
A systematic literature review on activity recognition with context-awareness techniques for mitigation of disasters

Fatai Idowu Sadiq

Faculty of Computing,
Universiti Teknologi Malaysia,
81310, Skudai, Johor Bahru, Malaysia
and
Department of Computer Science,
Ambrose Ali University,
P.M.B. 14, Ekpoma, Edo-State, Nigeria
Email: fisadiqsc-dpt@aauekpoma.edu.ng

Ali Selamat*

UTM-IRDA Digital Media Center of Excellence,
Universiti Teknologi Malaysia,
81310, Skudai, Johor Bahru, Malaysia
Email: aselamat@utm.my
and
Faculty of Computing,
Universiti Teknologi Malaysia,
81310, Skudai, Johor Bahru, Malaysia
and
Centre for Basic and Applied Research,
Faculty of Informatics and Management,
University of Hradec Kralove,
Rokitanskeho 62, Hradec Kralove, 50003, Czech Republic
*Corresponding author

Roliana Ibrahim

Faculty of Computing,
Universiti Teknologi Malaysia,
81310, Skudai, Johor Bahru, Malaysia
Email: roliana@utm.my

Abstract: This paper presents a systematic literature review on context-awareness using activity recognition for mitigation of disasters. Disaster preparedness and the method of mitigating it or reduction of causalities are very important in order to save human lives. The review process went through four phases which are: planning, collation, filtration of dataset to get the most relevant materials, exploration and the description based on findings was carried out. The problem with the existing studies are: technology used lack the capacity to give feedback to potential victims, inability to move

the technology around, high false negative alarm, low sensor quality for crowd behaviour monitoring. This study used smartphones which is handy and has potential of being carried about. The analysis of the result was presented under three major headings which are publication by year, techniques, results and findings. The result shows the research gap in the area of disaster mitigation and the further study suggests the need to extend context-aware framework for stampede prediction based on activity recognition accuracy used in the previous study.

Keywords: systematic literature review; SLR; activity recognition accuracy; ARAC; smartphone sensor; crowd disasters; context-awareness.

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Biographical notes: Fatai Idowu Sadiq is a Lecturer from the Ambrose Alli University, Ekpoma, Nigeria. He obtained his BTech/Maths and MTech in Computer Science in Minna and Akure. He obtained his MBA and PGD in Education from the Ambrose Alli University, Ekpoma and MPhil in Computer Science in Nigeria. He is currently a PhD Scholar at the Faculty of Computer Science and Information Systems in the Universiti Teknologi Malaysia. He is a member of Nigeria Computer Society (NCS), Computer Professional, Registration Council of Nigeria (CPN) and IEEE Computer Society. His research interest are context-aware computing, human activity recognition and disaster management. He has published number of papers in both local and international journals

Ali Selamat is currently a Chief Information Officer and Director of Centre for Information and Communication Technologies at the Universiti Teknologi Malaysia (UTM), Malaysia. He is also a Professor in the Software Engineering Department at the Faculty of Computing, UTM. He is nominated as the Vice Chair of the IEEE Computer Society Malaysia since 2014. Currently, he is the editor of *International Journal of Intelligent Information Database Systems (IJIIDS)*, Inderscience Publisher. His research interests include software engineering, software process improvement, software agents, web engineering, information retrievals, pattern recognition, genetic algorithms, neural networks and soft computing, computational collective intelligence, strategic management, key performance indicator and knowledge management.

Roliana Ibrahim is the Head of the Information Systems Department, Faculty of Computing, Universiti Teknologi Malaysia (UTM). She received her BSc (Honours) in Computer Studies from the Liverpool John Moores University, MSc in Computer Science from the Universiti Teknologi Malaysia and PhD in the field of Systems Engineering from the Loughborough University. She has been an academic staff at Information Systems Department since 1999 and was previously a Coordinator of BSc in Computer Science (Bioinformatics) Program and Master of Information Technology (IT Management). She is one of the editorial board members of *Journal of Information Systems Research and Innovation (JISRI)* and the Guest Chief Editor of *Special Issue on Science and Technology, Jurnal Teknologi UTM*.

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1 Introduction

Context-aware computing (CAC) is a software that examines and reacts to individual's changing context (Schilit et al., 1994). It is aware of the user's state and surroundings and helps to adapt to its behaviour (Satyanarayanan, 2002). The term context-awareness is a property of mobile computing. It originates as a term from ubiquitous or pervasive computing, which is an act of linking changes in the environment with a computer system that is static. The current target of pervasive computing is how to integrate intelligent agent that is capable of knowledge empowerment and reasoning to understand the local context and shared information to support intelligent application and interfaces (Chen et al., 2003). Contextual characteristics experience changes when the need arises, by using contextual commands, automatic contextual reconfiguration and context triggered actions (Schilit et al., 1994). Any CA software will use any of these contextual commands, automatic contextual reconfiguration or context trigger actions to execute the expected task to carry out its function. Present researches with context-awareness application are common in campuses (Hsieh et al., 2007), industries (Docter et al., 2007), healthcare services (Varshney, 2009), other organisations and the entire world. Previous research has shown that CA personalised activity assistance will act as a commonplace in the future. The smart home activities of daily living of people, particularly elderly or disabled can be monitored and analysed for personalised CA assistance living to be offered. Hence, activity recognition (AR) has emerged as a critical research issue related to the realisation of intelligence in pervasive environments. Context is any information that can be used to characterise the situation of an entity. The entities include identity (who); activity (what); time (when); and location (where) in other words, W4 can be used to describe context in Abowd et al. (1998). However, contextual information and any intelligent behaviour in an environment need to be relevant to the user's context and ongoing activities. This concept led to AR that has emerged as a final research issue related to the successful realisation of intelligent pervasive environments. A systematic literature review (SLR) is a product of evidence-based software engineering (EBSE) concept proposed in Kitchenham and Charters (2007). It is simply a "standard for identifying, evaluating and interpreting all available research that is relevant to a specific question, topic area or phenomenon of interest." However, some studies present a literature review on CA related research in Hong et al. (2009), Orwat et al. (2008), Naismith et al. (2004), Ye et al. (2012) and Al-Bashayreh et al. (2013), from our best knowledge, only Al-Bashayray et al. (2013), have conducted a systematic review of CA mobile patient monitoring framework (CMPMF). Al-Bashayray et al. (2013) adapted the 'concept matrix technique' to synthesise previous studies to develop a research agenda, which further provides foundation for extending the state-of-the-art research by filling the gaps identified. The study also categorised previous work using factors of strong context-aware application framework (FSCAF), to enhance the design of CMPMF and to address the problems encountered in mobile patient monitoring systems (MPMS) using wireless sensors, in the field of biomedical informatics with patients having chronic diseases; like hypertension, diabetes and epilepsy; in respect of vital sign, medication treatment and disease symptoms. A good scientific and quality literature review is a prerequisite for doing excellent and creative research. Quality research is quality because it promotes and expands our horizon and the overall perception on the research. Given this fact, this paper explores context-awareness techniques using AR related research

with focus on mitigation of disasters in particular, crowd disaster. In the light of this, this study carries out SLR in context-awareness with a focus on AR domain that addresses problems in a crowd disaster mitigation (DM) related topics.

Q1 What are the existing context-awareness techniques with AR?

Q2 What are the strength and weaknesses?

Q3 What are the contributions in context-awareness with AR related to DM?

Q4 Which of this CA AR researches address crowd disaster?

Meanwhile, the main contributions of this paper are outline as follows:

- 1 To present a SLR to identify the context-awareness techniques associated with AR.
- 2 To highlight the strengths and weaknesses of the identified techniques using the SLR.
- 3 To investigate and identify the context-awareness techniques with AR appropriate to solve disaster related problems thereby extending the study of Ramesh et al. (2014) and similar study found in literature.
- 4 To highlights the techniques that can help to mitigate crowd disaster to minimise loss of lives and properties in the 21st century and beyond.

The remaining sections are as follows, Section 2 explains flow of the research methodology adopted and research questions (RQs) used in this study. In Section 3, results and statistics of primary studies are presented. Section 4 discussed the various results to provide answers to the RQs (Q1–Q4), Section 5 discusses research activity issue in a crowd DM, existing context-awareness frameworks, important research gaps with future research directions and Section 6 presents the conclusions.

1.1 Problem background

Previous studies focus on the use of close circuit television (CCTV), as far back as 1995 (Schilit et al., 1994), video camera in computer vision (Satyanarayanan, 2002), for monitoring people in the crowd to provide safety measure, in the case of unforeseen incidents. These techniques suffer from:

- 1 no feedback to the potential victims of danger
- 2 the device which is normal position in a specific location due to their static nature
- 3 devices are hardly found among people except in rear cases like smartphone.

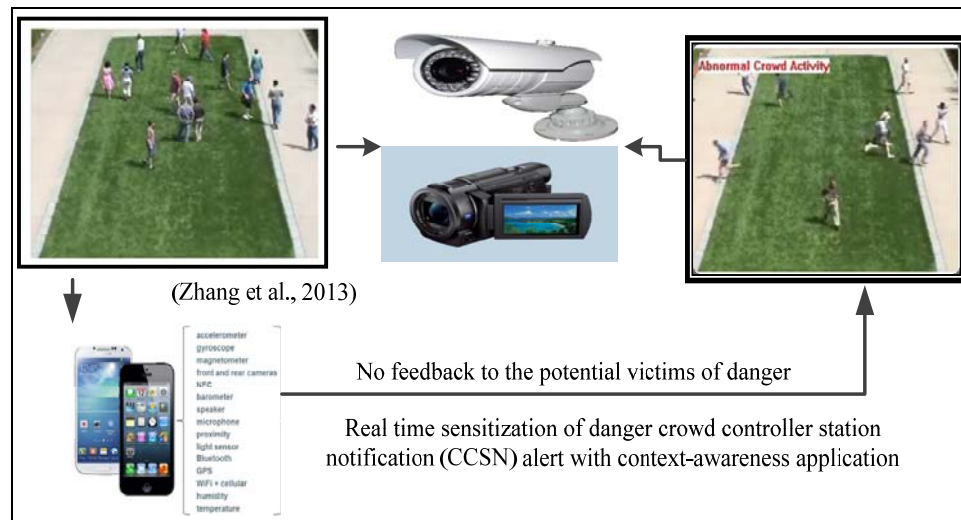
Wireless sensor network was remarked to be suitable for better situational awareness of crowd monitoring in a crowded area (Chen et al., 2003). Despite this, it has limitations such as:

- 1 high false negative alarm (FNA)
- 2 low sensor quality for crowd behaviour recognition.

