



1 **A spatial decision support system for enhancing resilience to floods.**
2 **Bridging resilience modeling and geovisualization techniques**

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12
13 **Abstract**

14
15 In the context of climate change and increasing urbanization, floods are considerably affecting urban areas. The
16 concept of urban resilience may be an interesting means of responding to urban flood issues. The objective of this
17 research is to propose a spatial decision support tool based on geovisualization techniques and a resilience
18 assessment method. The goal is to localize the level of resilience modeled in different territories. The methodology
19 proposed consists in integrating three resilience indicators applied to a case study in Avignon (Provence Alpes
20 Côte d'Azur Region, France) and the use of geovisualization techniques: using GIS for data processing and
21 analysis, visualization, mapping and model processing. The methodology integrates decision-making by
22 identifying characteristics capable of improving urban resilience and facilitating its understanding using a visual
23 tool. The results demonstrate the usefulness of modeling resilience using geovisualization techniques to identify
24 the potential for local resilience, integrate local stakeholders into a process of clarifying the concept through the
25 contribution of visualization, and consider easier access to this concept based on data analysis, processing and
26 visualization through the design of maps.

27
28 **Key words:** Geovisualization, Urban resilience modeling and mapping, flood risks

29
30 1- Introduction

31 1.1 *Issues and background*

32
33 The context of climate change has led to an increase in disasters, among which urban floods are
34 considered the most damaging, accounting for 43.4% of climate-related disasters over the period 1998-2017
35 (Wallemacq and House, 2018). At present, the European Environment Agency ranks France third among European
36 countries affected by natural hazards over the period 1980-2017 (European Environment Agency, 2019), as 33%
37 of its municipalities were affected with "*an estimated annual cost of around 250 million euros*" (Lhomme, 2012).
38 The Mediterranean region is among the most vulnerable in France, with an average of 10 deaths per year caused
39 by floods. 42% of the population of the Vaucluse Department live in areas at risk from floods and it ranks first
40 among departments exposed to flood risk, in comparison to the national average of 11% of the population living
41 in flood risk areas in 2009. With 147/151 municipalities in the department affected by floods, Vaucluse is
42 extremely vulnerable to this growing risk.
43



44 To address this growing risk, the concept of resilience has been included step by step into risk
45 management strategies, worldwide, as it offers a systemic approach to and analysis of risks, their issues, territories,
46 populations and management services (Bakkensen et al., 20). The concept of resilience can be defined as “*the*
47 *ability of a system, community or society exposed to hazards to resist, absorb, accommodate to and recover from*
48 *the effects of a hazard*” (UNISDR-United Nations International Strategy for Disaster Reduction, 2009). Although
49 and despite a significant increase in the use of the concept and its positive opportunities for risk strategies, concrete
50 progress towards operationalization is still needed (Klein et al., 2003). The objective of this research is therefore
51 to propose an approach to address this lack of operability. While some studies have been carried out in Europe
52 to operationalize the concept of vulnerability through indicators (Opach and Rød, 2013), few of them mention
53 resilience. When such is the case (Lhomme et al., 2013; Suárez et al., 2016), it is essentially from a technical and
54 organizational angle, but without considering the social and therefore systemic dimensions of the territory
55 concerned.

56 This research therefore aims at using the concept of resilience in a practical and understandable manner
57 at the city level, with the design of a spatial decision support system. The originality of the methodology is justified
58 by the collaborative approach taken, characterized by a socio-economic partnership with the City of Avignon and
59 its urban services. By combining the experiences of managers and politicians with scientific advances, the
60 approach aims at addressing the challenges and limitations of the concept of urban resilience in the face of flood
61 risk. The result of joint design, the spatial decision support system is being tested in the Avignon area in response
62 to more risky situations. Spatial decision makes it possible to establish a link between scientific advances and local
63 knowledge and practices. This spatial decision support system involves redefining the criteria for resilience and
64 measuring the potential for resilience (Frazier et al., 2013). It aims at overcoming:

- 65 - theoretical obstacles, by designing indicators to assess resilience;
- 66 - methodological issues by representing the potential for resilience through mapping tools used to
67 provide stakeholders with a medium capable of making them aware of the concept, integrate it into
68 their risk management strategies and transform it into concrete and applicable actions.

69 Meeting the challenges of operationalizing resilience therefore involves rethinking modeling and mapping
70 practices as well as focusing on understanding the concept, adopting it and integrating stakeholders into the
71 resilience process.

72

73 1.2 Research Focus

74

75 We adopt the viewpoint that promoting techniques to make resilience operational can be achieved by
76 collaboration and visualization methods. Getting people from different backgrounds to interact (Callon et al.,
77 2001), enriches discussions, encourages the expression of opposing viewpoints on the same subject, and makes it
78 possible to be both more measured and more incisive in a specific field. Resilience is therefore a subject that
79 requires the confrontation of views, and scientific and local knowledge (Radhakrishnan et al., 2017). We therefore
80 propose to develop strategies to operationalize resilience so that they are constructed jointly with the city's actors,
81 allowing their direct investment. Rather than taking a top-down approach, our goal is to create a common
82 discussion around resilience issues to initiate constructive dialogues to overcome the biases of each group of
83 stakeholders (Jacobs et al., 2005; Moser, 2005; Næss et al., 2006; Patt and Dessai, 2005). In addition, we consider



84 that techniques translating a fuzzy concept into a practical spatial decision support system - such as
85 geovisualization and modelling - would promote stakeholder involvement and understanding of the related issues
86 and thus lead to adapted decision-making. The motivation of the article is to demonstrate that combining certain
87 geovisualization techniques with resilience modeling will contribute to better understanding of the concept, and
88 lead to its operationalization and translation into tangible strategies at the local level.

89 We defend the hypothesis that defining resilience criteria and translating them visually for implementation in
90 an easy-to-use tool will promote and better integrate resilience techniques in view to managing urban floods. By
91 carrying out a municipality scale study and combining a collaborative methodology and GIS resilience modeling
92 to develop a geovisualization tool, we hope to clarify the concept, and ensure its understanding and adoption by
93 urban planners in their approaches to urban dynamics. In the first section we present a state of the art of resilience
94 modeling and geovisualization techniques in the field of climate risk management, and then the methodologies
95 chosen for this research. Finally, we present the first application of this research and its results in Avignon (France).
96 Finally, we discuss these initial results.

97

98 2. Resilience modelling and geovisualization techniques for risk management: a state of the art

99 2.1 *The resilience concept and modeling approaches*

100

101 As the concept of resilience is multidisciplinary, its definition and application as a risk management
102 strategy is extremely complex. In order to move towards its operationalization, it is necessary to build an analysis
103 model to address the concept. Several studies have attempted to build analysis models to define indicators or a
104 specific baseline (Bakkensen et al., 2017; Fox-Lent et al., 2015).

105 The 100 Resilient Cities (100RC) consortium was launched by the Rockefeller Foundation in 2013. The
106 purpose of the 100 Resilient Cities consortium is to help cities around the world become more resilient to the
107 physical, social and economic challenges of the 21st century. 100RC supports the adoption and integration of a
108 vision of resilience that includes not only disasters - earthquakes, fires, floods, etc. - but also the tensions that
109 weaken the urban area on a daily or cyclical basis. Resilience is defined as the ability of individuals, communities,
110 institutions, businesses and urban systems to survive, adapt and evolve, regardless of the types of chronic stresses
111 or shocks they may encounter. A holistic approach is advocated. 100RC has built a framework defining the
112 characteristics of urban resilience (Fig.1).

113



114

115

Figure 1: 100 Resilient Cities Framework, (100 Resilient Cities, s. d.)

116

117 The definition of resilience via these indicators allows identifying criteria for resilience in a territory or within a
118 population. It allows launching discussion around an initially fuzzy concept. However, it does not allow visualizing
119 criteria or resilience potentials at the local level (100 Resilient Cities, n.d.). Mapping is non-existent and the
120 absence of tangible data makes it difficult for local populations and actors to appropriate the concept, understand
121 it, and reproduce it.

122

123

124

Another study focused on identifying resilience capacities applied to urban networks. It led to the creation
of the DS3 (Spatial Decision Support System) model (Serre, 2018). Three resilience capacities were defined to
study resilience (Serre, 2018)), namely resistance, absorption, and recovery (Fig.2).

