BEYOND HASHTAGS: COLLECTING AND ANALYSING CONVERSATIONS ON TWITTER

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Abstract

In this paper we examine a series of techniques to enhance the collection and analysis of conversations on Twitter. We start from the position of seeking to understand how ordinary discussions about particular issues or controversies are unfolding on the social media platform. A key limitation of studies of topics on Twitter that rely on searching for a keyword or hashtag is that they may miss important sections of conversations around issues that do not match the keywords selected (Rambukkana, 2015, Bruns & Burgess, 2015). The Tracking Infrastructure for Social Media Analysis (TrISMA) (Bruns, Burgess & Banks et al., 2016) captures tweets of 2.8m Australian users on a continuing basis providing a comprehensive dataset that we can use to find reply chains which do not include the hashtag or keyword we are studying. Many existing methods of exploring Twitter data do not present conversation chains in a linked format (Bruns, 2012, Grant, Moon, & Busby Grant, 2010); we investigate how using network visualization might help researchers better understand the qualitative content and context of conversations.

Keywords: Twitter, social media, big data, issue mapping, Uber

Introduction

In this paper, we examine a series of techniques to enhance the collection and analysis of conversations on Twitter. We start from the position of seeking to understand how ordinary discussions about particular issues or controversies are unfolding on the social media platform. A key limitation of studies of topics on Twitter that rely on searching for a keyword or hashtag is that they may miss important sections of conversations around
issues that do not match the keywords selected (Rambukkana, 2015, Bruns & Burgess, 2015). The Tracking Infrastructure for Social Media Analysis (TrISMA) (Bruns, Burgess & Banks et al., 2016) captures tweets of 2.8m Australian users on a continuing basis providing a dataset that we can use to find reply chains which do not include the hashtag or keyword we are studying.

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Identifying tweets

We develop a similar method to Lorentzen and Nolin’s (2015) approach to collect a broader set of conversations on Twitter. Using the TrISMA dataset, we first identify all tweets that match a given keyword, and then also identify all tweets that respond to or were responded to by such keyword tweets. We trace these linked conversations recursively, to extract the full chains of conversations in which a search keyword has been mentioned at least once. Where Lorentzen and Nolin (2015) were limited by considering conversations based on the Twitter in-reply-to chains in a single direction from their hashtag tweets or their 5,000 most active participants, we are able to recursively move both ways along the reply chain, including tweets that are replies to tweets we have identified. This is not possible using the Twitter API. Our study is limited by the boundaries of the Australian Twittersphere as defined by TrISMA; conversation chains will be broken if they extend to accounts outside the TrISMA dataset, even if subsequent replies occur inside it.

Analysing reply chains

This paper will present the results of our initial case study of Australian controversies over the legitimization of the Uber ride-sharing service in Australia. We seek to collect all tweets that form part of the conversation around the keyword ‘uber’ in the Australian twittersphere since 2013, and zoom in from there to analyse specific controversies. We conduct a comparative study of two datasets – an original set that contains only tweets that match a keyword, and an expanded set that includes all conversations that reference those initial tweets. The first dataset – tweets matching the keyword ‘uber’ from January 2013 through to November 2015, comprises 176,941 tweets. The second dataset, which also includes tweets we can identify as part of the conversation around the first (using in_reply_to_tweet_id), comprises an additional 30,210 tweets.

The conversation chains are visualised as a directed network and a range of network metrics are calculated, including those used by Lorentzen and Nolin (2015), to allow comparison with their results.

We then apply issue mapping to social media analysis (Marres, 2015; Marres & Moats, 2015) to evaluate the advantages and limitations of these methods for extracting conversations from Twitter data. While work on issue mapping has largely used web data to trace the relationships among actors, themes and objects in the discussion of
matters of concern (Rogers & Marres, 2000; Severo & Venturini, 2015), we will use social media data to study how issue publics emerge through their engagement with specific controversies or themes surrounding Uber. Issue mapping using digital methods and informed by controversy analysis provides a useful methodology to account for popular and everyday modes of online participation in matters of concern (Burgess & Matamoros-Fernandez, 2016).

Our initial results show that the conversations we collected differ significantly from the tweets that match our keywords. From a preliminary textual analysis of small samples of tweets, we note that the tweets that contain the keyword ‘Uber’ seem to be reflect themes and frames dominated by the tech media, lobbyists and campaigners while the tweets that are found through our reply chain approach appear to be much more diverse. They include debates explicitly about the controversy by established players in the issue. Importantly, they also include the much less readily visible ordinary voices of individuals in conversations that are not highly politicised.

We seek to investigate the hypothesis that network metrics and visualisations of conversations may assist in mapping the actors, themes and objects involved in Twitter discussions of matters of concern. We will seek to understand how including conversations that do not explicitly mention ‘uber’ or related keywords may be able to add a richness to the data that was previously invisible. We suggest that this method appears to enable richer analysis of ordinary voices around a controversy. Not only it highlights some additional issues as the controversy plays out, but it also appears to surface new actors who do not necessarily seek to participate in visible public debates by using a common hashtag. We think that these discussions are of great value in identifying new issues and actors that help understand public communication on social media. Not only does this method highlight some additional issues as the controversy plays out, but it also appears to surface a greater proportion of everyday discussions by ordinary users who do not necessarily seek to participate in visible public debates by using a common hashtag. We think that these discussions will often be of very great value in understanding ordinary discussions around controversial issues.

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References


