Multi-criteria decision making for flood risk management: a survey of the current state-of-the-art

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Abstract

This paper provides a review of Multi-Criteria Decision Making (MCDM) applications to flood risk management, seeking to highlight trends and identify research gaps. Totally, 128 peer-reviewed papers published from 1995 to June 2015 were systematically analysed and classified into the following application areas: (1) ranking of alternatives for flood mitigation, (2) reservoir flood control, (3) susceptibility, (4) hazard, (5) vulnerability, (6) risk, (7) coping capacity, and (8) emergency management. Additionally, the articles were categorized based on the publication year, MCDM method, whether they were or were not carried out in a participatory process, and if uncertainty and sensitivity analysis were performed. Results showed that the number of flood MCDM publications has exponentially grown during this period, with over 82% of all papers published since 2009. The Analytical Hierarchy Process (AHP) was the most popular technique, followed by Technique for Order Preference by Similarity to an Ideal Solution (TOPSIS), and Simple Additive Weighting (SAW). Although there is greater interest on MCDM, uncertainty analysis remains an issue and is seldom applied in flood-related studies. In addition, participation of multiple stakeholders has been generally fragmented, focusing on particular stages of the decision-making process, especially on the definition of criteria weights. Based on the survey, some suggestions for further investigation are provided.

1 Introduction

Floods can be regarded as one of the most costly natural hazards both in developing and developed countries all over the world (Balica et al., 2013; Uddin et al., 2013). According to the Emergency Events Database (EM-DAT), these processes were the most frequent natural disaster between 2000 and 2014, causing at least 85,000 fatalities and affecting about 1.4 billion people. Apart from the loss of lives and physical dam-
age, floods resulted in approximately USD 400 billion in damage since 2000 (CRED and OFDA, 2015).

In order to mitigate these processes, and thereby the loss of life and damage to properties, a set of flood reduction measures need to be taken. The decision-making process related to flood risk management, especially in the prevention and emergency phases, tends to be rather complex and uncertain (Akter and Simonovic, 2005; Kenyon, 2007). Part of this complexity arises from the involvement of multiple stakeholders, each one with different views, background knowledge, interests, and frequently with competing objectives (Evers, 2008). In addition, the exact flood magnitude and damages are generally unknown and surrounded by considerable uncertainties (de Kort and Booij, 2007). As a consequence, making these decisions can rarely be solved with intuition alone. Thus, flood risk management requires the use of decision support tools, which can consider multiple stakeholders’ views, objectives, trade-offs, feasible alternatives and evaluation criteria.

Flood risk management can benefit from the use of multi-criteria decision making (MCDM). These methods provide targeted decisions, as they can handle the inherent complexity and uncertainty of such problems as well as the knowledge arising from the participation of several actors (Yan et al., 2011; Zagonari and Rossi, 2013). They can enhance the quality of decisions, by making the process more explicit, rational, and efficient, leading to justifiable and explainable choices (Mateo, 2012a). Furthermore, MCDM promotes the role of participants in the decision process, facilitates compromise and group decisions, and provides an adequate platform for stakeholders to communicate their personal preferences (Pohekar and Ramachandran, 2004). The combination of these characteristics enables the development of real participatory processes, which are crucial for the implementation of successful and long-lasting flood management programs (Affeletranger, 2001).

Therefore, MCDM provides a powerful tool for flood management and has received a great deal of attention in solving such problems, not only from researchers but also decision makers and practitioners outside the scientific community. Since the mid-90s,
MCDM approaches have been successfully applied to select the best strategies for flood risk mitigation, helping to optimize the allocation of available resources (e.g. Ghanbarpour et al., 2013; Tkach and Simonovic, 1997). In recent years, researchers have introduced MCDM to assess the flood risk and coping capacity (e.g. Guo et al., 2014; Roy and Blaschke, 2015; Yang et al., 2013).

Several authors have reviewed MCDM techniques in various fields of study previously. For example, Mendoza and Martins (2006) revised MCDM applications to forest and natural resources management. Stewart (1992) made a theoretical review by identifying potential advantages and pitfalls in the usage of various MCDM methods. Hajkowicz and Collins (2007) analysed over 134 papers in the field of water resource planning and management, focusing on problems such as water policy evaluation, strategic planning, and infrastructure selection. More recently, Estévez and Gelcich (2015) presented a concise literature survey, exploring the challenges behind participatory MCDM in marine conservation. However, despite practical experiences and methodological advances, there is no comprehensive literature review that explores the use of MCDM for flood risk management.

Hence, we believe that there is a need for a systematic survey to consolidate and synthesize recent research conducted in this area. Therefore, this paper aims to provide a state-of-the-art literature review regarding the application of MCDM as a decision support tool for flood risk management, seeking to assess emerging trends and identify issues for future investigation. In addition, it attempts to provide a better understanding of the current status of how participatory MCDM is being conducted and the way uncertainties are included in the decision-making process. With this review, we attempt to answer the following questions:

1. which flood risk management problem has further used MCDM approaches?
2. where was the research undertaken?
3. which MCDM method was most commonly applied?
4. were multiple stakeholders explicitly included in the decision-making process?
5. to which extent these studies applied uncertainty and sensitivity analysis?

We hope that this review will serve as a useful and ready source of information for scholars and practitioners working with MCDM and flood risk management.