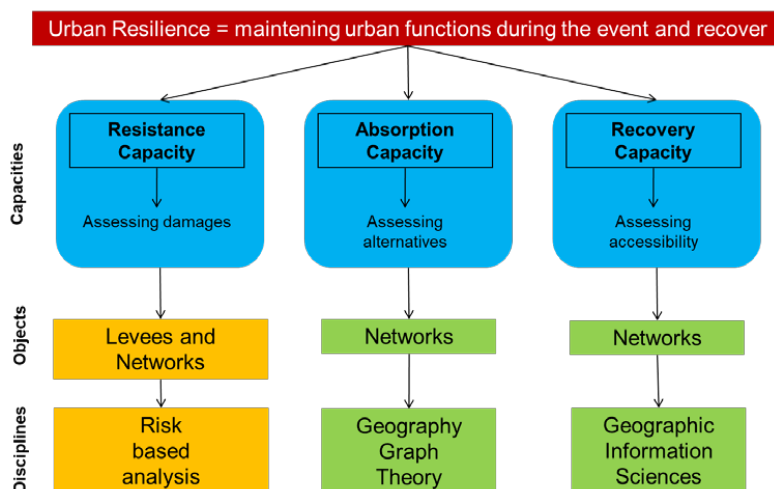


Figure 2: DS3 Model (Serre, 2016)

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126

127 The resistance capacity is necessary to determine the material damage of the networks. It is a given that
128 the more damaged a network is, the slower and more difficult it will be to return it to effective service. The results
129 of the damage analysis make it possible to measure this damage, and determine the interdependencies between the
130 various components of the networks.

131 Absorption capacity represents the alternatives available to the network following a failure. The idea is
132 to highlight solutions to maintain service continuity despite floods, operating in degraded mode.

133 Finally, the recovery capacity represents the time required to retrofit the networks until reaching a full
134 level of service.

135

136 The DS3 model can be used to identify factors that would lead to increased urban resilience, highlighting
137 the importance of urban networks and critical infrastructures. This technical approach focuses mainly on urban
138 networks. However, cities comprise many factors, such as social dynamics, urban interactions and technical
139 components, leading to additional indicators that must be monitored (Serre and Heinzlief, 2018).

140

141 These two approaches attempt to address the biases of conceptualization and modeling resilience. But, in
142 the first approach of the concept and data visualization, there is nothing evident about how the results should be
143 processed and explored. In the second approach, the exploration of the results is visible through the application of
144 the methodology, notably in a case study on Hamburg (Serre et al., 2016). Nonetheless, the exploration and
145 analysis of the data is not accessible to the public concerned, limiting their understanding and appropriation of the
146 method. Moreover, this approach analyzes the territory only through urban networks and not with the other
147 components that shape it.

148 The objective of this work is therefore to model and operationalize resilience as comprehensively and
149 exhaustively as possible. The aim is to analyze it at the local level in order to advise stakeholders and lead to
150 decision-making that integrates resilience in risk management strategies

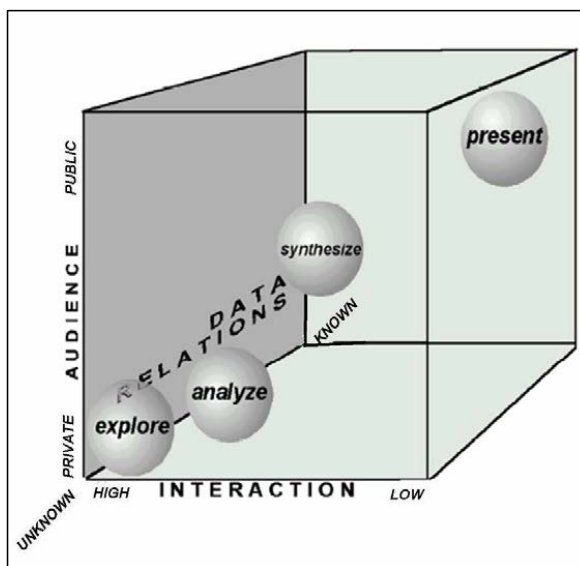
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152 2.2 Geovisualization techniques: added values in risk management processes

153 For many years, risk mapping was one of the main methods used to analyze, represent, and examine the
154 multiple characteristics of risks and risk management strategies (Barroca and Serre, 2018). However, new methods
155 have been introduced such as Geographic Information Systems (GIS), and scientific information visualization
156 (Kraak, 2003). GIS gives access to voluminous and heterogeneous tools like databases and graphic applications to
157 establish interactions between data and maps. These interactions can be visualized through an interface used to
158 explore the characteristics of data. The adaptation of scientific visualization to mapping was initially called
159 "geographic visualization" and then "geovisualization" (Maceachren and Kraak, 1997). Geovisualization is
160 defined as "the set of visualization tools that allow interactive exploration of geolocated data in order to build
161 knowledge without assumptions a priori" (Maceachren and Kraak, 1997). Geovisualization includes fields such as
162 scientific visualization, mapping, image processing, knowledge extraction, and GIS.

163 Therefore, geovisualization is a synthesis approach applied to GIS techniques that integrates practices
164 such as mapping, visualization, data and image analysis, by analyzing geospatial data (MacEachren and Kraak,
165 2001). This methodology offers the possibility of representing multidimensional, voluminous and heterogeneous
166 data. More specifically, geovisualization is mainly adapted to the representation and analysis of georeferenced
167 data. The mapping exercise is divided into several objectives: explore, analyze, synthesize and present. Geovisual
168 tools must be adapted to these different uses. The different tools currently available can be differentiated by three
169 criteria. The first is the audience, which can range from the "general public" with little knowledge of
170 geovisualization issues to experts with good knowledge of the subject. The second is the degree of interactivity
171 offered by the geovisualization tool. The last criterion is knowledge of the data, which varies from the domain of
172 the known to the domain of the unknown (MacEachren and Kraak, 1997). The 4 uses of geovisualization can
173 therefore be placed inside a cube (Fig.3). Each axis of this cube (x,y,z) represents one of the 3 criteria previously
174 mentioned (audience, interactivity, data relations).



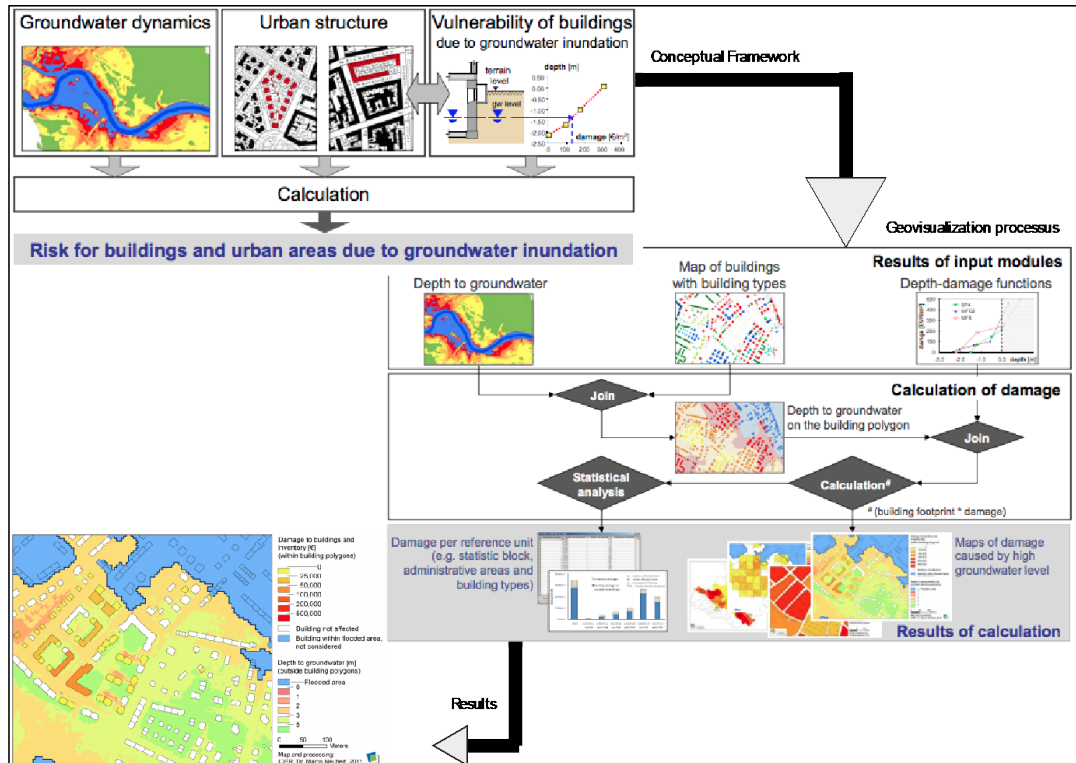
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176

Figure 3: The Map-Use-Cube (Maceachren and Kraak, 1997)



177 This data representation (Donolo, 2014) – also called “virtual science” - allows constructing,
178 reconstructing, representing and interpreting scientific issues (Yasobant et al., 2015). The fact of representing
179 spatiotemporal data in different forms provides better understanding of the different phenomena involved,
180 resulting in either better dissemination of the information or better decision-making. This methodology allows
181 exploring hypotheses, sharing arguments, developing solutions and, most importantly, building common
182 knowledge around the same issue (MacEachren, 1997).

