Use of social media for e-Government in the public health sector: A systematic review of published studies

Aizhan Tursunbayeva\textsuperscript{a,b}, Massimo Franco\textsuperscript{a}, Claudia Pagliari\textsuperscript{b,*}

\textsuperscript{a} Department of Economics, University of Molise, Via Francesco De Sanctis, 1, Campobasso 86100, Italy
\textsuperscript{b} eHealth Research Group, Usher Institute for Population Health Sciences and Informatics, The University of Edinburgh, Teviot Place, Edinburgh EH8 9AG, UK

\textbf{A R T I C L E   I N F O}

Keywords: e-Government, Public health, Social media, eHealth

\textbf{A B S T R A C T}

Although the intersection between social media and health has received considerable research attention, little is known about how public sector health organizations are using social media for e-Government. This systematic literature review sought to capture, classify, appraise and synthesize relevant evidence from four international research databases and gray literature. From 2441 potentially relevant search results only 22 studies fully met the inclusion criteria. This modest evidence-base is mostly descriptive, undisciplinary and lacks the theoretical depth seen in other branches of e-Government research. Most studies were published in the last five years in medical journals, focus on Twitter and come from high income countries. The reported e-Government objectives mainly fall into Bertot et al.’s (2010) categories of transparency/accountability, democratic participation, and co-production, with least emphasis on the latter. A unique category of evaluation also emerged. The lack of robust evidence makes it difficult to draw conclusions about the effectiveness of these approaches in the public health sector and further research is warranted.

1. Introduction

Governments worldwide are beginning to harness the Internet and related Information and Communications Technologies (ICT) in an effort to address citizens’ desire for greater information access, institutional transparency, participative decision-making and access to public services. One channel through which these objectives are being pursued is social media, which include off-the-shelf networking sites, such as Facebook, microblogging services, such as Twitter, and information dissemination platforms, such as YouTube (Porumbescu, 2016).

International surveys indicate that four out of five countries now have a national information portal containing links to government social media accounts on platforms such as Facebook and Twitter (UN, 2016). This interest in social media is being driven by the promise of e-Government to “enable stakeholders and government to communicate, collaborate, and engage in governance” (Oliveira & Welch, 2013, p. 397). These stakeholders include, but are not limited to, citizens, employees, non-profit organizations and other arms of government, as described by the taxonomy of social media interactions first developed by Fang (2002).

The health sector represents a critical area of governmental responsibility in most countries, accounting for a major proportion of national spending, equivalent to 9.9% of global Gross Domestic Product in 2014 (World Health Organization (WHO), 2014). Like other parts of the public sector, government departments of health, national agencies charged with monitoring, protecting and improving population health, and state-funded healthcare delivery organisations are under increasing pressure to engage with the e-Government agenda and it is likely that many are using social media specifically in order to do this. While there is a growing body of literature examining social media in health contexts; including aspects of public health communication, promotion and surveillance (e.g. Velasco, Agheneza, Denecke, Kirchner, & Eckmanns, 2014) little has been written about their use for enabling e-Government objectives (see Franco, Tursunbayeva, & Pagliari, 2016 for a discussion). Indeed, it is only recently that scholars have begun to explicitly link the concepts of e-Government, public health and social media; for example, Andersen, Medaglia, and Henriksen (2012) drew on e-Government theories in an exploratory study of the value impacts of social media for the Danish public health system and barriers to achieving these. Given the priority many governments are placing on digital services and the investments being made in social media engagement in the health sector, policy-makers and managers stand to benefit from a timely synthesis of relevant evidence, to guide future practice. Such a synthesis would also add value to the academic e-Government literature, in which healthcare is relatively underrepresented, compared with other public sectors. Our
study aimed to address this deficit by using the rigorous “systematic review” technique to identify, classify, critically appraise and synthesise the corpus of published research evidence relevant to the adoption, use and impacts of social media for e-Government in the public health sector. In doing so we recognised that relevant articles may not explicitly use all of these terms but it may nevertheless be possible to discern an implicit e-Government agenda from studies on the use of social media for delivering public health services (e.g. Thackeray, Neiger, Smith, & Van Wagenen, 2012). In order to facilitate our searches and study interpretation, we drew on the framework developed by Bertot, Jaeger, Munson, and Glaisyer's (2010), which delineates four classes of social media interactions in the public sector, summarised as democratic participation, co-production, crowdsourcing and transparency/accountability, and Fang's (2002) e-Government taxonomy, both of which are described in detail in the Research methods section.

To the best of our knowledge, this is the first systematic literature review to have specifically investigated the adoption and use of social media by public health organizations, taking the perspective that they are also part of government (Salinsky, 2010).

2. Research methods

2.1. Systematic literature review approach

This form of literature review uses “a systematic, explicit, and reproducible method for identifying, evaluating, and synthesizing an existing body of completed and recorded work produced by researchers, scholars and practitioners.” (Fink, 2010, p. 3). This approach was originally developed as a means of synthesising medical research evidence, but is increasingly used in other fields, such as social, policy and business studies (Stead, Gordon, Angus, & McDermott, 2007). In contrast to other types of literature review (e.g. narrative reviews and scoping reviews), systematic reviews focus on specific research questions with narrow parameters; are guided by inclusion/exclusion criteria set at outset (e.g. topics, settings, study types); extract data only from included studies; evaluate the quality of those studies, and base their conclusions largely on the evidence relating to the initial research question(s) (Armstrong, Hall, Doyle, & Waters, 2011; Holeman, Cookson, & Pagliari, 2016). In order to ensure a transparent and replicable process, we followed the “Preferred Reporting Items for Systematic Reviews and Meta-Analyses” (PRISMA) guidelines (Dekker & Bekkers, 2015; Moher, Liberati, Tetzlaff, & Altman, 2009).

2.2. Search strategy

The search strategy and inclusion criteria were informed by a scoping review, which helped to define the concepts of public health (mindful of international differences), e-Government and social media, and the nexus between them (see Franco et al., 2016).

