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# Hashtags Functions in the Protests Across Brazil

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#### **Abstract**

In this article, we discuss the communicative functions of hashtags during a period of major social protests in Brazil. Drawing from a theoretical background of the use of Twitter and hashtags in protests and the functions of language, we extracted a sample of 46,090 hashtags from 2,321,249 tweets related to Brazilian protests in June 2013. We analyzed the hashtags through content analysis, focusing on functions, and co-occurrences. We also qualitatively analyzed a group of 500 most retweeted tweets to understand the users' tagging behavior. Our results show how users appropriate tags to accomplish different effects on the narrative of the protests.

## **Keywords**

hashtags, functions, Twitter, protests, content analysis

### Introduction

During June 2013, Brazil was swept by one of the biggest wave of protests since the end of the dictatorship period in 1984 (Singer, 2013). These protests started as manifestations against the rise of the bus fare in Porto Alegre and Sao Paulo, organized by the Passe Livre (Free Fare) Movement (Saad-Filho, 2013). They were, however, sparse until June 13, when they gained momentum due to reports of police brutality in Sao Paulo. These reports, which showed injured protesters and journalists, triggered a substantial increase in media coverage. The images of this police brutality quickly spread through social media, leading to new and larger protests. On June 14 and 20, more than 2 million people gathered in the streets of more than 200 cities in Brazil, around several demands such as the end of corruption, the suspension of FIFA's Confederation and the World Cup, better public transportation system, and the end of law projects (Saad-Filho, 2013; Singer, 2013).

During these protests, social media and specially Twitter played a key role by allowing protesters to organize and spread their own narratives. Twitter also provided a platform to mobilize other users from different locations through hashtags (Bastos, Recuero, & Zago, 2014), helping spread the word from the people on the streets to social media audiences and often back again to the streets. However, Twitter's role in the mobilization of the people and the spread of the protests in the country is still unclear. In this article, we aim to investigate a part of this role, focusing on the narratives

created and changed by the usage of hashtags during the events. We aim to explore the roles played by different Twitter hashtags and their combination. Our objectives are directed to understand (a) the types and communicative functions of the hashtags used during the protests, (b) how co-occurrences of hashtags depict different meanings and functions, and (c) the tagging behavior of users as the events unfolded. To address these objectives, we collected more than 2 million tweets related to the protests using a set of 35 hashtags and keywords connected to the Brazilian protests. The data were analyzed using a combination of quantitative and qualitative methods, from content analysis to qualitative observation, as we will further explain.

## **Theoretical Background**

Several works have discussed the role of social media in helping organizing protest movements and reporting protests around the world. Some researchers have shown that social media provides new ways of social organization and allows for new forms of protesting activity (Castells, 2012; Malini

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& Antoun, 2013), particularly because social media lacks traditional mediators (Gutierrez, 2013) and enables new forms of participation (Penney & Dadas, 2014), creating "new" decentralized and emergent movements (Castells, 2012). Toret (2012) argued that such movements have hybrid characteristics that turn personal issues into collective political processes, aggregating different political motivations into large social movements, which is similar to what happened in Brazil, where demands were not clear and the reasons why people were in the street were plural (Saad-Filho, 2013).

Segerberg and Bennett (2011) argued that globalization is partially responsible for these characteristics of what they called "personalization of collective action." For the authors, social movements became more individualistic in nature, where people engage themselves in causes related to their own lifestyles, personalizing politics. This phenomena is directly connected to the choice of communication strategies (such as Twitter) and the decentralized nature of contemporary protests (Castells, 2012), as well as the stronger role of personal networks in these events (Segerberg & Bennett, 2011). These characteristics are also related to the reduced presence and role of conventional political organizations (parties, unions, etc.) because digital communication platforms play this part (Segerberg & Bennett, 2012).

The role of social media as an alternative source of information, acting in a peer-to-peer basis in mobilizing users, was also explored. Tufekci and Wilson (2012) interviewed protesters in Egypt and showed that Facebook and social media provided an alternative source of information for people under the authoritarian regime in the country, along with the means to organize themselves, thus allowing these ideas to spread. Similarly, Storck (2011) argued that social media provided alternative press and awareness for people in Egypt and contributed with political mobilization. Social media thus allows discourses that are not shown in traditional media to arise and spread, mobilizing users around these different topics, creating and spreading new and multiple narratives (Poor, 2005).

## Twitter and Protests

There is a growing literature about the role Twitter is playing in these new protests around the world. Some authors (Shapiro, 2012) refer to them as "Twitter Revolutions," arguing they were in several ways triggered by Twitter users through Twitter. The tool enables user to participate in different ways, both by creating and spreading information directly to other users. Penney and Dadas (2014) discussed Occupy protests and explored these roles, arguing that the platform allows for (a) facilitating protests off-line, (b) live online coverage, (c) spread of information and hyperlinks, (d) expressing different opinions about the events, (e) getting involved in debates about the protests, (f) connecting with other activists, and (g) facilitating online actions. Gleason

(2013) studied the Occupy movement and the different opportunities Twitter creates for participation as well, particularly through hashtags. The author explored how the content created and shared by users help them learn about the movement, explaining the particular affordances in Twitter, such as multimodal content and real time search, for example, may be connected to the user's participation.

Researchers also examined the process of recruitment through social media. Gonzalez-Bailon, Borge-Holthoefer, Rivero, and Moreno (2011) examined how people were influenced in Twitter during the mobilizations in Spain in 2011 and found that there are different topological position for early protesters and influencers that trigger the information cascades. Lotan et al. (2011) analyzed the information flows during Tunisian and Egyptian Revolutions created by different actors and showed how Twitter not only amplifies their voices but also allows the narratives to be co-constructed. Twitter has also played a role in creating more localized narratives and has amplified the voices of those protesting on-site (Bastos et al., 2014; Croeser & Highfield, 2014).

The literature also includes research focusing on the role of hashtags during political turmoil. Vallina-Rodriguez et al. (2012) discussed the "Indignados" movement in Spain. The authors partially credit the rapid spread of the protests in Spain to Twitter and the use of hashtags. Segerberg and Bennett (2011) explained these "revolutions" have different ecologies and illustrate their argument with the analysis of two different hashtags showing the different "protests" they depict. Beyond connecting the narratives, however, the literature has not explored systematically the roles different types of hashtags play during political events. In the next section, we will briefly cover previous research about general tagging behavior and hashtags in particular.