For reader’s convenience, the remainder of this paper is structured as follows. In Sect. 2, the basic features of the most used MCDM methods are briefly described. Section 3 outlines the search strategy, and the procedure used to classify the literature. Section 4 covers the discussion of the outcomes and provides answers to the research questions. Finally, Sect. 5 presents concluding remarks, limitations and recommendations for further research.

2 Overview of multi-criteria decision making methods

MCDM is a broad term used to describe a set of methods that can be applied to support the decision-making process by taking into account multiple and often conflicting criteria through a structured framework (Cinelli et al., 2014). Since the 1960s, dozens of MCDM techniques have been developed, which can be classified in a number of ways (Mendoza and Martins, 2006). Table 1 provides an outline of the fundamental properties of the MCDM methods analysed throughout the paper. A comprehensive and detailed description of the theoretical foundations of these techniques alongside with their main strengths and weaknesses can be found in Triantaphyllou (2000) and Ishizaka and Nemery (2013).

3 Framework for systematic literature review

3.1 Search strategy

A comprehensive literature review was undertaken, aiming to identify peer-reviewed papers that apply MCDM to flood-related problems. With this scope in mind, the sys-
tematic quantitative approach outlined in Pickering and Byrne (2014) was used since this method is explicit, reproducible and has fewer biases when compared to traditional narrative reviews. To ensure that potentially relevant papers were not missed, six major library databases were systematically searched, including Scopus, ProQuest, Science Direct, SpringerLink, Emerald Insight, and Web of Science. Publications such as doctoral dissertations, book chapters, reports, and conference proceedings were not taken into account. Furthermore, only papers written in English were included. To find eligible papers in the mentioned databases, Boolean functions were applied to combine the following keywords:

Keywords ((Multi-criteria OR MCDM OR multi criteria decision making OR MCDA OR AHP OR analytic hierarchy process OR ANP OR analytic network process OR MAUT OR multi-attribute utility theory OR MAVT OR multi-attribute value theory OR ELECTRE OR TOPSIS OR MACBETH OR PROMETHEE OR NAIADE OR VIKOR OR weighted sum method OR simple additive weighting OR DSRA OR ORESTE OR DEMATEL OR goal programming) AND flood OR floods)

Distinct combinations of these terms were used, taking into consideration the syntax requirements of each search engine. For example, some databases allowed nesting the keywords, whereas others did not. Thus, different functions were applied for each database. When possible, only the abstract, title, and keywords were searched. This narrowed the search space substantially and enabled to exclude papers that mention the keywords only in the references or literature review sections.

These queries elicited over 1350 potentially eligible references published between September 1989 and June 2015. As most papers have been issued since the mid-90s, and in order to have a time-span long enough to arrive to consistent conclusions, 1995 was chosen as a starting date for this research. At first, the title, abstract and keywords were screened manually to exclude irrelevant references. After this pre-selection, the full-text of 207 selected papers was revised in detail. The paper was excluded if it was not specifically focused on flood management, or when the MCDM technique applied was not clear. Thus, 74 papers were found to be beyond the scope of this inquiry. In
addition, five papers were not made available through the library system. At the end, 128 papers met all inclusion criteria and were considered in the analysis.

The review covers articles published in 72 different journals, in several areas of knowledge, suggesting that a diversity of publishers shares an interest in flood risk management. Journals with most papers were Natural Hazards, followed by Natural Hazards and Earth System Sciences, Water Resources Management, and Stochastic Environmental Research and Risk Assessment, with 16, 11, 10 and six articles, respectively. The remaining journals account mainly for one or two published papers each.

3.2 Classification scheme

Following the selection, all included papers were classified according to some key domains: publication year; area of application; country of application; MCDM method; whether it was or was not carried out in a participatory process; participatory techniques applied; and if uncertainty and sensitivity analysis were performed. With regard to the MCDM method, only approaches that were used thrice or more have their own category, whilst the rest were grouped in “others” class. In terms of research area, the papers were classified based on the overall emphasis of the application discussed. Totally eight types of MCDM applications were identified, as follows:

1. ranking of alternatives for flood mitigation: comprises the selection of the best combination of structural and/or non-structural mitigation solutions from a set of potential alternatives to reduce flood impacts and magnitude;

2. reservoir flood control: consists in selecting operational options among a range of alternatives to ensure safe operation of reservoirs during high inflow events, aiming to reduce the intensity floods to acceptable levels;
3. susceptibility assessment: involves the rating of the terrain units according to their propensity to floods without considering its probability of occurrence or return period;

4. hazard assessment: consists of qualitative or quantitative assessments of the probability of occurrence of potentially damaging floods of a certain magnitude in a given area within a specific period of time (Dang et al., 2011);

5. coping capacity assessment: comprises the evaluation of the ability of people, organizations and systems, using available skills and resources, to face and manage adverse conditions and emergencies resulting from floods (UNISDR, 2009);

6. vulnerability assessment: refers to articles that assess the propensity of exposed elements such as human beings, their livelihoods, and assets to suffer adverse effects when impacted by floods (UNISDR, 2009);

7. risk assessment: consists in analysing potential flood hazards combined with existing conditions of vulnerability that together could potentially harm exposed people, property, services, livelihoods and the environment (UNISDR, 2009);

8. emergency management: the papers in this class are concerned with the organization and management of resources and responsibilities for addressing all aspects of emergencies, in particular, preparedness and response steps (UNISDR, 2009).