Four international research databases, covering the health, technology, business and social science disciplines (Cochrane Library, MEDLINE, Scopus and Web of Science Core Collection), were interrogated on July 12, 2015. The broad search query was constructed as follows: (“e-government” OR “government” OR “department” OR “organization” OR “agency” OR “hospital” OR “clinic”) AND (“social media” OR “Facebook” OR “Twitter” OR “YouTube”) AND (“health” OR “healthcare”).

In addition to academic databases, we searched WHO reports and working papers (via WHO’s Institutional Repository for Information Sharing) utilizing the same keywords used to search the online academic databases.

The reference lists of articles included in the final set were searched by hand (an approach also known as “snowballing”), as a means of checking for additional studies that may not have been indexed in the online research databases (Yeager et al., 2014).

2.3. Article screening and selection

All outputs were stored in EPPI-Reviewer 4 software, where they were first screened independently by the first author, based on their titles and abstracts. Full text versions of articles appearing to meet the inclusion criteria were obtained for further screening. The third author iteratively checked samples of the assessed articles to ensure consistency with the inclusion and exclusion criteria. This allowed for ambiguities or uncertainties to be discussed and addressed at an early stage, so that consensus could be reached between reviewers. Remaining disagreements were referred to the second author for arbitration.

2.3.1. Inclusion criteria

• Academic or commercial (consultancy) research with a primary focus on the adoption and use of social media by public sector health organizations, at the regional or national levels, for interacting and enabling transactions with other governmental bodies, businesses or citizens, as part of a broader “e-Government” agenda. For example, studies focusing on social media adoption by government departments of health, regional health authorities, government-funded healthcare delivery organisations or national public health agencies.

• Studies published in any language between January 1, 2004 and July 12, 2015. The year 2004 has been chosen as a starting point, since this was when Facebook, the most widely used social media website, was created.

2.3.2. Exclusion criteria

• Studies focused on private sector health organizations.

• Studies focused on individual departments or specialties within public sector health organizations, such as emergency departments, cardiology services or diabetes clinics; for example, to enable a social support group, network with professional colleagues or send targeted messages to patients. This review, in contrast, concerns activities undertaken at the wider organizational level and aimed at enabling information exchange or transactions between public health organizations and other parts of government, citizens or businesses (e.g. Gazley & Guo, 2015).

• Studies primarily focused on the use of social media for health surveillance or research.

• Studies published before January 1, 2004.

The specific study inclusion and exclusion criteria are shown in Textbox 1.

2.4. Critical appraisal of study quality

As per systematic review requirements, the quality of the included studies was rated using the Critical Appraisal Skills Programme (CASP, 2013) checklist, which was slightly modified by adding a “not clear” (0.5) option for each item to the standard “yes” (1) or “no” (0) (These modifications are common in systematic reviews; for example, see Tursunbayeva, Bunduchi, Franco, & Pagliari, 2016). The first author assessed all the eligible studies, while the third author independently assessed a random sample in order to appraise inter-rater consistency and resolve any ambiguities. This exercise revealed only very minor discrepancies, therefore further secondary assessment by the third author focused only on studies that the first author was unsure of. The table derived from the quality assessment exercise is shown in Appendix A.

2.5. Data extraction and thematic analysis

The first author extracted information from all eligible studies with
the help of a research assistant, using a structured form containing the following fields: study authors, publication year; journal discipline; setting (type of organization, country/region in which the study was conducted, year when the study was conducted); type of social media used; stated objective for using social media; research purpose/questions; theoretical basis; study design; main findings; conclusion/comments. This extracted information was then verified by the other two authors.

Extracted study findings were firstly coded using the modified taxonomy of Fang (2002): Government-to-Citizen (G2C); Citizen-to-Government (C2G); Government-to-Business (G2B); Business-to-Government (B2G); Government-to-Government (G2G). Fang's categories of Government-to-Nonprofit (G2N) and Government-to-Employee (G2E) were eliminated, as a first reading revealed that none of the qualifying studies mentioned these. A separate category of Government-to-Professionals (e.g. clinicians, managers) also emerged. In addition to identifying the stakeholders involved, we identified the originator of the interaction (e.g. public health organizations or citizens).

Various models have been proposed for interpreting social media interactions in the public sector (e.g. Mergel, 2013; The White House, 2009). We chose to adopt the framework used by Bertot et al. (2010), which provided a convenient means of categorizing study findings according to their objectives and intended outcomes. The categories, as described by the authors, are as follows:

- **Democratic participation and engagement**, through which social media technologies are used to involve the public in government decision processes, to foster participatory dialog and policy development and implementation.
- **Co-production**, through which governments and the public jointly develop, design, and deliver government services to improve service quality, delivery, and responsiveness.
- **Crowdsourcing solutions and innovations**, through which governments seek innovation through public knowledge and talent to develop innovative solutions to large-scale societal issues. To facilitate crowdsourcing, the government shares data and other inputs so that the public has a foundational base on which to innovate.
- **Transparency and accountability**, through which government is open and transparent regarding its operations to build trust and foster accountability” (Bertot et al., 2010).

The results of this coding exercise were later compared with the Digital Public Service Innovation Framework of Bertot, Estevez, and...
3. Results and discussion

2441 results were generated by the search strategy and 1845 titles and abstracts remained after removing 596 duplicates. Of these titles/abstracts, 229 qualified for full text review, 73 due to their potential relevance and 156 because there was insufficient information in the title or abstract to be able to judge this. After examining the full texts and excluding articles not appearing to meet the inclusion criteria, 45 publications remained. 21 of these were excluded at the data extraction stage due to only describing social media adoption rates by public health organizations, without specifying the purposes or consequences of these uses, or because they were found to be irrelevant to e-Government.