183 Consequently, these characteristics and advantages make geovisualization an interesting methodology for
184 studying risk management. Crisis management is, indeed, a concrete example where it is useful to use visual, map-
185 based tools to integrate, assess and apply multisource geospatial information and data (MacEachren et al., 2004).
186 Indeed, in a context of climate change and related uncertainties, modeling or simulating disasters such as floods is
187 becoming increasingly complex. Current techniques are limited in the face of the complexity of floods, particularly
188 because of the multiple reasons, sources and causes of disasters (Leskens et al., 2014; Löwe et al., 2018), as they
189 are essentially used to model urban planning projects or response strategies to cope with the increase in the
190 occurrence of such events. Many studies have used geovisualization to analyze the complexity of flood risks,
191 whether to analyze flooding from the perspective of risk, for instance expected damage (Meyer et al., 2009; Ward
192 et al., 2011), hazard, such as duration, velocity, water depth, etc. (Schumann et al., 2009), management strategy
193 (de Moel et al., 2015), at the national (Burby, 2001), regional (Elmer et al., 2012; Gaslikova et al., 2011;
194 Vorogushyn et al., 2012), and local (Aerts et al., 2013; Apel et al., 2009; Gerl et al., 2014) levels, and even on the
195 built scale (Fig.4), with, for example, FReT (Flood Resilience Technologies) (Schinke et al., 2016; Golz et al.,
196 2015).
197



198
 199 *Figure 4: Links between resilience modeling and geovisualization techniques – at the buildingscale (Schinke et al., 2012)*

200

201 To summarize, geovisualization helps to explore data using visual geospatial representations to imagine
 202 hypotheses, solve problems and co-construct scientific knowledge (Kraak, 2003). Therefore, geovisualization
 203 methodology improves territorial knowledge and leads to tools such as decision support systems, by making
 204 possible dialogues between users and stakeholders and promoting collaborative approaches. In the field of risk
 205 management, it is essential to defuse subjects of tension, in order to present a risky situation objectively.

206

207 *2.3 Making urban resilience operational through geovisualization techniques*

208

209 Although several methods can be used to model risk characteristics, such as hydraulic modeling (Ernst et
 210 al., 2010) and geomorphological parameters (Bathrellos et al., 2012), it is quite difficult to model such fuzzy
 211 concepts like resilience and vulnerability, despite the common use of the latter in risk management. While the
 212 implementation of resilience policies and the design of resilient cities is desirable, assessing resilience and
 213 implementing it is complex. Several researchers have examined the difficulty of defining, implementing and
 214 evaluating urban resilience, usually through a geovisualization approach.

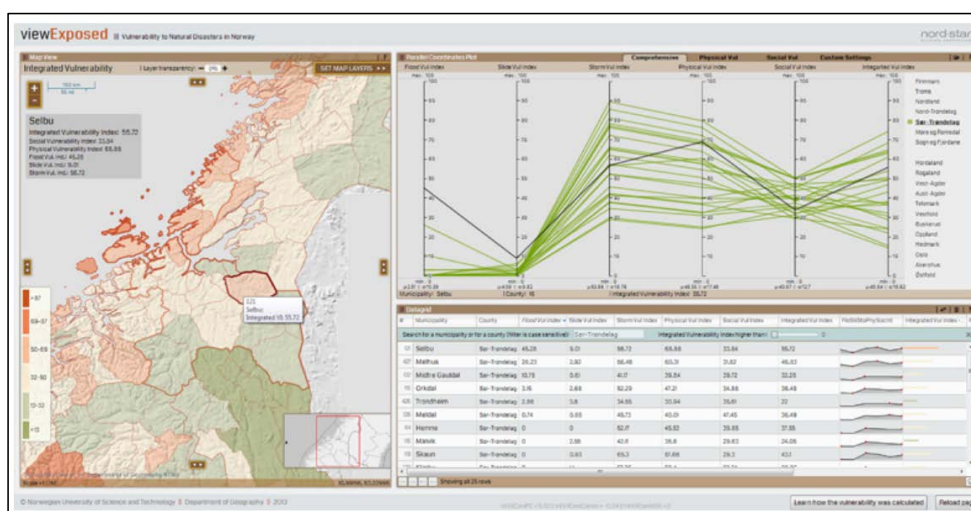
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216 Cutter and Finch (2008) presented SoVI, a tool providing a county-level (USA scale) comparative metric
 217 of social vulnerability to natural hazards, based on socioeconomic and demographic profiles. The aim of SoVI is
 218 to illustrate the geographic patterns of the USA, by defining social vulnerability as the sensitivity of a population
 to natural hazards and its ability to respond to and recover from them. Using several maps and in view to improving



219 emergency management, SoVI identifies which areas of the American territory are more or less vulnerable and
220 why. Geovisualization techniques improve understanding of the concept of vulnerability and help urban managers
221 to localize vulnerable areas and variables.

222 Based on SoVI, a Norwegian study examined the vulnerability of territories to climate change
223 (Opach and Rød, 2013). To avoid an increase in local and national vulnerability, the researchers built a
224 ViewExposed tool (Fig.5) whose objective was to inform local authorities about the most vulnerable areas of
225 Norwegian territory and the causes of this vulnerability.



226
227 *Figure 5: The ViewExposed Interface (Opach and Rød, 2013)*

228 The ViewExposed tool focuses on Norwegian municipalities' exposure to natural hazards and
229 the capacity of local populations to resist them. This interface tool was designed for professionals, local elected
230 officials and local residents. It is the result of collaboration between scientists and local experts via workshops.
231 Although focused on the concept of vulnerability, this tool also integrates the response of local managers and
232 actors to natural disasters, corresponding to the first step in resilience integration.

233 Another team of researchers and insurers developed a tool to help people in the Nordic countries to protect
234 themselves and prepare for climate risks. The main target users are private landowners, but this tool can also be
235 used by land-use planners and property managers. The tool, VisAdapt (Johansson et al., 2016) is intended as a
236 guide on how to prepare for climate events liable to affect individual homes. It is very simple to use, so every
237 citizen can employ it. The obvious interest of this tool is that it allows addressing local inhabitants directly by
238 proposing solutions to adapt to natural risks linked to climate change.

239
240 These tools have the merit of proposing operational instruments to obtain a clearer idea of the vulnerability
241 concept involved. The main scale is above all the national scale which, despite major advances in the visualization
242 and knowledge of vulnerable zones, does not always lead to decision-making by local actors and managers, since
243 the scale is sometimes too broad for actions. Beyond the spatial scale, the choice of data and tools for processing,
244 analysis and visualization have not been designed for non-expert audiences. Data is not always freely accessible,



245 nor are the processing and representation tools. Some tools are intended only for professionals while others point
246 to the need to open the results to a wider audience. In addition, the data are not accessible and downloadable,
247 which makes the methodology difficult to adopt and reapply in other territories outside the scope of expertise of
248 the research team. The limits are therefore divided between the choice of spatial scale, the free and accessible
249 nature of data and tools, and the non-integration of local actors, and thus the assurance of their understanding of
250 the tools and concepts used. Here, we intend to overcome these limitations by proposing the approach we have
251 developed.

252

253 3. Methods: linking resilience modeling and geovisualization techniques

254

255 The objective of this research consists in:

- 256 - making the concept of resilience more understandable through the construction of 3 indicators to
257 define and measure resilience;
- 258 - producing mapping results to quantify and visualize the results obtained;
- 259 - designing a comprehensive method including choice of data, processing and analyses for local actors,
260 by mobilizing geovisualization techniques;
- 261 - mapping the results to support decisions in favor of resilience to floods.

262

263 *3.1 A framework for defining resilience data?*

264

265 To analyze urban territories including their complexity through the prism of resilience, it was necessary
266 to define their issues and challenges, their dynamics, material and immaterial interactions, and their structures that
267 impact on the functioning of urban space. It is essential to understand the city as a system (Gardner, 2016). These
268 urban systems, like any living organism, are complex and hierarchical. Some studies have explored the impact of
269 rapid urbanization, leading to complex territorial responses and the lack of suitable reactions. In parallel,
270 challenges can increase when the country's gross domestic product (GDP) decreases. Nevertheless, in some case
271 studies, urbanization has been shown to have other results and is one of main elements to be taken into account
272 when building response capacity to risks (Garschagen and Romero-Lankao, 2015). This response capacity can be
273 determined by flood preparedness (Chinh et al., 2016), government implication and risk governance (Garschagen,
274 2015). Studying the city in the face of risks and its resilience capacity requires considering different spatial scales
275 of interactions and challenges. Therefore, several questions must be asked: Who is vulnerable/resilient? What?
276 When? What elements could limit the impacts of a crisis like a flood event? Are they efficient before, during and
277 after a flood?