Presently, the research has shifted to the use of context-awareness computing and wireless sensor network (Hsieh et al., 2007). Because of the evolution of smartphone

sensing and the power of inbuilt sensors with the tendency to monitor and sensitise potential victims of danger for any unforeseen incidents that can cause a disaster during an emergency with feedback to reduce risk associated with such occurrences. Figure 1 depicts the overview of scenarios based on the issues that led to existing studies (Docter et al., 2007; Varshney, 2009) that led to the existing studies.

Figure 1 Flow of crowd monitoring system issues (see online version for colours)



2 Methodology

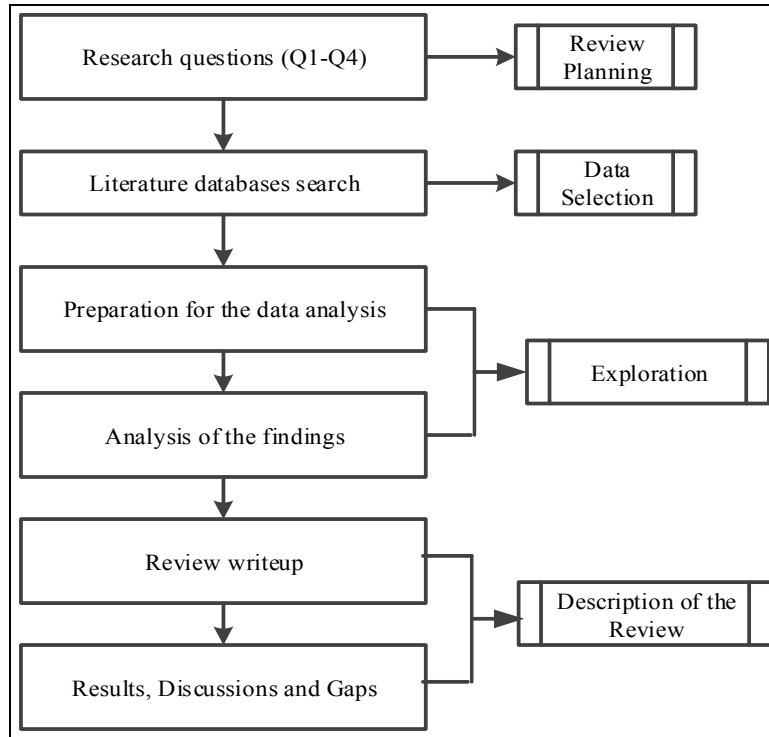
The various stages of a SLR are employed in this paper. Following the first stage presented in conference refer to Sadiq et al. (2016). The steps comprise of planning, conducting and documenting the review (Kitchenham and Charters, 2007; Achimugu et al., 2014). Each step involves combinations of simple procedures that will guide to achieve the goal of this study. The paper focus on the investigation of several issues concerning research on context-awareness and the extent to which it has been used to handle disaster management. It also focuses on SLR using combined approach and guidelines adapted from Naismith et al. (2004), Glass et al. (2002) and Musumba and Nyongesa (2013). The authors of the said study utilised four phases method to plan, collate, explore and described the literature-based discoveries. In this study, the focus is to employ a SLR approach to selected related publications and explanation on the under listed RQs.

- Q1 What are the existing context-awareness techniques with AR?
- Q2 What are the strength and weaknesses?
- Q3 What are the contributions in context-awareness with AR related to DM?
- Q4 Which of this CA AR researches address crowd disaster?

2.1 SLR approach

The study proceeds on SLR using the design flow showing six main steps in a way that each of them is important to the proposed study.

Figure 2 Design flow of the SLR approach



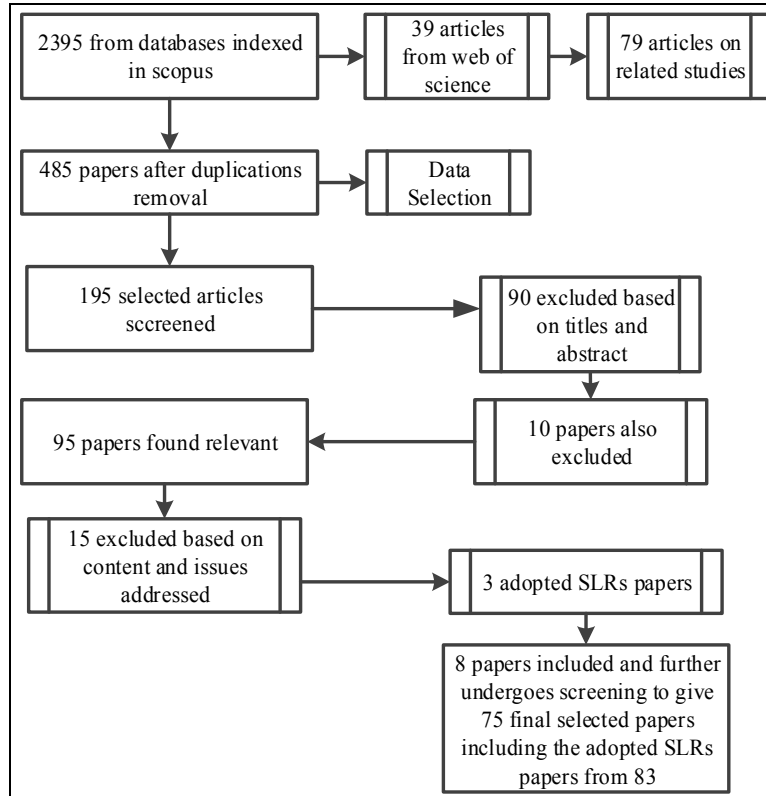
2.1.1 Search process

SLR deals with a comprehensive search of all relevant sources about the subject of discussion. However, the search process used in this study is classified into two main search stages. Search:

- Step 1 A thorough search was launched on the four electronic database sources and the returned results (papers) were assembled as a set of prospective papers from high-quality journals and the web of science.
- Step 2 The reference lists of all quality papers were read to pick additional quality papers, if any, combine them with the relevant ones.

The design flow following the first stage of this research presented in is as shown in Figure 2.

The summary of the search process and criteria for selecting the selected primary studies depicts Figure 3.

Figure 3 Flow of the SLR study databases and adopted search criteria

2.2 Search terms

The under listed search terms were used for obtaining relevant studies as a result of the topic of discussion. The terms are as follows: “context-awareness, CA technique and shortcoming of CA techniques, AR and CA and mitigation of crowd disasters.” The search terms were further modified as follows:

- 1 Identifying the key terms in the RQs formulated in Section 2.
- 2 Identifying different spellings and substitutes of key terms.
- 3 Accessing the index terms of relevant publications.
- 4 Embedding the Boolean OR in alternative spellings and synonyms.
- 5 Embedding the Boolean AND between links of key terms.

2.2.1 Study selection

From the first search stage, 2,395 prospective studies were realised. Next, the titles of these studies were used to scrutinise and collate relevant studies. This task was necessary

to eliminate duplicate and irrelevant studies. Consequently, 95 relevant studies were selected. After that, the keywords and references on each selected study were perused to determine important studies that might have been missed out during the initial search process. This effort led to the identification of 15 additional studies that were further screened out of the initial search process that took the tally of the selected studies to 75 from 83 plus SLR papers. The 75 primary studies (Table 3) may be as a result of the domain of focus in this study though other disasters-related works were found in the literature, they do not address the topic under consideration. Finally, the quality criteria stipulated in Table 1 used for scrutiny were applied to the 75 relevant studies discovered, thus form the final selected studies that are capable of providing answers to the formulated set of RQs.

Table 1 Inclusion and exclusion criteria

<i>Inclusion criteria</i>		<i>Exclusion criteria</i>	
a	Published papers written in English language	a	Published papers not written in English language
b	Papers that focus on context-aware and activity recognition or address disaster issues	b	Papers that do not answer any of the research questions
c	Relevant papers that are published previously until 2017	c	Papers without proper reference information
d	Papers with publishing details	d	Paper not yet published but in press with reference criteria
e	All published papers whose contents are related to the research questions formulated	e	Duplicate papers were excluded (papers that are similar in content or address same topics or have the same contributions)

2.2.2 *Scrutiny*

From Figure 3, 2,395 prospective studies were obtained during the first search process. Therefore, scrutiny was necessary to streamline the studies to relevant ones. First, the title of each study was considered then, their contents were briefly studied. Meanwhile, all the papers that do not reflect the topic of discussion or are incapable of addressing any of the formulated RQs were expunged from the relevant studies list. Also, only studies written and published in English language from peer-reviewed journals referred conference proceedings, workshops, book chapters, IEEE published and relevant journals were considered for inclusion in the list of relevant studies. However, when multiple copies of the same paper appeared, the most complete, recent and improved journal is included in the search processes while the irrelevant articles which does not address problem considered in this study are excluded.

2.2.3 *Quality assessment*

The quality assessment score shown in Table 2¹ utilises a score criteria described in this section. Publication topics, CA research that addresses general context interactions issues such as acquisition, modelling and inference or research on AR or those that focus on DM issues, such publication is assigned score = 1. Literature which discussed both CA and AR; CA and focuses on DM; or AR and focuses on DM, such publication(s) is

assigned score = 2. The paper that dealt with CA and AR and focuses on DM related issues such paper is assigned score = 3. The entire scores are then summarised in each column and used to determine the quality of the paper(s). Since all were summed up to 100%; the average quality score obtained from Table 2 is 33.3%. Any score below 33.3% is of low quality, hence, research in such area is low, thus opens the opportunity for further research. In addition, publications which address either one of CA or AR or DM only are $59 * 1 = 59$ (55.14%); those with both CA and AR, CA and DM or AR and DM are $12 * 2 = 24$ (22.43%) and those that addressed CA with AR and DM are $8 * 3 = 24$ (22.43%).