4 Results and discussion

This section presents a systematic analysis of 128 peer-reviewed papers published between 1995 and June 2015. To help readers extract quick and meaningful information, the obtained results are summarized in various charts and tables. The complete list of the papers reviewed, including their classification scheme, is provided in the Supplement.
4.1 Trends by year of publication

In an attempt to model the evolution of MCDM approaches in time, the data gathered were organized by year of publication and fitted through a polynomial regression model (Fig. 1). As can be seen from this figure, there has been a continuing growth in the number of flood MCDM studies from 1995 to June 2015. In fact, over 82% of the compiled papers have been published since 2009. Until 2004, the use of these tools was equal or less than one paper per year. Surprisingly, from 2010 to 2013, the numbers of studies have dramatically increased, from 5 to 22 papers. Accordingly, it can be estimated that in the coming years, these numbers will keep growing. This trend is pronounced, with a $R^2$ value of 94%, indicating that MCDM based approaches have a good vitality and acceptance for flood risk management.

A reason for the increasing number of publications could be a reflection of a growing awareness of natural disaster prevention and reduction policies. Secondly, the availability of easy-to-use and inexpensive MCDM software packages may also be an influencing factor. Alternatively, this increase may just match a general rise in published papers related to flood events as a whole. For example, according to the Web of Science citation reports, the number of papers with the keyword “flood risk” has increased exponentially from 1995 to 2015, with a similar trend as found out in this review.

4.2 Trends by area of application

During the last two decades, ranking alternatives for flood mitigation was the most widespread flood management topic, with more than 22% of all applications (Table 2). These studies focus mainly in selecting traditional engineering measures to reduce flood risk (e.g. Azibi and Vanderpooten, 2003; Tkach and Simonovic, 1997). Nevertheless, in recent years, they emphasize not only the so-called structural measures, which are still relevant, but also incorporate a wide range of non-structural options such as the development of evacuation plans, enforcement of building codes and insurance schemes (e.g. Zagonari and Rossi, 2013; Chitsaz and Banihabib, 2015).
The second most common theme was risk assessment (21.11%), followed by vulnerability and hazard analysis, both with 15.00% of all applications. In this regard, it should be pointed out that several papers evaluate simultaneously the vulnerability, hazard and risk (e.g. Lee and Chung, 2007; Zou et al., 2013; Wu et al., 2015). Only a reduced number of papers used MCDM as a decision support tool in reservoir flood control and emergency management problems. This is probably because managing emergencies, both in rivers and reservoirs, is a complex task, requiring effective coordination and communication among involved teams as well as reliable information regarding the current situation of emergency (Shan et al., 2012).

In order to have a complete vision of works published through the time, Fig. 2 presents the temporal breakdown of the different flood topics. As can be seen, flood risk management recently shifted its main focus from ranking alternatives for flood mitigation toward a risk-based view, which includes the assessment of risk and its components. This finding is in agreement with a worldwide trend, where disaster prevention is emphasized over assistance or relief, and evaluating the risk becomes a key element (World Bank, 2006). Another interesting result is that coping capacity studies are quite new in comparison to other topics, with the first paper published in 2009. In addition, the graph reveals that since 2010 the trend in the other flood problems remains fairly stable. This diversity of application areas shows MCDM’s flexibility to support decision making in all stages of the flood management cycle.

4.3 Trends by country of application

Totally, 38 countries on all populated continents have contributed to this survey (Table 3), showing that the spread of MCDM techniques is truly global. China accounts for 19.40% of all applications, what is not too surprising. Indeed, similar results were obtained by other MCDM review papers (e.g. Jato-Espino et al., 2014). In contrast to previous surveys (e.g. Govindan and Jepsen, 2015), Germany and South Korea were found to be prolific in the use of MCDM tools. Surprisingly, South American countries such as Brazil, Colombia, and Venezuela, which are severely affected by floods (CRED 6698...
and OFDA, 2015), were not represented in the literature. The limited use of MCDM in these countries could be explained by restrictions, such as lack of expertise, resources and technology. Half of the MCDM studies were conducted in Asia, followed by Europe (35.07 %), North America (8.21 %), Africa (3.73 %) and finally by Australia and South America, each with 1.49 % of all applications. Therefore, it is clear that when we analyse the findings of the present study, we are providing a predominantly Asiatic and European view of flood risk management.

Furthermore, only three papers report cross-country investigations (e.g. Ceccato et al., 2011; Evers et al., 2012; Almoradie et al., 2015). For example, Ceccato et al. (2011) analysed five case studies in Austria, Germany, India, Bhutan, and China. The authors found out that although the studied watersheds were characterized by distinct ecological, social and economic dimensions, the evaluation criteria selected by the stakeholders were rather similar. In this regard, multiple-case studies allow comparing findings, drawing parallels and examining differences across diverse cultural, environmental and governmental contexts.