In summary, 24 publications representing 22 separate studies (see Table 1) were included in the final analysis. The stages of selection are illustrated in Fig. 1, using a PRISMA flow chart (Moher et al., 2009). Further explanation is given in the detailed legend shown inTextbox 1.

3.1. Publication characteristics

All of the 24 qualifying articles were published within the last five years (between 2011 and 2015), peaking in 2014 when 9 were published (see Fig. 2). Data collection for the 22 studies represented in the articles was mostly undertaken between 2009 and 2014, as shown by the gray bars in Fig. 2.

We observed that, on average, almost two years typically passes between the period of data collection and the publication of results, although research on social media represents a “rapidly changing landscape” (McCorkindale & DiStaso, 2014). This suggests that the conventional academic literature may be lagging behind as a source of relevant information on social media in health.

Almost all (n = 22) of the publications included in the final analysis were journal articles. The other two were conference papers. Journal articles were initially classified into subject areas according to the taxonomy used by the Scimago Journal ranking portal (Scimagojr, 2016), for example, computer science, medicine, or business, management and accounting, and then using the broader disciplinary categories of medicine, ICT and social science; the latter also encompassing business and management. One article (Donelle & Booth, 2012) from a journal not covered by Scimagojr was manually assigned to the medicine category. This analysis revealed that 15 articles were published in medical journals, 2 in social science journals, one in an ICT journal, and the remaining four in inter-disciplinary journals: two in social science and ICT and two in social science and medicine. This mainly undisiplinary and medical focus suggests that research in this area is still academically siloed, which may reflect university incentives to publish in high impact specialties journals.

3.2. High level aims of the included studies

Although all of the 22 eligible studies had unique research aims and questions, it was nonetheless possible to group them into the following two classes:

- Studies focused on describing approaches to social media use (S1; S2;...
Table 1
Characteristics of studies meeting the inclusion criteria.

<table>
<thead>
<tr>
<th>#</th>
<th>Authors, year (discipline; SJR)</th>
<th>Study aims</th>
<th>Country (income group)</th>
<th>Organizational or institutional focus (number)</th>
<th>SM type (data year)</th>
<th>Study design</th>
<th>Study quality</th>
<th>Classes of e-Government [Bertot et al., 2010 taxonomy]</th>
</tr>
</thead>
<tbody>
<tr>
<td>S1</td>
<td>Bhattacharya, Srinivasan, &amp; Polgreen, 2014 (Medicine; SJR = 1.39)</td>
<td>To investigate factors associated with engagement of U.S. Federal Health Agencies via Twitter</td>
<td>U.S. (high)</td>
<td>Federal Health Agencies (130 accounts of 25 agencies)</td>
<td>Twitter (2012)</td>
<td>Mixed method</td>
<td>7</td>
<td>Democratic participation; transparency and accountability; others (G2G; G2C)</td>
</tr>
<tr>
<td>S2</td>
<td>Donelle &amp; Booth, 2012 (Medicine; not in SJR)</td>
<td>To analyse the content of health-related Twitter posts according to the Public Health Agency of Canada’s Determinant of Health framework. To examine the influence of socio-political factors including government health reforms</td>
<td>North America (focused mainly on U.S. and Canada) (high)</td>
<td>General public and other stakeholders. Governmental healthcare reform (N/A)</td>
<td>Twitter (2009)</td>
<td>Qual.</td>
<td>6.5</td>
<td>Democratic participation; others (G2G)</td>
</tr>
<tr>
<td>S3</td>
<td>Steele &amp; Dumbrell, 2012 (N/A) Dumbrell &amp; Steele, 2013a (ICT; SJR = 0.11) Dumbrell &amp; Steele, 2013b (N/A)</td>
<td>To analyse the characteristics of information disseminated via Twitter by Australian public sector health organizations and explore how citizens participate in onward dissemination</td>
<td>Australia (high)</td>
<td>Government health organizations (n = 16)</td>
<td>Twitter (2012)</td>
<td>Mixed method</td>
<td>6</td>
<td>Transparency and accountability; democratic participation (G2C; G2G; G2P)</td>
</tr>
<tr>
<td>S4</td>
<td>Glover et al., 2015 (Medicine; SJR = 1.8)</td>
<td>To examine the extent to which hospitals utilize SM and whether user-generated metrics on Facebook correlate with hospital quality measures</td>
<td>U.S. (high)</td>
<td>Hospitals (n = 315 performing better than and n = 364 performing worse than the U.S. national rate on 30-day readmissions)</td>
<td>Facebook (2011-2012)</td>
<td>Quant.</td>
<td>6.5</td>
<td>Evaluation (G2G)</td>
</tr>
<tr>
<td>S5</td>
<td>Greaves et al., 2014 (Medicine; SJR = 2.59)</td>
<td>To examine whether tweets sent to NHS hospitals in England contain information about care quality and whether Tweet sentiment is associated with patient survey data and standardised mortality rates</td>
<td>UK (high)</td>
<td>NHS Hospitals (n = 75)</td>
<td>Twitter (2012-2013)</td>
<td>Mixed method</td>
<td>7</td>
<td>Evaluation; transparency and accountability (G2G)</td>
</tr>
<tr>
<td>S6</td>
<td>Harris, Mansour et al., 2014a (Medicine/Social Science; SJR = 5.77)</td>
<td>To analyse citizens’ use of a public-health agency’s Twitter hashing to report food poisoning incidents</td>
<td>U.S. (high)</td>
<td>Chicago Department of Public Health (N/A)</td>
<td>Twitter (2013)</td>
<td>Mixed method</td>
<td>3.5</td>
<td>Co-production (G2G; G2C)</td>
</tr>
<tr>
<td>S7</td>
<td>Harris, Moreland-Russell et al., 2014b (Medicine; SJR = 1.65)</td>
<td>To analyse public responses to proposed e-cigarette regulations on Twitter, by volume, content, networks</td>
<td>U.S. (high)</td>
<td>Chicago Department of Public Health (N/A)</td>
<td>Twitter (2014)</td>
<td>Mixed method</td>
<td>6</td>
<td>Democratic participation (G2G, B2G)</td>
</tr>
<tr>
<td>S8</td>
<td>King et al., 2013 (Medicine; SJR = 0.73)</td>
<td>To investigate the role of Twitter in informing, debating and influencing opinion on health policy, how SM sentiment reflects opinion polls and which users have the most influence on SM</td>
<td>UK (high)</td>
<td>English NHS (N/A)</td>
<td>Twitter (2011-2012)</td>
<td>Mixed method</td>
<td>6</td>
<td>Democratic participation (G2G, P2G; B2G)</td>
</tr>
<tr>
<td>S9</td>
<td>Lachlan, Spence, Edwards, Reno, &amp; Edwards, 2014 (Social Science/ICT; SJR = 1.65)</td>
<td>To evaluate the speed with which agencies update their disease postings on Twitter influences the public’s perception of their credibility</td>
<td>U.S. (high)</td>
<td>Analogue to the Centers for Disease Control (N/A)</td>
<td>Twitter (N.S.)</td>
<td>Quant.</td>
<td>4.5</td>
<td>Transparency and accountability; evaluation (G2C)</td>
</tr>
<tr>
<td>S10</td>
<td>Lee &amp; Kwak, 2012 (Social Science; SJR = 1.38)</td>
<td>To inform an open government maturity model for SM-based public engagement, using case studies with five U.S. government healthcare agencies</td>
<td>U.S. (high)</td>
<td>Food and Drug Administration; U.S. Department of Health and Human Services; Centers for Medicare and Medicaid + others (N/A)</td>
<td>Twitter, Facebook, YouTube + others (N.S.)</td>
<td>Qual.</td>
<td>5.5</td>
<td>Transparency and accountability; co-production; others (G3C, C2G)</td>
</tr>
<tr>
<td>S11</td>
<td>Liu &amp; Kim, 2011 (Social Science; SJR = 0.8)</td>
<td>To compare how public health organisations framed the 2009 H1N1 pandemic via SM compared with traditional media</td>
<td>U.S. (high)</td>
<td>Department of Health and Human Services, Centers for Disease Control, WHO + others (N/A)</td>
<td>Facebook, Twitter (2009)</td>
<td>Quant.</td>
<td>6</td>
<td>Transparency and accountability (G2C)</td>
</tr>
<tr>
<td>S12</td>
<td>McGaughey et al., 2014 (Social Science/ICT; SJR = 0.88)</td>
<td>To evaluate the impact of SM on organizational performance</td>
<td>U.S. (high)</td>
<td>Hospitals (n = 106)</td>
<td>Facebook, Twitter, YouTube + others (N.S.)</td>
<td>Quant.</td>
<td>6</td>
<td>Evaluation (G2G)</td>
</tr>
</tbody>
</table>