2.3 Data collection

After the paper inclusion and exclusion criteria process, the collated papers which represents dataset of the primary studies are presented in Table 3. In the data collection stage, the authors' extracted different attributes from the primary studies. The extracted attributes from various papers detail are as follows:

- publisher (IEEE, Springer, ScienceDirect, ACM, other journal)
- type of publication (conference, book chapter, workshop, journal)
- published (journal or conference name)
- publication name
- country
- year of publication
- classification based on research activity
- classification based on techniques
- classification based on research contribution
- classification based on problem domain.

2.3.1 Classification approach

Scientific research papers are normally categorised based on their methods and approach. Glass et al. (2002), Musumba and Nyongesa (2013) and Ramesh et al. (2014)² have identified context acquisition, sensing, modelling, inference (Lee et al., 2014; Schmidt, 2013; Champiri et al., 2015) as contending issues in context-awareness services using context interaction. The classification used in this study is employed from similar SLR research articles such as Ampatzoglou and Stamelos (2010), relating the whole idea to the concept of research of interest conducted in this study. The key scientific approach categories explained in Glass et al. (2002) includes empirical, exploratory and descriptive approaches. Empirical reveal findings of direct or indirect observation of physical or actual subjects; exploratory approach assists in the determination of the best research design, the data collection method and selection of subject; it is used under a situation where the research problem is not clearly defined. On the other hand, a descriptive paper normally describes a system such as tool or methods.

Table 2 Quality assessment

<i>References</i>	<i>1</i>	<i>2</i>	<i>3</i>	<i>References</i>	<i>1</i>	<i>2</i>	<i>3</i>
C1	✓			C35	✓		
C2	✓			C36		✓	
C3	✓			C37		✓	
C4	✓			C38		✓	
C5	✓			C39		✓	
C6		✓		C40		✓	
C7	✓			C41	✓		
C8	✓			C42	✓		
C9	✓			C43	✓		
C10	✓			C44		✓	
C11	✓			C45	✓		
C12	✓			C46		✓	
C13		✓		C47		✓	
C14	✓			C48		✓	
C15	✓			C49	✓		
C16	✓			C50	✓		
C17	✓			C51	✓		
C18	✓			C52	✓		
C19*			✓	C53	✓		✓
C20	✓			C54	✓		
C21	✓			C55	✓		
C22	✓			C56	✓		
C23			✓	C57	✓		✓
C24	✓			C58	✓		
C25	✓			C59	✓		
C26	✓			C60	✓		
C27	✓			C61	✓		
C28	✓			C62	✓		
C29			✓	C63	✓		✓
C30			✓	C64	✓		✓
C31	✓			C65	✓		
C32		✓		C66	✓		
C33	✓			C67	✓		
C34	✓			C68–C75	✓		

Note: * is the baseline line study that led to the motivation of this SLR in disaster mitigation presented in this paper.

Table 3 Study dataset

S.N.	Publisher	Citation_type	Publication_name	Author_country	Year	Method	Approach
C1	Other	Journal	<i>Network and Heterogenous Media</i>	Switzerland	2011	Experiment	Empirical
C2	Other	Book	<i>Atlantis Press Review</i>	UK	2010	Survey	Empirical
C3	Elsevier	Journal	<i>Computer Vision & Image Understanding</i>	Germany	2012	Survey	Descriptive
C4	IEEE	Journal	<i>Integrated System</i>	Pittsburg	2006	Experiment	Experimental
C5	Elsevier	Journal	<i>Computers in Industry</i>	Australia	2014	Experiment	Exploratory
C6	Springer	Journal	<i>ICHIIT</i>	Republic of Korea	2011		Descriptive
C7	Other	Journal	<i>Sensor Fusion</i>	Republic of Korea, Turkey	2014		Exploratory
C8	Springer	Journal	<i>Lecture Notes in Computer Science</i>	USA	2004		Empirical
C9	Elsevier	Journal	<i>Expert Systems with Applications</i>	Turkey	2012	Experiment	Descriptive
C10	Springer	Journal	<i>Virtual Reality</i>	USA	1998	Case study	Descriptive
C11	Elsevier	Journal	<i>Systems & Software</i>	Taiwan	2014		Descriptive
C12	Elsevier	Journal	<i>Expert Systems with Applications</i>	South Korea	2009	Case study	Exploratory
C13	Elsevier	Journal	<i>Pattern Recognition Letters</i>	United Kingdom	2014		Empirical
C14	Elsevier	Journal	<i>Systems & Software</i>	Spain	2013		Exploratory
C15	IEEE	Conference	<i>IEEE International Conference on Systems & Cybernetics</i>	China	2009		Empirical
C16	IEEE	Journal	<i>IEEE Transaction</i>	Korea	2011	Case study	Descriptive
C17	IEEE	Journal	<i>Transaction</i>	Washington	2010		Empirical
C18	ACM	Journal	<i>Context-Aware Application</i>	USA	2012	Case study	Descriptive
C19*	Elsevier	Journal	<i>Ad Hoc Networks</i>	India	2014		Empirical
C20	Elsevier	Journal	<i>Context-Aware Application</i>	China	2012		Descriptive
C21	IEEE	Workshop	<i>Workshop on Wireless Local Networks</i>	Canada	2013	Survey	Descriptive
C22	Elsevier	Journal	<i>Expert Systems with Applications</i>	Taiwan	2011	Experiment	Empirical
C23	Elsevier	Conference	<i>Energy Procedia</i>	Korea	2012		Descriptive
C24	Elsevier	Journal	<i>Swarm & Evolutionary Computation</i>	India	2014		Empirical
C25	Elsevier	Journal	<i>Disaster Risk Reduction</i>	UK	2013	Survey	Descriptive

Table 3 Study dataset (continued)

S.N.	Publisher	Citation_type	Publication_name	Author_country	Year	Method	Approach
C26	IEEE	Conference	<i>Pervasive Computing & Communication</i>	Queensland	2004		Exploratory
C27	Elsevier	Journal	<i>Science of Computer Programming</i>	Brazil	2013		Descriptive
C28	Elsevier	Journal	<i>Expert systems with Applications</i>	Republic of Korea	2012	Experiment	Empirical
C29	Elsevier	Journal	<i>Pattern Recognition</i>	Australia	2015	Experiment	Empirical
C30	Elsevier	Journal	<i>Information Fusion</i>	France	2015		Empirical
C31	ACM	Journal	<i>IMCOM</i>	South Korea	2013		Exploration
C32	IEEE	Journal	<i>Sensors</i>		2014		Empirical
C33	Elsevier	Journal	<i>Pattern Recognition</i>	Italy	2014	Experiment	Empirical
C34	IEEE	Conference	<i>Context-Aware Application</i>	Iran	2009		Descriptive
C35	Springer	Journal	<i>Next Generation Data Technology</i>	UK	2011	Survey	Descriptive
C36	Springer	Journal	<i>Lecture Notes in Computer Science</i>	Japan	2013	Experiment	Empirical
C37	Springer	Journal	<i>User Centric Technologies & Applications</i>	Spain	2011		Descriptive
C38	Springer	Journal	<i>Multimedia Tools</i>	Belgium	2014		Empirical
C39	Springer	Journal	<i>Lecture Notes in Computer Science</i>	Spain	2011		Empirical
C40	Springer	Journal	<i>Transactions</i>	China	2007	Survey	Descriptive
C41	Springer	Book	<i>Context-Aware Sensing</i>		2006		Descriptive
C42	Springer	Conference	<i>Lecture Notes in Computer Science</i>	Singapore	2007	Experiment	Empirical
C43	Springer	Journal	<i>Artificial Intelligence Review</i>	Republic of Korea	2012	Survey	Descriptive
C44	Springer	Journal	<i>Pervasive Ubiquitous Comp.</i>	Portugal	2013	Survey	Descriptive
C45	Springer	Book	<i>Lecture Notes in Computer Science</i>	France	2013		Empirical
C46	Springer	Journal	<i>Ambient Intelligent Human Comp.</i>	Portugal	2014	Case study	Exploratory
C47	Springer	Journal	<i>Lecture Notes in Computer Science</i>	China	2013	Case study	Exploratory
C48	IEEE	Workshop	<i>Context-Aware Crowd Sensing</i>	France	2013		Empirical
C49	Springer	Journal	<i>Pervasive Ubiquitous Computing</i>	Italy	2011	Experiment	Empirical
C50	ACM	Journal	<i>Human Computer Interaction</i>	New York, Georgia, Atlanta	2001	Case study	Descriptive

Table 3 Study dataset (continued)

S.N.	Publisher	Citation_type	Publication_name	Author_country	Year	Method	Approach
C51	Elsevier	Journal	<i>Pervasive & Mobile Computing</i>	Greece	2015	Experiment	Descriptive
C52	ACM	Workshop	<i>International Workshop on Mobile Sensing</i>	Turkey	2012		Empirical
C53	IEEE	Workshop	<i>Context-Aware Application</i>		1994		Descriptive
C54	IEEE	Journal	<i>IEEE Transaction</i>	Japan	2007		Empirical
C55	IEEE	Journal	<i>Activity Recognition Applications</i>	Germany	2008		Empirical
C56	Elsevier	Journal	<i>Systems & Software</i>	USA	2010		Exploratory
C57	Elsevier	Journal	<i>Systems & Software</i>	Belgium, China	2011	Case study	Descriptive
C58	Elsevier	Journal	<i>Computer Standards & Interfaces</i>	Taiwan	2014		Empirical
C59	Elsevier	Journal	<i>Computer Standards & Interfaces</i>	Taiwan	2014		Exploratory
C60	Elsevier	Journal	<i>Expert Systems with Apps.</i>	Spain	2012		Descriptive
C61	ACM	Journal	<i>Ubiquitous Computing</i>	USA	2014	Experiment	Empirical
C62	IEEE	Conference	<i>International Conference Proceeding on Multimedia & Ubiquitous Computing</i>	Korea	2007		Descriptive
C63	IEEE	Journal	<i>Acceleration Based Activity</i>	China	2010		Empirical
C64	Elsevier	Journal	<i>Expert Systems with Apps.</i>	Republic of Korea	2014		Empirical
C65	Elsevier	Journal	<i>Medical Eng. & Physics</i>	Ireland	2014		Empirical
C66	Elsevier	Journal	<i>Pattern Recognition Letters</i>	UK	2011		Empirical
C67	Elsevier	Journal	<i>Expert Systems with Applications</i>	Spain	2014	Experiment	Empirical
C68	IEEE	Conference	<i>Pervasive Computing & Technologies</i>	Italy	2011	Experiment	Empirical
C69	Elsevier	Journal	<i>Accident Analysis & Prevention</i>	Germany	2017	Case study	Empirical
C70	Elsevier	Journal	<i>Computer Communications</i>	Brazil	2017	Case study	Exploratory
C71	Elsevier	Journal	<i>Knowledge Based Systems</i>	Italy	2017	Experiment	Empirical
C72	Elsevier	Journal	<i>Journal of Parallel & Distributed Comp.</i>	China	2017	Experiment	Empirical
C73	Elsevier	Journal	<i>Artificial Intelligence in Medicine</i>	Italy	2016	Experiment	Empirical
C74	Elsevier	Journal	<i>Knowledge Based Systems</i>	Israel	2016	Experiment	Empirical
C75	Elsevier	Journal	<i>Knowledge Based Systems</i>	Taiwan	2017	Experiment	Empirical

