## Environmental disclosure and sentiment analysis: state of the art and opportunities for public-sector organisations

Disclosure and sentiment analysis

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#### Abstract

Purpose – Because of the expansion of the internet and Web 2.0 phenomenon, new challenges are emerging in the disclosure practises adopted by organisations in the public-sector. This study aims to examine local governments' (LGOs) use of social media (SM) in disclosing environmental actions/plans/information as a new way to improve accountability to citizens to obtain organisational legitimacy and the related sentiment of citizens' judgements.

**Design/methodology/approach** — This paper analyses the content of 39 Italian LGOs' public pages on Facebook. After the distinction between five classes of environmental issues (air, water, energy, waste and territory), an initial study is performed to detect possible sub-topics applying latent Dirichlet allocation. Having a list of posts related to specific environmental themes, the researchers computed the sentiment of citizens' comments. To measure sentiment, two different approaches were implemented: one based on a lexicon dictionary and the other based on convolutional neural networks.

**Findings** – Facebook is used by LGOs to disclose environmental issues, focussing on their main interest in obtaining organisational legitimacy, and the analysis shows an increasing impact of Web 2.0 in the direct interaction of LGOs with citizens. On the other hand, there is a clear divergence of interest on environmental topics between LGOs and citizens in a dialogic accountability framework.

**Practical implications** – Sentiment analysis (SA) could be used by politicians, but also by managers/ entrepreneurs in the business sector, to analyse stakeholders' judgements of their communications/ actions and plans on corporate social responsibility. This tool gives a result on time (i.e. not months or years after, as for the reporting system). It is cheaper than a survey and allows a first "photograph" of stakeholders' sentiment. It can also be a useful tool for supporting, developing and improving environmental reporting.

Originality/value — To the best of the authors' knowledge, this paper is one of the first to apply SA to environmental disclosure via SM in the public sphere. The study links modern techniques in natural language



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Meditari Accountancy Research Vol. 29 No. 3, 2021 pp. 617-646 Emerald Publishing Limited 2049-372X DOI 10.1108/MEDAR-09-2019-0563 processing and machine learning with the important aspects of environmental communication between LGOs and citizens.

**Keywords** Facebook, Sentiment analysis, Natural language processing, Local governments, Environment, Organisational legitimacy, Dialogical accountability

Paper type Research paper

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#### 1. Introduction

New challenges emerge in communication practises in the public context: posts, messages and comments on social media (SM) platforms tell the story of everyday political and civil-society life. In recent years, there has been an explosion of freely opinionated texts. Web 2.0, blogs and microblogs have changed the way people communicate, moving from passive internet usage to an active approach with content creation. In 2018, the world population accounted for around 7 billion people, and the number of active SM (Facebook, Twitter and Instagram) users was around 3 billion; thus, around 43% of the active world population were using internet and Web 2.0 platforms to communicate.

The impact of this novel way of communicating for general purposes and also environmental fields determines the grow of researches measuring the sentiment included in the messages leading to the so-calling "Sentiment Analysis" (SA). In particular, this study analyses the use of SM for environmental disclosure by Italian local governments (LGOs). The reason for this choice lies in a previous study on the diffusion of sustainability reporting by Italian LGOs (Giacomini *et al.*, 2018), demonstrating that sustainability reporting practises among LGOs are declining. In 2009, 11% of Italian municipalities published a sustainability report, in 2013 only 4%. The main motivation declared by LGOs was the search for cost reductions that represented a barrier to sustainability report preparation in terms of time and staff availability (Giacomini *et al.*, 2018, p. 14). Linking these findings with the increasing use of SM platforms leads to the recognition that new channels of communication have to be studied to understand the potential for new forms of environmental accountability (Russell *et al.*, 2017).

We chose to study LGOs because their level of government is closest to citizens' daily lives. LGOs' actions and plans have a direct impact on territories and people, and they also play a critical role in educating the local community to promote sustainability in their local area (Agenda 21 Action Plan, 1992). The concept of sustainable development as a "process of change in which the exploitation of resources, the direction of investments, the orientation of technological development and institutional change are all in harmony and enhance both current and future potential to meet human needs and aspirations" (Brundtland Report, 1987, p. 30) reveals a government's commitment to foster this change. Thus, LGOs play a crucial role and deserve more attention.

We chose Italy as an empirical setting for several reasons. Firstly, Italy has a Facebook usage rate in line with that of the main Western countries, and the Italian legal system provides numerous environmental functions for municipalities. Secondly, the number of large Italian municipalities is high, which favours a significant collection of data (Agostino, 2013). Thirdly, the authors are more familiar with the identification of Italian keywords in the environmental field, which is a relevant issue in textual analysis. Finally, there are already some studies on the use of Facebook in LGOs, confirming the validity of the empirical setting (Gesuele, 2016; Agostino and Arnaboldi, 2016; Manetti and Bellucci, 2016; Agostino, 2013).

Our focus on environmental matters is twofold. Firstly, LGOs have precise responsibilities in Italy, from waste disposal to air pollution control, to the decision to create new landfills; their awareness of environmental issues is therefore relevant. Secondly, environmental sustainability (ES) is seen as perhaps the most paradigmatic societal challenge requiring citizens' co-productive efforts to obtain a successful implementation of environmental policies (Bremer and Meisch, 2017)

as involvement of citizens is a key component of the European Union's 7th Environmental Action Programme (European Union, 2014).

In this context, SM, compared to traditional disclosure systems, might be seen as a powerful tool that gives citizens the opportunity to participate in governance practises (Mergel, 2012). Following Mergel (2013), governments can exploit SM technologies to increase transparency, support inter- and intra-organisational collaboration and enable innovative forms of public participation and engagement. The use and support of SM platforms in the construction of dialogic accounting systems is the framework used to analyse the interactions between LGO disclosure and citizens' participation or reactions (Manetti and Bellucci, 2016; Manetti et al., 2016; Bellucci and Manetti, 2017).

Recent research focuses on the new concept of *e*-government, analysing the interaction between government and citizens on SM at different levels, such as political parties, national and LGOs, with respect to different topics (see Section 2). However, as far as we know, no study has been conducted in investigating the interaction between citizens and LGOs on environmental disclosure via SM by exploring the sentiment of these communications.

The analysis was conducted by analysing Italian LGOs' official pages on Facebook and the relative interaction with the public. The methodological technique is SA, which involves the study of the emotions contained in a textual message. This can be a very powerful tool for different fields such as marketing campaigns (Rambocas *et al.*, 2013) or event detection (Sakaki *et al.*, 2010).

Our research aims to understand the potential contributions of an SM platform in the construction of a dialogic accountability system and the underlying purpose of that approach. As far as we know, this is the first time SA has been applied to public organisations to study the interests and emotions of citizens towards LGOs' disclosure on environmental issues. This contributes to the call for studies on new challenges in disclosing systems (Lodhia, 2018; Parvez et al., 2018; Russell et al., 2017; Lai and Stacchezzini, 2019; Suárez-Rico et al., 2019; Giacomini et al., 2020).

The paper is structured as follows: Section 2 briefly reviews the main literature on SM in the public sector and describes the theoretical background. In Section 3, the paper defines the steps of the analysis and the models applied. Section 4 highlights the main findings of this study. Section 5 offers discussions of the results, and Section 6 concludes the paper.