(continued on next page)
Table 1 (continued)

<table>
<thead>
<tr>
<th>#</th>
<th>Authors, year (discipline; SJR)</th>
<th>Study aims</th>
<th>Country (income group)</th>
<th>Organizational or institutional focus (number)</th>
<th>SM type (data year)</th>
<th>Study design</th>
<th>Study quality</th>
<th>Classes of e-Government [Bertot et al., 2010 taxonomy]</th>
</tr>
</thead>
<tbody>
<tr>
<td>S13</td>
<td>Neiger, Thackeray, Burton, Thackeray, &amp; Reese, 2013 (Medicine; SJR = 1.65)</td>
<td>To examine how LHDs use Twitter to share information, engage with followers and promote action, and whether use varies depending on size of population served</td>
<td>U.S. (high)</td>
<td>LHDs (n = 210)</td>
<td>Twitter (2012)</td>
<td>Quant.</td>
<td>7</td>
<td>Smaller LHDs: Co-production; transparency and accountability; bigger LHDs: transparency and accountability; co-production (G2C)</td>
</tr>
<tr>
<td>S14</td>
<td>Richter, Muhlestein, &amp; Wilks, 2014 (Social Science/Medicine; SJR = 0.43)</td>
<td>To examine hospital characteristics associated with SM use and to examine how U.S. hospitals’ use Facebook</td>
<td>U.S. (high)</td>
<td>Hospitals (n = 471)</td>
<td>Facebook, Twitter, YouTube + others (2012–2013)</td>
<td>Quant.</td>
<td>6</td>
<td>Co-production; transparency and accountability (G2C; G2G; G2B)</td>
</tr>
<tr>
<td>S15</td>
<td>Shai et al., 2015 (Medicine; SJR = 1.06)</td>
<td>To examine the use and impact of SM on 2-way communication between consumers and public-sector food safety or nutrition agencies</td>
<td>UK and Ireland (high)</td>
<td>Government food safety agencies and food-related health promotion organizations (n = 5)</td>
<td>Facebook, Twitter, YouTube (2012–2013)</td>
<td>Qual.</td>
<td>5.5</td>
<td>Evaluation; transparency and accountability; others (G2G; G2C; G2B)</td>
</tr>
<tr>
<td>S16</td>
<td>Street, Hennessey, Watt, Hiller, &amp; Elshaug, 2011 (Medicine; SJR = 0.85)</td>
<td>To examine whether SM analysis can elucidate community perspectives, media framing and sociopolitical issues around disinvestment in existing health technologies (assisted reproduction)</td>
<td>Australia (high)</td>
<td>Australian Government through Universal health insurance program - Medicare (N/A)</td>
<td>Facebook + others (2010)</td>
<td>Qual.</td>
<td>6</td>
<td>Democratic participation (G2C; G2G)</td>
</tr>
<tr>
<td>S17</td>
<td>Thackeray et al., 2012 (Medicine; SJR = 1.37)</td>
<td>To assess the extent to which SHDs are using SM, which SM applications are used most often and how often SM is used to engage audiences</td>
<td>U.S. (high)</td>
<td>SHDs (n = 50)</td>
<td>Twitter Facebook + others (2011)</td>
<td>Mixed method</td>
<td>6</td>
<td>Transparency and accountability; others (G2C; G2G)</td>
</tr>
<tr>
<td>S18</td>
<td>Thackeray, Neiger, Burton, &amp; Thackeray, 2013 (Medicine; SJR = 1.65)</td>
<td>To discover whether SHDs are primarily using Twitter for one-way information sharing or community engagement, and how this compares with nonprofit organizations</td>
<td>U.S. (high)</td>
<td>SHDs (n = 39)</td>
<td>Twitter (2012)</td>
<td>Mixed method</td>
<td>6.5</td>
<td>Transparency and accountability; co-production; democratic participation (G2C)</td>
</tr>
<tr>
<td>S19</td>
<td>Van de Belt et al., 2015 (Medicine; SJR = 1.65)</td>
<td>To identify the value of SM in monitoring healthcare quality and safety through user reporting of incidents and risks</td>
<td>Netherlands (high)</td>
<td>Dutch Healthcare Inspectorate (N/A)</td>
<td>Twitter Facebook + others (N.S.)</td>
<td>Mixed method</td>
<td>5.5</td>
<td>Evaluation (G2G)</td>
</tr>
<tr>
<td>S20</td>
<td>Vangetti et al., 2014 (Medicine; SJR = 0.32)</td>
<td>To ascertain how many Local Health Authorities and public hospitals have a presence on Facebook, Twitter or YouTube, how well these are known to the public, and how they engage with citizens via SM</td>
<td>Italy (high)</td>
<td>Local Health Authorities (Aziende Sanitarie Locali) (n = 149) and public hospitals (n = 96)</td>
<td>Facebook, Twitter, YouTube (2012)</td>
<td>Mixed method</td>
<td>5.5</td>
<td>Transparency and accountability (G2C; G2G)</td>
</tr>
<tr>
<td>S21</td>
<td>Verhoef, Van de Belt, Engelen, Schoonhoven, &amp; Kool, 2014 (Medicine; SJR = 1.65)</td>
<td>To analyze the relationship between SM and quality of care</td>
<td>Review: N/A Consultation: (high)</td>
<td>Review: N/A Consultation: Dutch public health agencies (N/A)</td>