278 These questions allowed us to establish three resilience indicators to study technical, urban and social
279 resilience (Heinzlef et al., 2019). The methodological choice of using indicators was based on several arguments.
280 The first one is that by defining and characterizing an abstract concept, indicators allow sensitizing both the
281 scientific community and the public to complex subjects (Prior and Hagmann, 2014). In addition, resilience
282 indicators can make an important contribution to assess a community's needs and goals while helping it to develop
283 resilience strategies (Cutter, 2016). These indicators, useful when creating a strategy, are also important for
284 monitoring the decision-making process. Finally, an essential benefit of using such indicators is that they can act



285 as driving factors for risk management, by including the concept of resilience clearly and more holistically (Linkov
 286 et al., 2014).The objective is to analyze the different social, urban and technical components (Serre and Heinzlief,
 287 2018) of the area concerned (Tab.1). The indicators were designed after adapting the Baseline Resilience Indicators
 288 for Communities (BRIC) methodology (Cutter et al., 2008; Patil et al., 2008; Singh-Peterson et al., 2014).

<i>Resilience indicators</i>	<i>Variables</i>	<i>Sources</i>	<i>Impact on resilience</i>	<i>References</i>
Social resilience indicator	<i>Population Structure</i>			
	00-02 years old	INSEE	Negative	(Morrow, 2008); (Cutter, 2010); (Opach et Rod, 2016)
	25-39 years old	INSEE	Positive	(Morrow, 2008); (Cutter, 2010); (Opach et Rod, 2016)
	more than 80 years old	INSEE	Negative	(Morrow, 2008); (Cutter, 2010); (Opach et Rod, 2016)
	<i>Professional situation</i>			
	Active 15-64 years old	INSEE	Positive	(Tierney et al., 2001)
	Unemployed 15-64 years old	INSEE	Negative	(Tierney et al., 2001; Tierney, 2014)
	<i>Habits</i>			
	Active people 15 years or older not using transport	INSEE	Positive	
	Active people 15 years or older, using public transport	INSEE	Positive	
	<i>Insurances</i>			
	Health insurance beneficiaries	INSEE	Positive	(Heinz Center 2002)
	Beneficiaries of CAF allocations	INSEE	Positive	(Heinz Center 2002)
	<i>Education</i>			
Exit before the 3 rd grade	INSEE	Negative	(Norris et al. 2008), (Morrow 2008)	
Bac +2 and better	INSEE	Positive	(Norris et al. 2008), (Morrow 2008)	
Urban resilience indicator	<i>Buildings</i>			
	Number of main residences built before 1919	INSEE	Positive	(Mileti, 1999); (Cutter, 2010), (Opach et Rod, 2016)
	Number of main residences built from 1919 to 1945	INSEE	Negative	(Mileti, 1999); (Cutter, 2010), (Opach et Rod, 2016)
	Number of main residences built from 1946 to 1970	INSEE	Negative	(Mileti, 1999); (Cutter, 2010), (Opach et Rod, 2016)
	Number of main residences built from 1971 to 1990	INSEE	Negative	(Mileti, 1999); (Cutter, 2010), (Opach et Rod, 2016)
	Number of main residences built from 1991 to 2005	INSEE	Negative	(Mileti, 1999); (Cutter, 2010), (Opach et Rod, 2016)
	Number of main residences built from 2006 to 2010	INSEE	Positive	(Mileti, 1999); (Cutter, 2010), (Opach et Rod, 2016)
	<i>Critical Infrastructures</i>			
	Defense	SIRENE	Positive	(Sylves, 2007); (Cutter, 2010)
	Fire and rescue services	SIRENE	Positive	(Sylves, 2007); (Cutter, 2010)
	Hospital activities	SIRENE	Positive	(Opach et Rod, 2016)
<i>Economic dynamics</i>				



	Tourist and other short-term accommodation	SIRENE	Positive	(Tierney, 2009)
	Creation of new companies	SIRENE	Positive	
	Removal of companies	SIRENE	Negative	
Technical resilience indicator	<i>Diversity of networks</i>	Municipality data	Positive	(Bambara, 2014); (Balsells et al, 2015)
	<i>Network accessibility</i>			
	Accessibility of networks by public road within a 100m radius	Municipality data	Positive	(Cutter, 2010); (Opach et Rod, 2016) ; (Lhomme et al., 2013)

Table 1: Example of data selection, sources and references

289

290

291 3.2 Resilience processing

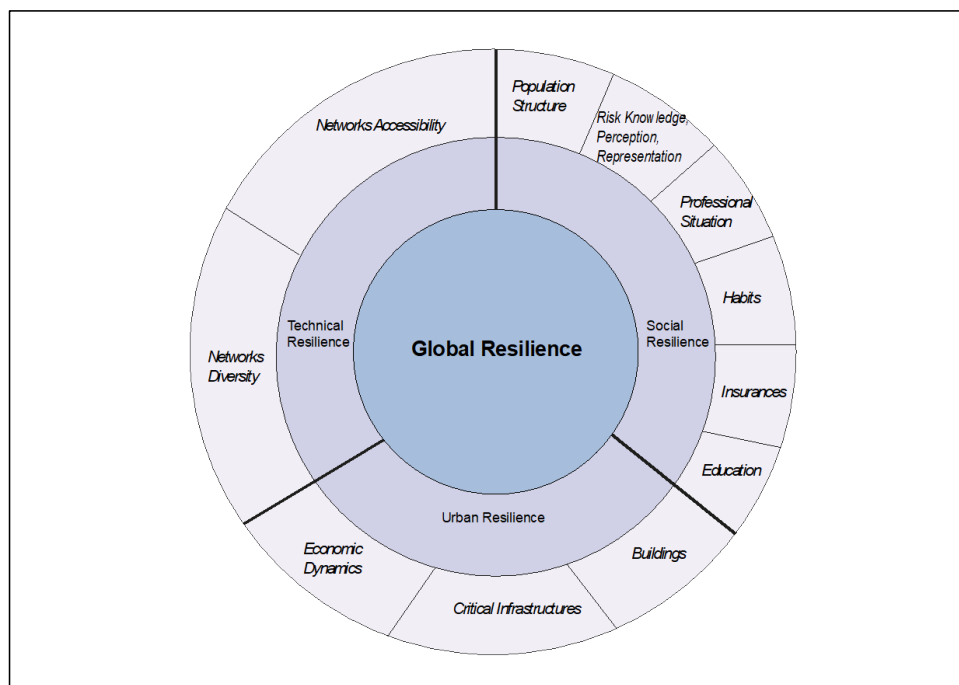
292 The advantages of using geovisualization techniques to remove barriers to resilience are:

293 - analyzing heterogeneous and geolocated data;

294 - supplying a visualization based on the most recent scientific advances;

295 - extracting, producing and sharing data with innovative layouts.

296 To address these three pillars, we propose to clarify some of the resilience criteria defined above (Fig.6).



297

Figure 6: Resilience characteristics (Serre & Heinzl, 2018)



298 We argue that analysis of resilience at the local level is facilitated by using open access data. On the other
299 hand, data processing and analysis become more understandable for local actors when tools are chosen that
300 highlight the visualization.

301 3.2.1 Data used for resilience assessment

302 We chose to use mainly open data which we acquired via the INSEE service of the French Ministry of
303 the Economy and Finance (National Institute of Statistics and Economic Studies), whose function is to collect,
304 analyze and disseminate data. Since we wanted to analyze urban resilience at the community scale as finely as
305 possible, we chose to analyze the IRIS (Islets Grouped for Statistical Information) scale which constitutes the basic
306 building block for the dissemination of infra-communal data.

307 The concept of "open" and accessible science has been developed to strengthen dialogue and commitment
308 among scientists and the local population around common issues and problems, by creating a language and
309 vocabulary understandable to everyone. While there are obvious limitations to Open Data - security, privacy and
310 property protection - it is nonetheless accepted that using ideas and knowledge freely is a universal right. This is
311 why we chose to use a data source whose access, use and downloading are free, to ensure not only the
312 reproducibility (Jovanovic et al., 2018) of the methodology, but also to participate in the education and
313 communication of the concept of resilience.

314 In addition to the INSEE INSPIRE database, we used the data from the SIRENE database (INSEE) in
315 Open Source. The SIRENE database is an INSEE service used to identify all the characteristics of companies and
316 establishments. The information provided gives a precise idea of the company's activity, its date of creation, etc.
317 These data were used for the urban and technical resilience indicators to demonstrate economic, urban and
318 technical dynamism.