### 4.4 Trends by MCDM method

Results showed that AHP and its family of methods have been by far the most used approaches (Table 4). One reason for this might be that its structure is straightforward, flexible and easily understandable (Cinelli et al., 2014). Thanks to these characteristics, it can be adapted to different problems without requiring previous knowledge from the analyst. Moreover, several software packages incorporate AHP (e.g. DE-CERNS, ExpertChoice, MakeltRational, Super decisions), including GIS (Geographic Information System) software’s (e.g. ArcGIS, Idrisi, ILWIS). The second most employed method was TOPSIS, closely followed by SAW. These results, with a few differences and similarities, were confirmed by other MCDM review papers such as Jato-Espinó et al. (2014) and Broekhuizen et al. (2015) that ranked AHP as the first and TOPSIS as the second method with more applications.
Note that the sum of the applications (165 items) in Table 4 does not match the number of papers (128 items) since some articles used several MCDM techniques to analyse differences in scoring and ranking. For example, Chitsaz and Banihabib (2015) compared seven MCDM tools and concluded that ELECTRE III stood superior to select flood management options. On the other hand, Chung and Lee (2009) employed five methods and found out that there is no clear methodological advantage to any of the considered techniques. Apart from comparative studies, several researchers have combined two MCDM approaches to complement each other (e.g. Margeta and Knezic, 2002; Lee and Chung, 2007; Zhou et al., 2014). For instance, Zhou et al. (2014) applied AHP to assign relative weights to each criterion and TOPSIS to rank the risk. Overall, 106 out of 128 papers (82.81 %) used one MCDM method while 12.50 % used two, 3.13 % used three and 1.56 % applied more than three.

The survey also showed that MCDM techniques are not applied only in a stand-alone mode, but are commonly extended and combined with soft computing technologies, including fuzzy set theory (e.g. Chen and Hou, 2004; Guo et al., 2014), artificial neural network (e.g. Radmehr and Araghinejad, 2014; Liu et al., 2014), and tools such as SWOT (Strengths, Weaknesses, Opportunities and Threats) analysis (e.g. Vafaei and Harati, 2010; Miyamoto et al., 2014). Moreover, there are also numerous hybrid methods, developed to address gaps in classical techniques (e.g. Yang et al., 2013; Shams et al., 2014). This suggests that MCDM is versatile, enabling researchers to combine effectively it with different tools according to the requirements of the decision to be taken.

Overall, AHP is the most prominent MCDM method in all application areas, except for reservoir flood control (Table 5). The primary reason for the popularity of AHP for mapping the risk and its components is that the implementation of this technique within the GIS environment is straightforward, enabling the users to quickly derive the weights associated with criteria map layers (Malczewski, 2007). For reservoir flood control, miscellaneous methods such as fuzzy hybrid approaches were the preferred techniques. This is probably because reservoir operations involve a large number of uncertain fac-
tors that can be properly addressed by fuzzy set theory. Additionally, TOPSIS is highly popular for ranking alternatives for flood mitigation, which emphasizes the effectiveness of this technique to deal simultaneously with conflicting objectives.

Although the most widespread MCDM methods were used at least once, no study has used DEMATEL (Decision Making Trial and Evaluation Laboratory), DRSA (Dominance-based Rough Set Approach) or ORESTE (Organization, Rangement Et Synthese De Donnes Relationnelles). It could be that these methods are relatively difficult to apply when compared to classical approaches, especially when numerous criteria are involved. However, some of them were specifically developed to address limitations of traditional techniques.

### 4.5 Trends regarding stakeholders’ involvement

Flood risk management decisions may be designed without the direct participation of multiple stakeholders. However, they cannot be implemented without them (Affeletranger, 2001). Therefore, flood management decision making should be ideally carried out in a participatory process, where the knowledge and preferences of interested actors are integrated into the process from the beginning. According to Evers et al. (2014), this creates trust among decision makers and stakeholders, which often lead to a successful implementation of the chosen measures.

The survey revealed that 65 (50.78 %) studies have explicitly acknowledged the involvement of multiple actors in the decision-making process. Policy makers and experts were the most participated stakeholders. This was expected since they are often responsible for the selection and implementation of chosen measures and have a broad knowledge of the problem of interest. Additionally, some papers mentioned the involvement of local community members (e.g. Kandilioti and Makropoulos, 2012; Sahin et al., 2013; Roy and Blaschke, 2015). According to Affeletranger (2001), the consideration of community members’ opinion may improve their resilience, as well as their response capacity when confronting natural disasters.
Nevertheless, participation was generally fragmented and restricted to consultation at specific stages, such as the selection of evaluation criteria (e.g. Haque et al., 2012) and the definition of criteria weights (e.g. Kienberger et al., 2009; Sahin et al., 2013). This segmentation may be related to methodological and time constraints since participatory decision making is time-consuming and costly, particularly when the decisions are made in a group where careful facilitation is required.

Crucial aspects of the decision-making process like the definition of objectives, identification of the alternatives, and estimation of its consequences were usually constrained to analysts and experts, which inhibit the achievement of genuine participation. Only in exceptional cases, the input from the stakeholders was a critical element in the entire process (e.g. Ceccato et al., 2011; Evers et al., 2012). For example, Ceccato et al. (2011) developed a methodological proposal aimed at strengthening the communication and collaboration within the scientific community and local actors for flood management decision making. The authors applied the NetSyMoD (Network Analysis – Creative System Modelling) framework (Giupponi et al., 2008), where the identification of relevant stakeholders, definition of the problem, establishment of objectives and criteria as well as the selection of alternatives are conducted in a participatory process.

Another interesting result is that only four studies seek to obtain consensus (e.g. Haque et al., 2012; Lee et al., 2013, 2014, 2015), in which participants take decisions by agreement rather than by majority vote or averaging approaches. Nevertheless, enhancing mutual understanding for consensus building is essential for a long-lasting and successful flood management program, especially for selecting alternatives for flood mitigation and emergency management. It allows decision makers to derive meaningful solutions that fulfil their own needs while at the same time satisfying the requirements of other actors, legitimating the participation as a learning process to solve complex problems.