<td>Facebook + others (N/A)</td>
<td>Review + consultation with public health agencies Quant.</td>
<td>6.5</td>
<td>Evaluation (G2G)</td>
</tr>
<tr>
<td>S22</td>
<td>Yamaguchi et al., 2013 (Medicine; SJR = 1.39)</td>
<td>To examine the effect of SM activity on the collection of signatures opposing government health reform (reimbursement of traditional medicine)</td>
<td>Japan (high)</td>
<td>Japanese Society of Oriental Medicine opposing the medical policy of Government Revitalization Unit (N/A)</td>
<td>Twitter (and an Internet Forum) (2009)</td>
<td>Quant.</td>
<td>6</td>
<td>Democratic participation (G2G)</td>
</tr>
</tbody>
</table>

a Classified according to the World Banks (2016) Country and Lending Groups. Abbreviations: SM = social media; SJR = Scientific Journal Rankings of Scimago Journal ranking portal; Qual. = qualitative; Quant. = quantitative; N.S. = not specified; N/A = not applicable; NHS = National Health Service; LHD = Local Health Department; SHD = State Health Department.
3.3. Country, units of analysis and social media types studied

Twenty studies were specific to a particular country. All of these studies were conducted in high-income countries (see Table 1): 12 in the U.S., three in the UK (one including Ireland), two in Australia, and one each in Italy, the Netherlands and Japan. One study was a scoping review of international literature, exploring the role of social media and rating sites as tools for understanding quality of care (S21). We identified only one study examining the use of social media by public health organizations for e-Government in more than one country, namely the US and Canada (S2), although broader international comparisons of e-Government exist in the literature (e.g. Mickoleit, 2014). Importantly, no studies meeting the inclusion criteria took place in low- and middle-income countries (LMIC), despite there being a high need for government transparency and accountability in many of these regions (World Justice Project, 2015) and the potential of e-Government to support sustainable development (UN, 2016). Recent literature reviews have nevertheless documented innovative uses of social media in LMIC to support aspects of e-Government in the public health system (Holeman et al., 2016). Taken together, these results suggest a need for new international and interdisciplinary research to shed light on how the appropriateness and usefulness of e-Government approaches using social media may differ across political, socioeconomic and cultural contexts.

A diverse range of public-sector health organisations were examined in the included studies. Five studied specific hospitals (S4; S5; S12; S14; S20) two studied state-level public health departments (S17; S18), two studied a large urban department of public health (S6; S7) and two studied local health departments (S13; S20). The remaining studies focused on other types of public health organization, as shown in Table 1.

Half of the 22 studies focused on a single social medium; mostly Twitter (n = 10), and one on Facebook (n = 1). Others studied several platforms in parallel: Facebook and Twitter (n = 1); Facebook, Twitter and YouTube (n = 2); Facebook, Twitter or YouTube plus another social medium outwith our inclusion criteria (n = 8). For studies in the latter category we extracted and coded only the findings related to Facebook, Twitter or YouTube, as per our inclusion criteria.

Twitter was the social medium described most frequently as a means of enabling health organizations to pursue goals around Democratic Participation, Transparency and Accountability, or Co-production, and was also mentioned as having potential to address the other e-Government objectives described in the Digital Public Service Innovation Framework (Bertot et al., 2016). However, this also reflects the dominance of Twitter in the corpus of studies. For the Evaluation purposes that emerged in our research (see Section 3.7.3 for details), Twitter and Facebook were used equally (see Table 1 for further details). Very few of the included studies provided separate findings for the use of YouTube in the context of e-Government in public health (e.g. it was usually mentioned under the generic umbrella of social media), despite YouTube reportedly being one of the most commonly used social media by government overall (e.g. Abdelsalam, Reddick, Gamal, & Al-shaar, 2013).