#### 2. Theoretical background

The section is divided into two paragraphs: the first illustrates a literature review on the Italian Governments' SM communications and interactions, and the second presents the theoretical framework applied in this paper.

#### 2.1 Literature review on government social media interactions

Several investigations were conducted to evaluate the online communications of sample governments (Agostino, 2013; Hofmann *et al.*, 2013; Reilly and Hynan, 2014; Zavattaro *et al.*, 2015; Bonsòn *et al.*, 2015; Bellstrom *et al.*, 2016; Reddick *et al.*, 2017; Haro-de Rosario *et al.*, 2018). Table 1 summarises previous work that analysed government–citizen interaction on SM. As Table 1 shows, a wide range of analyses was conducted around the globe, either focussing on the full list of LGOs in a specific country (Sandoval-Almazan and Gil-Garcia, 2012), government departments (Unsworth and Townes, 2012) or government institutions from specific cities (Wirtz *et al.*, 2018).

However, as argued by Reddick *et al.* (2017), a lack of research on the potential role of public feedback still exists. *E-participation* is defined as citizens' voluntary participation in public administration through online platforms, which encourages two-way communication between governments and its subjects (Reddick *et al.*, 2017).

A recent study proposed by Bonson *et al.* (2015) highlights that regarding the public's e-participation, Facebook shows a higher level of citizens' engagement with local government, especially when resident's posts are promoted. In fact, as Table 1 shows, most of the work on citizens and government interactions used Facebook as a data source because it provides the opportunity to examine direct government disclosure as well as member reactions in relation to the specific item. Another widely used SM platform is Twitter (Alam and Lucas, 2011; Hubert *et al.*, 2018); however, some researchers have conducted analyses merging various information sources. For instance, Agostino (2013), Ellison and Hardey (2014) and Wirtz *et al.* (2018) also include data from YouTube and Flickr.

Mergel (2012) proposed a set of three features to evaluate these interactions in the public sector: transparency, participation and collaboration. Transparency in SM can be obtained by a one-way communication style that pushes information in the name of education. Some metrics to evaluate transparency are posts' likes, for example. Participation involves two-way communication encouraging SM visitors to frequent the government's website. Metrics to evaluate participation involve click-through, page views and number of comments. Collaboration is related to citizens' content co-creation, and is then based on direct interaction with governments. The public should be able to see that their input has influenced government policies. A way to measure collaboration involves analysing shares and comments. The work of Zavattaro et al. (2015), for example, analysed the aforementioned Mergel's points in US local government entities via Twitter.

Initial studies (Sandoval-Almazan and Gil-Garcia, 2012; Ellison and Hardey, 2014) performed survey analysis to ask LGOs directly about the use and importance of SM for citizens' participation, while more recent researchers collected data from the Web and mostly applied descriptive statistics to analyse the interactions (Reilly and Hynan, 2014; Bellstrom *et al.*, 2016). Few works adopt textual analysis and topic modelling (Arunachalam and Sarkar, 2013; Bonsòn *et al.*, 2015), while different researchers also investigated the public's sentiments (Hofmann *et al.*, 2013; Zavattaro *et al.*, 2015; Hubert *et al.*, 2018; Durahim and Coşkun, 2015; Arunachalam and Sarkar, 2013; Naiknaware *et al.*, 2017). The literature review (Table 1) also reports research proposed by Reilly and Hynan (2014), who explored sustainability topics on SM, but with respect to corporations and not to government. Reddick *et al.* (2017) focussed on textual data related to the Solid Waste Management Department of San Antonio City, Texas.

To the best of our knowledge, ours is the first work that analyses disclosure of environmental matters on SM by LGOs.

### 2.2 Government-citizen interaction on social media: an organisational legitimacy perspective or dialogic accountability?

The SM interactions between LGOs and citizens on sustainability issues is a nearly new field of study, as the previous paragraph demonstrates, and the most common theoretical background applied in literature is the organisational legitimacy perspective (Etter *et al.*, 2018; Knox, 2016).

One-way communication between LGOs and citizens on sustainability issues is represented by a voluntary disclosure that traditionally has sustainability reporting as the outcome. The literature recognises the legitimacy concept as the theoretical background to explain public organisations' choice to disclose sustainability issues (Dowling and Pfeffer, 1975; Navarro Galera *et al.*, 2014), where "organisational legitimacy is generally defined as the social acceptance of organisations and their actions" (Etter *et al.*, 2018, p. 61). The process of constructing organisational legitimacy could be seen as a path by which an organisation communicates with its evaluators to meet their expectations in terms of standards and norms. However, this process cannot be considered in a one-way linear

Study	Language <sup>a</sup>	Data source <sup>b</sup>	$Target^c$	Citizens interactions	LGO activity	Model <sup>d</sup>	SA	Data year	Data size <sup>e</sup>
Alam et al.	ENG	TW	CG Int	X	X	DA	I	ı	709 Docs
Hand et al.	ENG	FB	CG Int	×	×	DA	I	2010	2116 Docs
Kavanaugh et al.	ENG	FB, TW	Egov	ı	×	DS, TA, IW	I	2010	34 Acc
Sandoval-Almazan et al.	$_{ m SP}$	Website	CG Int	X	×	Survey	I	2009	$108 \mathrm{Acc}$
Unsworth et al.	ENG	TW	Egov	I	X	NA	I	I	I
Agostino	П	FB, TW, YT	Egov	I	X	DS	I	2012	119 GO
Arunachalam et al.	ENG	TW, FB, FL, YT	C-G Int	×	ı	TA, TM	×	2012	41K DOC
Hofmann et al.	DE	FB	CG Int	X	×	MM-A	×	2012	35K Docs
Ellison et al.	ENG	TW,FB, FL, YT	Egov	I	X	Survey	I	2011	$352 \mathrm{Acc}$
Reilly et al.	ENG	FB, TW	CSR	X	I	DS	I	I	16 Docs
Bonson et al.	I	FB	Egov	I	X	TM, DS	I	2013	3K Docs
Durahim et al.	TR	TW	WB	X	I	SS	×	2013	500K Docs
Zavatarro et al.	ENG	TW	CG Int	X	X	SMTAS	×	2013	17K Docs
Bellstrom et al.	SWE	FB	CG Int	X	X	DS	I	2015	$110  \mathrm{Docs}$
Naiknaware et al.	ENG	TW	CG Int	X	I	$_{ m SA}$	×	2016	60K Docs
Reddick et al.	ENG	FB	CG Int	X	X	DLSA	I	2012 - 14	620 Docs
Hubert et al.	$_{ m SP}$	TW	CG Int	X	X	VA	×	2017	12K Docs
Haro-de Rosario et al.	$_{ m SP}$	FB, TW	CG Int	X	X	$_{ m CM}$	I	2014	18K Docs
Wirtz $et al.$	ENG	*	CG Int	X	X	SM4F	I	I	$454 \mathrm{Acc}$
This work	П	FB	St CG Int	×	×	LDA, SA	×	2018	20K Docs

media four-factor model; SMTAS: social media analysis and tracking system; SS: SentiStrength, TA: text analysis; TM: topic modelling; and VA: visual analysis government interactions, and WB: well-being <sup>4</sup>Model – CM: corporate metrics; DA: data analysis, DLSA: double-loop learning thought social media; DS: descriptive analysis; IW: interview; LDA: latent Dirichlet allocation; MM-SA: multi-method analysis; NA: Network Analysis; SA: sentiment analysis; SM4F: social "Data Size – Acc. accounts, and Docs: documents" The authors analysed several SMs such as: Sina Weibo, Tumblr, Twitter, Flickr, Instagram, YouTube, Notes: <sup>a</sup>Language – DE: German, ENG: English; IT: Italian; SP: Spanish; and TR: Turkish <sup>b</sup>Data Source – FB: Facebook; FL: Flirck; YT: YouTube; and TW: Twitter Target - CG Int: citizens-government interactions; CSR: corporate social responsibility; Egov: e-governments; St CG Int: sustainable citizens-Facebook, Foursquare, Google+, LinkedIn and Pinterest

Table 1. Summary of related works on governments' SM interactions

direction because it compromises various "sources of legitimacy – evaluators" (Etter *et al.*, 2018, p. 63) that could have different criteria and perspectives.