Totally, 43 out of 65 studies provided unambiguous descriptions of the participatory decision-making techniques applied. Figure 3 shows that questionnaires (e.g. Giupponi et al., 2013; Taib et al., 2015) and face-to-face interviews (e.g. Deshmukh et al.,
2011; Jun et al., 2011) were the most applied tools. These methods allow for opinions to be conveyed without influence from dominant participants and are simple and fast to realize. On the other hand, the participants are not able to share and hear different perspectives through open dialogue, which is essential for achieving common agreement.

In this sense, Mendoza and Martins (2006) argue that group elicitation methods involving open discussion offer several advantages, including the consistency in the information obtained, and a better definition of the preferences. On the other hand, the results can be influenced by dominant stakeholders and noises in the responses (Hsu and Sandford, 2007). In this review, group elicitation methods such as workshops (e.g. Kenyon, 2007; Porthin et al., 2013), group meetings (e.g. Azibi and Vanderpooten, 2003; Marttunen et al., 2013) and focus group discussions (e.g. Rahman and Saha, 2007; Haque et al., 2012) were less applied.

Recently, researchers have introduced the Delphi technique to overcome shortcomings of conventional group elicitation methods regarding dominant individuals and time constraints. (e.g. Chung et al., 2014; Lee et al., 2014). This method provides anonymity to respondents, a structured feedback process, and is suited for consensus building (Hsu and Sandford, 2007). Additionally, it is advantageous when the stakeholders live some distance apart, and it is prohibitive to bring them together for a workshop or group meeting (Lee et al., 2013).

It is interesting to highlight that two studies reported the use of collaborative web-based platforms in which stakeholders are engaged in selecting and raking alternatives in an interactive way (e.g. Evers et al., 2012; Almoradie et al., 2015). These platforms have the potential to overcome hindrances in participatory MCDM such as the limitation of financial resources and stakeholders’ spatial distribution, providing full transparency of information and results. By taking this approach, the confidence in the decision-making process is increased as well as the level of acceptance of negotiated measures, which are crucial conditions for successful participatory flood risk management.
4.6 Trends regarding sensitivity and uncertainty analysis

Flood decision making is subjected to multiple sources of uncertainty, including the assessment of criteria weights, the parameters' uncertainties, and structural uncertainty (Broekhuizen et al., 2015). In addition, there are uncertainties associated with the inherent randomness of flood events (Merz et al., 2008), which, in principle, cannot be reduced. Thus, in order to improve the quality of decisions and verify the robustness of the model outputs, flood risk management should be based on a comprehensive assessment of the sensitivity combined with a thorough investigation of the uncertainties involved.

In this review, 93 (72.65 %) papers do not report any kind of sensitivity analysis, thereby ignoring the impact of changes in input weights on model results. The remaining articles (35 or 27.34 %) applied mainly one-way sensitivity analysis, where one criteria weight or performance score is modified at a time and the variation of the alternatives’ ranking is observed. If the induced variation does not change the rank order of alternatives, the decision is considered robust. This technique is intuitively appealing and requires little time, making it a practical way to assess the sensitivity. Even though this method is sufficient for most flood applications, the range over which weights are varied is normally arbitrarily defined, and the commutative impact of uncertainty is not considered. Hence, these drawbacks may lead to a biased view of the influence of uncertainty on the final decision (Broekhuizen et al., 2015).

Two papers performed Global Sensitivity Analysis (GSA) by applying the FAST (Fourier Amplitude Sensitivity Test) procedure, where two or more evaluation criteria are varied at the same time (e.g. Fernández and Lutz, 2010; Chen et al., 2015). Although GSA allows for the full uncertainty range of the criteria to be explored and analysed, it can become an extremely time-consuming task as a large number of criteria are included in the analysis. Additionally, four papers elaborated best and worst case scenarios to incorporate decision maker’s attitude to risk (e.g. Kandilioti and Makropoulos, 2012; Penning-Rowsell et al., 2013; Ghanbarpour et al., 2013; Alipour, 2015). Fi-
nally, two studies used a probabilistic approach (e.g. Yazdandoost and Bozorgy, 2008; Fernández and Lutz, 2010), which is the most rigorous form of sensitivity analysis. It requires the estimation of a maximum percentage that the actual criteria weight may differ from the estimated value.

Several authors have listed the uncertainty as a major drawback (e.g. Almoradie et al., 2015; Bana e Costa et al., 2004; Edjossan-Sossou et al., 2014; Godfrey et al., 2015). However, only eight (6.25 %) papers performed uncertainty analysis, in an attempt to describe the entire set of possible outcomes, together with their associated probabilities of occurrence. In situations where uncertainty is mainly due to randomness, the methods used were probability-based. This is the case of Qi et al. (2013) and Li (2013) that used Monte Carlo simulation to convert uncertainties in input criteria into probability distributions. Another approach applied was the Taylor’s series error propagation method (e.g. Fernández and Lutz, 2010), which analyses how the uncertainty in input data propagates through the model and affects its outputs. In addition, three papers assessed the uncertainty in a qualitative way, by describing its main sources (e.g. Cozannet et al., 2013) or by analysing the degree of confidence related to stakeholders’ opinion (e.g. Ceccato et al., 2011; Penning-Rowsell et al., 2013).

