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## Spatial aggregation of indicators in sustainability assessments: Descriptive and normative claims

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# 1 Spatial aggregation of indicators in sustainability assessments: descriptive and 2 normative claims

## 3 Abstract

4 Indicators are widely used in sustainability assessments. They serve both a descriptive function  
5 (i.e., assessing a situation or effects of potential changes) and a normative function (i.e., allowing  
6 the expression of value judgments). These functions are usually considered when identifying and  
7 using indicators. However, processes such as formalization, estimation, and customization are  
8 needed to produce tangible indicators. These processes and their influence on sustainability  
9 assessments are studied less often. We focus on spatial aggregation, a specific type of customization  
10 commonly used for landscape-scale and regional assessments. Using a database with 146 indicator  
11 profiles for water management, we investigated reasons for spatial aggregation choices, i.e. whether  
12 indicators based on spatially-explicit data are aggregated while under development or are provided  
13 to users in a disaggregated form. Although the literature assigns a descriptive function to spatial  
14 aggregation, our database shows that reasons underlying aggregation choices are more diverse.  
15 These reasons include highlighting differences, fitting to the scale of a process, fitting to criteria,  
16 recognizing a lack of knowledge, expressing social rationality, contextualizing information, and  
17 allowing different interpretations of the same indicator. Some of these reasons reflect the choice to  
18 expand or reduce the range of potential uses of an indicator, and therefore the potential for different  
19 viewpoints to confront each other. Hence, normative claims combine with descriptive claims when  
20 aggregating indicators, and even more so when customizing them. In general, the form of indicators  
21 merits more attention in the practice and theory of sustainability assessments.

## 22 Highlights

- 23 • Indicators are viewed as objects to describe and debate a situation.
- 24 • Indicators result from different information processes that are sometimes “hidden”.
- 25 • The process of spatial aggregation is investigated.
- 26 • Spatial aggregation choices provide a degree of leeway in interpreting indicators.
- 27 • Choices illustrate tension between the need for consistency and that for diversity.

## 30 1. Introduction

31 Sustainable management of natural resources requires governance that considers long-term  
32 dynamics and the spatial scale of the resource managed and that allows different actors to

33 participate in the decision-making process. Sustainability assessments combine tools that can assist  
34 decision-makers in this task (Ness et al., 2007). Most sustainability assessment tools require  
35 indicators, which can be used to assess a situation and measure progress towards sustainable  
36 development (Pires et al., 2017; Singh et al., 2012; Smeets et al., 1999) or are embedded in  
37 prospective methods to assess scenarios involving change or policy options (Leenhardt et al., 2012;  
38 Singh et al., 2012).

39 Several guidelines and methods exist to identify relevant sets of indicators for sustainability  
40 assessments (Alkan Olsson et al., 2009; Bockstaller et al., 2009; Kurka and Blackwood, 2013; Reed  
41 et al., 2006; Valentin and Spangenberg, 2000), and the variety of methods reflects the diversity of  
42 contexts in which indicators are used. For instance, within “governance contexts” (Hezri and  
43 Dovers, 2006), in which people outside of the political elite participate in the decision-making  
44 process, indicators cannot be produced according to a clearly identified audience (Hezri and  
45 Dovers, 2006). Considering this, several authors claim that identifying indicators should be a social  
46 learning process that involves multiple participants (Bell and Morse, 2004; Fraser et al., 2006;  
47 Valentin and Spangenberg, 2000).

48 The large number of studies on indicator identification could suggest that defining the indicator  
49 set is the only crucial step of information processing, from which comes results of the sustainability  
50 assessment. However, other processes occur between defining the relevant indicator set and  
51 evaluating a situation or option. For spatial decision support systems, Uran and Jansen (2003)  
52 noted that “output sometimes needs simplification, aggregation, structuring, or another form of  
53 processing in order for it to be used in a decision-making process. In some systems this is done  
54 automatically, or ‘hidden’, so the user is unaware of the fact that an evaluation step has been made”.  
55 This highlights the need to clarify and question the entire chain of processes required to develop  
56 indicators (and not only to identify them); this is the core motivation for this article.

57 More specifically, we focus on spatial aggregation, a specific process that occurs after an  
58 indicator is identified. Spatial aggregation entails changing fine-resolution data into coarser-  
59 resolution data (e.g., the entire landscape or region) to derive “meaningful” information. We  
60 distinguish “data” from “information” according to definitions of Pahl-Wostl et al. (2013): “‘Data’  
61 are symbols, such as the numbers produced by a temperature-measuring device, whereas  
62 ‘information’ places data in relation to some meaning that makes them useful (e.g., impacts of July  
63 temperature on the yield of a certain crop).” Spatial aggregation is often considered to serve a  
64 descriptive purpose: spatial aggregation choices (e.g., which aggregation pathways, which spatial  
65 resolution) can depend on the characteristics of the model used (Faivre et al., 2004; Jansen et al.,  
66 2009), on the process the indicator intends to describe (Alkan Olsson et al., 2009), or on the  
67 expected assessment scale (Chopin et al., 2017). Spatial aggregation clearly differs from indicator

