



Encountering #feminism on Twitter

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Structure of this presentation: Focus on Methodology

- Aims of the project (Viv)
- Methodology:
 - ■Background (Viv)
 - ■What we did, how we did it, limitations (Daniel)
- Findings (Daniel)
- Future work (Viv)
- Questions



Aims

- Began with a realisation and a claim...
- ◆ But what is feminism? Does the idea of a '4th wave' (or any wave) make sense? How is feminism being expressed, and by whom? And what is the reaction to this? Led to another question -
- What can an analysis of social media (in this case, use of Twitter) tell us about feminism today?
- A series of sociological questions but using AI techniques to explore them



Methodology: Background

- ◆ Two cultures: social work academics with an interest in feminism & discourse analysis, and researchers from a social media analysis team.
- Funding from SSPS research grant
- A challenging enterprise How to analyse conversations, at scale?
- ◆ Building cultural bridges How do sociologists & data scientists communicate with one another when we have little shared language? We've learnt a lot about what we can and <u>cannot</u> answer...



Methodology 1: The Data

- ◆ Tweets between Dec 2012 and May 2014, using 1% Twitter 'sprinkler' data: 1 million+ messages a day
- Filtered the data against a list of topic keywords led to a dataset of 500,000 messages







Methodology 2: Analysis

- ♦ How can we code a data-set of ½ million messages?
- ◆ Tweets were coded ('tagged'):
 - 1. We designed tagsets around the project's reseach questions: topic, position, message-type, tone
 - 2. Each of our 5 research team tagged 100 tweets.
 - 3. Discussed and refined the tagsets
 - 4. Final tagged data: 800 position tags, 1,000 topic tags. Just 0.02% of the data!
 - 5. AI classifier trained on the tagged tweets.
 - 6. Pattern of AI errors measured on the tagged data
 - 7. AI classifier auto-tagged all the remaining tweets
 - 8. Error correction applied



Tagging Process

- Our initial tagsets included:
 - Feminist: 1st wave / 2nd wave / 3rd wave / 4th wave
 - Tone: angry / calm / joking
- We iterated adjusting where things weren't working
- Unsuccessful:
 - "Rare" topics, e.g. children
 - Tone was too subjective to judge.
 - 1st wave, 2nd wave: could not assess from a tweet.
 - Message type: mostly inform-explain-proclaim.
- Successful:
 - The main topics: rape, abortion, violence
 - Pro / anti feminism



Limitations/issues

- How representative is the data?
- Coding tagging is difficult & time-consuming
- ◆ AI for individual messages is not very accurate
 - but we can use error correction.



Findings

- Pro and anti feminism camps can be distinguished with reasonable accuracy – confirming that debate on Twitter is often polarised
- Feminism on Twitter does not easily fit with the waves of feminism model.
- Rape is the dominant topic, followed by abortion.
 - Geo-political events mostly did not affect the topics (notable exception: Wendy Davis).
- High-level of casual mysogyny: e.g. tweets about rape made up about 70% of our data, mostly casual use.
 - The proportion is largely constant over time
 - What does this tell us?



Future work: This Project

- Trace specific topics/campaigns & their critiques over time: e.g. '#endfathersday' and '#yourslipisshowing'
- How could we improve the method used?
 - More tagging needed for really robust results:
 - More training data improves AI accuracy
 - Provide a larger sample size for error correction
- More work to ensure consistency of tagging:
 - Were we each tagging in the same way?
 - Writing fuller descriptions of each tag might help
- Interestingly, these issues of validity & reliability in coding are shared with other research projects.



Future work: Methodology

- Streamline the development of good tagging
 - A 'best practice' guide for big-data tagsets on the basis of our experience?
- ◆ Improve software support for academic research
 - Report confidence intervals
 - Measure team tagging consistency
- ◆ Shorten the research loop: high-level question → precise question → data analysis → interpretation → further questions → ...
 - The "research loop" with social media data is already fast, compared with interviews or surveys.
 - But it could be faster still.



QUESTIONS?

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