### Hashtags and Tagging Behavior

Hashtags are keywords or brief sentences posted on Twitter preceded by the "#" sign. They are used to contextually mark Twitter's conversations around a certain topic (boyd, Golder, & Lotan, 2010). However, their appropriation by Twitter users also suggests other contexts and functions. Because Twitter falls into the definition of a "social tagging system" (Marlow, Naaman, boyd, & Davis, 2006), it allows users to participate in creating and sharing a particular resource through tags. These resources can be multiple because their meaning is socially constructed. Therefore, hashtags have several functions beyond marking context.

Huang, Thornton, and Efthimiadis (2010) studied tags in Twitter and Delicious and classified them as organizational and conversational types. While the first one focuses on organizing resources, the second is focused on conversation, the tag itself carrying an important part of the message. According to the authors, this second group is more prevalent in Twitter's tagging behavior. Further work provides

even more ideas on the tagging behavior in Twitter. Page (2012) discussed hashtags as visibility mechanisms, not just contextual markers. Recuero, Amaral, and Monteiro (2012) showed that fans use tags to publicly support their idols and gain visibility. Finally, Diaz-Aviles, Siehndel, and Naini (2011) argued that tagging behavior in Twitter is more related to filtering and directing content to specific audiences, where the tag itself is part of the message. Thus, hashtags have an important role in creating visibility for a message and also, can be the message themselves, not only marking context but also changing and adding content to the tweet.

In the previous section, we reviewed the literature on protest movements organized through social media and how the latter, specially Twitter, arguably influences events on the ground. The role hashtags play in these events and how tagging behavior works is, however, less studied. Bastos, Raimundo, and Travitzki (2013) claimed some hashtags in protests have different roles such as "pamphleteering," which relates directly to the visibility characteristic discussed by previous authors also adding a layer of mobilization to the tag (similarly to what Vallina-Rodriguez et al., 2012, argued). Barash and Kelly (2012) discussed the different roles of political hashtags and showed that different semantic tags were used in different ways by Twitter communities in Russia, having thus different dynamics. However, none of these works focused on hashtags from a linguistic and communicative point of view, discussing how their functions may help understand their usage. These works shed light into the role of hashtags during political protests. However, to the best of our knowledge, no systematic attempt to categorize the different communicative roles played by hashtags has been presented. In the following section of this study, we discuss these roles in detail by relying on theories of language.

## Functions of Language

Roman Jakobson (1960) is a central reference in the Prague Linguistic Circle. He created a model of the communication process that ties together both communication and linguistic functions. The model is a classic work in linguistics and has been deployed in a range of disciplines. Although it is criticized by many authors (Shapiro, 2012), the model provides a starting point to discuss how hashtags create meaning and how their different functions influence tagging behavior online.

Jakobson's (1960) model of the functions of language is built upon six elements that are necessary for every communication to occur. Every communication between two actors needs (a) context, (b) sender, (c) receiver, (d) contact, (e) common code, and (f) message (Figure 1). Each of these elements may have a preeminent role on the process, and each communication may have a different function. Thus, there are six key functions to the language: the *referential* function (where the context is the focus of the interaction), the *emotive* or "expressive" function (where the focus is on the

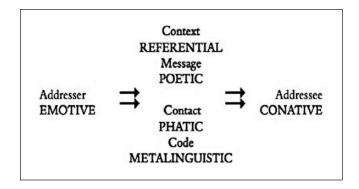


Figure 1. Jakobson's model.

sender and what he or she wants to express), the *conative* function (where the focus is on the receiver and the action the sender wants him or her to take), the *phatic* function (focusing on the contact/channel aiming at understanding whether the channel works), the *metalingual* function (which focuses on the code used), and finally, the *poetic* function (which focuses on the message itself). These functions help the analyst to understand what is important in each message and how the message is conveyed. To Jakobson, a single act of communication may have several functions, albeit one is typically more preeminent.

In social media context, Radovanovic and Ragnedda (2012) discussed the phatic function related to the need to communicate and to keep in touch with other people in Facebook and Twitter posts. In this work, they classified types of phatic posts, showing how users constantly use this type of information to check whether others are there and whether they are visible. Miller (2008) also discussed the importance of this function in online media in general. However, literature focusing on these functions on Twitter is still forthcoming.

Jakobson's model is appropriate for the purposes of this study because it pays particular attention to language in mediated environments and addresses the role played by the channel as a fundamental part of the communication process. Social media is a channel and language has to be negotiated so users can make sense and appropriate the medium for their own purposes. Contentious communication is no exception to this. Jakobson's model also helps us to unveil how hashtags may play a bigger role than simply being a contextual (boyd et al., 2010) or an organizational and conversational function. When used for activism and political participation, hashtags may assume different roles and have different communication functions to help users make sense and influence others in times of turbulence.

## Method

During the Brazilian protests of June 2013, we monitored and collected a data set of 2,321,249 tweets related to the events. The period of data collection comprehends June 13 to

20, which was the most active period of the protests (Bastos et al., 2014; Singer, 2013). The original data set used in this article was thus created through the archival of 35 keywords and hashtags related to the protests using the open-source platform yourTwapperKeeper (2012). Data collection also relied on keywords used by protesters and observers, such as "protest" and "protests," "manifestation" and "manifestations," and so on (see the appendix). Relevant keywords were subsequently added to the corpus as new entries to the open-source platform yourTwapperKeeper. A more detailed account of the procedures used for data collection is provided elsewhere (Bastos et al., 2014).

As previously discussed in the article, our primary objective is to describe the role hashtags played during the protests in Brazil. This objective is divided into three specific goals: (a) to understand the types and functions of hashtags used during the protests, (b) to examine how co-occurrences of hashtags depict different meanings and functions, and (c) to describe tagging behavior of users during the protests. To discuss these objectives, we first describe the data collected and rely on content analysis to classify the functions of hashtags and their co-occurrences (Krippendorff, 2013). To this end, we subsampled the data and applied a qualitative coding procedure to the 500 most retweeted tweets in the data set that identified tagging behavior within co-occurrences of the categories of hashtags. The sample size was defined according to the possibilities of human coding and aimed at providing an overview of the context where hashtags were used.

## Data

In the original database of 2,321,249 tweets, 56.3% (1,306,847) of them did not contain any hashtag. There were 567,623 (24.5%) with one hashtag and 19.2% (446,779) with two or more hashtags. The distribution of hashtags per tweet is shown in Table 1. The total number of occurrences of hashtags in the database was 1,868,427, and the total number of unique hashtags was 77,074.