319 Data from the city cadaster (MAJIC) were also used to complete the Open Data database. These data are
320 considered sensitive and owned exclusively by municipalities. It is therefore essential to create a partnership with
321 a city and its GIS services.

322 3.2.2 Method and tools for resilience assessment

323 After selecting the raw data, data were transformed and normalized with a theoretical orientation. In order
324 to understand the frequency of each variable, each item of raw data has been transformed into percentages
325 (Equation (1) and Equation (2)).

$$326 \quad \text{Social Resilience} = \frac{\text{Variable}}{\text{IRIS Population}} \quad (1)$$

327

$$328 \quad \text{Urban resilience} = \frac{\text{Variable}}{\text{IRIS AREA}} \quad (2)$$

329

$$330 \quad \text{Technical resilience} = \frac{\text{Variable}}{\text{IRIS AREA}} \quad (3)$$



331 Nevertheless the weighting is 1 for all the variables (Holand et al., 2011). This single weighting is explained by
332 the willingness to avoid disparities between the variables (Fekete, 2009), since some of them are sensitive and
333 subjective. Indeed, we have no theoretical references (Esty et al., 2005) and there is no practical experience
334 (Fekete, 2009) on which to determine weights that are mostly subjective. Besides, to apply such weights does not
335 necessarily reflect decision makers' and urban planners' priorities and realities (Cutter et al., 2010). Nonetheless,
336 since this approach puts forward a participatory and collaborative methodology, readjusting the weight of these
337 variables with regard to managers' perceptions is entirely justified and in line with the current approach.

338 Following this process, it was necessary to determine a normalization. Normalization allows adjusting a
339 series of values (typically representing a set of measurements) according to a transformation function to make
340 them comparable with certain specific reference points. We proceeded with a Min-Max standardization (Casadio
341 Tarabusi and Guarini, 2013) to obtain a positive resilience impact variable, Equation (3), and a negative resilience
342 impact variable, Equation (4), where each variable is decomposed into an identical range between zero (worst
343 rank) and one (best rank), to create indicators with similar measurement scales, and to compare them.

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$$\frac{x - \min(x)}{\max(x) - \min(x)} \quad (3)$$

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$$1 - \frac{x - \min(x)}{\max(x) - \min(x)} \quad (4)$$

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350 The choice of processing tools was influenced by the availability of Open Source tools in order to uphold
351 transparency and collaborative approaches as well as the availability of such tools to all stakeholders. To create
352 the computer script, we used a tool, the Feature Manipulation Engine (FME) to extract, transform and load raw
353 data (Extraction Transformation Loading). Its interface allows visualizing each step of the processing, from
354 loading raw data (INSEE files) to choosing variables, while integrating the resilience formula to finally obtain the
355 results. Although this tool has a cost, it is nevertheless used by GIS practitioners on a large scale, nationally and
356 internationally.

357 Several steps (Fig.7) were necessary to set up the computer script, integrate the input data, create a
358 geometry, generate the processing and forecast an overall resilience value. The output data is in SpatialLite format
359 (sl3 format), which is a spatial extension of SQLite and provides vector geodatabase capacity. This format can be
360 understood by many processing, visualization and mapping software applications including QGIS.

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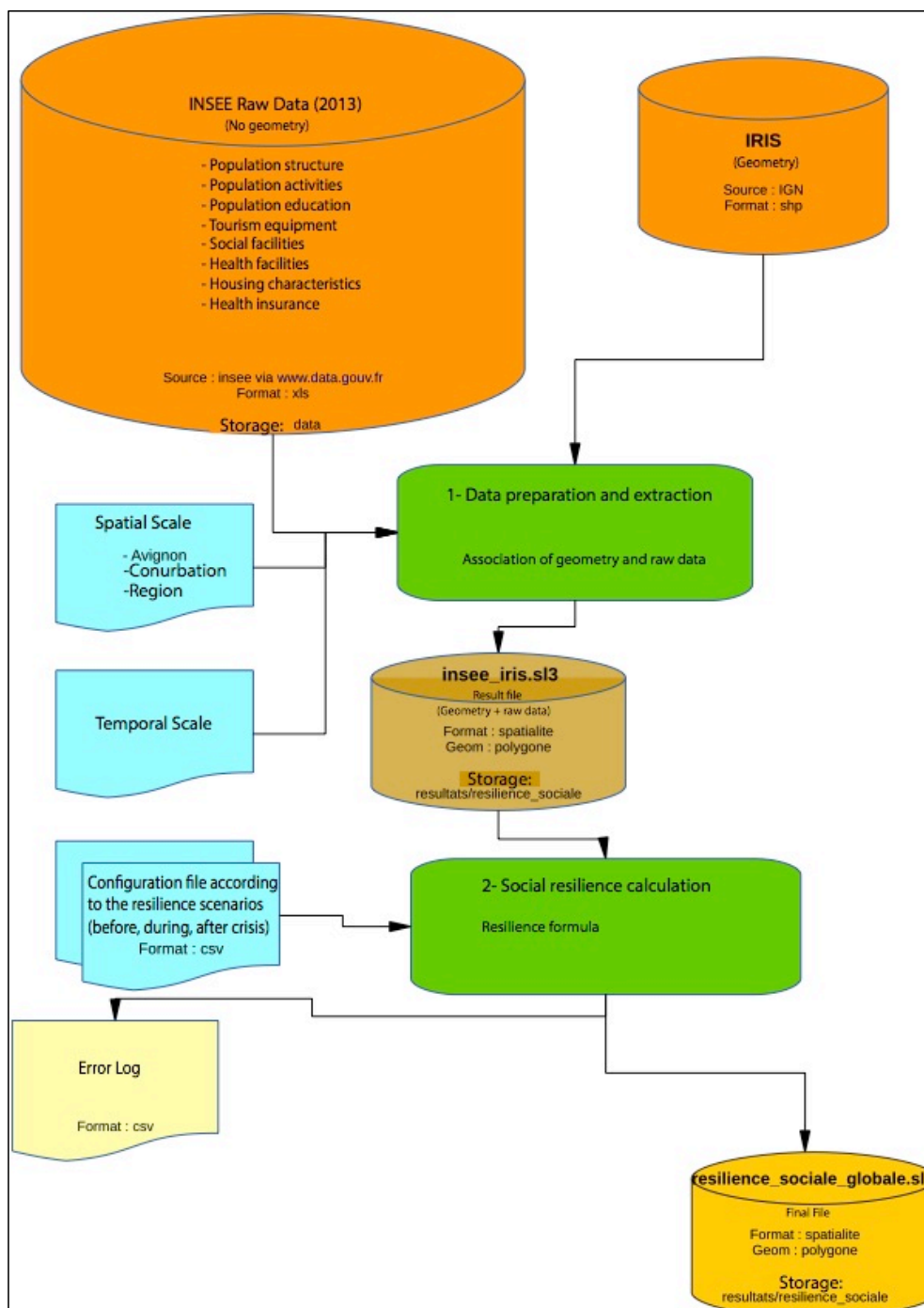


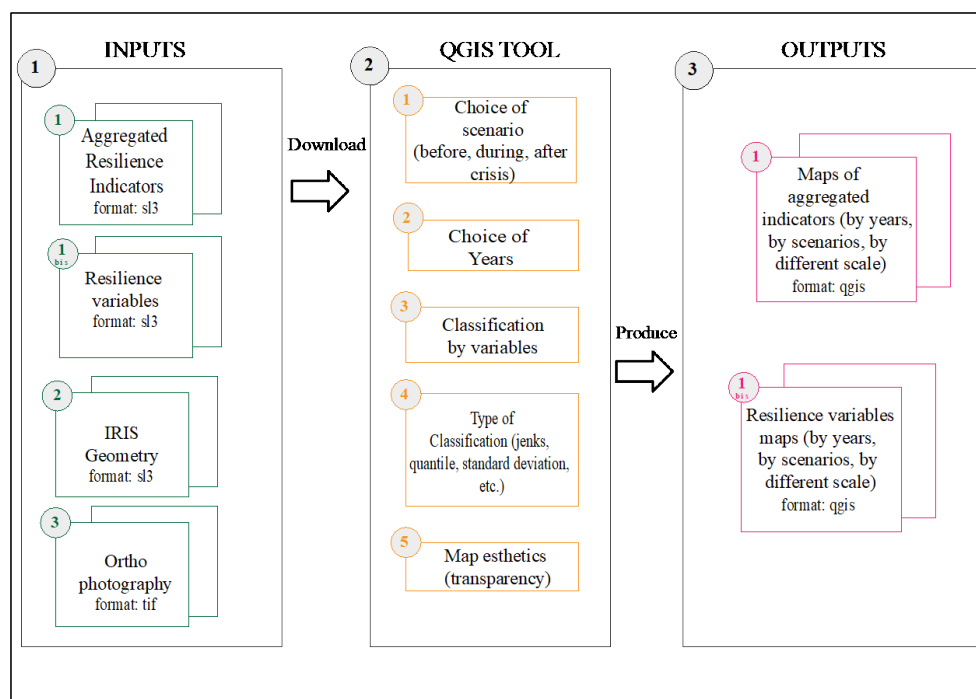
Figure 7: Details of the social resilience assessment process

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364 Once the computer processing was completed, the visualization and analysis work was done via a GIS, namely
 365 the QGIS software (Fig.8). It allows the automatic spatialization of data according to data variables or variables



366 resulting from relationships between objects, and finally the use of graphical tools to visualize and differentiate
367 data (sizes, colors, distances).