3.4. Research designs and study quality

Almost half of the studies used mixed methods (n = 10). Seven employed quantitative designs (e.g. quantitative content analysis, descriptive statistics, etc.), while four used qualitative designs. One study was a literature review.

None of the qualifying studies received a maximum score of 8 on the quality assessment scale, although more than half were ranked as being of good quality (scoring 5.5–6) and seven as high quality (scoring 6 +) including three mixed methods studies, one literature review, one qualitative and two quantitative studies. Studies which received lower quality ratings (scores below 5) did so because they had not adequately justified their research design, did not clearly state the value of their research findings for practice and future research or did not consider sources of potential bias (see Appendix A). The articles were mostly published in high quality journals according to the Scimago Journal Rank Indicator (Scimagojr, 2016), the lowest being Applied Mechanics and Materials (SJR = 0.11) (Dumbrell & Steele, 2013a) and the highest being Morbidity and Mortality Weekly Report (SJR = 5.77) (Harris, Mansour et al., 2014a).

3.4.1. Software used to extract data from social media

In addition to formal research designs, several articles reported using specific software to extract or analyse data from social media, mostly from Twitter. These included Twitter Streaming Application Programming (S1; S5; S12; S13; S18), The Archivist (S2), twitterR package for R version 2.15.2 (S7) and NodeXL (S7) for data extraction, and SentiStrength (S1) and TheySay Ltd. for data analysis (S5). Some studies chose to extract and categorize Twitter data manually, after finding that programs such as Twitonomy, TweetVolume, TweetStats, “did not provide the necessary functionality for region-wide, domain-wide or ‘tweet meaning’-based data capture and categorization” (S3). One study (S17) used the data capture software SnagIt to make screenshots, while two studies (S15; S16) used Nvivo by QSR International to analyse data posted on multiple social media.

3.5. Theoretical frameworks

Over half of the studies (n = 13) did not specify any theoretical perspective. In the remaining papers, theories were cited as explanatory/theoretic frameworks, e.g. Rogers’ diffusion of innovations theory (S11); as practical/guiding frameworks, e.g. the Public Health Agency of Canada’s Determinants of Health framework (S2), the Rand Public
3.6. Type and direction of social media interaction

All of the studies either focused on, or mentioned in their findings, the interaction between public sector (government) health organizations and citizens (G2C or C2G). Others described public health organizations interacting with businesses (B2G), other governmental departments (G2G) and professionals such as clinicians (G2P). Some described citizens actively engaging in policy discussions with public health organizations (C2G) or discussing with one another (C2C). The direction of interactions between public health organizations and various stakeholders described in the studies is illustrated in Fig. 3, including uni-directional and bi-directional forms. This also indicates a relative paucity of studies examining public health organizations' interactions with businesses (B2G) or B2G), other governmental departments (G2G) and relevant professional groups (G2P or P2G), which would benefit from further research.

3.7. Reasons for using social media

We used Bertot et al.'s (2010) framework for classifying the social media interactions referred to in the studies. However, we found a high degree of overlap between Bertot et al.'s categories of Crowdsourcing solutions and Co-production and, for this reason, we merged the two. We also added a new category which emerged as a separate theme, concerned with the Evaluation of public health services by citizens, via feedback, comments or suggestions posted on social media. Table 2 summarises the studies according to the adapted categories of e-Government (Bertot et al., 2010) and social media interactions (Fang, 2002).

3.7.1. Transparency and accountability

Transparency and accountability (Bertot et al.'s (2016) transparent category) were the main reasons cited for public health organizations’ interaction with the aforementioned stakeholders (see Table 2). In these cases, information sharing was primarily between government and citizens, in both directions: G2C (e.g. S1) and C2G (e.g. S5).

Transparency and accountability mainly involved using Facebook, Twitter or YouTube in order to post information about the organization itself (e.g. staff members, services, accreditation), to provide updates on ongoing activities (e.g. news, job openings, events, projects) or to increase awareness of their Open Data resources. However, studies described this as a largely one-way interaction, where public health organizations provide and stakeholders receive information. One study observed that small public health organizations were more likely to post tweets about themselves, although large public health organizations tweet more in general (S13). Another study described how a transparent approach to resolving patient problems via social media could help to improve health organizations' public image (S15).

3.7.2. Democratic participation

Democratic participation (Bertot et al.'s (2016) participatory category) was the next most frequently cited reason for public health organizations’ use of social media in e-Government (see Table 2). Articles described social media as multi-disciplinary, non-hierarchical meeting places where citizens and professionals could share information and refine or reinforce their own views (S8). This allowed stakeholders to voice opposition or support for proposed health legislation or reforms, whilst enabling public health organizations to “listen” to and understand their views, as well as to disseminate information about the proposals in question.

Thus one study found that governmental health organizations used social media to disseminate health policy news more often than other health-related organizations, such as not-for-profits (S3), possibly reflecting a greater requirement for civic engagement. Although none of the included studies reported that citizens' comments on social media had directly influenced health policy, several authors pointed out that social media are widely used by policy makers and may play a significant role in informing government decisions. For example, a study in the US showed how a municipal public health department used social media to understand citizens' views about proposed e-cigarette regulation (S7). Another, from the UK, found that a diverse range of stakeholders had engaged with information about health and social care reforms on Twitter and that negative sentiment towards these reforms echoed those found in public opinion polls (S8).

Social media can also be used by advocacy groups to influence health policy; for example, a Japanese study found that it increased the number of online signatures collected in a campaign to oppose reforms to the reimbursement of traditional medicines (S22).