From the original set of tweets, we selected a subset covering the 1,040 most frequent hashtags (the ones that appeared more in the total number of unique tweets) which accounted for 85% of the total of unique occurrences (Figure 2) with a total occurrence of 1,605,816. These hashtags were manually categorized by two independent coders based on their linguistic and communicative perceived functions according to Jakobson's (1960) model using Content Analysis (Krippendorff, 2013). Content Analysis is a method for textual analysis based on classification and codification of elements and parts of texts, often described as "quantitative," "objective," and "systematic" (Neuendorf, 2002, p. 1).<sup>2</sup> Jakobson argued that a single message may have all functions present. For analytical purposes, thus, Hébert (2011) explained the analyst needs to establish a hierarchy between the functions by "identifying the dominant function" (p. 1). Our criteria

Table 1. Distribution of Hashtags per Tweet in the Data Set.

Hashtags	Tweets	%
0	1,306,847	56.3
1	567,623	24.5
2	234,395	10.1
3	112,434	4.8
4	51,923	2.2
5	23,964	1.0
6	12,129	0.5
7	6,763	0.3
8	2,956	0.1
9	1,155	0
10	584	0
11	238	0
12	146	0
13	37	0
14	17	0
15	13	0
16	12	0
17	8	0
18	1	0
19	4	0

were based on the question, "What is the purpose of this message?" which coders applied by consulting the usage of each tag in the tweets from the data set. We thus qualitatively consulted at least 10 tweets per tag to get a better understanding of their usage. To check for reliability of the classification between the coders, we used Krippendorff's alpha, which was .76 (Krippendorff, 2013) with a reliability of 82.3%, which is considered good.

In addition, as we can see in Table 1, there are many tweets with more than one hashtag. To better explore hashtag functions, we also analyzed co-occurrences of hashtags. For this second part of the analysis, we extracted a random sample of 45,000 tweets from the total number of tweets with two or more hashtags (446,776). The sample includes 46,090 unique pairs of different hashtags and a total of 118,906 co-occurrences (Figure 3). Of these co-occurrences, 1,270 were used at least 10 times and are responsible for 47% of all co-occurrences. The few hashtags appearing in this subset that were not categorized in the previous analysis were coded using the same methodology.

Finally, we also analyzed qualitatively the 500 most retweeted tweets that used hashtags from the original data set, with a total of 278 hashtags co-occurrences. We analyzed the tweets focusing on the (a) meaning of the messages and (b) general context of the tweet through a case study (Creswell, 2006), aiming to provide more context to the quantitative analysis. This group of tweets was the base for the discussion of users' retweeting behavior and will provide the examples we will further use in this article. Tweets were anonymized. All the tweets we analyzed in this work were originally composed in Brazilian Portuguese<sup>3</sup> and translated

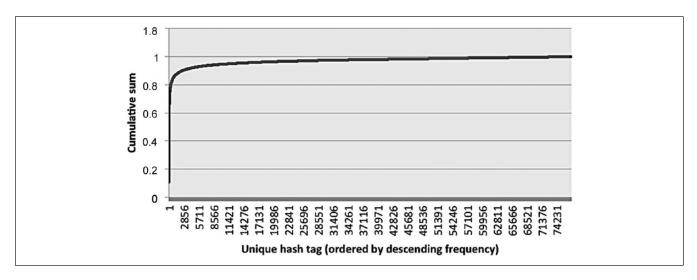


Figure 2. Cumulative distribution function of total number of hashtags in the database.

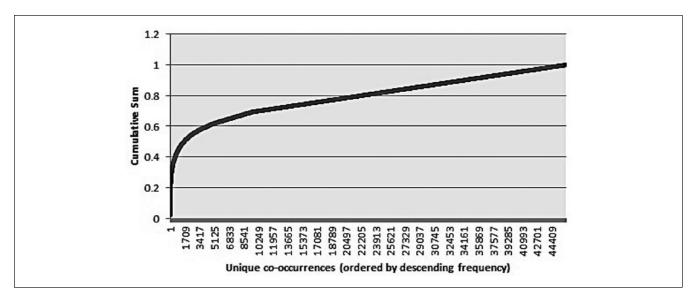


Figure 3. Cumulative sum of co-occurrences of hashtags.

to English for the purposes of this report. As native Portuguese speakers, we translated every hashtag and tweet aiming at keeping the original meaning and context. Thus, tweets such as "#cometothestreet Fight for a fairer Brazil #changebrazil" were originally "#vemprarua Lute por um Brasil mais justo #changebrazil." We kept the words that were originally in English and translated the remainder from Brazilian Portuguese into English.

## **Analysis**

## Types of Hashtags

We classified the hashtags according to their function on Jakobson's (1960) proposal of language functions. In this

section, we present the analysis focused on hashtags that presented agreement between coders during the classification process.

a. **Referential hashtags** were considered the hashtags that address the context of what was happening, frequently trying to locate the event being discussed. We found in the data set several similar hashtags with small variations, such as #protestosp, #protestorj, and #protestopoa. Each tag refers to "protest" (protesto) followed by the event's geographical location (sometimes a city or a state in Brazil). Some other contextual hashtags were less specific, especially the ones directed to a foreign audience (e.g., #brazil, as opposed to #brasil, the Portuguese word). One example is "RT @

userC: FGV releases the wi-fi password in Paulista! Password: 548 [link] #protestosp." The hashtag localizes to which protest the message is directed. Paulista Avenue was the place where the majority of the protests in Sao Paolo happened. Because the lines got too busy during the events, protesters frequently asked buildings around the avenue to share their wi-fi passwords allowing the users to live tweet the events. Other examples are "RT @userA #protestorj the police started to use tear gas to stop people from getting to the protest!" and "RT @userB: Tomorrow, 5pm, Sé Square! #protestosp." In both cases, the localization is key for the audience to understand the context of the message.