Thus, the approach of disclosing "something" to "someone" in a monological way has to be compared to a dialogic accountability system (Brown, 2009; Brown and Dillard, 2013a, 2013b; Manetti and Bellucci, 2016; Bellucci and Manetti, 2017), where "the rights and responsibilities of the constellation of constituencies...affected by an account provider's actions" (Dillard and Vinnari, 2019) are taken as starting points. In the dialogic process, the identification of who these constituencies are, and their interests and needs for information, comes not from the organisation but is a recognition from outside towards the institution. In other words, it could be seen as a sort of "networking engagement" which is different from the stakeholder engagement that is focused from the organisation to the outside, but it is like a civil society request for information towards the entity's actions. In that scenario, there could be different methods and tools that enhance dialogical approaches, as shared platforms, "scenario workshops, deliberative mapping, multicriteria analysis, open space technologies, Q methodology and dissensus conferences" (Bellucci et al., 2019 citing Brown and Dillard, 2014).

Platforms could be seen as one of the tools supporting the dialogical accountability system (Bellucci and Manetti, 2017). Thus, with the advent of SM platforms, citizens have the chance to express judgments/opinions/needs/interests about organisations online (Blankespoor, 2018). This opportunity for comment creates a shared platform where the actions, plans and information of the institutions are under scrutiny in a networked perspective. Through SM "the voices of those most negatively affected by corporate – organisations - activity – that - are notably absent" (Everett, 2004, p. 1079) could express their sentiment in the social context, otherwise not engaged in the process, or the "judgements in social media can be considered as contributing to the coconstruction of organizational legitimacy" (Etter et al., 2018, p. 64). The freedom of writing to and answering posts on SM platforms means, in other words, that everyone can write or post his/her point of view, perceptions, reactions without restrictions and it might be seen as a "potential democratization of online arenas" (Etter et al., 2018, p. 68). The consideration of SM as a potential public arena also requires underlining and considering the limits that SM has. First of all comes the circulation and diffusion of fake news (Allcott and Gentzkow, 2017; Shu et al., 2017; Srauy, 2015). Furthermore, SM is primarily a product designed to monetize users' activities, which can affect the visibility and dissemination of content, and is therefore not impartial (Gillespie, 2014). The same problem does not only concern SM but almost all cyberspaces (Dahlberg, 2005). For example, fake news could influence the virtual debate on a certain issue, making the results of the discussion unreliable as a support for dialogic accountability. In other cases, discussion policies on SM can inhibit free discourse on a relevant topic.

The dialogical accountability approach could have two purposes:

- The first is a deliberate and legitimate view where "stakeholder engagement is necessary for defining the general consensus among diverse stakeholders or inside a specific category" (Bellucci and Manetti, 2017).
- (2) The second is an agonistic view where pluralistic needs are taken into account as various requests for information, dealing with each differently.

The first approach links legitimacy to a rational argument in a public sphere based on Habermas' "ideal speech situation" (Habermas, 1989). This means that SM platforms could have the characteristics of Habermas' "ideal public sphere" because they have the "the potential to increase the symmetrical communication" (Knox, 2016, p. 483) thanks to:

sentiment

Disclosure and

- the equal opportunity to communicate;
- the same possibility to start a new dialogue, to continue an old one or to express
  oppositions, explanations, etc.; and
- also to inform sentiments or emotions (i.e. likes and hates).

Moreover, the communication is bidirectional, instant or not, and it is transparent that SM is a place for sharing knowledge (Mergel, 2012). In Habermas' ideal of authentic communication, SM characteristics could create a "communicative rational justification of validity claims by public administrators to citizens, unlike rhetoric" (Knox, 2016, p. 488).

The first purpose of dialogic capabilities is to create a convergent consensus over actions, plans and projects throughout a transparent and constructive discussion in a public sphere. Organisational legitimacy could be seen as the outcome of this process. Otherwise, the second purpose of dialogic accountability is focused on an agonistic model of participation, called "agonistic dialogic accounting" by Brown (2009), Dillard and Yuthas (2013) and Brown and Dillard (2013a, 2013b).

Mouffe's agonistic pluralism theoretical frame describes a pluralistic arena where stakeholder engagement is the collection of various standpoints that public organisations/politics must take into account in their plans from an agonistic, not antagonistic, perspective (Mouffe, 1999, 2000). This means that citizens with divergent opinions or values are not enemies but adversaries, as Mouffe wrote in 2000:

[...] modern democracy's specificity lies in the recognition and legitimation of conflict and the refusal to suppress it by imposing an authoritarian order. This requires providing channels through which collective passions will be given ways to express themselves over issues, which, while allowing enough possibility for identification, will not construct the opponent as an enemy but as an adversary.

SM could have the potential to be one of the possible channels that the public could use to gather this different sentiment, as SM platforms could give the opportunity to express different and contrasting opinions and emotions to a public and wide arena, even if they cannot be considered a "perfect" one for agonistic pluralism because of the limitations mentioned above (fake news, private nature, digital divide, etc.). The agonistic scenario aims to create a wider engagement between parties that results in the reasoning of means and ends in a democratic decision-making process (Bellucci and Manetti, 2017; Vinnari and Dillard, 2016).

The second scenario considers the different "passions" and "to mobilize those passions towards democratic designs" (Mouffe, 2000), while the first aims to take into account the sentiments but, strictly, not incorporate them in the systems' strategies.

Finally, both purposes have a dialogical approach where the engagement and consideration of citizens, community or even non-experts seems to be as relevant as the information and facts disclosed (Brown and Dillard, 2013a, 2013b). In fact, literature calls for new accounting processes that consider the "polyvocal citizenship perspective" (Gray, 1997), the *pluralistic nature of contemporary society* (Bellucci *et al.*, 2019, Dillar and Ruchan, 2005) and for a more constructionist approach in the social accounting system (Thomson and Bebbington, 2005) that to "engender . . . . More importantly, societally relevant and meaningful solutions" (Dillard and Vinnari, 2019).

In accordance, this paper aims to contribute to the debate on dialogical accountability using SA applied to public organisations to understand the public's interests and emotions towards LGOs' disclosure of environmental issues, to explore the potential contribution of SM platforms in the construction of a dialogic accountability system and the underlying purpose of that approach.