Apart from uncertainty and sensitivity analysis, fuzzy set theory is widely combined with AHP, TOPSIS and CP to handle uncertainty and incomplete information about the decision situation. For instance, Lee et al. (2013) integrated TOPSIS and fuzzy set theory to fuzzify the weighting values and all criteria maps. In the same sense, the approach proposed by Yang et al. (2012) combines AHP and triangular fuzzy number to assess the flood risk and its components. Fuzzy set theory is widespread in MCDM due to its intuitiveness and computational requirements. Nevertheless, some studies have shown that fuzzy AHP does not provide better results than regular AHP since the judgments in AHP are already fuzzy (Saaty, 2006). Therefore, the additional complexity of utilizing fuzzy numbers may be unnecessary in some cases.

Finally, it is relevant to note that some MCDM methods explicitly account for uncertain input criteria scores. For instance, ELECTRE and PROMETHEE adopt the pseudo-
A criterion model that introduces indifference and preference thresholds. Likewise, MAUT considers imprecise data input with probabilistic approaches (Cinelli et al., 2014). Also, AHP allows the generation of an inconsistency index, which can be considered as an indirect measure of the uncertainty in the criteria weighting step.

5 Conclusions

5.1 Summary

This study has presented a systematic literature review of 128 papers that apply MCDM to flood-related problems, aiming to provide an overall picture of what has motivated researchers and practitioners in 38 different countries over the past 21 years. Our findings suggest an increasing interest in flood MCDM since 2009, as compared to the previous 14 years. A wide range of application areas was identified, with most papers focusing on ranking alternatives for flood mitigation, followed by risk and vulnerability assessment. This highlights the utility of MCDM as a decision supporting tool in all stages of the flood management process. Nearly 85% of the applications were conducted in Asian and European countries, mainly in China, Germany and South Korea. Overall, AHP was the most widespread method, indicating that other methods may be overlooked. About half of the studies have acknowledged the involvement of multiple stakeholders. However, participation was fragmented and focused on particular stages of the decision-making process. Most papers rely on the use of questionnaires and interviews to capture stakeholders’ perspectives, with few applications seeking to obtain consensus. In addition, shortcomings remain in handling the uncertainty. In this sense, it is interesting to recall that only 35 and 8 papers have conducted sensitivity and uncertainty analysis, respectively.
5.2 Recommendations for future research

This review enabled us to identify gaps in the knowledge of MCDM for flood risk management regarding several aspects. First, classical MCDM approaches such as MAUT, MAVT, PROMETHEE, and DEMATEL were overlooked. Almost half of the reviewed applications used AHP to elicit criteria weights, which is a relatively easy and flexible method, requiring fewer skills than other tools. In this sense, exploring the implications of methodological differences in existing MCDM methods for flood risk management is an interesting research challenge. Similarly, future research can focus on understanding advantages and limitations of each method for handling different sources of uncertainty.

Secondly, there were surprisingly few studies that effectively considered stakeholders’ participation throughout the entire decision-making process. Therefore, greater rigour in endorsing an active participation in all stages of the decision-making process should be undertaken, in order to increase the feasibility and subsequent implementation of chosen measures. Future research could be directed towards developing web-platforms to elicit stakeholders’ preferences, aiming to reach consensus in a simpler and easily accessible way. In addition, this course of action can be combined with other participatory techniques such as cognitive mapping, Delphi technique, and voting theory. Conversely, it should be noted that intensive participation is time consuming. Thus, in real-life applications, trade-offs have to be made between the available resources and the expected outcomes of the MCDM process.

The third challenge, and perhaps the most relevant research gap, refers to fully consider the uncertainties around peoples’ judgments. Although uncertainty in MCDM is not a new problem and significant improvements have been made over the last decades, it remains a major open issue. Previous studies suggest that properly addressing the uncertainties can substantially improve MCDM applications, assisting stakeholders to make better decisions. Potential exists to apply Bayesian framework methods (e.g. Bayesian networks and Dempster–Shafers’ theory), possibility theory,
and evidence theory. Additionally, innovative approaches may be required to account for special characteristics in developing countries, where limitations in resources and technology exist. However, regardless of the uncertainty method applied, considering all sources of uncertainty in the decision-making process might not be a feasible task (Mowrer, 2000). Nevertheless, it is essential to identify many sources of uncertainty as possible, and attempt to reduce or handle them.

Lastly, a significant gain can be made if flood MCDM applications are able to consider climate and socioeconomic changes, which have potential to aggravate existing risks. This has been tackled in a recent study by Giupponi et al. (2013) that assessed the flood vulnerability within the broad context of climate change adaptation.

### 5.3 Limitations

There are multiple caveats that should be taken into consideration when interpreting the results obtained in this review. One of the main limitations is that the papers’ quality was not taken into account since they had all been published in peer-reviewed journals. Thus, some of the applications were superficial, while others were detailed, including intensive stakeholder participation, validation of results, and probabilistic-based uncertainty and sensitivity analysis. Some studies were carried out with real data, involving real decision makers and stakeholders, while others discussed hypothetical applications or were secondary studies that re-examined empirical work. A future review paper can address this limitation. In this sense, it would be interesting to apply heuristic checklists as the one proposed by Beecham et al. (2008), which can be used as a guide to assess the overall quality of a study.

In addition, defining the flood application area for each paper turned out to be an exhaustive and subjective process, especially when it came to distinguishing between susceptibility, hazard and risk assessment. There is a misunderstanding about these terms in the literature, which are used in slightly different ways by researchers with different backgrounds. Although each of these terms has its own definition within the field of physical geography, they are often used interchangeably. Thus, in some cases,
it was hard to define a clear line for when it was susceptibility, hazard or risk. Where possible, the term used by the authors was respected.