- b. We classified as **expressive/emotive** hashtags focused on the sender's emotions, thoughts, or opinions. This group includes hashtags aggressively targeting the Brazilian President Dilma Rousseff—for example, #foradilma (#dilmaout) and #chupadilma (#suckit-dilma). Hashtags in this group also include demands of the protesters such as #contraoaumento (#against-theraise) and #passelivre (#freefare). One example is "RT @userD: #cometothestreet #dilmaout." In the example, the tag functions as a way to show the user's opinion.
- c. Conative hashtags were mostly used in the imperative tone, such as #vemprarua (#cometothestreet). Others asked followers to do something, as "#compartilhe" (#share). We classified as "conative" hashtags aiming at mobilizing protesters and directed at causing an action by the receiver. Conative tags were heavily used during the protests. One example is "RT@userF: Friends with the same goal: Brazil's future! #gobrazil [link]." The tag here is used to motivate other users to also protest.
- d. Finally, we also found **metalingual hashtags**, which were used to show or convey the code. In this category, we found signature hashtags, such as the ones used by popular media to identify their content (e.g., #r7, #bbc, and #oglobo). We also found hashtags used to describe the content type of an attached link, such as #instagram, #picture, or #video. One example is "RT @TerraoAoVivo: RJ protester is shot with lethal ammunition in front of the Assembly [link] #TerraAoVivo." The hashtag refers to the vehicle that published the information.
- e. Poetic hashtags were not found in the data set.
- f. The phatic category was not found in the data set, even though it is not uncommon to find hashtags such as "#goodmorning" or "#hello" in Twitter, which would be classified under this category. We believe the lack of phatic and poetic hashtags can be explained by the context of the protests and the function performed by Twitter in mobilizing people and covering what was happening on the ground. Because Twitter only

Table 2. Number of Hashtags and Occurrences per Type.

Туре	Occurrences	Unique	Total/unique
Conative	730,497	215	3,397
Referential	401,270	339	1,184
Emotive	110,397	140	789
Metalingual	5,149	11	468
Poetic	0	0	0
Phatic	0	0	0
Others	64,256	164	382

Table 3. Occurrences of Hashtags Pairs for Each Category.

Pair	Occurrences	Unique	Unique/total
СС	19,998	272	74
CR	15,471	336	46
RR	7,498	203	37
CE	5,901	170	35
CM	1,768	38	47
RE	1,728	86	20
EE	631	32	20
RM	536	21	26
MM	217	2	109
EM	148	3	49

Note. CC = conative-conative; CR = conative-referential; RR = referential-referential; CE = conative-emotive; CM = conative-metalingual; RE = referential-emotive; EE = emotive-emotive; RM = referential-metalingual; MM = metalingual-metalingual; EM = emotive-metalingual.

allows 140 characters, it seems that most users would focus on sharing the message with the effect they had in mind, rather than using other tags to create different effects. Poetic hashtags would create attention to the message itself, not the receiver as phatic would also create attention to the channel, not the effects desired.

There were also incomplete hashtags, hashtags from other protests, and hashtags not at all associated with the protests (frequently from spammers). We discarded these data because our main focus in this article is to discuss protests happening in Brazil (see Table 2).

The proposed classification, once validated, was further applied to the second step that we detail next. In cases where a hashtag performed more than one function, we classified them under the strongest one (Jakobson, 1960). (Table 3 shows the total number of occurrences and unique occurrences of hashtags in each category.) The frequency of these different hashtags shows that the tags had more than a contextual function during the protests. On average, each unique hashtag occurs from a few hundred times to thousands of times, showing that hashtags are often co-opted by users. Conative hashtags were by far the most frequent category by total number of occurrences, followed by Referential and Emotive. Unique occurrences show a similar pattern, but it

**Table 4.** Co-Occurrences of Hashtags Among the Most Retweeted Tweets.

Co-occurrences	Total
CR	123
RR	80
CC	35
RE	28
RM	23
CM	1

Note. CR = conative-referential; RR = referential-referential; CC = conative-conative; RE = referential-emotive; RM = referential-metalingual; CM = conative-metalingual.

can be seen that Referential tags are more diverse than the other categories, showing the importance of adding a localization context to the tweets. Emotive tags were also used, showing tags as forms of sharing an opinion. The presence of these functions may help understand Twitter as a public sphere for voicing one's opinion, for mobilization, and for debate.

## Co-Occurrences of Hashtags and Tagging Behavior

We observed that hashtags often occurred together on the same tweet. Thus, we classified the co-occurrences of hashtags by combining each hashtag's function. We chose to classify the co-occurrences in pairs as most tweets include less than three tags (Table 1). Based on these results, we created the notation: referential-referential (RR), referential-emotive (RE), conative-referential (CR), conative-emotive (CE), conative-metalingual (CM), metalingual-metalingual (MM), emotive-metalingual (EM), and referential-metalingual (RM). Table 4 summarizes the categorized data. In this section, we will present each type of co-occurrence and further discuss the user tagging behavior of each through qualitative examples obtained in the data set. Examples for this discussion were provided by the qualitative analysis.

Group **conative-conative** (CC) is the most preeminent in the data set of hashtags. Conative hashtags are designed to mobilize users and frequently use the imperative form. When used in pairs, they reinforce and complement each other for example, #vemprarua (cometothestreet) and #vemprajanela (cometo-thewindow). The vast majority of tweets in the sample include imperative sentences. Tweets such as "#cometothestreet Fight for a fairer Brazil #changebrazil" and "Show your dissatisfaction! Don't shut yourself down! #cometothestreet #dilmaout #shoutwithoutfearBrazil" directly address the audience through the usage of imperative verb form. In these cases, the hashtags are strengthening the meaning of the imperative mood. Also, the usage of exclamations and slogans in this category was frequent. For example, many tweets used parts of the national anthem as a sentence such as "Thou will see that a son of thine doesn't

flee from battle #wakeupBrazil #cometothestreet." In other cases, the tweet was multimodal, with a picture of the protests and two conative tags ("RT @username oh my god #changebrazil #Thou willseethatasonofthinedoesn'tfleefrom battle - link [for a picture of a huge amount of people in the protest in Rio]"). As we can see, the usage of the hashtags in this case adds meaning to the original tweet. It does not only pamphleteer but also urge people to take to the streets and to "wake up." In this category, we also observed tweets that only contained the hashtags, such as "#thouwiltseethatasonofthinedoesntfleefrombattle #cometothestreet." These tweets comprise their entire meaning on the tags used, also strengthening the mobilization status of the message. Even though not all tweets in this category had imperative sentences, the usage of a combination of CC hashtags added a meaning layer of mobilization to the message. Because of this, tweets like "Tomorrow will be another day #thegiantwokeup #changebrazil" or "This is it! #cometothestreet #cometothewindow4" have different meanings with and without the hashtags. These tweets had thus very clear mobilizing characteristics directed to the audience, urging the users to act and take part in the protests.