#### 3. Methods

To study the interactions between LGOs and citizens with regard to ES disclosure via SM, we applied some well-known techniques in the field of natural language processing (NLP) (Manning *et al.*, 1999) and text analysis (Zhai and Massung, 2016). The section is composed of two sub-sections. In Section 3.1, we describe how LGOs' posts were managed, while in Section 3.2, we focus on the analysis of the comments. Figure 1 describes the mind map of the approach followed in this paper.

#### 3.1 Analysis of local governments' Facebook posts

The first step of the analysis consists of evaluating LGOs' SM posts. The authors identified five different classes of environmental issues in the context of Italian municipalities: air, energy, territory, water and waste. Appendix 1 explains the classification process, sources and list of keywords used to identify SM posts for each of the five groups.

Starting from the raw Facebook posts, the authors cleaned the text message of numbers, stop words, punctuation and Uniform Resource Locator (URL) references. The text preprocessing was performed using the *nlth* module in Python. Then, to obtain an overview of the posts content, we analysed the word frequency. Thus, for each word in a document, the term frequency is computed as follows:

$$tf_{i,j} = \frac{n_{i,j}}{|dj|}$$

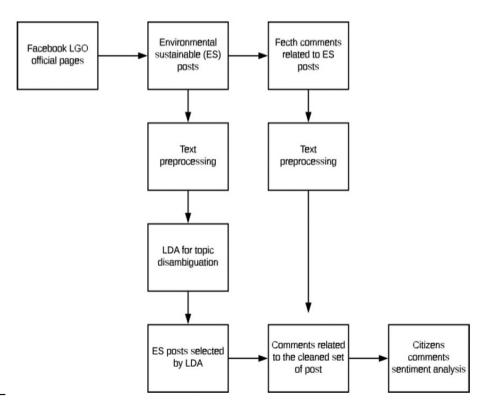


Figure 1. Methodology approach

where  $n_{i,j}$  is the number of occurrences of a word i in the document j, and  $d_j$  is the size of the document j expressed as the total number of words in the document. Analysis of word distribution might be seen as a basic step, but it is useful to understand which topics are included in the document and if there are any out-of-topic posts. In this work, it was crucial to understand whether misleading topics were found in the five classes. To remove posts and relative comments not strictly related to our analysis, we performed topic detection by applying the well-known latent Dirichlet allocation (LDA).

LDA is one of the most common algorithms for topic modelling. It is a statistical method to find a mixture of words associated with each topic, and determine the mixture of topics that describe each document. Given the following terms (Blei *et al.*, 2003):

- a word w is the basic unit of text data, indexed by 1, . . ., V;
- a *document* is a sequence of N words denoted by  $w = (w_1, w_2, ..., w_N)$ , where  $w_N$  is the n-th word in the sequence; and
- a *corpus* is a collection of M documents denoted by  $D = w_1, w_2, \dots, w_M$ .

It is common to model each document w as a distribution  $\Theta(w)$  over K topics, and each topic  $z_j, j = 1, \ldots, K$  as a multinomial distribution  $\Phi(j)$  over the set of words W. To discover the set of topics in each document in a corpus D, it is necessary to estimate  $\Phi$  and  $\Theta$ . LDA assumes the following generative process for each document w in a corpus D (Blei et al, 2003):

- Choose  $N-Poisson(\beta)$ ; and
- Choose  $\Theta Dir(\alpha)$ .

For each of the N words  $w_N$ :

- choose a topic  $zn \sim multinomial(\Theta)$ ; and
- choose a word  $w_N$  from  $p(w_N|z_n,\beta)$ , a multinomial probability conditioned on topic  $z_n$ .

In this work, before implementing LDA, the text was stemmed. Stemming is an NLP technique used to reconduct words to their roots by removing affixes (prefixes and suffixes) (Manning *et al.*, 1999). Stemming was implemented using the *nltk* module available on Python to perform the Snowball Stemmer (Porter, 2001).

The only parameter required to compute LDA is the number of topics k. To identify k, we used the FindTopicNumber function implemented in the R package datuning. To determine the best number of topic k for each sustainable class i, the authors evaluated two metrics for  $k = 1, \ldots, \sqrt{number\ of\ posts\ for\ the\ i-th\ class}$ . The metrics used to evaluate the best value for k are:

- Griffiths and Steyvers (2004) metric: The authors evaluated the consequences of changing
  the number of topics k using the Gibbs sampling algorithm to obtain samples from the
  posterior distribution over z at several choices of k (minimisation criterion).
- Arun et al. (2010) metric: The measure was computed in terms of symmetric KL-divergence of salient distributions derived from the matrix factor, and it was observed that the divergence values are higher for a non-optimal number of topics (maximisation criterion).

Once the LDA has been implemented for each environmental class (air, energy, territory, water and waste), the authors dropped the topics not strictly related to the research purpose. Thus, the authors identified the final sample of LGOs' posts and their relative comments by discarding the misleading ones.

3.2 Citizens' comments analysis

The method applied to study citizens' comments and interactions with LGOs is SA. It is also called opinion mining and uses computational methods to automatically analyse human opinions, sentiments and evaluations of entities such as products, services and organisations (Liu, 2012). SA typically focuses on one specific domain at a time, such as hotels (Shi and Li, 2011), movies (Tripathy *et al.*, 2017) or financial markets (Krishnamoorthy, 2018).

There are two main sentiment classification methods: lexicon and machine learning (ML)-based. A lexicon is a special dictionary in which words are assigned to sentiment scores (Ghosh and Kar, 2013; Kumar et al., 2016). The main advantage is that, once a lexicon is built, a fast unsupervised sentiment classification is achieved by summing the overall word scores. Thus, there is no need for labelled data, Moreover, in some cases, specific adhoc lexicons are created for a target domain (Grabner et al., 2012). However, lexicons tend to produce lower performances when compared with supervised machine-learning approaches. Thus, ML is widely used for SA (Pang et al., 2002; Salvetti et al., 2004; Pouransari and Ghili, 2014). Sentiment classification studies initially explored simpler feature engineering (e.g. Ngrams or bag-of-words) and machine-learning algorithms (e.g. naive Bayes [NB] and support vector machines [SVMs]). After 2014, recent text classification advances, such as word embedding and deep learning (Goodfellow et al., 2016), were naturally incorporated into SA works (Ortigosa et al., 2014; Lai et al., 2015; Conneau et al., 2017). Recently, Zola et al. (2019) proposed a cross-source cross-domain sentiment classification method based on a domainadaptation approach which aims to classify the sentiment of unlabelled data from Web 2.0, passing through other already labelled data from Web sources.

Table 2 reports a brief summary of previous works in sentiment classification considering different domains from websites to social networks. The columns: L – lemmatisation, S – stemming and P – part-of-speech (POS) tagging identify specific type of text pre-processing commonly used in NLP.

This paper describes SA of citizens' Facebook comments. In comparison to other SM, Facebook comments do not have any limits on the number of characters. Moreover, as mentioned above in the Introduction, a wide range of analysis for LGO–citizen interactions has previously focussed on Facebook, which is, to date, the SM platform with the highest number of daily active accounts, and thus the SM with the highest worldwide coverage (www.statista.com/statistics/272014/global-social-networks-ranked-by-number-of-users/).