The exclusion of non-English literature can also be understood as a limitation (Behzadian et al., 2010). The results of our preliminary searches showed that several MCDM French school authors have published in French language journals. Furthermore, there are a significant number of research papers published in German, Chinese and Korean. Thus, it should emphasize that, when feasible, searches using multiple languages are advantageous.

Nevertheless, despite some weaknesses, this paper is the first to present a state-of-the-art literature review on flood MCDM. The survey has highlighted gaps, challenges and trends in this area of knowledge. Therefore, we believe this paper can provide valuable information for guiding future research and applications. It is hoped that this review will serve as a ready reference for researchers and practitioners working with flood risk management and MCDM.

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References


Evers, M.: Decision support systems for integrated river basin management: requirements for appropriate tools and structures for a comprehensive planning approach, University of Hanover, Hanover, 2008.


### Table 1. Description of different MCDM methods included in the review.

<table>
<thead>
<tr>
<th>Abbreviation</th>
<th>Method</th>
<th>Description</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>AHP</td>
<td>Analytic hierarchy process</td>
<td>Structured technique for analysing MCDM problems according to a pairwise comparison scale, where the criteria are compared to each other</td>
<td>Vaidya and Kumar (2006)</td>
</tr>
<tr>
<td>ANP</td>
<td>Analytic network process</td>
<td>Generalization of the AHP method which enables the existence of interdependencies among criteria</td>
<td>Saaty (2004)</td>
</tr>
<tr>
<td>CP</td>
<td>Compromise programming</td>
<td>Method based on the use of different distance measures to select the most suitable solution</td>
<td>Ballestero and Bernabeu (2015)</td>
</tr>
<tr>
<td>ELECTRE</td>
<td>Elimination et choix traduisant la réalité</td>
<td>Group of techniques addressed to outrank a set of alternatives by determining their concordance and discordance indexes</td>
<td>Figueira et al. (2013)</td>
</tr>
<tr>
<td>MAUT</td>
<td>Multi-attribute utility theory</td>
<td>Method where decisions are made by comparing the utility values of a series of attributes in terms of risk and uncertainty</td>
<td>Wallenius et al. (2008)</td>
</tr>
<tr>
<td>MAVT</td>
<td>Multi-attribute value theory</td>
<td>Simplification of MAUT that does not seek to model the decision maker’s attitude to risk</td>
<td>Belton (1999)</td>
</tr>
</tbody>
</table>
Table 1. Continued.

<table>
<thead>
<tr>
<th>Abbreviation</th>
<th>Method Description</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>PROMETHEE</td>
<td>Preference ranking organization method for enrichment of evaluations</td>
<td>Behzadian et al. (2010)</td>
</tr>
<tr>
<td>TOPSIS</td>
<td>Technique for order preference by similarity to an ideal solution</td>
<td>Behzadian et al. (2012)</td>
</tr>
<tr>
<td>VIKOR</td>
<td>Method that employs aggregating functions and focuses on determining compromising solutions for a prioritization problem with conflicting criteria</td>
<td>Mateo (2012b)</td>
</tr>
<tr>
<td>SAW</td>
<td>Simple Additive Weighting Tool, also known as weighted sum method, that aims to determine a weighted score for the alternatives by adding each attribute multiplied by their weights</td>
<td>Abdullah and Adawiyah (2014)</td>
</tr>
</tbody>
</table>
Table 2. Distribution of applications by flood risk management topic.

<table>
<thead>
<tr>
<th>Area of application</th>
<th>N</th>
<th>%</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ranking of alternatives for flood mitigation</td>
<td>41</td>
<td>22.78</td>
</tr>
<tr>
<td>Risk assessment</td>
<td>38</td>
<td>21.11</td>
</tr>
<tr>
<td>Vulnerability assessment</td>
<td>27</td>
<td>15.00</td>
</tr>
<tr>
<td>Hazard assessment</td>
<td>27</td>
<td>15.00</td>
</tr>
<tr>
<td>Susceptibility assessment</td>
<td>21</td>
<td>11.67</td>
</tr>
<tr>
<td>Coping capacity</td>
<td>11</td>
<td>6.11</td>
</tr>
<tr>
<td>Reservoir flood control</td>
<td>8</td>
<td>4.44</td>
</tr>
<tr>
<td>Emergency management</td>
<td>7</td>
<td>3.89</td>
</tr>
<tr>
<td>Total</td>
<td>180</td>
<td>100</td>
</tr>
</tbody>
</table>
Table 3. Distribution of applications by country of application.