![Fig. 3. Direction of public health organizations' interactions with stakeholders.](image-url)
While opportunities for advocacy may offer public benefit by giving a voice to civil society organisations, social media also present opportunities for certain groups or individuals to gain influence in ways that may be seen as unrepresentative, unfair or even antidemocratic. For example, in the aforementioned study of NHS reforms (S8), tweets from newspapers and celebrities were disproportionately represented and the authors caution that Twitter should be regarded as a place for sharing views rather than a forum for genuine debate (King et al., 2013). The use of social media channels for corporate lobbying represents a particular challenge for e-Government in the context of public health. For example, Harris, Moreland-Russell et al. (2014b) analysed Twitter activity in response to information disseminated by the Chicago Public Health Department regarding local proposals to regulate electronic cigarettes as tobacco products. They describe how the department’s Twitter account was rapidly inundated with hundreds of messages arguing against the legislation. Most of these were found to originate outside the area and many bore hallmarks of corporate “astroturfing”, in contrast to the smaller number of local postings, which were more in favour of the proposal (S7).

Overall, these findings are consistent with other e-Government research demonstrating the ability of social media to engage and enable citizens to participate in the policy making process (Mergel, 2013), whilst also highlighting the need for vigilance in identifying cases of misuse.

3.7.4. Co-production

Co-production (Bertot et al.’s (2016) co-created category) was the least reported use of social media by public health organizations for e-Government (see Table 2), echoing previous analyses of the e-Government literature that have shown the dominance of one-way interaction between governmental organizations and citizens (Riarh & Roy, 2014). Several authors mentioned that very few organizations use social media in order to ask external stakeholders to do something to benefit their organizations (e.g. S13; S14). Nevertheless, some describe organizations actively soliciting the collaboration of citizens or patients, such as requesting volunteers (e.g. S18) or collecting suggestions on how to improve services, as undertaken by the U.S. Centers for Medicare and Medicaid and the Food and Drug Administration (S10). One study observed that Local Health Departments (LHD) asked their followers to do something for the organization more often if they were small, compared with larger LHDs. Possible explanations offered by the authors include the limited capacity of small LHDs to provide a wide range of services and a sense of familiarity or cohesion that might be more common within rural communities served by smaller health organisations (S13).

3.7.5. Other uses

Several authors (e.g. S1, S2; S15; and S17) considered other ways in which data from social media interactions might be used to inform health organisations, which to some extent reflect the remaining, arguably future-focused, categories described by Bertot et al. (2016) as anticipatory, personalised, context-aware and context smart, although these were not explicitly articulated. For example, in their “open government maturity model for social media-based public engagement” - which was informed by case studies of healthcare administration agencies (S10) - Lee and Kwak (2012) describe the potential for data collected from public health organizations’ online portals, mobiles and social media to feed analytics that support rapid and timely decision-making, virtuous cycles of public engagement and collaboration, and continuous quality improvement within public health organisations, although they point out that these functions had yet to be integrated at the time of writing.

4. Challenges and limitations

Our background research to inform the review protocol (Franco et al., 2016) highlighted the challenges involved in defining the scope of the public health sector, given national differences in health system structure and financing, and helped to clarify this for the purposes of our inclusion and exclusion criteria. Nevertheless, the studies yielded by our search strategy reported a diversity of public health organisa-
tions, some clearly labelled as such and others only evident through further reading. For example, we included one study of a food standards agency based on a separate definition of these as governmental agencies with a public health remit (S15).

Our search results contained many studies that simply reported social media adoption rates by public health organizations (e.g. Bermúdez-Tamayo et al., 2013; Griffis et al., 2014) or described social media interactions between public health organizations without specifying the reasons for those interactions (e.g. Harris, Choucair, Maier, Jolani, & Bernhardt, 2014). These were excluded from our analysis since they did not explicitly link social media to e-Government. Likewise, we excluded studies that described social media practices within specific healthcare units, since these typically involve the provision of information or support to specific groups of patients, rather than addressing e-Government objectives at the wider organizational level, which is the focus of our systematic review. We also excluded studies using social media data as a tool for understanding specific illnesses or patient communities, which could arguably be classed as a type of e-Government “listening” activity but is more akin to digital health surveillance or eHealth research (e.g. Pagliari & Vijayakumar, 2016). Although these studies fall outside the scope of our systematic review there is doubtless relevant knowledge to be gained from synthesising them, and other reviewers may wish to do so.

Our coding framework would have benefitted from documenting whether authors considered the ethical implications of the social media uses they describe. Given the difficulty of disguising citizens’ and patients’ identities on social media, simply mining these data for research or to inform public services raises ethical challenges, particularly when consent has not been or cannot be obtained, as in sentiment research or to inform public services raises ethical challenges, particularly when consent has not been or cannot be obtained, as in sentiment research or to inform public services raises ethical challenges, particularly when consent has not been or cannot be obtained, as in sentiment research or to inform public services raises ethical challenges, particularly when consent has not been or cannot be obtained, as in sentiment research or to inform public services raises ethical challenges, particularly when consent has not been or cannot be obtained, as in sentiment research or to inform public services raises ethical challenges, particularly when consent has not been or cannot be obtained, as in sentiment analysis from Twitter data. There would be value in further research to assess the extent to which these practices are consistent with privacy laws and policies, and acceptable to stakeholders.

Finally, all of the included studies were conducted in high-income countries, despite the fact that our searches covered international databases, including WHO. We are aware, from a separate review, that cases of social media use for aspects of e-Government in the public health sector of LMIC exist, although these would not typically be classed as research (see Holeman et al., 2016 for discussion). Other approaches to evidence capture, such as scoping reviews of innovation projects and expert consultations, may therefore be necessary to uncover this activity.

5. Summary and conclusions

To the best of our knowledge, this is the first systematic review to have captured, appraised and synthesised the corpus of research evidence on the use and impacts of social media for e-Government in the public health sector. Its key messages are summarised in Textbox 2.

