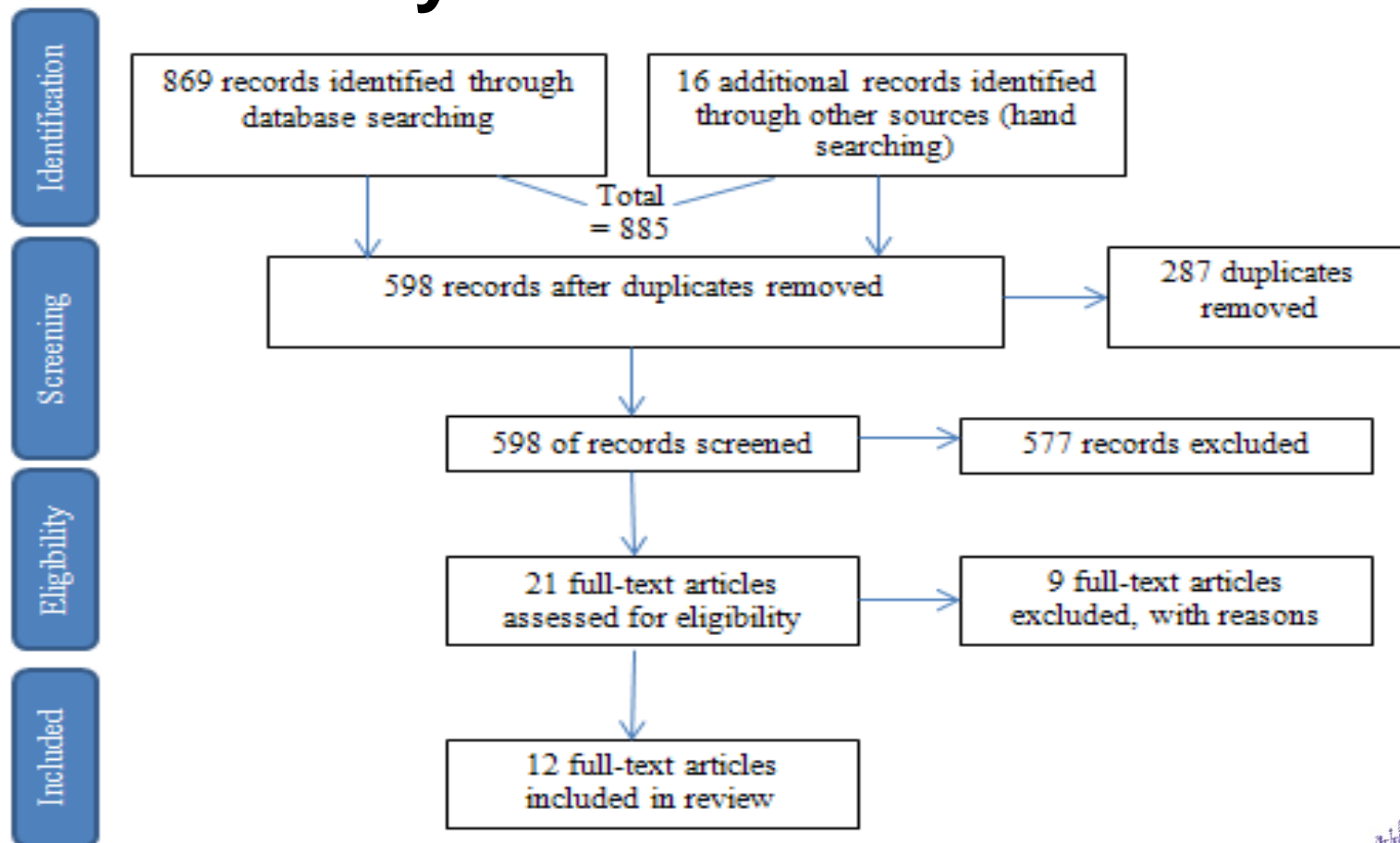
Search Strategy

- Literature search conducted June-11
- Searched articles on 3 categories (*Disease, Medium, Methodology*)
 - Keyword in which article was indexed
 - Title and abstract of article
- Search terms chosen to reflect our objective
 - Conducted generalized search in Google to further inform search term selection