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Figure 8: QGIS interaction architecture for resilience assessment.

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The map is therefore a decision-making tool in the sense that it represents and filters a mass of data and makes them accessible and comprehensible. But the production of a map cannot in itself be considered a spatial decision support system: its value first depends on the consistency and reliability of the information collected upstream, then on its structuring and effective readability.

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4- Testing the resilience model and geovisualization process in Avignon

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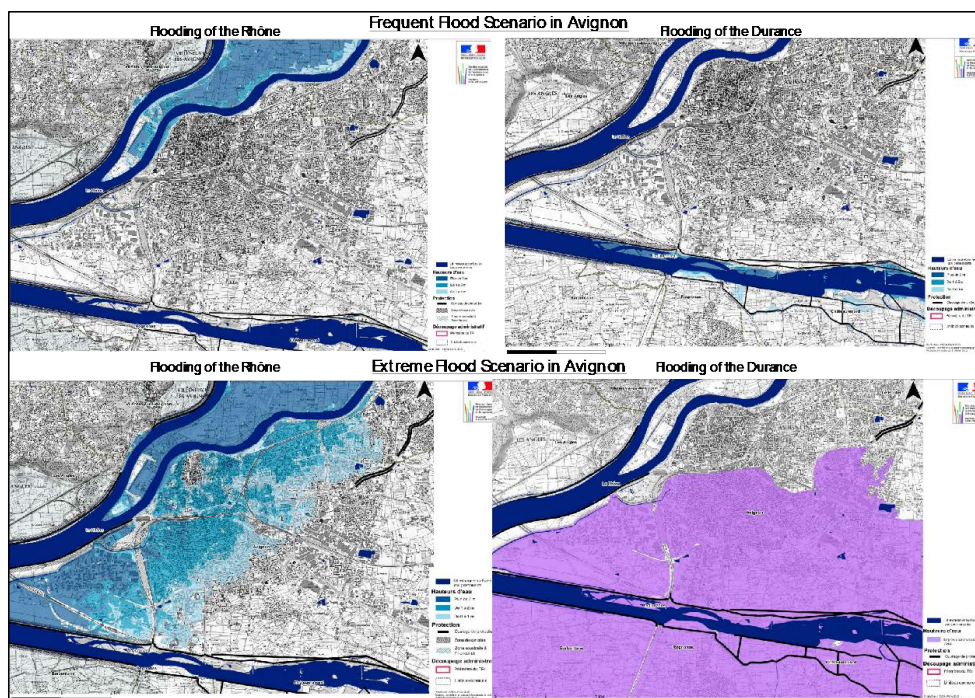
4.1 Avignon flood issues

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Avignon, the chief administrative center of Vaucluse, is faced with flood risks due to its proximity to the confluence of the Rhône and Durance rivers (Fig.9). The island of Barthelasse, the largest river island in Europe, is the area of Avignon most affected by the Rhône's floods. It serves as a buffer between the city and the Rhône, and serves to absorb floods. The few existing dikes protect the island from low floods, but it is still floodable, as shown by the 2-meter floods in 1993, 1994, 2002 and 2003.

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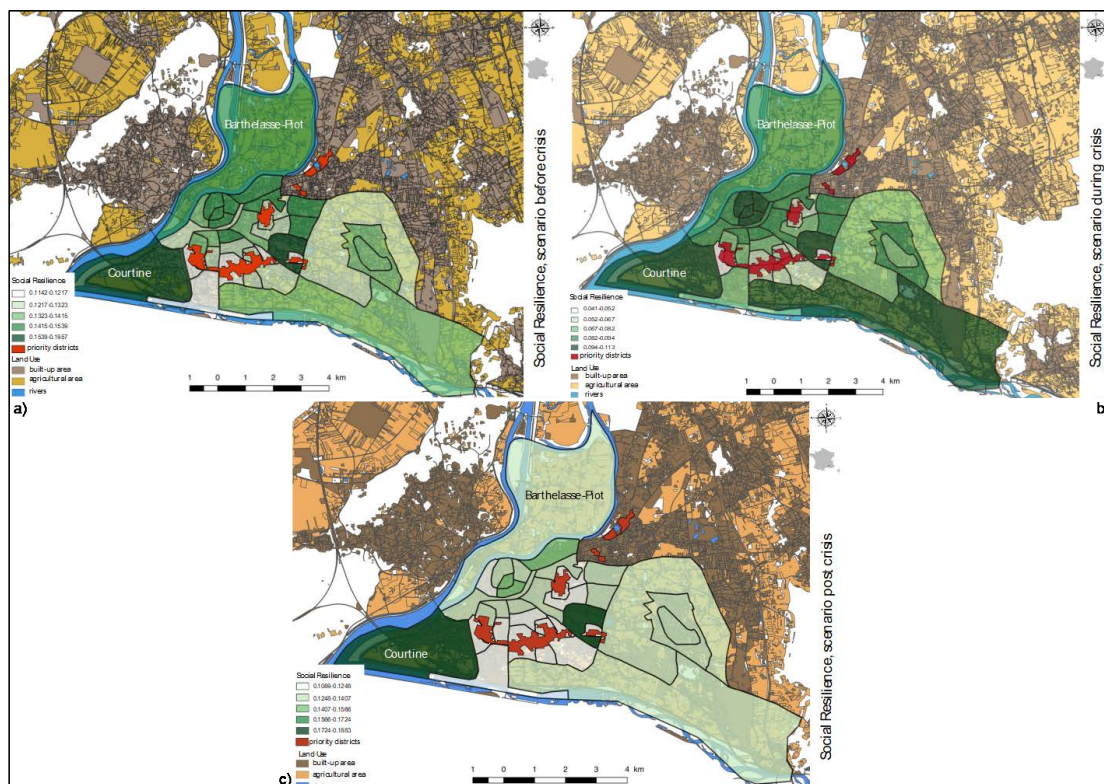
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Figure 9: Frequent and extreme flood scenario in Avignon, (Heinzle et al., 2019) inspired by @ DREAL PACA

388 Therefore, a spatial decision-support system that integrates resilience in practice would be helpful for a flood-
389 prone community. It was developed in partnership with the Avignon city council GIS Department. This
390 collaboration took place at several levels, both in the involvement of local actors in the study, in the data exchange
391 process, and in the choice of processing tools, to ensure and improve the re-usability of the methodology once the
392 study has been completed. The final choice of resilience variables, data processing, and their final visualization
393 was made in constant collaboration with the city's technical services, to ensure that data and their analyses were
394 shared and understood.