**CR** is the second most preeminent group among the data set. This combination adds to the conative hashtag a role of context. Frequently, these co-occurrences were related to one localized hashtag (e.g., #protestorio) together with an imperative one (e.g., #vemprarua/#cometothestreet). This goes to show that Twitter had not only a mobilizing role among Brazilian users but also a localized one, which was important for local manifestations to share their message. Hashtags were used not only to mobilize but also to motivate users in their own localities to engage in the protests. Referential tags also played a role in tying together the protest narratives of different locations. Tweets from this category were often tweets sent "live" from the site of the protests, such as "Hello people who are honking because we closed the antonio carlos. We're sorry for the inconvenience. We are changing the country #protestbh #cometothestreet." Another example is "#protestsp #cometothestreet Raposo Tavares closed by protesters!"6 Other tweets with this type of co-occurrence expressed the user's opinion about the protest, such as "THIS" IS BEAUTIFUL! #protestsp #cometothestreet." In this case, the tweet doesn't only represent an opinion, but also the subject the user is talking about (the protest in Sao Paulo) and the mobilization hashtag (come to the street). In this category, we also found several image links accompanied by one conative and one referential hashtags (multimodal tweets). In such cases, the message works both as an incentive for users to attend the protests and as a context marker for the picture. This category of co-occurrences thus seems to act to create context for the information shared and also motivate other users to join the events. Therefore, CR tweets are frequently used with hashtags to add a layer of localization and mobilization to the message in a specific location, trying to coordinate different protests and different participants through the localization (contextual hashtag).

The third group is the **RR**. These hashtags were mostly seen on tweets with a more informational role. They were used to mark context of the narratives shared on Twitter. Sometimes, the hashtags used reinforced the same context (e.g., #RJ and #protestori both refer to the same area, the city of Rio de Janeiro, and appear together rather frequently). Tweets with these tags probably aimed to reach an audience that used several referential tags to refer to the same event. One example is the following tweet: "#RJ Attention people who live in the center please open your wifis, the 3G is off! #protestoRJ' (tweet sent during a protest in Rio de Janeiro). Both tags, #RJ and #protestoRJ, refer to the same event, in the same place. Another example is "The congress was invaded, police is alert, take care people! Two protection lines from the police #protestsp #sp17j #cometothestreet." In this case, while the user chose two referential hashtags ("protestsp" and "#sp17j"—Sao Paulo, June 17), he also chose to use a conative tag. Hashtags in this category were used mostly to provide only context to the tweet, often narrating the protests on-site and were among the most retweeted tags. Referential tags were often used together to amplify the narrative to other networks and to reinforce the localization/context of the information published. Although this may also happen in other categories, RR reinforced the localization of tweets, which in Brazil, given the size of the country and the number of protests happening at the same time in different cities hundreds of miles away, seemed to be important.

The CE is a group of co-occurrences also used to mobilize other users. This group includes hashtags associated with demands (e.g., #passelivre/#freefare) that co-occurred with conative hashtags. In this case, users would tweet their demands and opinions and use a conative hashtag to make reference to it. This category included most of the demands, with a few opinions also occurring. For example, tweets such as "Even people who couldn't leave home participated. #freefare #thepeoplewokeup [link]." While the tweet depicts the users' opinion, it also has a demand within (#freefare). Hashtags were often used in this sense. Another example was the tag "#hurtsineveryoneofus" that was heavily used after June 13, when the images of the police truculence started circulating both in social media and in traditional media sites. Tweets such as "#Cometothestreet #hurtsineveryoneofus" often with a link for a picture or video of the violence were also found in the case. These types of tweets focus on mobilizing the viewer through opinions and demands.

The **referential-conative** (**RE**) group includes hashtags about the context and voicing users' demands. For example, #partiu (#gone) and #protestosp not only indicated that the user was going to the protest but also added to the bigger picture of the narrative aggregated by the referential hashtag. While this category contained demands and opinions, it also added a referential layer to the tweet. This referential function often surpassed the emotive one. For example, tweets from the media used emotive hashtags to identify the protests in the beginning. One example is "*In the video, journalists*"

say 'we're press!' and the police officers fire [link] #protestsp #freefare." Another example is "Unique demands, clear and specific: repealing the increase #mpl #repealingtheincrease." The tweet uses the #mpl (Movimento Passe Livre—Free Fare Movement) tag to contextualize whom it is talking about and the "repealing" tag to demand the fare rise to be repealed by the government. Hashtags also had their meaning changed by these co-occurrences. One example is the following tweet "#SupportDilma is this serious? People that don't know how 'comfortable' it is to use metro and bus today or the public hospitals are angry! #protestrj."

The CM group mostly had signature hashtags posted with conative ones. The group Anonymous, for example, posted several tweets with conative tags and their signature that pledged others to take to the streets (e.g., #AnonymousBrasil and #vemprarua/cometothestreets). Several media groups also used this combination of tags. Tweets in this category were frequently from the alternative media using hashtags as signatures along with conative tags as a way to motivate other people. Anonymous frequently posted messages with these types of hashtags, for example, "Hacked #Anonymous #ChangeBrazil." In the example, we have two hashtags cooccurring, a signature one (metalingual) and a conative one. We also observed that when users retweeted updates from media outlets, they would often include conative hashtags, creating a layer of mobilization over the information shared. "#cometothestreetmanaus example, '@portalr7: Government reveals increase in the expenses for the World Cup 2014 [link] #r7." The original tweet was "Government reveals increase in the expenses for the World Cup 2014 [link] #r7," which contained only a metalingual hashtag. By retweeting and adding the "conative" tag, users change the meaning of the original tweet, using the information to mobilize others. Another example is "RT @r7 Crowd covers downtown Rio in white [link] #R7//There is space for many more #goRJ." The original tweet contained a link and a metalingual hashtag—the signature of the media vehicle (R7). The user retweeted the message with the conative hashtag "#goRJ" added.

Emotive-emotive (EE) was a category with demands and opinions. Many tweets, for example, demanded the resignation of the Brazilian President Dilma Rousseff with hashtags such as #foradilma (dilmaout) and #chupadilma (suckitdilma) in the same message. Others added demands and opinions such as #Globomente (Globolies-Globo is the largest media group in the country). Several tweets among this category had only hashtags such as "#repealtheincrease and #gonetofightformyrights." They often explained not only what the user was doing but also her motivations to take to the protests and her demands. Another example is "#SupportDilma No, no . . . I prefer #suckitdilma." The hashtag "#supportdilma" appeared during the protests, as a response to the large spread of the "#suckitdilma" tag which dominated the Brazilian Trending Topics for some time. In the example, the user shows both hashtags to indicate which

one she supports, changing the meaning of support of the first tag by adding the second one. These tags also appeared often with other groups of tags, especially conative ones. For example, "The city is OURS! #Spwillstop #freefare #repeal-theincrease"; tweets such as the example would not only depict the demands but also pamphleteer meanings, trying to convince other users to take to the streets.