RESULTS

Study Selection Process:

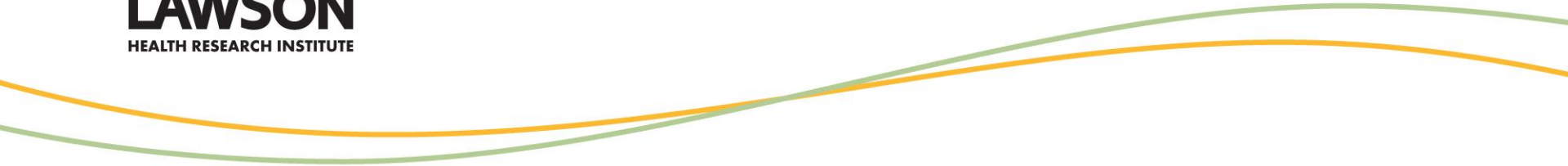


Studies Included:

Author	Date Published	Country	Methods	Summary of Findings
Achrekar et al.	2011	US	SNEFT – Aggregated anonymous data stored in repositories; web crawler	4.7 million tweets High correlation between tweets and CDC data ($r = 0.984$)
Chen et al.	2010	US	SNEFT – Aggregated anonymous data stored in data repositories; web crawler	2 million tweets Tweet patterns influenced by media Very little misinformation found in tweets
Chew & Eysenbach	2010	Canada	Infovigil; mine & analyze tweets; over 8 mos	2 million tweets Tweet patterns influenced by media Very little misinformation found in tweets
Eysenbach	2011	Canada	Infovigil; mine & analyze tweets; over 8 mos	2 million tweets Tweet patterns influenced by media Very little misinformation found in tweets
Corley et al.	2009	US	Evaluated blog posts; over 3 mos	44 million tweets Sig. Correlation between posts and CDC data ($r = 0.626$)
Corley et al.	2010 a, b	US	Evaluated weblogs, microblogs, social media items; English	High correlation between posts and CDC data ($r = 0.626$)
Culotta et al.	2010	US	Developed regression model; evaluated tweets over 10 weeks	Multiple regression model out-performs simple regression ($r = 0.78$)
de Quincey & Kostkova	2010	UK	Code developed to retrieve tweets; over 1 wk	135, 428 tweets Analyzed content of tweets for swine flu
Lamos & Christianini	2010	UK	Developed monitoring tool; ILI; Flue-score created	160,000 tweets/day High correlation between flu-scores & weekly HPA reports Independent of language Geo-located
Lopes et al.	2009	Portugal	Developed web-crawler: Data Collector; H1N1	Collected tweets from 6 EU countries 700 pertaining to H1N1
Signorini et al.	2011	US	Track tweets; H1N1; over 34 days; plot spikes of public interest	951,697 tweets – 34 days 4 million tweets – 3 mos No sustained interest in vaccine-related issues

Summary of Results:

- All studies mine & analyze
- 2009 onwards
- Design
- ILI & H1N1
- Medium
- High correlations between posts and national data



DISCUSSION / CONCLUSIONS

Summary: Effect of infodemiology...

Advantages	Disadvantages
➤ Provides real-time, relevant info	➤ Textual data can be hard to interpret
➤ Improves public access to health info	➤ Data may not reflect entire population
➤ Operating costs can be low	➤ Geo-location of data hard to confirm
➤ Opportunity for “Mash-ups”	

Study Limitations:

- Diffuse search
- Depended on hand-searching

What's next?

- Identify target audiences
- Determine user interactions
- Role of evidence-based medicine tools



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Questions?