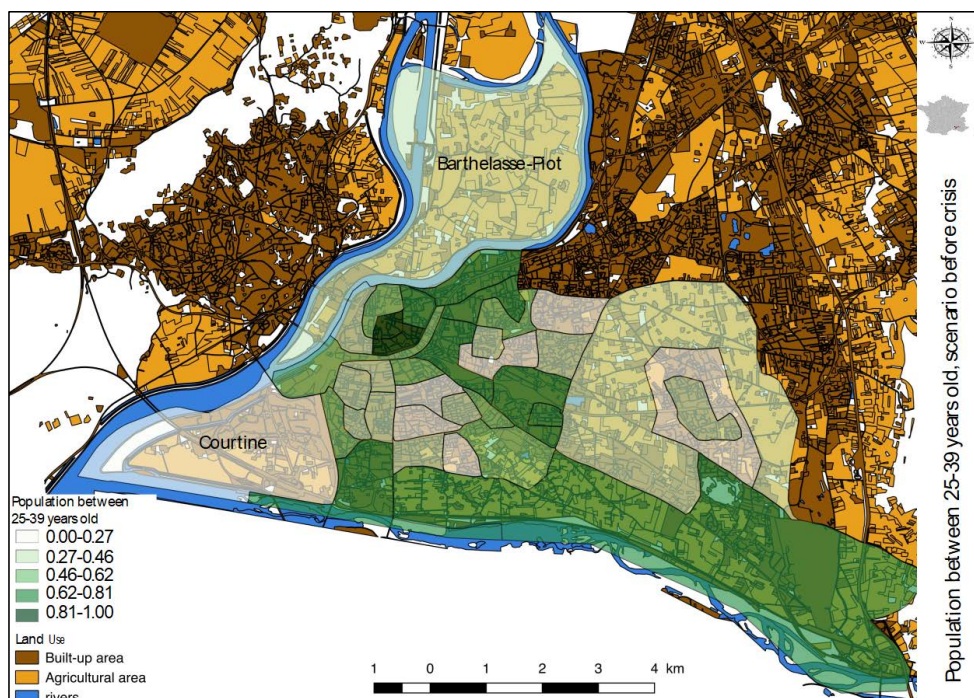
395 *4.2 Resilience to flood assessment in Avignon: a few results*

396 The city was divided into a local scale – IRIS scale – to visualize which areas are resilient or not. The indicators
397 - social (Fig.10), urban and technical- and each variable (Fig.11) included in the model can be visualized.
398 Therefore, it is easier to perceive which variables improve resilient capacities, and which areas have developed
399 these variables or not. As each indicator is independent from each other, it is easier for politicians and managers
400 to work on variables with low levels of resilience and identify areas to be redeveloped and / or reintegrated in
401 urban dynamics.



402
403 *Figure 10: Social Resilience Indicator-Multi scenario, Avignon scale (IRIS scale analysis). The left-hand map identifies the*
404 *most resilient areas (greener) according to the social resilience characteristics before a crisis; the right-hand map identifies*
405 *the most resilient areas according to the social characteristics during a crisis; the map below identifies the most resilient*
406 *areas according to the social characteristics after a crisis - Open Database License, "ODbL" 1.0.*

407
408 The difference between the three maps in Figure 10 is explained by the different scenarios considered before,
409 during and after a crisis. Not every variable is included in every scenario. For example, age variables are important
410 both before (preparation, knowledge of risk, etc.), during (understanding of the situation, ability to move, etc.),
411 and after the reconstruction process. On the contrary, whether or not individuals have a job does not play a role
412 during the crisis but is decisive afterwards, in order to rebuild and relaunch an activity.



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Figure 11: Population between 25 and 39 years old-scenario before crisis, Avignon scale (IRIS scale analysis). The map above identifies the value of the population variable 25-39 years old according to the total IRIS population before a crisis - Open Database License, "ODbL" 1.0.

417

418 Figure 11 shows the location of individuals aged 25 to 39 years, with a segment of the population
419 potentially more resilient, before, during and after. Indeed, they can have a risk culture beforehand, act and
420 survive during, and restart an activity after the disruptive event. They are more prevalent in the city center and
421 in the South and South East. This is mainly due to the location of the two universities, in the city center and outside
422 the city walls, which favors student accommodation and low-cost housing.

423



Resilience indicators/ IRIS	Variables	Impact on resilience	Barthelasse	Courtine
Social resilience indicator	<i>Population Structure</i>			
	00-02 years old	Negative	0.08	0.28
	25-39 years old	Positive	0.35	0.23
	more than 80 years old	Negative	0.67	1
	<i>Professional situation</i>			
	Active 15-64 years old	Positive	0.69	1
	Unemployed 15-64 years old	Negative	0.40	0.38
	<i>Habits</i>			
	Active people 15 years or older not using transport	Positive	0.59	1
	Active people 15 years or older, using public transport	Positive	0.24	0.08
	<i>Insurances</i>			
	Health insurance beneficiaries	Positive	0.50	0.92
	Beneficiaries of CAF allocations	Positive	0.19	0.93
	<i>Education</i>			
	Exit before the 3 rd grade	Negative	0.89	0.02
	Bac +2 and better	Positive	0.55	NULL
Urban resilience indicator	<i>Buildings</i>			
	Number of main residences built before 1919	Positive	0.05	0
	Number of main residences built from 1919 to 1945	Negative	0.89	0.99
	Number of main residences built from 1946 to 1970	Negative	0.95	1
	Number of main residences	Negative	0.92	0.97



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	built from 1971 to 1990			
	Number of main residences built from 1991 to 2005	Negative	0.02	0.02
	Number of main residences built from 2006 to 2010	Positive	0.01	0.26
	<i>Critical Infrastructures</i>			
	Defense	Positive	NULL	NULL
	Fire and rescue services	Positive	0.12	0.34
	Hospital activities	Positive	0	0.75
	<i>Economic dynamics</i>			
	Tourist and other short-term accommodation	Positive	0.43	0.51
	Creation of new companies	Positive	0.05	0.29
	Removal of companies	Negative	0.93	0.64
Technical resilience indicator	<i>Diversity of networks</i>	Positive	0	0.85
	<i>Network accessibility</i>	Positive		
	Accessibility of networks by public road within a 100m radius		0	0.17

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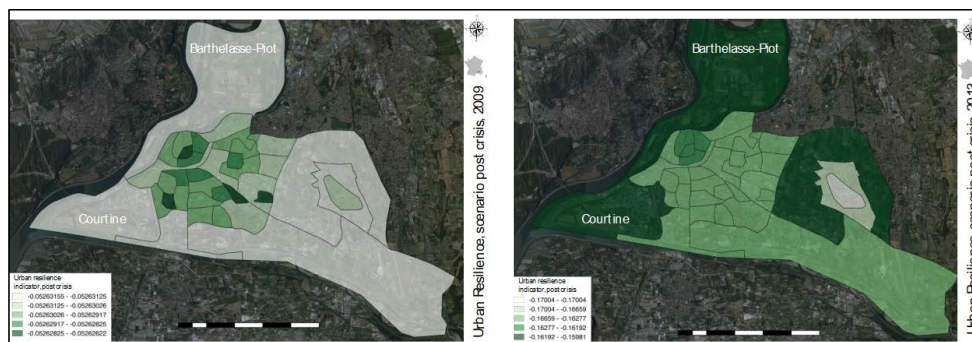
Table 2: Variable values - scenario before crisis

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Table 2 presents the value of each variable according to the pre-crisis scenario for the IRIS of Barthelette and Courtine (Tab.2) after Min-Max standardization. These values illustrate the representativeness of each variable in the territory, and make it possible to understand the social and spatial dynamics at the IRIS scale for Barthelette and Courtine. This detailed analysis, carried out on a variable-by-variable basis, allows engaging in a discussion with local actors in an attempt to reintegrate neighborhoods at the margins of territorial functioning, in order to work on the integration of urban resilience in the face of daily territorial stresses and when confronted by a more exceptional event such as a flood.

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Regarding the urban resilience indicator, INSEE data are available from 2009 to 2013, thus making it possible to perform a multi-date analysis over several years (Fig.12) and gain understanding of urban evolutions and resilience trends. For instance, certain elements have evolved, such as the proportion of tourist accommodation, and surgical and hospital activities, thereby increasing resilience capacities.

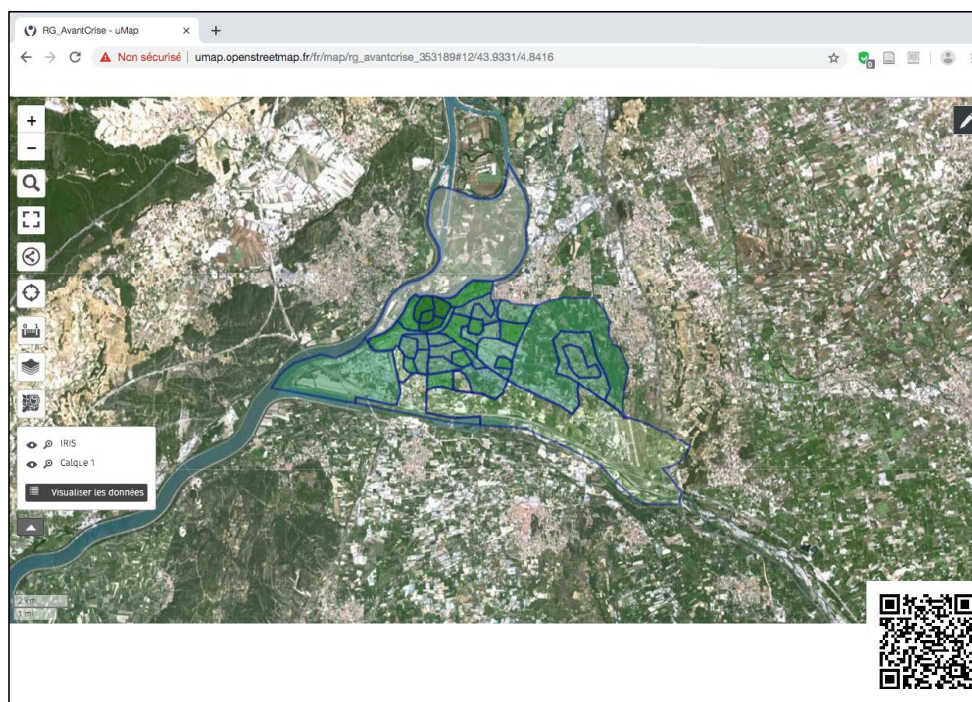


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Figure 12: Comparative analysis between 2009 and 2013 for the Urban indicator, post crisis scenario, Avignon scale (IRIS scale analysis). The map on the left identifies the most resilient areas according to the urban characteristics after a crisis in 2009; the map on the right identifies the most resilient areas according to the urban characteristics after a crisis in 2013 - Open Database License, "ODbL" 1.0.