However, people tend to use a lot of slang (Golia and Zola, 2019) on blogs and microblogs. This freedom in writing involves many problems for sentiment computation. In this work, the authors decided to apply two SA approaches: lexicon database (LD) and ML. The two approaches were tested separately to propose a double sentiment verification overcoming the drawbacks of the two methods reported in Section 3.2. In fact, if LD approaches are fast, they are not able to catch the so-called cyber-slang, while ML methods of computation are more expensive but offer greater flexibility.

Before implementing any SA, the comments text was first cleaned to remove stop-words and URLs as previously done for the LGOs' posts. Moreover, the authors performed a POS tagging to identify and remove comments that are composed only of "tags". In SM, tags identify a "call", where a user *a* "calls" user *b* or *n* users, therefore using a POS tagger we can identify comments including only proper nouns and remove them, as a tag does not involve any kind of feeling/sentiment. To perform the POS tag, the authors used the *RDRPOSTagger* (Nguyen *et al.*, 2014) library developed in the R software. The *RDRPOSTagger* supports Italian language and is faster in tagging than other POS taggers, such as *Treetagger* (Schmid, 2013), available in Python. Once the comments have been cleaned, the study also distinguishes

Study	Language <sup>a</sup>	Data <sup>D</sup>	J	S	Ь	L S P Algorithm <sup>c</sup>	Nr classes	Data size
Pang et al.		IMDb	1	- 1	×	N.gram+NB, N.gram+SVM, N.gram+ME	3	16K rev
Blitzer et al.		Am	I	1	1	TOS	2	8K rev
Bolleaga et al.		Am	×	1	×	FE+L1LR	2	76K
Ortigiosa et al.		FB	I	1	×	L, NB, 148, SVM	2	3K
Dos Santos and Gatti	ENG	IMDb, TW	I	1	1	WE+CNN, CE+CNN	2	12K, 80 K
Kumar et al.		Am	I	1	1	NB,LR,SW	2	ı
Tripathy et al.		IMDb	I	1	1	N-gram+NB, N-gram+SVM, N-gram+ME, N-gram SGD	2	50K
Conneau et al.	щ	* various	I	I	I	Ce+VDCNN	2-14	11M
Zola et al.	ENG, IT	Am, FB, TR, TW	I	×	×	We+NB, We+SVM, We+MLP, We+CNN	2,3	1.3M

TR: Tripadvisor reviews; and TW: Twitter 'SA method - Ce: character embedding: CNN: convolutional neural network; FE: feature extraction; J48: decision tree. L. lexicon information; L1LR: L1 regularized logistic regression; LR: logistic regression; ME: maximum-entropy; MI: mutual information; NB: naive Bayes; SGD: stochastic gradient descent; SVM: support vector machine; SCL: structural correspondence learning; SW: SentiWordNet (Baccianella et al., 2010); VDCNN: very deep convolutional neural network; W2V: word to vec; and We: word embedding The authors used different sources: news data, DBPedia, Yelp reviews, Yahoo Answer and Amazon reviews Notes: "Language – ENG: English; CH: Chinese; IT: Italian; and SP: Spanish Data source type – Am: Amazon reviews; FB: Facebook; IMDb: movies reviews;

**Table 2.** Summary of related works on SA

between citizens' comments and LGOs replies to stakeholders. The sentiment is then computed only with respect to citizens' comments.

Lexicon database. The LD used in this analysis is the *mrc* lexicon proposed by Mohammad and Turney (2010). They created the word-emotion association lexicon through Amazon Mechanical Turk. The initial *mc* lexicon was developed for English language and later extended to other languages such as Italian. Mohammad and Turney (2010) used *Rogets Thesaurus* as the source for terms, annotating 14,000 unique English words. The words are then separated into eight categories corresponding to eight emotional and semantic clusters denoted as joy–trust, sadness–anger, surprise–fear and anticipation–disgust, composed of four opposite pairs. Eight features under the *mrc* lexicon are joy, sadness, anger, fear, trust, disgust, surprise and anticipation. The authors implemented an LD SA using the *syuzhet* package in R software.

Machine learning: There are a large number of SA models based on ML or deep-learning techniques. In our research, we performed the analysis applying the Zola et al. (2019) algorithm, which is mainly based on a domain adaptation logic in which sentiment from well-known Web sources is useful for defining the feelings within SM sentences. The authors investigated four ML models: NB, SVM, deep multilayer perceptron (MLP) and convolutional neural networks (CNNs) for sentiment classification of SM messages on Facebook and Twitter. The best model is based on a CNN where the textual strings are handled by word embedding (Mikolov et al., 2013). For the Italian case, before applying the word-embedding, the text was stemmed (Manning et al., 1999). An example of stemming in Italian is the word "bellissima" that is synthetized by its root "bell". The sentiment derived is both binary (negative and positive) and multiclass (negative, neutral and positive), and the original research has been conducted both in English and Italian. We decided to apply Zola et al.'s (2019) algorithm because the data are from Facebook Italian pages and, in the original paper, the results pertinent to Facebook comments in Italian showed good performance.

#### 4. Findings

#### 4.1 Data samble

The sample of Italian LGOs evaluated is reported in Table 3. The study analysed the Italian municipalities with a population of more than 100,000. However, not all Italian municipalities with more than 100,000 citizens are in the sample, as some cities do not have an official Facebook page. As reported in Section 3.1, the authors identified five environmental topics among the overall LGOs posts (Appendix 1).

The textual data in this work are directly downloaded from the Facebook LGOs' public pages using Python script. For each LGOs' page, the authors recorded information about the posts written by the LGOs as well as the likes, shares and comments expressed by citizens. The period varied according to the LGOs' SM activity. The ending date for every LGO is the end of July 2018. Table 3 reports the municipalities included in the sample and for each class date of the oldest post and the number of posts published between the respective date and 30 July, 2018. The total number of posts considered in the analysis was 5,939.

#### 4.2 Results

As described in Section 3.1, the authors started from the distribution of words for each of the five ES classes. This step is fundamental to understanding if there are some possible out-of-topic arguments in the data. Table 4 reports the first ten words extracted by the *tf-idf* matrix for each class.

For the air sustainability class, Table 4 shows that the most frequent words are related to pollution from road traffic. The energy topic is mainly related to energy savings and renewable energy. For the territory topic, the most frequent words refer to parks and urban green areas, while for the waste class, the major attention is door-to-door recycling, which