According to Wohlin et al. (2012), three different methods for empirical investigation are used to evaluate new techniques, methods and tools. The investigation includes types of experiments, surveys and case studies. Experiments are criteria for exploring the relationships such as identifying the correlation between two variables. Surveys are suitable forms of investigation when the method of study is in use for a while. While case studies are used for similar situations like that of experiments, but in this aspect they are observational studies (Wohlin et al., 2012; Kitchenham et al., 1995). The empirical validation types identify for papers listed in Table 3 are summarise in Table 4. A similar empirical classification used in Dybå and Dingsøy (2008) were adopted where a relevant concept in the domain is considered.

Table 4 Empirical methods in the selected study

<i>Empirical method</i>	<i>Explanation</i>
Case study	A situation where-by sets of people are asked to perform certain tasks in a controlled environment is an experiment. In the end, results are generated from inspecting the task outcome.
Experiment	Sets of people are used to participate in questionnaire filling directly or indirectly at times it may be through the internet.
Survey	Results are derived from the valid answer supplied by respondents to the questionnaires.

Source: Ampatzoglou and Stamelos (2010)

2.3.2 Data analysis

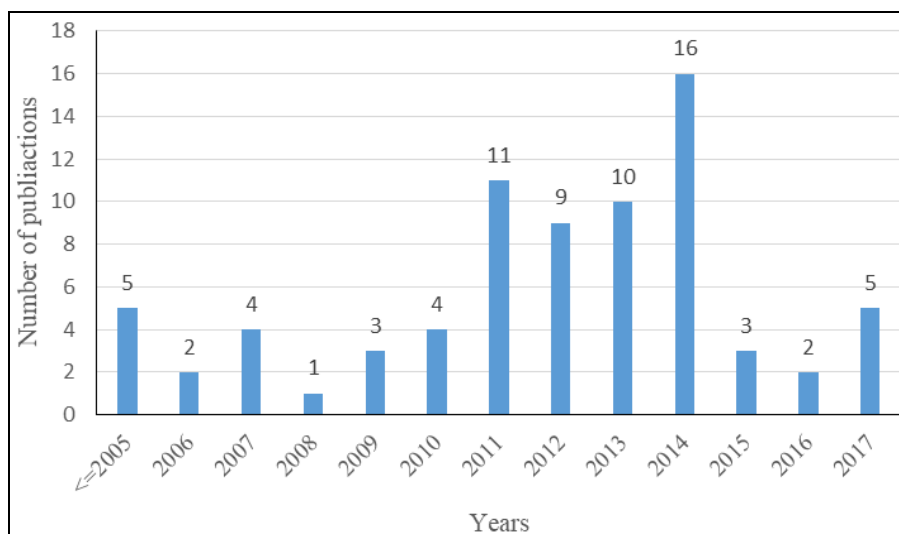
The data collection specified in Section 2.3 and summarised in Table 3 are statistically analysed to investigate the following:

- The number of publications (primary studies) showing techniques used as published per year addresses Q1.
- The number of researches each country and continent produced in the last 12 years also addressed Q1.
- The strength and weaknesses of the identified techniques concerning approaches used in CA AR researches that investigates DM specifically addresses Q2.
- The number of studies that shows contributions mapped into research focuses and types is addressed in Q3
- The number of studies that employed each research approach related to disaster issues with their methods addresses Q4.

In Figure 4, the data analysis was conducted based on the selected primary studies. From the study, 2014 had 16, as the highest publications, 2017 has five and the lowest is 2008 having one. Most of the reviewed papers were on context-awareness and activity recognition (CAR) but only a few of them addressed issues in crowd disasters, while some discussed other disaster issues. The reason for context-awareness was because many fields are faced with the issues of context acquisition, sensing and modelling. Most importantly, the emergence and penetration of smart mobile devices has given rise to the

development of CA system which utilises sensors to collect relevant data about users to improve different users services (Unger et al., 2016). Similarly, contextual information has equally influenced building of systems to meet users' needs more efficiently and practically (Lee et al., 2017). In addition, the reasons highlighted, by Unger et al. (2016); Lee et al. (2017), CA is applicable to many other domains including adaptation of human robot behaviour by Nessi et al. (2016) and the prevention of road accident through Vehicle safety actuators by Böhmländer et al. (2017). The use of CA application and its techniques has been a hot research area in recent time meanwhile, the research is still growing.

Figure 4 Number of selected primary studies by year of publications (see online version for colours)



3 Results

This section presents the results of statistical analysis using the dataset of the primary studies shown in Table 3. It is important to state that description of context-interaction issues using AR accompanied with the discussion of RQs in Section 2 are described in this section:

- Q1 What are the extent of CA research that is related to AR?
- Q2 What are the strength and weaknesses of the existing context-awareness techniques with AR?
- Q3 What are the contributions in context-awareness with AR related to DM?
- Q4 Which of those context-awareness with AR researches addresses issues on crowd disasters?

Context-awareness is a field that promotes pervasive computing research through the use of smartphone device with the aid of inbuilt sensors. Context-awareness has been found to be useful in crowd monitoring and other disaster-related domain such as vehicle safety on the roads due to its ability to offers feedback using context-awareness when programmed with sensors on mobile Android applications (Böhmländer et al., 2017; Ramesh et al., 2014). Figure 4 earlier presents the number of selected publications by years from 2005 up to 2017 based on research domain under review as stated in Section 2. Figure 4, also provides information on the number of articles recorded by each type of publication. Specifically, journal, conference, workshop and book chapters as shown in Table 5. The years of publication considered are between 1994 to 2017, but articles that belong to 1994–2004 are recorded as (≤ 2005) using the primary studies as showed in Table 3. The most active research publication channels for CAR studies distribution published based on publishers name are ScienceDirect popularly known as Elsevier journals with 35 (46.7%), Springer 17 (22.7%), IEEE 15 (20.0%), ACM 5 (6.7%) and Google scholar 3 (4.0%) as shown in Table 5. However, 75 primary studies were identified during the search processes. Out of the 75, 60 were journal articles, eight were conference papers, four were workshop papers and three were contributions from book chapters. Research related to CA and DM was just three publications. This section describes the results of the studies based on the RQs formulated. Table 3, Figures 4 and 5 form the basis of discussions for RQs Q1–Q4.

Table 5 Number of publications per publisher

<i>Publisher name</i>	<i>#Publications</i>	<i>Percentage (%)</i>
ScienceDirect	35	46.7
Springer	17	22.7
IEEE	15	20.0
ACM	5	6.6
Google scholar	3	4.0

4 Discussions of results

To address Q1, Figure 5 presents the number of research activities by country of the research based on the corresponding author's name within the periods of 12 years using the details in Figure 4 and primary studies presented in Table 3.

Details of published papers are showed in Figure 4 and countries of research activities by authors are presented in Figure 5. Analysis in Table 6 addresses the RQ Q2, it revealed that 35 techniques were reported in this study from 1994 to 2017. The 35 techniques identified are shown in Table 6 alongside with their strengths, weaknesses and sources base on the SLR analysis. The outstanding research gaps from this study summarised in Figure 6(b), where less attention was paid to metric contribution, architecture and model. Although, many researches have been done on the framework, but less numbers of research were found on DM. Further research initiative in a crowd monitoring using AR is still possible (Abowd et al., 1998).

Table 6 Existing context-awareness techniques, strengths and weaknesses

<i>Technique proposed</i>	<i>Strengths</i>	<i>Weaknesses</i>	<i>References</i>
Bluetooth location network (BLN)	It interacts with users that have other devices such as a Bluetooth-enabled terminal or any mobile terminal and a Bluetooth badge	Only relied on sensor device but cannot cater for a non-sensitive object in the environment	González-Castaño et al. (2005)
Fuzzy petri nets (FPN)	It provides an opportunity for sensing contextual information about learner and the learning content with the help of metadata using English learning as a focus	Cannot handle sensors related to other subjects matter	Jian et al. (2008)
Ontology	It provides knowledge reuse, sharing, logic-inference, syntactic and interoperability services	Developing the ontology is a major challenge at the level interface for the system.	Neves et al. (2014)
Fuzzy weighted average	The technique used learning strategies and pedagogical framework to enable students to participate in context-awareness synchronous learning environment	Unable to cater for students learning in asynchronous learning environment does not support the actors in the environment	Huang et al. (2008)
Efficient fuzzy weighted average (EFWA)	It employed FWA algorithm with the combination of $O(n \log n)$ computation time to generate weighted average result	It does not allow a linear relationship	Böhmländer et al. (2017)
Ontology rule-based matching	This build technique ontology for the context-aware mobile Chinese language learning (CAMCLL) approach to understanding the location and specific domain from context acquisition using interpretable machine mechanism	It is language specific in a particular domain	Al-Mekhlafi et al. (2009)
Signal of wireless LAN	It uses localisation in a building with the help of transactions, using algorithm for pairs of service set identifier and signal strength	It is restriction to only indoor system	Noh et al. (2012), Wang et al. (2012)
Context modelling	It provides a high level of abstraction for context information. It does this in different ways for modelling context for example key-value pairs and context category	It does not permit reusable architecture/mechanism for context acquisition. Thereby prevents room for adaptation	Mostefaoui et al. (2004), Henriksen and Indulska (2006), Chiti et al. (2012)
Crowd sensing	Employed user context sampling function by using minimum sampling interval for costs	It has a minimum coverage size which it cannot exceed	Carreras et al. (2013)
Data mining	Employed data synthesis with a sequential pattern discovery and association rule mining to acquire the candidates' data	It is prone to inaccuracy of data and possible misuse of information may occur	Kwon et al. (2004), Viswanathan et al. (2012)