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After reflection on the visibility of the results, information sharing, and a neutral and collaborative approach, we are considering making our work accessible to inhabitants by developing a website to continue the risk communication process on flood risks and strengthen the geovisualization process. This website, which is currently subject to reflection, developed with the creation of interactive maps accessible via a web link (<http://u.osmfr.org/m/353189/>) and a QR code (Fig.13).



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Figure 13: Web link Global Resilience - Before Crisis Scenario Map – QR code - Open Database License, "ODbL" 1.0.



481 Concerning the contribution of geovisualization used to promoter dialogue on the issue of territorial
482 resilience, workshops were organized to develop interaction around the maps produced and the database provided
483 and accessible to stakeholders. These workshops provided an opportunity for scientific experts and critical
484 infrastructure managers as well as decision-makers in risk management strategies to exchange views in order to
485 support the reflection process and foster long-term collaboration. These maps and this new database allowed the
486 actors to extract new knowledge from the decision support tool, especially theoretical knowledge provided by the
487 maps and consistent with the database. This knowledge is both current but also part of a long-term construction,
488 since the data evolve as a function of INSEE production.

489
490 5- Discussion

491 This research is at the crossroads of resilience modeling and geovisualization practices based on visualization,
492 data processing, mapping and also the decision support process. Rather than focusing on technological
493 developments, this work attempted to reflect on the accessibility of the methodology and its appropriation by local
494 stakeholders. The results are expressed through maps illustrating the potential for social, urban and technical
495 resilience at the community level. It therefore takes into account a large number of dimensions in making the
496 concept of resilience operational.

497 Several improvements are already being considered to overcome the limitations of this work. Concerning the
498 question of tools, willingness to switch entirely to free tools led to reflection on abandoning the FME tool. The
499 project to build a QGIS plugin is under study in view to increasing accessibility. The advantage of the plugin
500 would be to make the computer script behind the methodology completely free, accessible and downloadable.

501 Another improvement to consider would be to test the approach in other territories, either by developing a
502 partnership of the same scope or by switching the entire process to open data. At present, this analysis can be
503 performed at the scale of the Sud Provence-Alpes Côte d'Azur region (Fig. 14) and at the scale of France, but only
504 for the social resilience indicator.

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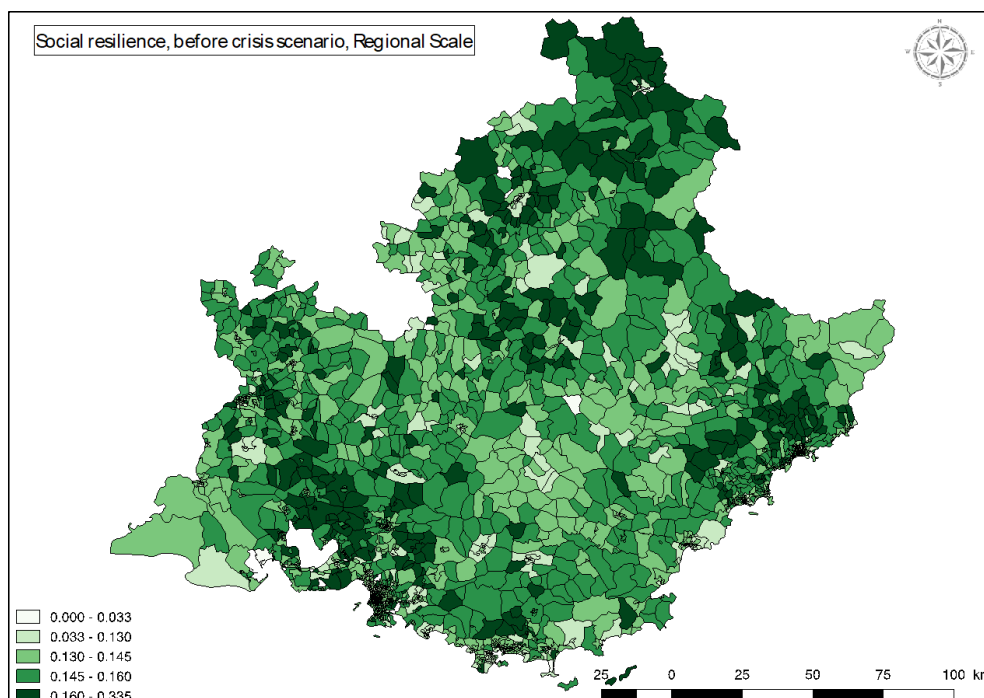


Figure 14: Social resilience, before crisis scenario, Provence-Alpes-Côte d'Azur regional scale - Open Database License, "ODbL" 1.0.

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Following theoretical modeling and visual, cartographic and geovisualized production work, further development included the organization of workshops to question users on their understanding and use of the tool and the results. These workshops took place with critical infrastructure managers and made it possible to (re)launch the debate around the issue of resilience and thus to build knowledge without hypothesis a priori (Maceachren and Kraak, 1997) around a tool for visualizing a concept that is difficult to put into practice.

This modeling and cartographic production work based on geovisualization has made it possible to rethink the issue of urban resilience. The mapping results led to workshops to review and compare the methodology with the reality of the territory and risk management practices. This work is part of a broader dynamic and reflection on the question of operationalizing resilience. Based on the results of this decision-making tool designed to operationalize urban resilience, a more global project is now under construction. It is thus planned to use these results to build an urban resilience observatory that will be tested on the island territories of French Polynesia. This will provide an opportunity to merge representations of risks, territories and techniques for data processing, production and analysis, visualization, and collaboration with local actors.

6- Conclusion



529 This article proposed a methodology intended to clarify the concept of resilience in the context of increasing
530 urban flooding. This methodology is divided into two stages. First, the modeling and analysis of the concept of
531 resilience through the formulation of three definitions and measurement indicators in order to approach resilience
532 in an exhaustive way on the basis of social, technical and urban criteria. Secondly, we used geovisualization
533 techniques (mapping practices, visualization, data processing and analysis, map processing) to build a spatial
534 decision support system accessible and understandable to local stakeholders in the Avignon community. This
535 spatial decision support system sought to provide a simple and accessible methodology to quickly verify and
536 analyze information for decision-making. The aim is to use the principles of visualization of geovisualization to
537 widely disseminate map results in order to improve resilience culture. The contributions and innovations of this
538 work are therefore of several kinds:

- 539 - the design of a spatial decision support system with and for local actors;
- 540 - the design of a resilience model;
- 541 - the use of open access data to enhance INSEE data and match the knowledge of local actors;
- 542 - the use of tools to highlight the visualization of data processing: FME and QGIS;
- 543 - the use of free and easy to use tools to perform advanced mapping processing;
- 544 - the implementation of dialogue between local experts and actors through visual and understandable
545 cartographic production.

546

547 The advances achieved have made it possible to map resilience at the local level, ensure that local actors are
548 understood, and that the methodology is accessible to non-experts and reproducible. The method therefore focuses
549 on the accessibility promoted by geovisualization techniques rather than on technicality.

550

551 While some limitations have been observed - in particular regarding the non-exhaustiveness of open access
552 tools, the need to include local actors from the outset and changes of scale - many perspectives are already being
553 considered for the future. The first step has been taken to switch all the tools to open access via the development
554 of a QGIS plugin. In addition to the response to the tools, this plugin will also integrate the reflections of different
555 actors in order to develop the tool using the feedback expressed. Regarding the issue of scale, the need to go
556 beyond the national framework was expressed through reflection on the use of Open Street Map data. Finally,
557 regarding the form of this spatial decision-making tool itself, work is in progress to develop it by setting up a
558 Resilience Observatory for the island territories of French Polynesia. Studies and analyses are being carried out to
559 this end.

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561

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