om uly																														(pai
Water No. of post (from oldest date to July 2018)	1	I	ı	I	I	I	I	I	4	I	I	I	I	I	I	17	15	I	I	10	I	1	I	I	I	I	I	I	I	(continued)
Oldest o post*	ı	I	ı	1	ı	I	ı	I	11/6/2015	ı	ı	ı	ı	ı	ı	12/9/2016	10/29/2012	ı	ı	3/8/2014	ı	7/13/2016	ı	ı	ı	ı	ı	ı	Ι	
Waste No. of post (from oldest date to July 2018)	45	24	104	ı	154	ı	18	13	26	47	115	132	ı	196	17	174	ı	32	96	ı	81	144	358	ı	62	ı	12	ı	I	
N Oldest ol post*	2/7/2014	12/7/2015	1/15/2013	ı	5/26/2010	1	1/11/2015	10/19/2011	1/1/2013	4/14/2010	12/7/2011	2/17/2015	ı	4/20/2009	11/7/2016	10/18/2016	ı	7/16/2010	5/6/2010	ı	2/11/2014	7/21/2010	7/30/2012	ı	8/2/2010	ı	7/14/2010	ı	I	
Territory No. of post (from oldest date to July 2018)	44	82	93	I	20	I	69	127	124	37	142	119	18	16	21	22	93	13	47	45	93	12	37	101	57	152	2	17	I	
Te N Oldest of post*	1/10/2014	1/25/2016	12/20/2012	I	11/9/2009	I	11/19/2012	5/10/2011	12/21/2012	8/7/2012	11/29/2011	2/21/2015	11/15/2011	10/11/2010	11/3/2016	10/14/2016	3/22/2013	6/4/2010	11/19/2010	9/9/2011	5/25/2014	10/7/2013	3/5/2013	2/21/2015	11/29/2010	10/23/2014	11/28/2014	12/2/2014	Ι	
Energy No. of post (from oldest date to July 2018)	9	4	I	ı	က	ı	ı	I	ı	I	28	ı	ı	ı	I	ı	53	I	I	∞	6	I	I	ı	ı	I	I	ı	I	
Oldest post*	9/25/2014	2/17/2016	: I	ı	7/29/2010	ı	ı	I	ı	ı	10/31/2013	ı	ı	ı	ı	ı	10/30/2012	ı	ı	3/27/2014	5/9/2014	ı	ı	ı	ı	ı	ı	ı	I	
Air No. of post (from oldest date to July 2018)	ı	1	8	I	26	1	15	22	2	7	20	9	14	4	ı	21	18	12	30	ı	1	41	34	∞	21	∞	I	8	28	
Air No. c Oldest oldes post*	1	3/13/2017	1/19/2013	ı	7/30/2009	I	3/17/2015	6/13/2012	3/26/2013	9/30/2011	12/1/2011	10/1/2015	11/2/2010	4/16/2011	I	11/10/2016	10/30/2012	12/17/2013	5/26/2011	I	12/27/2017	6/19/2912	5/31/2013	4/1/2015	2/26/2012	12/3/2015	I	3/5/2015	12/18/2013	
s097	Ancona	Andria	Bari	Bergamo	Bologna	Bolzano	Brescia	Cagliari	Catania	Ferrara	Firenze	Foggia	Forlì	Genova													Е.		Ravenna	

**Table 3.** Municipalities analysed

(from o July													
Water No. of post (from oldest date to July 2018)	I	6	I	I	I	I	I	I	∞	I	10	5	I
Oldest o	I	1/19/2012	ı	I	ı	ı	ı	ı	9/23/2013	ı	1/8/2015	9/3/2014	I
Waste No. of post (from oldest date to July 2018)	36	13	131	1	74	138	I	ı	131	ı	177	125	133
Oldest o	2/10/2015	6/20/2014	4/6/2010	ı	3/25/2011	5/5/2011	ı	I	10/16/2009	I	9/10/2012	11/18/2009	5/20/2011
Ferritory No. of post (from oldest date to July 2018)	7	179	38	က	73	35	16	I	26	99	72	38	39
Oldest post*	12/24/2016	3/22/2011	10/20/2011	4/18/2017	6/6/2011	1/4/2012	4/3/2014	I	9/24/2009	10/22/2013	7/26/2012	1/19/2012	6/8/2011
Energy No. of post (from oldest date to July 2018)	I	47	I	92	I	I	I	ı	I	I	I	I	I
Oldest o	I	2/11/2010	I	3/8/2013	I	I	I	I	I	I	I	I	Ι
Air No. of post (from oldest date to July 2018)	I	7	23		1	က	I	ı	39	2	1	33	2
Oldest o	I	Reggio 3/9/2010 Emilia	4/6/2010	5/24/2013	4/7/2017	11/18/2011	I	I	2/11/2011	2/6/2015	3/30/2016	5/4/2012	12/28/2015
s097	Reggio Calabria	Reggio Emilia	Rimini	Roma	Salerno	Sassari	Sicuracusa	Terni	Torino	Trento	Trieste	Venezia	Vicenza

Note: \* Data are expressed in the format: day/month/year

	th	tions								lity	
Water	Englis	intervenction	street	water	sewer	building	network	children	hours	municipa	press
	Italian	lavori	via	acdna	fognario	palazzo	rete	bambini	ore	comune	stampa
Vaste	English	garbage	street	collection	door	municipality	recycling	city	hours	municipal	service
Wa	Italian	rifiuti	via	raccolta	porta	comune	differenziata	cittá	ore	comunale	servizio
Territory	English	municipality	green	environment	street	city	public	municipal	hours	park	project
	Italian	commue	verde	ambiente	via	città	pubblico	comunale	ore	parco	progetto
ergy	English	Street	City	Solar	Energetic	Municipality	Save	Trees	square	Green	New
丑	Italian	via	cittá	solare	energetico	comme	risparmio	alberi	piazza	verde	nuovi
Air	Italian English Italia	municipality	euro	aspestos	pm10	air	measures	city	vehicles	days	tomorrow
	Italian	comune	euro	amianto	pm10	aria	misure	cittá	veicoli	giorni	domani

**Table 4.** Words' distribution among sustainable classes

has involved Italian municipalities in recent years. Concluding with the water class, the LGOs' posts are more linked to warnings of work in progress and traffic problems because of water network maintenance.

These findings reveal that there are specific interests in each class. Thus, if we consider, for example, territory, it means we are talking about parks and urban green spaces. This clarifies the meaning that LGOs attribute to sustainability issues.

After evaluating the most frequent words, it was necessary to select topics among LGOs' posts, especially for classes that showed misleading words with respect to ES. Topic detection and classification were performed using LDA (see Methods and Appendix 1), which was performed for all five classes. As reported in Section 3.1, to perform LDA, it is necessary to define the number of hidden topics k. To define the best k value for each class, we performed the Griffiths and Steyvers (2004) and Arun *et al.* (2010) metrics. The results of the two metrics and of the entire LDA process are reported in Appendices 2 and 3.

The LDA allowed out-of-topic posts to be discarded. After these re-elaborations, it was possible to identify the comments of the citizens related only to strictly environmental posts. Before passing through the comment's analysis, it was necessary to distinguish the citizens' messages from the LGOs' replies. Figure 2 reports the frequency of LGOs' posts related to ES issues (without the out-of-topic ones), the distribution of the correspondent citizens' comments and the LGOs' replies. The percentage of citizens' comments is computed proportionally to the amount of LGOs' posts for each class, and the LGOs' reply distribution is computed proportionally to the number of citizens' comments.

These findings reveal the real SM interactions between LGOs and citizens towards each sustainability class. The contribution clearly identifies the direct interactions among the five classes and provides an opportunity to understand if there is a relation, and its frequency. After investigating the degree of interaction between LGOs and citizens on SM, we studied how the public reacts; in other words, what is the sentiment of residents regarding their LGO's disclosure on ES?

As reported in Section 3, SA was performed using two different approaches:

- LD: Use of the nrc lexicon that supports Italian language.
- ML: Use of the domain-adaptation algorithm proposed in Zola et al. (2019).