Table 6 Existing context-awareness techniques, strengths and weaknesses (continued)

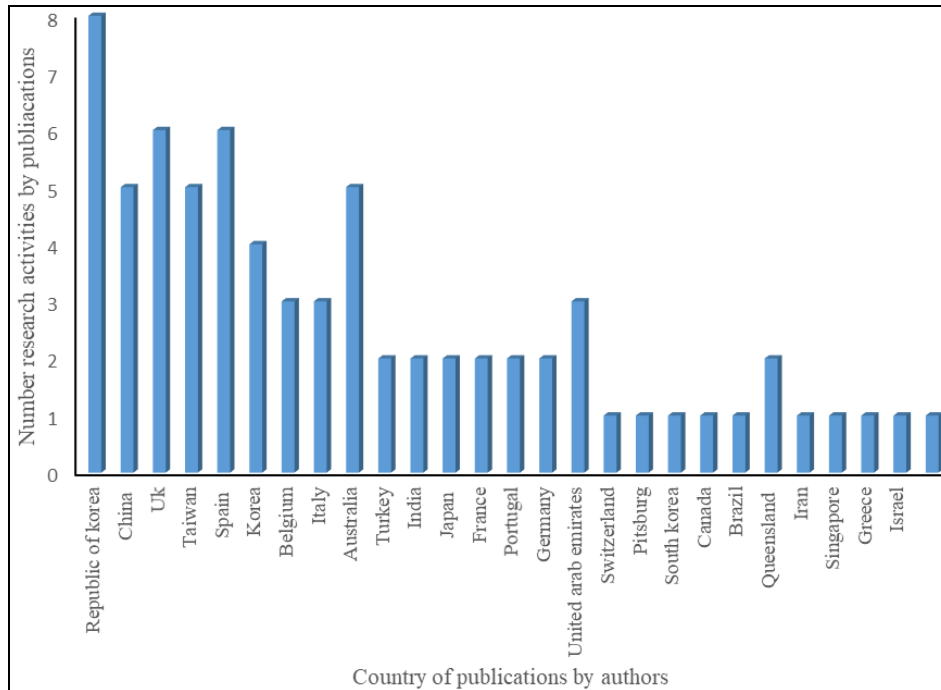
<i>Technique proposed</i>	<i>Strengths</i>	<i>Weaknesses</i>	<i>References</i>
Machine learning	It permits mobile computer that is capable of knowing particular user's states and learn instantly to adjust its settings based on past knowledge/experience. Example of algorithms employed are decision tree (DT) specifically (J48), random forest (RF), naive Bayes (NB), sequential minimal optimisation (SMO) and K-nearest neighbour (KNN)	It relied on experience. Therefore, it may not be suitable for just any system	Ramesh et al. (2014), Krause et al. (2006)
Sigmoidal-shaped	Provide dynamic assistance to the operator of robotics during cooperation in a specified scenario	Stiffness ellipsoid of the manipulator for addressing human-like behaviour	Nessi et al. (2016)
Context-aware Recommender system	Extraction of data from rich set of mobile sensors to infer unexplored user contexts in an unsupervised manner	Addition of context to recommendations' system difficulties	Unger et al. (2016)
Sensor-based	Detection of potential collisions to trigger safety actuator before the occurrence of road accident	Limited to certain scenarios which must be specified within coverage time	Böhmländer et al. (2017)
Agent-based simulation, social force model, motion history image	Each of this technique has been employed with success for crowd evacuation at the point of recovery	Each of them has no connection with context-awareness or activity recognition focus but have potential to solve disaster related problem	Zaharia et al. (2009), Mehran et al. (2009), Alqayssi and Sasi (2013)
Group-based (GCHAR)	Efficient group-based context-aware classification for human activity recognition to exploits hierarchical group-based scheme to improve classification efficiency	Human activity recognition scheme classification accuracy is relatively low	Cao et al. (2017)
Artificial neural network	This technique use a solid cluster algorithm to elicit model for context-awareness in a sensor network	It requires a greater computational burden and is prone to over fitting	Huang et al. (2008), Petersen et al. (2008)
Hidden Markov models (HMM)	This technique allows very rich sequence of sub-activities to be recognised without the system having to learn separate HMM for each sequence	It takes longer time in term of memory computation for a sequence of length n	Mehmood and Jawawi (2013), Chiti et al. (2012)
Time domain (TD)	A statistical technique used for feature extraction from raw sensor data of activity recognition. An example of TD features are mean, standard deviation, root mean square to mention a few	No criteria to know the appropriate TD features, thus some salient features that can improve accuracy were not considered in activity recognition model for disaster related problem	Khan et al. (2014), Ramesh et al. (2014)
Frequency domain (FD)	The extraction of the coefficient of sensor signals of TD features using Fast Fourier Transformation (FFT)	It is very difficult to know the frequency domain features to be extract from the sensor's raw data at initial stage.	Khan et al. (2014), Ramesh et al. (2014)
Time domain and frequency domain (TDFD)	The tendency to produce higher performance in term of accuracy, precision and recall using the combination of TDFD extract features	Sometimes it generates poor performance as the result always depend on the quality of raw sensor data of activity recognition	Khan et al. (2014), Ramesh et al. (2014)

Table 6 Existing context-awareness techniques, strengths and weaknesses (continued)

<i>Technique proposed</i>	<i>Strengths</i>	<i>Weaknesses</i>	<i>References</i>
Bayesian network	Used for categorisation of the associated sensor's reading with a probabilistic graphical model in direct acyclic graph	Does not recognised the role of the place where a user is located therefore is not suitable for context-awareness	Fabiana et al. (2013)
Bayesian approach to sensor-based	Classify contexts of a mobile device users in their normal daily activities with audio features measured in a home scenario	It is computationally expensive and indoor specific	Gonzalo et al. (2011)
K-nearest neighbourhood	It employs classification to predict correct context of the user at each time step in a scenario using metric distance as a model	Only suitable for data at runtime and it cannot handle large data	Kose et al. (2012)
Model driven development (MDD)	This technique used machine – readable model at various level of abstraction as its main artefacts. It permits development of services by separating context information from business logic	It cannot handle contextual adaptation at runtime for context-aware service	Mostefaoui et al. (2004), Kwon et al. (2004), C53
Hierarchical temporal memory (HTM)	It provides possible usage for context identification, thereby facilitating context-aware service for mobile user	It only caters for mobile phone context data in real time	Roggen et al. (2011)
Service directory mechanism	This technique employed query to perform searching of the most preferred service containing relevant information for users' context in a centralised environment	It cannot handle service of information for users' context in a decentralised peer-to-peer environment	Kose et al. (2012)
Fuzzy adaptation process	This technique apply fitness function to set policies corresponding to services candidates for n context attributes to be monitored by the policies as input and generate ranked list of policies p as output with distance-based algorithm	This technique does not cater for large spectrum of applications and domains	Ning et al. (2011)
Maximum likelihood method	This technique used normalisation and regression formula assign value for student and examination to determine analysis of students results and learning context so as to know the students' performance in examination through support learning context-aware and auto-notification system	Cannot handle contextual activity in the system apart outside the examination problem	Pedro et al. (2013)
Designed-based	This technique used a Likert scale of 25 questions in a questionnaire to evaluate students satisfaction with independent sample t-test for 10 students both male and female of equal size to determine learning satisfaction on the developed context-awareness system for elementary Chinese learning	It is a language specific and few sample size used may require more investigation	Seungwok and Hee (2012)

Table 6 Existing context-awareness techniques, strengths and weaknesses (continued)

<i>Technique proposed</i>	<i>Strengths</i>	<i>Weaknesses</i>	<i>References</i>
Natural language	It employ mapping of sensor values domain and user values domain with subdomains such as: specific, generic and aggregated in order to offer mobile assistant capable of resolving ambiguity that users can face during interpretations of rules when defining high-level context by tagging other users, objects or any events, thereby facilitating the adoption of context-aware computing in everyday life	It is only suitable for mobile device. In ability to differentiate users contextual information	Kwon et al. (2004)
Decision support system	It employs optimal model and logistics strategy algorithm with the help of mathematical model to assist the actors for human monitoring	Does not support real time for unforeseen situation which often result to disaster	Meditsko et al. (2015), Phan (2014)
Collaborative filtering	This technique is an alternative to information filtering algorithm, it provides users with effective tools that can help to discover difficult items that ordinarily may be impossible	As the no of items and users grow it will suffer serious scalability problem which may not be applicable to crowd scenario	Kitchenham and Charters (2007), Jian et al. (2008)
Artificial intelligence	It employed dynamic artificial neural network dimension to offer personalised services to users by computing context-weight as a relationship between context and user profile	Difficult for human control. It is only suitable when the system is intelligent inclined	Huang et al. (2008), Davies et al. (1995)
Neural network-based WLAN	This technique employs signal strength features for training the employed back-propagation neural networks for positioning learner's location	The optimal sample parameter cannot be determined and it is not suitable for physical context	Hsieh et al. (2007), Huang et al. (2008)

Figure 5 Research activities by authors' country name (see online version for colours)