While most publically-funded health organizations are beginning to use social media in ways that are consistent with e-Government objectives, our review shows that few published studies have explicitly linked these concepts. Of those that exist, most focus on social media as a channel for organization-citizen interaction (dissemination and feedback), rather than other forms of stakeholder-to-stakeholder interaction described in Fang’s (2002) e-Government taxonomy. The specific e-Government objectives described in these studies are broadly compatible with the categories proposed by Bertot and colleagues in 2010. In this regard, the strongest emphasis is on facilitating the transparency/accountability of public services and enabling democratic participation/engagement, while active co-production by citizens appears less frequently, likely reflecting the healthcare sector’s prioritization of evidence-based medicine and policy. A separate category of “evaluation” also emerged as a distinct theme, involving the use of social media to actively solicit or passively listen to citizens’ opinions on the quality of public health services, alongside studies evaluating the potential of social media to yield this information. This evaluation category represents a potential addition to Bertot’s model and warrants further study. While our analysis pre-dated Bertot et al.’s (2016) Digital Service Innovation Framework, it is easy to see how some of the observed uses of social media in public health could inform “smart, anticipatory” approaches; such as monitoring nascent indicators of reputational risk to inform rapid quality improvement activities. With advances in data science the potential for automated social media analytics to drive adaptive “learning health systems” in the future is considerable (Krumholz, 2014). At the present time, however, research and practice involving social media for e-Government in the public health sector is relatively immature compared with other areas of e-Government research. This is reflected in the lack of theoretical depth which we have observed in this literature, the dominance of descriptive analyses and the absence of multi-disciplinary and international studies which could shed light on important contextual influences. We recommend investment in new interdisciplinary research to better articulate the value proposition for social media as a facilitator of e-Government in public health organizations and to build evidence of their uses, impacts and contextual mediators, drawing on multiple stakeholder perspectives. Such evidence will be vital for guiding managers and policymakers as to the most cost-effective, appropriate and responsible uses of these approaches in this essential, but increasingly resource-constrained, public sector.

Conflicts of interest

Not known.

Funding

The corresponding author is a collaborator on three RCUK-funded research programmes relevant to this article: The Farr Institute for Health Informatics Research (Scotland) MRC, Grant number MR/K007017/1; The Administrative Data Research Centre for Scotland, ESRC, Grant number ES/L007487/1; The Science and Practice of Social Machines EPSRC, EP/J017728/1. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Acknowledgements

We thank research assistant Stefano Di Lauro for his contribution during the data extraction stage of this systematic review. We also thank anonymous reviewers for their valuable recommendations, which helped to strengthen our paper.

Appendix A. Studies according to methodological quality assessment criteria (CASP, 2013)

<table>
<thead>
<tr>
<th>Category (Questions)</th>
<th>Considerations</th>
<th>Yes</th>
<th>Not clear</th>
<th>No</th>
</tr>
</thead>
<tbody>
<tr>
<td>Research objectives</td>
<td>Is there a rationale for why the study was undertaken?</td>
<td>S1; S2; S3; S4; S5; S7; S8; S9; S6</td>
<td>S10; S11; S12; S13; S14; S15; S16; S17; S18; S19; S20; S21;</td>
<td></td>
</tr>
</tbody>
</table>
Textbox 2. Key messages.

- 4 international research databases and 1 source of gray literature were systematically searched, using key words, to identify studies focused specifically on the use of social media for e-Government in the public health sector.
- Out of 2441 search results only 22 studies matched the eligibility criteria.
- These studies date exclusively from the last 5 years, come from high income countries and were published in academic journals (mostly medical).
- The studies are mainly descriptive, unidisciplinary and atheoretical, although scored well on methodological quality criteria.
- Twitter was the most commonly studied social medium.
- Most studies focused on interactions between public health organizations and citizens, rather than between other e-Government stakeholders, as described by Fang (2002), although some fell into a new category of Government-to-Professionals.
- The e-Government objectives for which social media were being deployed mostly related to Bertot et al.’s (2010) categories of transparency and accountability (openness) and democratic participation (consultation and feedback), with a lesser emphasis on co-production (collaboration). Evaluation (e.g. of organizational performance) also emerged as a unique theme.
- More interdisciplinary research is needed to understand how public health organizations are using social media for e-Government, to articulate their pathways to impact, to evaluate their effectiveness in achieving e-Government objectives and to examine the contextual factors influencing each of these.
- While systematic reviews are highly focused and prioritise published research, a broader scoping review would be useful for documenting further examples of social media use for e-Government in different public health settings internationally.
both for and against the researcher’s arguments, been demonstrated? Has the researcher discussed the credibility of their findings (e.g., triangulation)? Are limitations of the study discussed explicitly? Are the findings discussed in relation to the original research questions? Are the conclusions justified by the results?

Value of the research (Is the study of value for research and practice?)

Does the researcher discuss the contribution the study makes to existing knowledge or understanding? Does the research identify new areas in which research is necessary? Does the researcher discuss whether or how the findings can be transferred to other populations, or consider other ways in which the research can be used?

References


Dumbrell, D., & Steele, R. (2013b). Twitter and health in the Australian context: What types of information are health-related organisations tweeting? (pp. 2666–2675). -.


S15; S16; S17; S18; S19; S20; S21; S22

**Aizhan Tursunbayeva** is a PhD Scholar at the University of Molise (Italy) and a visiting doctoral researcher at the University of Edinburgh, studying information systems in healthcare under the supervision of the other co-authors at the Universities of Molise and Edinburgh. She previously worked as an HR manager for HSBC.

**Massimo Franco** leads the Organisational Studies and Human Resource Management Group at the Department of Economics, Management, Society and Institutions, University of Molise, Italy.

**Claudia Pagliari** leads the Global eHealth Programme at the University of Edinburgh’s Usher Institute for Population Health Sciences and Informatics and is a member of the Administrative Data Research Centre for Scotland and the Farr Institute of Health Informatics Research (Scotland).