68 aggregation, which involves condensing several indicators into a smaller number of indicators.  
69 Indicator aggregation is known to embed normative values and thus go beyond describing social-  
70 ecological processes (Böhringer and Jochem, 2007; Mayer, 2008). For instance, ecological  
71 economists have demonstrated that using compensatory or non-compensatory rules to aggregate  
72 indicators respectively entails a weak or strong conception of sustainability (Garmendia and  
73 Gamboa, 2012; Martinez-Alier et al., 1998). However, since ecological economics is as unfamiliar  
74 with spatial decision problems (Allain et al., 2017) as spatial planning is with non-equivalent  
75 descriptions of a problem (Ramsey, 2009), spatial aggregation of indicators is not studied as a  
76 normative process.

77 We consider spatial aggregation both a descriptive process (i.e., translation of information in a  
78 formal system) and a normative process (i.e., in which actors express value judgments). To explore  
79 how descriptive and normative aspects are combined when considering a sustainability issue with a  
80 spatial dimension, we used a database of indicator profiles built to compare water management  
81 options.

82

## 83 **2. Definitions and theory**

### 84 **2.1 Defining “indicator”**

85 The term “indicator” has multiple definitions (Abbot and Guijt, 1998), and most authors use a  
86 finalist or functionalist definition, in which indicators are defined according to their purpose.  
87 Consequently, the nature of indicators remains ambiguous: i.e., whether indicators are ideas,  
88 variables, objects, variable values, etc.

89 In quantitative assessments, an indicator set is generally displayed in a table containing the names  
90 of variables and some additional attributes, such as definitions, scales, calculation methods, and  
91 units. In this sense, indicators are variables, and a situation is assessed according to their values  
92 (Chopin et al., 2017). However, from a deliberative perspective, “indicators” include outputs of  
93 analysis (“The species is not impacted by gear, as a secondary involuntary catch, in any significant  
94 way”), variables (“Gross efficiency of the catch (catch/net P.P.)”) or ideas for analyses  
95 (“Length/Frequency analysis of catches”) (examples from Douguet et al., 2010). In addition,  
96 Meadows (1998) stated that “Indicators can take many forms. They don’t have to be numbers. They  
97 can be signs, symbols, pictures, colors.” Consequently, when indicators are identified in contexts  
98 with multiple actors, the term “indicator” represents any type of argument, be it a suggested  
99 analysis, a quantitative result, a color or a photograph, that empowers someone to assess or judge a  
100 situation.

101 Although the vagueness of the term “indicator” may be useful in deliberative contexts, for the  
102 purpose of this article, we consider a narrower definition: *an object with meaningful qualitative or*

103 *quantitative information that facilitates learning about a situation and forming a value judgment*  
104 *about it.* This definition assumes the existence of one or many indicator developers, i.e. those who,  
105 from a heterogeneous set of suggestions (of varying degrees of development) for potential  
106 indicators, create a set of formal objects (the indicators).

107

## 108 **2.2 Steps in indicator development**

109 Our analysis focuses on indicator development, a stepwise process of coding information that  
110 occurs between the processes of identifying potential indicators and using them (Fig. 1). We provide  
111 insights into indicator identification and use before describing intermediate processes involved in  
112 indicator development.

113 The literature on indicator identification highlights qualities that a “good” sustainability indicator  
114 should have (Reed et al., 2006), which are generally linked to the latter’s soundness and ease of use.  
115 The literature also contains many frameworks developed to derive indicator sets, such as goal-  
116 oriented frameworks (Alkan Olsson et al., 2009), multi-scale and systemic frameworks (Astier et  
117 al., 2012; Bossel, 1996), and ecosystem-service frameworks (de Groot et al., 2010). Methods to  
118 identify indicators are as varied as the frameworks. They can be divided between top-down (a  
119 generic framework transposed or adapted to a local context, e.g. Speelman et al. (2007)) and  
120 bottom-up approaches (indicator sets are derived from locally relevant issues, and generic  
121 knowledge is used to explore these issues, e.g. Fraser et al. (2006)). Both can involve experts and  
122 stakeholders, and can be implemented in a deliberative approach involving “extended peer  
123 communities” (Funtowicz and Ravetz, 1990) or in a prescriptive approach. Within deliberative  
124 approaches, developing an indicator “profile” (i.e., meta-information related to the scientific  
125 validity of an indicator and its relevance to the context in which it will be used) is considered a  
126 crucial element of knowledge quality (O’Connor and Spangenberg, 2008; Sluijs et al., 2008).