4.1 Trends in context-awareness techniques using RQ2

It is very rare, to see disciplines and industries where CA related applications are not applicable, as a result of context and its relationships with environmental and situational changes in our daily life. This statement support previous research in the literature from Schilit et al. (1994), Schmidt (2013) and Ramesh et al. (2014) with explanations provided for context and the origin of 'CAC' as proposed in Ampatzoglou and Stamelos (2010) also cited in Wohlin et al. (2012). Several concepts has been proposed in the area such as education (Glass et al., 2002; Kitchenham et al., 1995), building automation (Dybå and Dingsøy, 2008), healthcare (Unger et al., 2016), library management (Lee et al., 2017), movie industry (Nessi et al., 2016), business (Böhmländer et al., 2017), restaurant (Ramesh et al., 2014) and banking institution (González-Castaño et al., 2005) to mention just a few. Several techniques have been proposed for context-awareness related research either for services or applications-based research. The techniques include collaborative filtering (Jian et al., 2008), FPNs (Neves et al., 2014), artificial intelligence (Huang et al., 2008), ontology for contextual entity description (Glass et al., 2002; Al-Mekhlafi et al., 2009) and activity-based entity relationship (Noh et al., 2012). In addition to Glass et al. (2002), Jian et al. (2008), Neves et al. (2014), Huang et al. (2008), Al-Mekhlafi et al. (2009) and Noh et al., (2012), 27 additional techniques have been identified from literatures as part of methods that can be use when undergoing research using context-awareness computing, making a total of 35 techniques reported in this study from 1994 to 2017. The SLR conducted in this study, will further provide evidence in this research domain to know what has happened over the years, and beyond 2000–2007 as

presented in Kitchenham and Charters (2007) to date. Also, in Wohlin et al. (2012) and Wang et al. (2012), though both of them (Wohlin et al., 2012; Wang et al., 2012) did not indicate the years of coverage in their studies. The topics of concerns investigated are; CAC research with ‘context-aware systems’; ‘CMPMFs’ and ‘context-awareness in healthcare’. Though, the literature review presented in the previous study, are more general since they focused on how to identify the likely applications areas which existing studies have covered in one of the RQ employed in Section 2. However, this study is the first to specifically focus on SLR for disaster related problems using AR based on context-awareness. We are not aware of an existing SLR study in this domain from the literatures. Disaster management is everybody business in our society. Therefore, current research on DM is very essential to minimise risk in human lives to impact the growth and development of any nation through the safety of human lives and properties.

Research in the area of disaster management has grown rapidly over the years. However, most researchers have concentrated efforts on techniques such as agent-based simulation (Zaharia et al. 2009); social force model (Mehran et al., 2009); motion history image on crowd evacuation (Alqaysi and Sasi, 2013), for a reason, knew to them and the fact that favourable outcome was obtained from these techniques. The three techniques; agent-based simulation (Zaharia et al. 2009); social force model (Mehran et al., 2009); and motion history image (Alqaysi and Sasi, 2013) does not focused on the utilisation of sensor’s data using AR from mobile phone sensing (MPS) and context-awareness for mitigation of crowd disaster in their research.

Most importantly new trend of research on CA was remarked in Ramesh et al. (2014) as “the critical consideration of their research on context-aware computing and wireless sensor network (CAC-WSN) to improve the quality and reliability of sensing information by utilising the potential of CAC.” In Aloulou et al. (2015), it was stated that any decision make using CA systems are very crucial and have a meaningful impact on the users, as a result, context-awareness technique can help to realise accurate decision needs for monitoring real life situations in a crowd, during emergency to minimise risk and prevents danger pose to human lives. Therefore, as observed in Charith Perera et al. (2015), the benefits of the proposed solution with AR and context-awareness but not limited include the following:

- 1 the ability of people to get relevant information according to their need using CAC
- 2 sensors learns about human being through context
- 3 context is used for real-time decision support and activity, which has helped the smartphone to become smarter.
- 4 the future holds lots of promises for CA applications.

4.1.1 Context-related information that has been used for DM

Time domain (TD) is a feature extraction method conventionally used in AR research domain (Ramesh et al., 2014; Khan et al., 2014; Kawarehara et al., 2007). Example mean, standard deviation, (STD), minimum (min.), maximum (max.), range, variance, root mean square (RMS) and interquarter tile in Table 6, the TD features that has been extracted and used for AR specifically for prediction of crowd disaster are mean along x, y, z axes; STD along x, y, z axes; correlation coefficient (COR) of xy, xz, yz axes;

RMS of x, y and z axes (Ramesh et al., 2014). In addition to this, the frequency domain (FD) for RMS has equally been used. Consequently, the TD and FD metrics (mean, STD, COR, RMS and FFT_RMS) have produced good result in AR research (Helbing et al., 2007), outside their commonly used as feature extraction methods in the AR domain (Khan et al., 2014; Guiry et al., 2014). The features are of light weight in nature this potential makes the features suitable for smartphone with average memory size and processing speed. Therefore, the features require less battery consumption of smartphone device when used to monitor movement behaviour of people in a crowd scenario. Most importantly, some salient features capable of producing higher accuracy for early detection of danger using AR in a crowd, but shows good result in other domain not considered in the previous disaster domain research are among the challenges that this SLR equally identified as part of finding.

5.1 Focus of the research

The selected studies were divided into five research focus as shown in Figure 6(a) based on the research topics considered in this paper. The keywording method described in software engineering (Petersen et al., 2008) in combination with Mehmood and Jawawi (2013) were studied and apply to the research domain under consideration in line with issues, relevant area and problems. The problems include application, context acquisition, sensing, and modelling (context interaction), AR prediction, context reasoning and DM.

5.1.1 Application

This category mainly included papers that report facts about applying the concept in practice. For instance, papers that present model or method using existing approaches or proposal to develop a system or prototype applied to a specific scenario. These are found in nearly all literature but we only concentrate on those publications related to topic that focuses on problem of interest.

5.1.2 Context acquisition, sensing and modelling (context interaction)

The context acquisition, sensing and modelling are pertinent issues associated with context-awareness application referred to as context interaction. This includes studies that describe a novel method or model and any of those papers which propose a solution or contribute accordingly by addressing either of these issues in their research paper.

5.1.3 Activity recognition accuracy prediction

Activity recognition accuracy (ARAC) prediction includes data analysis research papers. The papers focused on a novel method, model, architecture or framework in human activity monitoring. They use classification algorithms in carrying out their data analysis mostly to determine ARAC.

5.1.4 Context reasoning

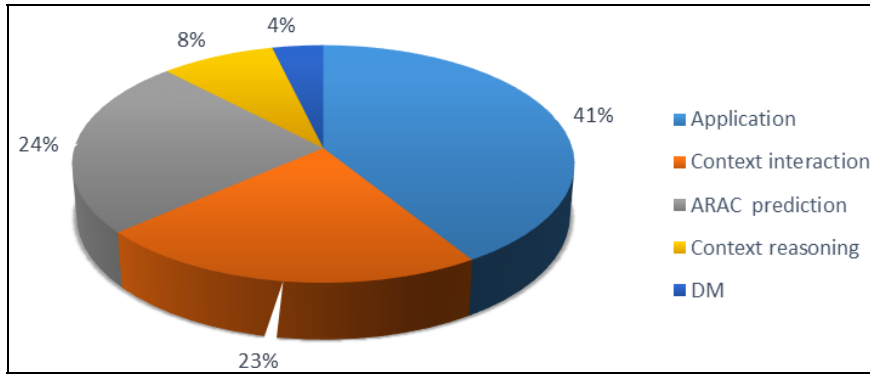
The context reasoning is papers that focused on novel method, model and having elements of clever technique such as artificial intelligence (Tuan et al., 2012), ontology

(Fernandez-Caballero et al., 2012). We identify some ontology-based study in connection with AR and statistical inference.

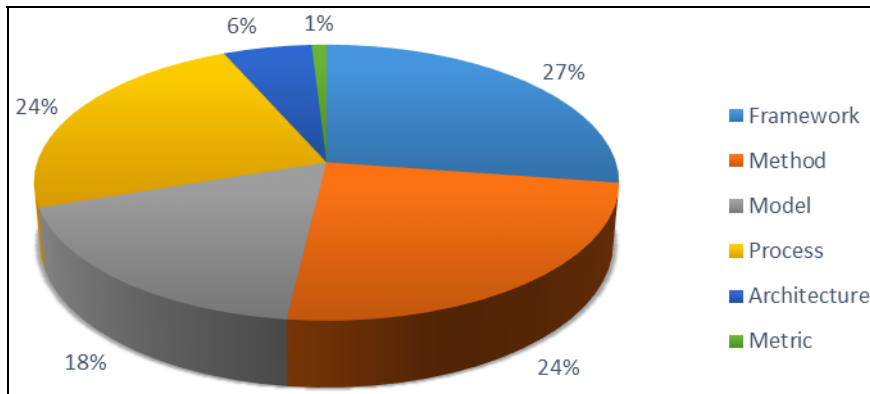
5.1.5 Disaster mitigation

This category of papers propose a unique model and utilised existing methods while other authors contribute frameworks (Abowd et al., 1998) by addressing problems relating to crowd and stampede (Ramesh et al., 2014) detection using the topic of concern in this study area. Figure 6(a) suggests that few of the selected primary studies from the entire publications addressed issues of DM based on crowd scenario and other related disaster control which are 4 (4%) of all the 75 publications found useful in this SLR. The summary of the analysis explained in Sections 5.1.1–5.1.5 depicts Figure 6(a).