RM was used by some media groups and also by alternative media to narrate operations and individual protests. One example is the group #midianinja (ninja media) that used their own name in one tag and the event they were posting about in the other #sp17j (an acronym for Sao Paulo, June 17). This category was composed mostly of media tweets. The referential hashtag was used to identify the context whereas the metalingual was used to sign the content. Tweets such as "ALERT: Protesters clogged the Marginal Pinheiros in Sao Paulo [link] #R7 #protestsp" used hashtags to sign and contextualize the tweet both by locating it and showing the source (R7 is a media group in Brazil).

The **EM** Tweets often included tweets from the media that were appropriated by users to show their opinions on the subject. Emotive tweets were also used to change the meaning of other hashtags, by adding an opinion to retweets. For example, "The world will protest! RT @UOL Berlin act supports SP [link] #UOL #dilmasuckit." The original tweet was "Berlin act supports SP [link] #UOL." In this example, we can see how users appropriated the original message with a metalingual hashtag (signature) by adding opinions (emotive hashtag—#dilmasuckit).

The **MM** Tweets was the least frequent category. Few cooccurrences were found, formed mostly by usage of several hashtags in the same tweet (again, Anonymous used often two or more hashtags such as #Anonymous and #AnonymousBrazil to reinforce the context and signature of their tweets). Some tweets simply mixed content from different media outlets and used both hashtags to identify the sources (others merely repeated information).

By observing the unique co-occurrences, it is possible to see that we now have CR as the most frequent category, followed by CC and RR. This is due to the number of different hashtags used to localize the protests and the importance to localize the narratives among several protests that occurred at the same time. Context was always changing and the most popular hashtags are fairly similar (e.g., hashtags following the method of protest+city). However, conative hashtags are less varied. The most popular tags remained being used throughout the entire period of analysis.

## Co-Occurrences in Most Retweets and Tagging Behavior

Finally, we qualitatively analyzed hashtag co-occurrences among the 500 most retweeted tweets. In this data set, we found 203 tweets with more than one hashtag, with a total of 307 co-occurrences (some tweets had more than two tags).

These co-occurrences were further analyzed and are discussed in this section (see Table 4). Even though there were more than two tags in some tweets, we analyzed tweets by co-occurrences of two by two categories as the previous analysis.

The most retweeted tweets were specially focused on the live narrative of the protests, frequently by protesters themselves, celebrities, and the media. Both referential and conative hashtags were the most used types, followed by metalingual (mostly media signatures) and emotive. The most retweeted tweet with co-occurrences was from a media outlet, "TerraAoVivo" (TerraLive), and said, "Manifestants sit down in Berrini (a street in Sao Paulo). Police also sits down and is applauded. [link] #Terralive #protestsp" (765 retweets). Hashtags in this tweet have different functions. The first one is a metalingual tag, a media signature. The second one localizes the event the tweet is covering (context). While the signature adds credibility to the information (it is from the live coverage of a media outlet), the second one adds context. Live tweeting is also an important evidence of how Twitter allowed users to create and share their own narratives of the protests, using hashtags to personalize them (as Segerberg & Bennett, 2011, 2012, discussed).

Hashtags co-occurrences were very similar to what we described before, in the larger data set. The majority of cooccurrences was within CR group (123 co-occurrences), such as "RT @userH It started! They are shooting people with rubber bullets in Rio [link to a media transmission] #cometothestreets #protestrj." As we pointed before, conative tags were used as motivation tags to inspire other users to go out. Thus, these tweets were often comments from people who seemed to be watching the manifestation and from people from the streets. Examples are "RT @userX The police is throwing bombs from helicopters! Is this a war? #protestbh #cometothestreets" (this tweet was sent by a group covering the manifestations live) and "RT @userR Riot police attacked the protesters and threw bombs even inside the Sao Cristovao station [metro]! #protestrj #changebrazil." This usage also points to the idea of personalization of collective action that Segerberg and Bennett (2011) discussed. As Twitter allows users to create personalized narratives to their own contexts where the protests are happening, the role of the tool seems to be increased as a key player in the organization and diffusion of the events. This enables users to create and share their own vision about the events (Poor, 2005).

The second most preeminent category was RR (80 co-occurrences). This category seemed to be populated by tweets from users in the streets live tweeting the event. For example, "RT @userR You won't believe, but right now there are no political [parties] flags in the manifestation! We expelled the PSTU [a Brazilian party]! #protestsp #protestbr." This tweet was sent by a tech celebrity in Brazil who was live tweeting the protest (several of his tweets are among the most retweeted). Another example is "RT @userT Police"

confiscating cell phones and cameras from protesters for no reason in Rio de Janeiro #protesrj #brazilprotests." This tweet was sent by a group who was also live tweeting the protests in Rio de Janeiro. Some of the co-occurrences in this category were redundant (two hashtags that described the same event—for example, #protestrj and #rj both refer to Rio de Janeiro and were used to localize the narrative). This may be due to the fact that users wanted to ensure that the audience following different tags would be able to receive their message or because users wanted to relate different protests in the same narrative. As protests in major cities were given more attention by the media and the audience, this strategy may have been used to make smaller protests more visible (Poor, 2005).

Category CC had 35 co-occurrences and was used among different tweets. For example, some people used these tags together with several other types of tags, probably in an attempt to add visibility to their tweet, such as "RT @userI And thou will see that a son of thine doesn't flee from battle! #protestsp #cometothestreets #thegiantwokeup #freefare [link]." Some others used the tags to narrate the protests from the streets such as "RT @userH Police is throwing bombs from inside the palace [the police headquarters] #cometothestreets #changebrazil #protestsp." We understand the difference found in this category to be related to how users focus on mobilizing other users at the same localization. Therefore, we believe both mobilization and localization of the narratives play an important role in this type of message, as pointed by Penney and Dadas (2014).

Category RE was next with 28 co-occurrences. Emotive tags, however, were not among the most used; they were present because of tweets with more than two tags. An example is "RT @userZ #allrevolutionsstartwithasparkle #downwithglobonetworkthepeoplearenotstupid #wakeupbrazil #thegianthaswakeup #protestsp #protestrj [link]." The same goes to the CE category, which has 16 occurrences due to tweets similar to the example.

Category RM mostly had media tweets; the metalingual hashtag was used both as a signature (RT @UOLNews It went down! The fare in Sao Paulo has decreased to 3 reals! [link] #protestsp #uolnews) and, sometimes, to describe the link in the tweet (RT @userK #live streaming from Largo do Batata [the place where protesters met in Sao Paolo]! #protestsp). This category was fairly present (23 co-occurrences) because among the most retweeted tweets, there were several from the media.