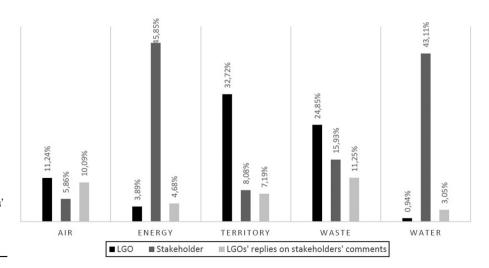


Figure 2.
Frequency
distribution of LGOs'
cleaned posts,
citizens' comments
and LGOs' replies

The combined use of these two approaches (the details of the feelings detected are in the Appendix) led to the quantification of public sentiment towards environmental disclosure in three macro-categories (negative, neutral and positive), as reported in Figure 3.

The prevailing sentiment is negative: for every environmental class, more than 50% of comments express a feeling of dissatisfaction. The lower value for positive emotion is related to the air class (19%), which is also the class with less public interaction (Figure 3). In contrast, the energy class has a greater number of positive comments (38%), and is the environmental class in which stakeholders are more active on Facebook. SA contributes to the evaluation of citizens emotions in SM interactions with LGOs and reveals the public's attitudes towards each sustainability class. That approach contributes to study "sentiments" of overall Italian citizens towards each municipality, as a big database.

#### 5. Discussion of findings

The first result is that LGOs disclose environmental topics using Facebook and they communicate mostly on territory and waste matters. Thus, they decided to use a social platform to explain their environmental efforts towards a green municipality with parks and public projects, and they would like to disclose more about waste collection and recycling issues. The second finding is that stakeholders are interested in energy and water matters instead; these two classes have the lowest frequency distribution for LGOs.

The choice of using a SM platform to discuss environmental issues – that are not compulsory even in the regulation system with very few voluntary sustainability reports (Giacomini *et al.*, 2018) – could be seen as a chance to talk to citizens on the subject in a shared "public sphere". As stated in the theoretical framework, the SM platform could be a "virtual" place where citizens interact and express their opinions towards LGOs in a networked perspective. This means that the LGOs' decision to create a Facebook public page and thereupon account for environmental facts could be seen as a first attempt to make a dialogical accountability approach because through SM, potentially every citizen could be part of the online arena. In other words, the publics' participation in those arenas is not called by LGOs – they are not specifically engaged in

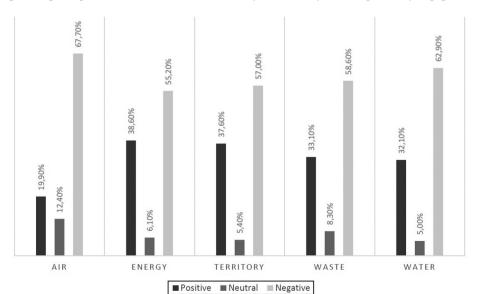


Figure 3.
SA by cross source cross domain CNN
model

the process of accountability by LGOs, but from outside, they express their interests and needs for information. According to the dialogic process, citizens could ask for information about the LGO's actions, directly and in instant or non-instant time.

In our findings, it is clear that stakeholders' comments are a request for information about energy and water, not the same as the LGOs' apparent priorities in accountability. The stakeholders' chance to call for energy and water disclosure is the first achievement of an "ideal public sphere" because the SM platform gives:

- citizens the opportunity equal to that of LGOs to discuss environmental issues; and
- to express a different stand on the relevance of environmental topics (Figure 2).

According to the literature (Section 2.2), there are two potential purposes of the dialogical process: deliberative and agonistic. The findings reveal that the number of LGOs' replies to citizens' comments is low (Figure 2), and the highest percentage concerns the waste issue, one of the most relevant from the LGOs' point of view. Thus, the municipalities' behaviour reveals a sort of "communicative rational justification of validity claims by public administrators to citizens" (Knox, 2016), but only on the topics they consider important. This approach can be viewed as a deliberative one, as there is bidirectional information, and the aim to create general consensus, *but* only on the topics that matter to the LGOs.

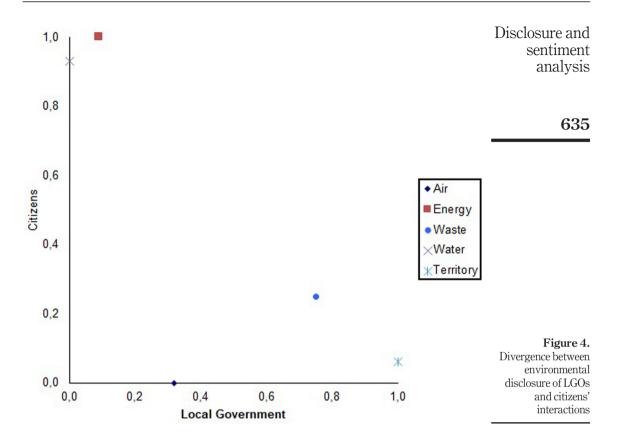
Another finding that confirms this dialogical approach is that SM platforms give citizens the opportunity to express their sentiments and emotions about environmental issues as they do in daily life. The first result of the SA, in fact, is that citizens express these feelings in their writing. The second finding in the SA is a critical stakeholder approach that, for every environmental matter, communicates "negative" passions (Figure 3).

One of the most significant classes for citizens is water. In that class, sentiment has one of the highest percentages of negative emotion. On the other citizens' principal topic, energy, their sentiment has the highest percentage of positive emotion. Thus, citizens seem to encourage a constructivist approach (Thomson and Bebbington, 2005) that could be seen as a step-in consensus towards that topic. Air arouses a very high percentage of negative emotion; otherwise, it is not one of the most relevant topics for LGOs or citizens. According to that, Mouffle's agonistic pluralism could be seen as a potential scenario in which findings are valorised by LGOs and citizens towards open discussions on sustainability behaviours, and future actions and plans in a participatory way. The previous result about air, in a hypothetical agonistic pluralism scenario (Mouffle, 2000), would be considered by LGOs to understand how to answer citizens claims for air discussion. For example, the municipality could account for the actions to be taken to reduce PM10 in the air (as this is one of the words in that topic, Table 4), even if this is not the most relevant topic for citizens; thus, it could not focus on the consensus way but open up the disclosure system around future actions and plans. At the time of our study, this was not happening. Moreover, the intrinsic limits of SM platforms are to be considered in this scenario.

To synthetize the divergence between environmental disclosure of LGOs and citizens' interactions, Figure 4 graphically describes the distance in the values of the frequency distributions of LGOs 'cleaned up posts and citizens' comments, reworked from 0 to 1.

#### 6. Conclusion

Recently, SM has transformed the scope of organisational communications and created opportunities for unprecedented interaction. At the same time, environmental disclosure is an important and growing area of research in accounting literature in light of environmental emergencies afflicting the planet and the continuing interest of stakeholders in this respect (Blankespoor, 2018). In this regard, this study appears to be among the first to examine environmental disclosure via Facebook made by the biggest Italian LGOs and the related stakeholders' sentiment.