Figure 6 (a) Research focus (b) Research contributions (c) Research types (d) Mapping flow of the research contributions and types based on the research focuses on the study (see online version for colours)

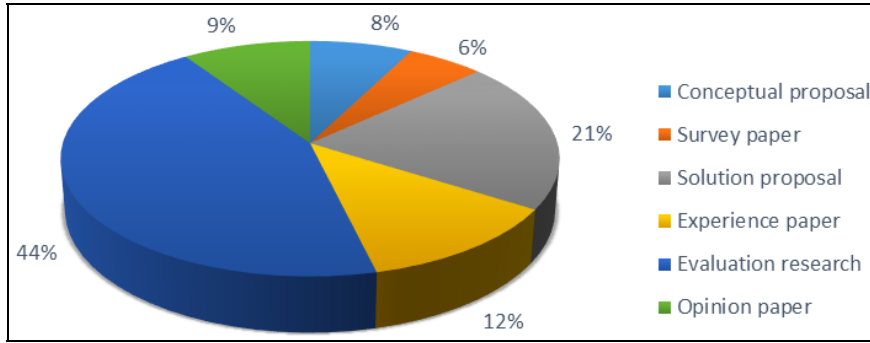


(a)

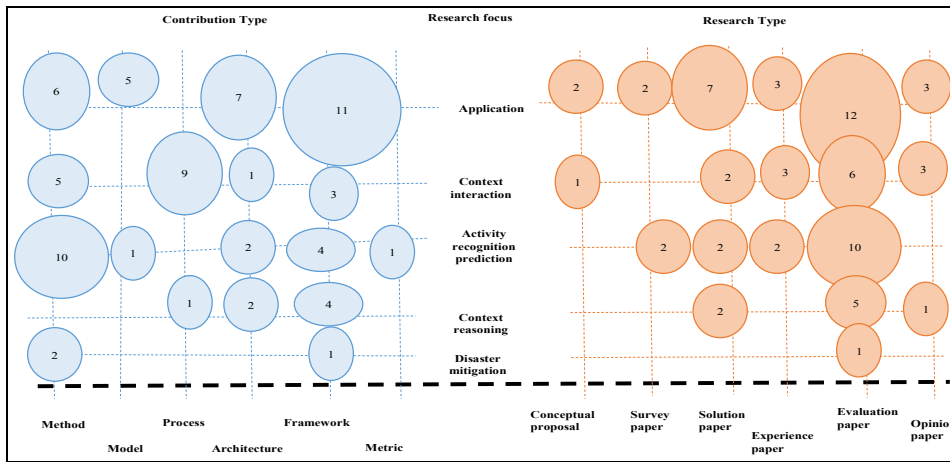


(b)

Figure 6 (a) Research focus (b) Research contributions (c) Research types (d) Mapping flow of the research contributions and types based on the research focuses on the study (continued) (see online version for colours)



(c)



(d)

5.2 Research contribution type (Q3)

This section provides answer to the Q3 as highlighted in Section 2.3.2. It is divided into six different categories and they are described in Section 5.2.1. Part of them appear as the title of the paper identified as a primary selected study.

5.2.1 Method

This describes contribution that specifically provides a novel method or improved approach or a specific way of how context can be composed or adapted for CA about human activity recognition (HAR). It is used while addressing some of the research focus.

5.2.2 Model

It describes contributions that present a conceptual discussion or comparative study, explore relationships, identify challenges, types of classification for CA and AR related topic.

5.2.3 Process

They are papers that extend the CA modelling or approaches and provide a detailed discussion on the overall CA system development process. For example, a paper that presents an abstraction used to describe context interaction of users and the behaviour of people in an environment.

5.2.4 Architecture

It describes contribution that focus on architecture support for CA research and AR either in the form of design building or structure that can be integrated into a framework.

5.2.5 Framework

It refers to a basic conceptual structure or description of a process that constitutes a complex task in a contextual relation form. It is explained in a model form or predefined diagram that addresses CA and AR scenario.

5.2.6 Metric

This set of papers describes contributions related to preprocessing and general form of metrics that is use when dealing with context recognition and related research activity. For example different standard techniques are discussed in the paper. The summary of the analysis explained in Sections 5.2.1–5.2.6 depicts Figure 6(b).

5.3 Research type

The research type presents the kinds of research done in the primary study. We employed the scheme proposed by Weiringa et al., (2006) and that of Mehmood and Jawawi (2013) for the classification of research type. Descriptions of research type identified in the primary study are classified through Sections 5.3.1–5.3.6.

5.3.1 Conceptual

The papers that focused on things that is in existence in a unique way. They did not solve the problem but proposed a theoretical framework that may include some form of enhancement by adding a new characteristic.

5.3.2 Experience

This kind of paper reports on personal knowledge revealing past ordeals in their real life project. It usually presents details on how the project was accomplished, as well as how it was done for publication.

5.3.3 *Evaluation*

This type of paper tries to examine a solution that has been applied practically. Such papers normally explore the practical implementation of solutions and come up with results using case studies or field studies.

5.3.4 *Solution*

This provides a solution to any problem mentioned or described in the paper. It is achieved by contributing a novelty result through technique or model.

5.3.5 *Survey*

These papers deal with a general view of any research topic, in general, form (in-depth). It is regarded as an investigation of what other researchers have done in a particular domain identify the problem and proper solution. In this case, we focus on CA and AR related topics with emphasis on DM.

5.3.6 *Opinion*

This type of paper reports on a personal view of the author on true satisfaction or dissatisfaction on a specific method, or technique. Such papers present their view or feelings on how certain technique(s) should be realised. For the purpose of RQ Q4, Figure 6(a) provides answers to the Q4, four (4%) addressed issues related to DM in a crowd out of all the 75 selected primary studies, Ramesh et al. (2014); automatic video analysis of crowd disaster (Krausz and Bauckhage, 2012); and crowded surveillance scenes (Leach et al., 2014), though only Ramesh et al. (2014) proposed CA framework based on crowd abnormality monitor (CAM) and employed AR with machine learning approach to determine stampede occurrence. Ramesh et al. (2014) further remarked that higher ARAC is a necessity for effective and reliable stampede prediction in a crowded area. However, Cao et al. (2017) recently further remarked that HAR schemes classification accuracy is relatively low as well. Therefore, it is imperative to identify causes for the low accuracy and proper solution to improve the accuracy to prevent negative consequence the low accuracy can cause in a real life. Research types is summarised in Figure 6(c).

6 Research activity on crowd DM issue (Q4)

In Figure 6(a), few of the selected primary studies from the entire publications addressed crowd disaster issues and other related disaster control which is four (4%) of all the 75 publications. This indication further explains why there is need for further research in the domain. Despite the overwhelming interest of researchers in CAC and AR, efforts in MPS for CAC with AR (Chao et al., 2010; Ramesh et al., 2014) is still very low. This prompted the recent development for the existing approach as earlier stated in Section 4.3.6. The distribution of other research focus showed in Figure 6(a) include, 44 (41%), 26 (24%), 24 (23%) and nine (9%) on application (Abowd et al., 1998; Kuo et al., 2014; Hong et al., 2009; Leach et al., 2014), ARAC, context interaction and

context reasoning (Mehta and Banati, 2014; Gonzalo et al., 2011; Pozo et al., 2011; Lee et al., 2007) respectively. Meanwhile, stampede is a major cause of crowd disaster (Illiyas et al., 2013). According to Davies et al. (1995), search from media account and research results reported 215 human stampede identified worldwide, from 1980–2007 with over 7,067 deaths and over 14,000 injured. This pathetic situation influenced the interest of researchers worldwide in finding ways to reduce these lingering dangers in our society. Although risk from human natural hazards cannot be eliminated completely, rather it can be minimised through a systematic approaches using disaster risk reduction (DRR) (Gomez et al., 2009). This effort prompted the different methods, models, techniques and frameworks proposed in one study or the other using the technology under consideration. Consequently, the mapping of the research contributions and research types using the research main focus is summarised in Figure 6(d).

6.1 Related crowd disaster researches

Human stampedes is an hazards from mass gathering events and it has witnessed less attention worldwide while formulating risk management strategies (Illiyas et al., 2013). Research has shown that the mass gathering (crowd) originated from religious festivity (Illiyas et al., 2013). Crowd monitoring originated in 1995 when CCTV were used as means of monitoring human activities in a crowd with pattern recognition techniques (Davies et al., 1995). The technique which faced with the following short coming such as manually coordination, with lots human effort (tedious and cumbersome) above utilised physical appearance of security officers. These limitations led the proposed wireless sensor networks with wireless communication techniques for better situational awareness of monitoring people in crowded places (Gomez et al., 2009). However, it suffered from a high rate of false alarm, and the use of low power sensors in the study due to low temperature and acoustic sensors used makes the technique less reliable and undependable. The limitations of this study witnessed the introduction of an emerging technology with MPS by Roggen et al. (2011). Despite the novel contribution and solution proposed in Roggen et al. (2011), their approach suffers some weaknesses. Thus, the weakness and the stampede incident that claimed life of over 102 devotees in the State of Kerala in Southern India and the constant deaths of more than 100 people per year due to crowd disaster motivates the proposed CAC-WSN for a crowd DM by Ramesh et al. (2014). The study seeks to address HAR accuracy, poor real time information and high rate of false alarm for effective stampede prediction in order to mitigate crowd disasters as identified in Ramesh et al. (2014). However, Ramesh et al. (2014), was the first to introduce CA application using AR to investigate HAR accuracy as a basis for effective stampede prediction in mitigation of crowd disasters by developing multi-context-fusion which suffers some drawbacks led the study in this paper. CAness application availability on smartphone devices is a technology that helps to sensitise users of activities going on around them (Anguita et al., 2014). It has been used in prediction of web analytics (Kiseleva et al., 2013).

6.1.1 Research gaps and recommendations for future research direction

In Figure 6(b), the contributions from the primary studies shown in Table 3 shows that 29 (27%) of all the publications are related to the framework. The distribution of the existing techniques, their strengths and weaknesses are summarised in Table 6. Less

attention was paid to crowd disaster-related research with context-awareness applications and less emphasis is on context reasoning. Further research has the potential to provide efficient, accurate timeliness of sense information (ATSI) for the critical situation, reduce FNA; improve ARAC; utilisation of context information (SCI); and employing the use of MPS for reliable ARAC for stampede prediction in a crowd scenarios are discussed in Sections 6.1.1.1–6.1.1.6. The findings are summarised as follows: in Figure 6(a), less attention was paid to crowd disaster related research with CA application and context reasoning. Figure 6(b) showed that less attention is equally paid to metric contribution, architecture and model. Though many approaches have been reported on framework, only few study focuses on DM.

6.1.1.1 Accurate and timeliness of sensed information

ATSI for critical situations is important to react to human safety in real time (Ramesh et al., 2014). Figure 6 indicates that only one of the selected studies takes cognisance of this important issue of human safety in literature. Also, Fruin (1993) identified the importance of time and information as a critical factor responsible for the causes of stampede in any crowded area in its paper titled ‘Causes and prevention of crowd disaster’ where he proposed the first model to described stampede accident. Based on the first model, one can deduce the fact that any CA framework with the suggested characteristics can adequately support a risk reduction for stampede occurrence to prevent crowd disaster in a crowded area. The solution in previous research on Muslim pilgrimage in Hall et al. (2009), Helbing and Mukerji (2012) and Luštrek and Kaluža (2009), once led to structural alternation and re-arrangement of the pilgrimage, which prevented deadly occurrence of stampede during the Moitinho de Almeida (2016) and Ramesh et al. (2014) study, is a welcome development that needs continuation of such research to meet up with the present dynamic and technological advancement using novel approach.