<table>
<thead>
<tr>
<th>Country</th>
<th>N</th>
<th>%</th>
<th>Country</th>
<th>N</th>
<th>%</th>
</tr>
</thead>
<tbody>
<tr>
<td>China</td>
<td>26</td>
<td>19.40</td>
<td>Finland</td>
<td>2</td>
<td>1.49</td>
</tr>
<tr>
<td>Germany</td>
<td>13</td>
<td>9.70</td>
<td>Italy</td>
<td>2</td>
<td>1.49</td>
</tr>
<tr>
<td>South Korea</td>
<td>10</td>
<td>7.46</td>
<td>Kenya</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Iran</td>
<td>7</td>
<td>5.22</td>
<td>Kuwait</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Greece</td>
<td>6</td>
<td>4.48</td>
<td>Vietnam</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>India</td>
<td>6</td>
<td>4.48</td>
<td>Taiwan</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Canada</td>
<td>6</td>
<td>4.48</td>
<td>Bhutan</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Malaysia</td>
<td>5</td>
<td>3.73</td>
<td>Switzerland</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Bangladesh</td>
<td>5</td>
<td>3.73</td>
<td>South Africa</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>USA</td>
<td>5</td>
<td>3.73</td>
<td>Poland</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>UK</td>
<td>4</td>
<td>2.99</td>
<td>Spain</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>France</td>
<td>4</td>
<td>2.99</td>
<td>Portugal</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Slovakia</td>
<td>3</td>
<td>2.24</td>
<td>Scotland</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Egypt</td>
<td>2</td>
<td>1.49</td>
<td>Serbia</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Turkey</td>
<td>2</td>
<td>1.49</td>
<td>Nigeria</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Japan</td>
<td>2</td>
<td>1.49</td>
<td>Chile</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Australia</td>
<td>2</td>
<td>1.49</td>
<td>Argentina</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Croatia</td>
<td>2</td>
<td>1.49</td>
<td>Romania</td>
<td>1</td>
<td>0.75</td>
</tr>
<tr>
<td>Austria</td>
<td>2</td>
<td>1.49</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>the Netherlands</td>
<td>2</td>
<td>1.49</td>
<td>Total</td>
<td>134</td>
<td>100.00</td>
</tr>
</tbody>
</table>
### Table 4. Distribution of applications by MCDM method.

<table>
<thead>
<tr>
<th>MCDM method</th>
<th>N</th>
<th>%</th>
</tr>
</thead>
<tbody>
<tr>
<td>AHP, fuzzy AHP, trapezoidal fuzzy AHP and ANP</td>
<td>70</td>
<td>42.42</td>
</tr>
<tr>
<td>TOPSIS, fuzzy TOPSIS and modified TOPSIS</td>
<td>22</td>
<td>13.33</td>
</tr>
<tr>
<td>SAW</td>
<td>21</td>
<td>12.73</td>
</tr>
<tr>
<td>Others (MACBETH, NAIADE, goal programming, etc.)</td>
<td>20</td>
<td>12.12</td>
</tr>
<tr>
<td>CP, spatial CP and fuzzy CP</td>
<td>10</td>
<td>6.06</td>
</tr>
<tr>
<td>ELECTRE I, II, III and TRI</td>
<td>7</td>
<td>4.24</td>
</tr>
<tr>
<td>MAUT and MAVT</td>
<td>7</td>
<td>4.24</td>
</tr>
<tr>
<td>PROMETHEE I and II</td>
<td>5</td>
<td>3.03</td>
</tr>
<tr>
<td>VIKOR and fuzzy VIKOR</td>
<td>3</td>
<td>1.82</td>
</tr>
<tr>
<td><strong>Total</strong></td>
<td>165</td>
<td>100</td>
</tr>
</tbody>
</table>
Table 5. Distribution of applications by MCDM method and area of application.

<table>
<thead>
<tr>
<th>Area of application/Number of applications</th>
<th>AHP</th>
<th>TOPSIS</th>
<th>SAW</th>
<th>Others</th>
<th>CP</th>
<th>ELECTRE</th>
<th>MAUT</th>
<th>PROMETHEE</th>
<th>VIKOR</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ranking of alternatives for flood mitigation</td>
<td>14</td>
<td>10</td>
<td>9</td>
<td>8</td>
<td>9</td>
<td>5</td>
<td>2</td>
<td>3</td>
<td>1</td>
</tr>
<tr>
<td>Risk assessment</td>
<td>27</td>
<td>10</td>
<td>5</td>
<td>6</td>
<td>1</td>
<td>1</td>
<td>3</td>
<td>1</td>
<td>2</td>
</tr>
<tr>
<td>Vulnerability assessment</td>
<td>21</td>
<td>3</td>
<td>5</td>
<td>4</td>
<td>1</td>
<td>1</td>
<td>2</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Hazard assessment</td>
<td>25</td>
<td>3</td>
<td>2</td>
<td>5</td>
<td>1</td>
<td>1</td>
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<td>0</td>
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<tr>
<td>Susceptibility assessment</td>
<td>18</td>
<td>0</td>
<td>4</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Coping capacity</td>
<td>8</td>
<td>4</td>
<td>2</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Emergency management</td>
<td>5</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Reservoir flood control</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>5</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>Total*</td>
<td>119</td>
<td>31</td>
<td>28</td>
<td>28</td>
<td>12</td>
<td>9</td>
<td>8</td>
<td>5</td>
<td>3</td>
</tr>
</tbody>
</table>

*It is important to highlight that some papers analysed two or more flood problems simultaneously by using the same MCDM method. Thus, the number of applications in Table 5 is higher than in Table 4.
Figure 1. Number of MCDM flood papers published over the period 1995–June 2015 (papers published in late 2015 are not included in the present review due to the limitation of reporting time).
Figure 2. Distribution of MCDM papers by application area between 1995 and June 2015.
Figure 3. Methods used to incorporate multiple stakeholders’ views in the decision-making process.

- questionnaires
- interviews
- workshops
- delphi technique
- group meetings
- web-based platform
- focus group discussions
- narrative analysis
- stakeholder analysis