Categories MM and CM were not present in the data set. We believe these categories were not present in this data set simply because of the role Twitter played during the protests.

Thus, co-occurrences of hashtags seem to be related to (a) the personalization of the messages narrating the protests, specially focusing on the mobilization and the narration of the events in each place where they occurred; and (b) strategies to relate different protests, maybe to gain visibility or even to create a more unified narrative. This may indicate the

Table 5. Functions of Hashtags During Protests.

Function	Function during protests	Examples
Conative	<ul><li>Mobilization</li><li>Pamphleteering</li></ul>	#cometothestreet #screamwithoutfearBrazil #changebrazil
Referential	<ul> <li>Localization</li> <li>Narrative organization</li> <li>Audience direction</li> <li>Live coverage</li> </ul>	#protestorj #protestobh #protestosp
Emotive	<ul><li>Demands</li><li>Opinions</li></ul>	#passelivre #foradilma
Metalingual	<ul><li>Signature</li><li>Characterize information</li></ul>	#R7 #instagram #AnonymousBrazil

hybrid nature of these events and how Twitter enables users to coordinate and act in different locations at the same time. This is especially important in a country so huge as Brazil.

### **Discussion**

In this article, we focused on exploring and discussing the functions of hashtags during the protests of June in Brazil. We examined the frequency of hashtags types and their cooccurrences and the tagging behavior of user. We found that during the protests, Twitter hashtags performed several communicative functions that were not only focused on context, as boyd et al. (2010) previously indicated as the main function for hashtags. Sometimes, the main function seemed to localize the tweet or to mobilize the audience toward the protests. By creating a complex ecosystem of tags, users change the original meaning adding different layers of functions, often overlapped between each other. Thus, different types of hashtags performed specific functions that changed the original meaning of the message, by adding layers of opinion, calls for mobilization, or assigning meanings unrelated to the original message (see Table 5).

The majority of hashtags used had both a conative function aimed at motivating and mobilizing users and a referential one aimed to localize the event (the CR category). This may be expected because these tags worked as pamphlets similarly to what Bastos et al. (2013) previously reported. However, the co-occurrence of conative tags with referential ones is something new. Probably because there were many protests occurring in several places at the same time (Bastos et al., 2014; Singer, 2013), this co-occurrence seem to be a way to focus the narrative and mobilize specific audiences within Brazil. Also, conative hashtags tend to co-occur with other conative hashtags, highlighting the mobilizing role of Twitter and acting as a vector to persuading more people to protest as pointed out by Vallina-Rodriguez et al. (2012).

Another category greatly used was the referential hashtags. Although they refer to context, the most used ones refer to particular contexts, often localized ones. Thus, their

role was not only to provide context but also to allow users to access information related to a particular context. This is a more organizational function in the sense discussed by Huang et al. (2010). Referential hashtags also played a very important role in live coverage of the events (one of Twitter's functions in protests, according to Penney & Dadas, 2014, by organizing the narrative toward different cities and audiences). Among the most retweeted tweets, referential tags were always present, helping to contextualize the tweets (often sent live from the site of the protests). When used with conative tags, the pair works as a pamphlet and mobilize specific areas. Because of these characteristics, referential tags were also important for live tweeting the events and amplifying the voice of protesters (Bastos et al., 2014; Croeser & Highfield, 2014). Referential tags could, thus, create separate streams of conversation to each protest happening in each state of city and help organize the information stream of local events (as we observed through the usage of the tags #Protest+city). Combined with other types of tags, referential hashtags could add the mobilization layer (when combined with conative tags), add the opinions and the demands layer (when combined with emotive tags), or simply add signatures or descriptions of the content (through the metalingual tags) both to people at home and in the streets.

Emotive hashtags were mostly about demands. It is important to notice that because these demands were showed through tags, they could gain visibility and support among other protesters in other parts of the country (Page, 2012) by the adoption or not of the tag. However, when used with other types of hashtags, such as conative, they acquire new functions, such as motivation and pamphleteering. So a tag such as "#dilmaout" put together with a "#cometothestreet" urges people not just to protest but to protest against Brazilian President Dilma Rousseff in a more broad context (not only in a localized protest). On the contrary when used with referential tags, emotive tags would tie the demand or the opinion to a local context, often creating a specific demand to a protest in a certain locality.

The functions Twitter provides protesters (Penney & Dadas, 2014) cannot happen all at the same time. For Twitter to work as a communication tool, users need to find a way to organize the flow of messages. Those two categories (referential and conative) were by far the most used, mostly because of the functions they create during an event like a protest. While referential tags organize the flow, conative tags keep people engaged and mobilized. Emotive tags showed support and added demands and critics to the conversation.

Finally, metalingual hashtags were mostly used to characterize the information posted (conveying the code) such as #instagram to indicate part of the information was in the site or to indicate the link posted was a video (#video). Other function was to "sign" the message, identifying the content provider. Media often used this function. When metalingual hashtags were used with others, they usually retained their

Table 6. Summary of Findings.

Co-occurrences	Functions during protests
CC	Mobilization
	<ul> <li>Motivation</li> </ul>
CR	<ul> <li>Live tweeting (on-site)</li> </ul>
	Localization
RR	<ul> <li>Context</li> </ul>
	<ul> <li>Localization</li> </ul>
	<ul> <li>Audience direction</li> </ul>
CE	<ul> <li>Motivation</li> </ul>
	<ul> <li>Demands</li> </ul>
RE	<ul> <li>Context</li> </ul>
	<ul> <li>Demands</li> </ul>
EE	<ul> <li>Reinforce opinion</li> </ul>
	<ul> <li>Show/reinforce demands</li> </ul>
	<ul> <li>Discuss/oppose demands</li> </ul>
CM	<ul> <li>Live tweeting (alternative</li> </ul>
	media)
RM	<ul> <li>Live tweeting (media)</li> </ul>
MM	Characterize information
	<ul> <li>Reinforce code information</li> </ul>
	<ul> <li>Signature</li> </ul>

Note. CC = conative-conative; CR = conative-referential; RR = referential-referential; CE = conative-emotive; RE = referential-emotive; EE = emotive-emotive; CM = conative-metalingual; RM = referential-metalingual; MM = metalingual-metalingual.

function but added a narrative focus (when used with referential tags) or a mobilization one (when used with conative tags).