The contribution of this research is twofold – first, it demonstrates that LGOs and citizens use SM platforms to discuss environmental issues regarding the municipality. This finding confirms what was previously stated by Blankespoor (2018): at its core, SM is another dissemination channel for organisational communication and, in our study, for environmental matters. However, the characteristics of this channel create different opportunities and risks for management than traditional channels. The low number of responses to public comments also suggests the need for a cultural change in public sector organisations that are still poorly equipped on the issue of participation and engagement. The role of the public administrator should change from that of a neutral expert towards becoming a facilitator of participation and engagement (Brainard and McNutt, 2010). Of course, Facebook is not the only channel for environmental disclosure, but there could also be others such as sustainability reports, municipal websites, conferences and public debates. Future research could compare all these tools to verify their relevance in the environmental communication field, and their strengths and weaknesses.

The second contribution is the statistically tested and validated picture of the relationship between LGOs' behaviour and citizens' sentiment towards environmental topics on a SM platform. The results could represent a dialogical process in its infancy:

Citizens have a public "virtual" arena in which they could express their interest and
opinions towards LGOs environmental communications, without restrictions or
directed requests, and they use the Facebook platform as a tool.

However, the divergence between LGOs' disclosure topics and those of the citizens could
categorise the purpose of this dialogic approach, as in its first attempt, in a deliberative one.

LGOs do not follow the different citizens' interests, but year after year they decide to communicate in the same direction. This paper considers these posts and comments in a longitudinal way, so the conclusion is that LGOs do not modify the content of their disclosure even if citizens call for something else over time. This could be related to a completely rational argumentation (Manetti and Bellucci, 2016; Knox, 2016) that recognises the way that LGOs behave in the organisational legitimacy process.

Otherwise, the potential of an agonistic pluralistic approach in dialogical accountability must be considered from the LGO perspective – citizens express mainly a negative sentiment, but in a different manner in relation to different topics. The study could also be replicated by each municipality individually to understand the importance the public gives to the topic and thus be aware of their needs – interest and sentiment, each differently. Thus, LGOs could use the same SA as this paper in a more constructive way, using SM platforms to co-construct plans and activities in real life.

In the authors' opinion, the co-construction of a co-accountability scenario to combat climate change and discuss environmental issues could be the goal of the next generation (i.e. millennials) who will be the actors in LGOs as well as citizens. Furthermore, SA could be used by politicians, but also by managers/entrepreneurs in the business sector, to analyse the stakeholders' judgement of their communications/actions and plans with regard to corporate social responsibility. This tool gives a result on-time (i.e. not months or years after, as for the reporting system). It is cheaper than a survey and allows first "photograph" of stakeholders' sentiments. It can also be a useful tool for supporting, developing and improving environmental reporting.

Future research aims to deepen these considerations through qualitative case studies to understand what lies behind the LGOs' disclosure behaviour and why they do not comply with citizens' environmental interests. (Maybe they do not examine that data and they communicate on the basis of perceptions?) The management of comments and interactions is interesting not only from the disclosure strategy aspect, but above all, it is important to understand whether online dialogue with the public actually influences LGOs' strategic choices in the environmental field.

As any piece of research, this paper has some limitations. It is confined to data obtained from a small number of LGOs that restrict generalisability to the sole world of LGOs or to other public sector organisations. Future researchers should emphasise the quantitative inquiry of this study for validation on a broader spectrum and include diverse LGOs of different sizes to ensure the generalisability of the findings in multiple settings and across countries. Furthermore, the limitations of SM – circulation of fake news, private ownership and digital divide – mean that SM cannot be a perfect public arena. While considering these significant limitations, their penetration into the population and their characteristics make SM platforms one of the tools most capable of facilitating dialogue with stakeholders. For this reason, they deserve to be carefully analysed by organisations and scholars.

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#### **Further reading**

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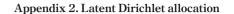
#### MEDAR 29,3

#### Appendix 1

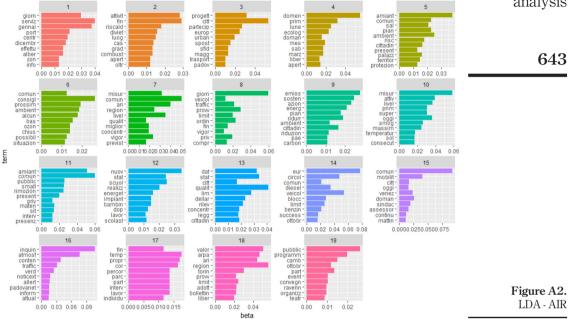
#### 642

# Limitation of environmental impact; Reduction of energy consumption; Energy production from renewable sources; Water supply; Recycled water; Wastewater Treatment; Environmental investments. 5 areas identified: 1. Air 2. Energy 3. Territory 4. Waste Simissions of CO2 and other altering climate gases; Conferment of urban waste to landfill; Urban air quality; Availability of parks; Contaminated sites; Areas with hydrogeological problems; Protected areas; Concern for the loss of biodiversity; Separate collection of municipal waste.

## Figure A1. Sources of the five environmental classes (Global Reporting Initiative guidelines and indicators for "Fair and sustainable wellbeing")



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643



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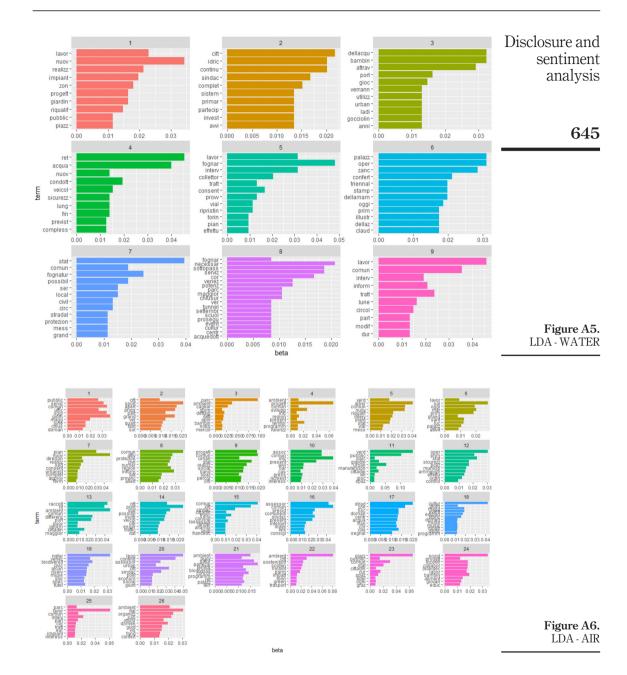
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Figure A4. LDA - WASTE



#### MEDAR 29,3

646	Environmental sustainability area	Topic excluded
Table A1. Topic excluded after LDA	Air Energy Waste Water Territory	1,6,17,19 1,2,4,6,7,8,10,11 3,4,5,7,8,9,11,13,19,21,22,24,26 1,3,4,5,9,11, 13,15,16,20,21 1,2,3

#### Appendix 3

		Air (%)	Energy (%)	Territory (%)	Waste (%)	Water (%)
	Anger	10.7	9.4	9.1	8.9	8.4
	Anticipation Disgust	14.6 10.0	12.7 6.8	14.6 8.3	14.3 12.1	12.4 9.0
	Fear	13.1	11.3	10.6	10.2	13.2
Table A2.	Joy	10.4	14.0	14.3	11.1	11.6
Lexicon dictionary	Sadness Surprise	14.2 6.1	14.3 6.8	13.6 6.4	15.7 5.3	17.4 6.9
SA SA	Trust	20.9	24.7	23.1	22.5	21.2

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