6.1.1.2 FNA for effective mitigation of crowd disaster

FNA can be determine by false negative rate (FNR). FNR is one of the common measurements in a classification evaluation in AR and data mining research (Gomez et al., 2009). FNR is the number of misclassified negative activity meant to be accurate but predicted wrongly in any classification analysis (Ramesh et al., 2014; Alvarez de la Conception et al., 2014). Investigation of event prediction before any terrible accident happen is very important (Fradi and Dugelay, 2015). Although, it is applicable to classification evaluation but is very relevant when we talk of detection of abnormal movement in a crowd scenario as well. To the best of the researcher’s knowledge only four studies related to crowd monitoring found in literature are shown in Figure 6(a) (Ramesh et al., 2014; Roggen et al., 2011). Most importantly one of the study employed feature extraction method using raw sensor data from AR for the prediction of crowd disaster based on ARAC performance (Ramesh et al., 2014). However, major limitation of existing study is high FNA noticed in Ramesh et al. (2014), which is attributed to the ARAC of 92% reported. The ARAC is inadequate for stampede prediction during emergency in a crowd scenario. As a result of the high FNA, there is needed to extend the research conducted by Ramesh et al. (2014), thus form the reason for the SLR conducted in this paper. For example, a case of someone who is almost fallen in a critical moment and such situation is reported to the security or stakeholder as sitting

or standing require proper classification; the implication of this may lead to accident or pose a risk to human life if adequate attention is not taken, this scenario is trigger by FNA.

6.1.1.3 Context interaction (sensing, acquisition, modelling)

In Figure 6(a) context, interaction showed 13 publications from the selected primary studies that dealt with important issues in CA application. These are challenging issues as far as CA researches are concern (Musumba and Nyongesa, 2013; Guiry et al., 2014). In addition, Ramesh et al. (2014) addressed CA issue and profer solution for mitigation of crowd disaster using AR in a crowd scenario. In recent years, lots of considerable interest in context sensing for pervasive computing are conceived Bouguessa et al. (2015), meanwhile only few are used in disaster related problem. Likewise context acquisition, inference and modelling that affect people in the society was attempted to address crowd disaster in Ramesh et al. (2014). This is obvious because most researches in this domain often address similar issues related to users' environment, computing and communication. Irrespective of the domain either of the issues plays major role. Moreso crowd activity has direct relationship with human, lives while using mobile devices for service within the environment. It is therefore, important, for research in this domain to take care of how these issues can be incorporated to improve on the existing studies for effective AR monitoring using CA, in order to facilitate research and development for mitigation of crowd disaster.

6.1.1.4 Investigation of ARAC

- ARAC: It has been said that intelligent pervasive environments and applications will influence future and living spaces in transforming human lives and impact the society (Chen and Khalil, 2011). AR which is a classification problem is known as an increasingly important determinant to the success of CA and personalised pervasive applications (Roggen et al., 2011; Chao et al., 2010; Ramesh et al., 2014). In Figures 6(a)–6(b), some of the publications employed AR by collecting data in real time using different choice of number of participant (student) to investigate ARAC (Roggen et al., 2011; Chao et al., 2010; Ramesh et al., 2014; Meditsko et al., 2015). However, Ramesh et al. (2014) employs the AR using 20 participants to mitigate crowd disaster whenever an emergency is sensed thus predict at what time would stampede occur using CA framework with the CAC-WSN. Lots of researches in AR (Anguita et al., 2012; Kwon et al., 2014) were found in the literature but Ramesh et al. (2014) is the first to utilise the 92% of ARAC reported using DT (J48) specifically for mitigation of crowd disaster. Crowd disaster is an important global issue aimed to be address through SLR presented in this paper to X-ray the research gap and the unsolved problem in disaster management research. Crowd disaster is an issue in the previous study demands further investigation to further enhance the method of addressing the serious menace in human lives in our society. For example, online approach permits access to each data point of activity in real time with the adequate update AR data, which is capable of detecting abnormal behaviour that can lead to stampede in a crowd prior to the emergency occurrence.

6.1.1.5 Mobile phone sensing

- MPS: Previous studies in crowd monitoring relied on video camera (Krausz Bauckhage, 2012); CCTV (Davies et al., 1995) and visual tracking (Sugimura et al., 2009). More than two billion people carry mobile phones (Fehske et al., 2011) which can easily help context-awareness application accessible to everyone. The ubiquitous devices with its increasing power of capturing, classifying and transmitting image, acoustic, location and other data, interactively or autonomously using adequate architecture, could act as sensor nodes and location-aware data collection instruments (Burke et al., 2006). The state-of-the-art study already emphasised the benefit of mobile phone integration with CAC as a necessity for making the devices capable of sensing the physical world, processing the current scenario, adapting and reacting to dynamic changes in the environment. However, the need to extend their research to explore the potential of MPS with AR to monitor people in a crowd. This MPS will aids reliable and effective prediction of crowd disaster in case of emergency (Ramesh et al., 2014).

6.1.1.6 Feature extraction

Over the years, feature extraction methods known as TDFD methods have shown good result in literature. Few authors who have conducted research using TDFD methods to extract features include (Ramesh et al., 2014; Luštrek and Kaluža, 2009) where they investigated 17 features comprises of mean, STD, correlation, SMA and entropy using support vector machine (SVM) for recognition performance. The such as (mean, STD, COR, RMS and FFT_RMS) were utilised and used in Ramesh et al. (2014). It features mentioned above. It is necessary for mean to be generated first before other statistical metric features can be extracted based on computed mean (Chen and Shen, 2017). The entire TD features were later transformed to FD using the FFT method.

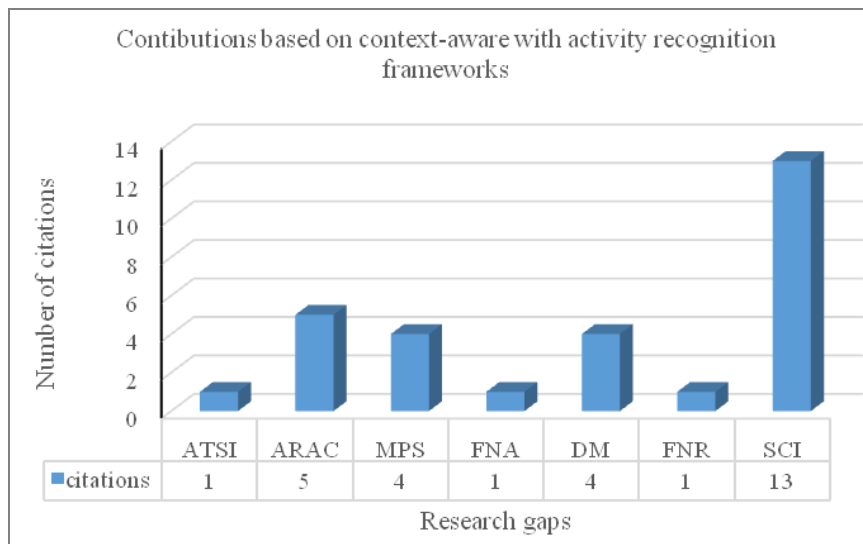
7 Context-related information not used earlier but newly suggested for DM

Generally, the statistical techniques specifically TD are light weight features. However, there is no standard of lay down criteria for features to be extracted in AR and data mining research. Therefore, it is desirable to extract as many features as possible to generalise the model under review based on the problem focus. In this study, minimum (min), maximum (max), median, variance (var) and range along x, y and z axes are suggested as reliable features has they have not been used before now in a disaster related issues especially for prediction of crowd disaster to minimise danger in an emergency. Furthermore, the newly suggested features are combined with (mean, STD, COR, RMS and FFT_RMS). As it is necessary for mean to be generated first before other statistical metric features can be extracted based on computed mean (Ramesh et al., 2014). The entire TD features were later transformed to FD using the FFT method.

7.1 Future research recommendation

Further research initiative in a crowd monitoring using AR is still possible with enhanced feature extraction and feature selection using minimum redundancy maximum relevance technique and information gain. Because, the techniques listed above can be used to identify some salient features not considered as important in the previous DM research found in literature earlier mentioned in the paper. Further research has potential to provide efficient ATSI, and reduce FNA, improve ARAC with FNR, utilisation of context interaction to employ the use of MPS for better accuracy of stampede detection probability and prediction in Figure 7. Consequently, future research has potential to investigate AR for human stampede as an online approach (real time) instead of offline batch update, to address inaccuracy in AR and time complexity issues as matter of necessity in this 21st century. Identifying TDFD features which has not been used in disaster domain but are capable of producing higher ARAC, recall, precision, f-score and FNR is desirable in this 21st century and globalisation era.

Figure 7 SLR research gap findings between years 1994 to 2017 (see online version for colours)



7.2 Feature selection

Feature selection is another relevant method which allows a dimensionality reduction of feature set numbers from the extracted feature in smartphone sensor data, used to monitor the individual activity of participant in a crowd. To avoid the high cost of computation, for the easy and fast decision process, the technique has a tendency to improve ARAC (Kandaswamy et al., 2013; Lin et al., 2015). The challenges are the problems discussed in Sections 6.1.1–6.1.1.6 which are results of SLR in Section 4 including the issue of inaccuracy in AR which has not been completely addressed. Examples of feature selection applied in AR are genetic algorithm, COR and information gain. However,

developing valid classification (or prediction) model with feature selection is usually the key to success in most data mining, AR and pattern recognition problems (Kandaswamy et al., 2013; Lin et al., 2015).

7.3 Conclusions

The study revealed a great gap in DM, which is very important worldwide for safety control measure in a small or large gathering across the country where the crowd is inevitable. Future research has the potential to investigate AR for the prediction of crowd disaster to extend the CA framework to minimise risk and loss of human lives, and to promote growth and development worldwide with improve ARAC in real time for efficient mitigation of crowd disaster with suggested solution identified as findings in the study of a SLR presented in this paper.

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Notes

- 1 C1–C75 are primary study dataset used in the SLR, details are in appendix.
- 2 C1–C75 are also used as citation in the text see appendix for detail.

Appendix

Primary studies

Details of study dataset for the SLR

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