Among the most retweeted tweets, we saw a similar pattern, with an emphasis to the categories CR (the most preeminent), RR, and a large number of RM. The RR category was mostly used by live users narrating the protests in an attempt to localize the events. The CR category was used by different users both to localize and to mobilize others to protest. Finally, the RM category was present mostly due to the presence of tweets from the media, which were largely retweeted. These results reinforce the categories from the other data sets.

A summary of these findings is shown in Table 6.

This usage of hashtags also adds a layer of personalization to each tweet related to the protests, helping users to create personalized actions, as argued by Segerberg and Bennett (2011). The usage of these hashtags analyzed through Jakobson's model also provides some insights on how people personalized their messages and the nature of collective action during the protests in Brazil. The larger number of CR and RR is evidence to the localization and mobilization as strategies for protests to happen in different cities and the role Twitter plays as an organization platform, which may have allowed the protests in Brazil to happen in the decentralized way they did (similarly to what Segerberg & Bennett, 2012, argued). Through analyzing the functions of hashtags, we may have a glimpse on the hybrid nature of

these events and how social media provides a platform for personalized communication.

### Conclusion

This study provided a detailed analysis about how hashtags present different meanings to tweets in political contexts. We showed that during the protests in Brazil, Twitter hashtags had several communicative functions. Using a frequency, co-occurrence, and interpretative analysis of tweets published during June 2013, we explored and discussed how different hashtags had different functions that added to or changed the meaning of tweets.

To this end, we proposed a classification of hashtags based on Jakobson's (1960) functions of communication and analyzed how users appropriate hashtags for political activism and participation. Based on this classification, we found that conative tags, which focused on mobilizing users, were the most frequent in the tweets. Referential tags, which referred to the different contexts of protest activity, such as cities where the events took place, were less common but more diverse.

We also identified that the largest amount of unique cooccurrences were between conative hashtags (which aimed at mobilizing users) and referential hashtags (which aimed at localizing the narrative). Brazil is a very large country, and there were several protests occurring at the same time; thus, hashtags were used to align different protests and connect them through this organizational narrative. Referential hashtags were important to the live coverage of the events. Conative hashtags were central to the narrative of the events not only because of their pamphleteering function but also because they tied together different local protests. CC cooccurrences were the majority of co-occurrences, thus highlighting how users mobilized each other through Twitter. CR co-occurrences show that this was also performed locally. Other types of hashtags were less frequent but also added information about demands and opinions. Hashtags were also often used to change the meaning of other tags or to add/ reinforce demands and opinions.

The classification proposed and the results reported in this study shed light on the affordances of Twitter platform and how users create and share meaning during political protests. The usage of hashtags is an important form of political activism and mobilization through different audiences and narratives. By understanding how users rely on these markers as part of their message, we can understand the role Twitter plays in instances of political unrest and the strategies users create to spread their message. In this sense, Jakobson's model provides an important contribution to understanding the function of the markers, which as shown in this study, are not simply contextual or narrative.

A caveat of this study is that the localization of the sample (Brazil) may have particularities related to how Twitter is used by the local user base. Nevertheless, the classification

proposed may be informative for other studies focusing on political protests or other communication settings. Types and roles of hashtags may vary in different countries and languages, and these aspects can be further assessed in future studies.

## **Appendix**

List of Twitter Hashtags and Keywords Associated With the Vinegar Protests.

1		Vinogay	Tweets
2       brazilianspring       358         3       catraca_livre       1,512         4       catracalivre       1,515         5       changebrazil       285,385         6       chupadilma       28,200         7       contraoaumento       7,167         8       mudabrasil       103,614         9       obrasilacordou       12,411         10       ogiganteacordou       181,511         11       passe_livre       41,810         12       passelivre       37,942         13       primaverabrasileira       4,460         14       protesto       2,210,304         15       protestobh       41,651         16       protestobr       30,683         17       protestoce       10,698         18       protestoce       10,698         18       protestopelotas       1,572         20       protestopoa       18,044         21       protestopa       18,044         21       protestop       133,260         23       revogaoaumento       2,243         24       saladuprising       839         25       sp13j		Vinegar	
3         catraca_livre         1,512           4         catracalivre         1,515           5         changebrazil         285,385           6         chupadilma         28,200           7         contraoaumento         7,167           8         mudabrasil         103,614           9         obrasilacordou         12,411           10         ogiganteacordou         181,511           11         passe_livre         41,810           12         passelivre         37,942           13         primaverabrasileira         4,460           14         protesto         2,210,304           15         protestobh         41,651           16         protestobr         30,683           17         protestoce         10,698           18         protestoce         10,698           18         protestopelotas         1,572           20         protestopoa         18,044           21         protestop         133,260           23         revogaoaumento         2,243           24         saladuprising         839           25         sp13j         6,366			
4         catracalivre         1,515           5         changebrazil         285,385           6         chupadilma         28,200           7         contraoaumento         7,167           8         mudabrasil         103,614           9         obrasilacordou         12,411           10         ogiganteacordou         181,511           11         passe_livre         41,810           12         passelivre         37,942           13         primaverabrasileira         4,460           14         protesto         2,210,304           15         protestobh         41,651           16         protestobr         30,683           17         protestobr         30,683           17         protestoce         10,698           18         protestope         13,683           17         protestope         18,044           21         protestopoa         18,044           21         protestorj         114,557           22         protestosp         133,260           23         revogaoaumento         2,243           24         saladuprising         839			
5         changebrazil         285,385           6         chupadilma         28,200           7         contraoaumento         7,167           8         mudabrasil         103,614           9         obrasilacordou         12,411           10         ogiganteacordou         181,511           11         passe_livre         41,810           12         passelivre         37,942           13         primaverabrasileira         4,460           14         protesto         2,210,304           15         protestobh         41,651           16         protestobr         30,683           17         protestoce         10,698           18         protestoce         10,698           18         protestopelotas         1,572           20         protestopoa         18,044           21         protestosp         133,260           23         revogaoaumento         2,243           24         saladuprising         839           25         sp13j         6,366           26         sp17j         953           27         tarifa         455,756           28			
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#### **Notes**

- 1. http://www.bbc.com/news/world-latin-america-22992410
- 2. There are, however, qualitative approaches to content analysis.
- 3. All authors of the article are fluent in Brazilian Portuguese.
- During the protests, users from the street urged people through social media to go to the windows of the buildings nearby to show support to the protest.
- Antonio Carlos is an avenue of Belo Horizonte, the capital of the Brazilian state Minas Gerais.
- 6. Raposo Tavares is a motorway in the city of Sao Paulo.

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