127 Management science studies of the performativity of indicators and their use as management  
128 tools in organizations and society (Desrosières, 1997; Espeland and Sauder, 2016) have inspired a  
129 literature on the uses of sustainability indicators. Some authors clearly distinguish indicator uses  
130 from misuses (Lyytimäki et al., 2013), suggesting that safeguards are required to avoid  
131 inappropriate uses. Other authors build on the principle that the users and uses of sustainability  
132 indicators are diverse. For instance, Hezri and Dovers (2006) distinguish five types of indicator uses  
133 for policy-making in a governance context: instrumental (to solve problems), conceptual (to  
134 increase understanding), tactical (as a “substitute for action”), symbolic (as “ritualistic insurance”)  
135 and political (to support a pre-defined position). Because we focus on using indicators for group  
136 deliberation rather than effective policy-making, we consider it more appropriate to distinguish a  
137 descriptive function from a normative function of indicators (see section 2.3).

138 Indicator development proceeds stepwise, with the potential for several iterations and feedback  
139 between steps (Fig. 1). After indicator identification (step 1), “formalization” (step 2) is necessary  
140 to help select variables and specify characteristics of future indicators. Indicator profiles result from  
141 this formalization, and they may be transferred to similar issues or serve as a source of inspiration  
142 for different issues. Next, “estimation” (step 3) produces raw data (e.g., values of the variables  
143 selected for the case study that are obtained through expertise, simulation or a data search).  
144 “Customization” (step 4) includes all activities that provide meaning and tractability to outputs of  
145 the estimation (Leenhardt et al., 2012), such as spatial aggregation, classification (aggregating into  
146 classes), creation of archetypes, and shaping (e.g., into a map, a graph). The result of customization  
147 is called an “indicator” (see section 2.1). The last step, indicator use (step 5), follows development  
148 of indicators. It has another nature: it produces knowledge, defined as information embedded in a  
149 context of interpretation (Pahl-Wostl et al., 2013). As emphasized by O’Connor and Spangenberg  
150 (2008), knowledge *based on indicators* and knowledge *about indicators* can be derived from  
151 indicator use. This knowledge about indicators feeds back into the previous steps of indicator  
152 development and can help to identify and develop new indicators and discard or modify other  
153 indicators.

154

### 155 2.3 Assessment and evaluation

156 Indicator use can refer to “assessment” or “evaluation” (Fig. 1). Although they are generally used  
157 synonymously, we use the terms to indicate two different activities. Assessment is a descriptive  
158 activity in which a user an indicator is compared to other estimated values or references. For  
159 instance, the workload on different farms may be measured or estimated, which provides a basis for  
160 *assessing* which farms have a workload that *surpasses* the national average. Evaluation is the  
161 process of assigning a value judgment to the information, such as whether the workload is  
162 *acceptable* or *fair*. In short, one *assesses* when stating that the workload is “high” but *evaluates*  
163 when claiming that it is “too high”. In contexts with multiple stakeholders, assessment allows for  
164 confrontation of different descriptions of a situation, while evaluation allows for confrontation of  
165 different value systems.

166 Because confrontations are essential to democratize knowledge-production and decision-making  
167 processes (Funtowicz and Ravetz, 1993), we consider that a “good” indicator should be useful both  
168 for assessing and evaluating a situation or alternative option. Deliberative sustainability assessment  
169 (Frame and O’Connor, 2011) and systemic sustainability analysis (Bell and Morse, 2004) are  
170 frameworks for identifying indicator sets that explicitly recognize this duality. However, these  
171 frameworks do not mention that the duality of indicators influences both indicator identification and  
172 all of the steps that transform indicators into usable forms (Fig. 1). More specifically, because

173 customization generates meaning, it influences indicator use and in return is influenced by the  
174 expected uses. Using the specific case of spatial aggregation, our analysis illustrates how normative  
175 considerations become relevant in the customization process.

176

## 177 **2.4 Spatial aggregation**

178 In defining spatial aggregation as the conversion of fine-resolution data into coarser-resolution  
179 data (e.g., the entire landscape or region) to derive “meaningful” information, we emphasize  
180 changes in scale and in meaning. Different methods can be used to aggregate spatial data (Chopin et  
181 al., 2017; Ewert et al., 2011), but discussing them in detail lies outside the scope of this article.  
182 Sustainability assessments at regional or landscape scales usually include spatial aggregation, since  
183 indicator developers tend to have wide access to spatially-explicit data at these scales.  
184 Consequently, the issue of spatial heterogeneity arises (i.e., should they use average values, quantify  
185 heterogeneity, or keep the data at the finest resolution?).

186 Systematic aggregation of spatial values into a single value is criticized as an oversimplification  
187 of the processes targeted (Scholes et al., 2013) or because of its opacity to decision-makers (Janssen  
188 et al., 2005; Uran and Janssen, 2003). Nevertheless, maintaining spatially disaggregated values,  
189 generally by displaying them on maps, is costly and introduces a high level of cognitive complexity  
190 that can be difficult for users to handle (Jankowski et al., 2001). Similarly, Uran and Janssen (2003)  
191 suggested that providing disaggregated data is useful when users have sufficient skill in processing  
192 spatial data or when they are guided in their interpretation.

193 However, while we recognize diverse uses and multiple users of indicators, it seems difficult to  
194 target the use of (at least some) indicators to a few specific users. Therefore, to enrich the debate  
195 over spatial aggregation, we assume that descriptive and normative uses cannot be separated and  
196 that users cannot be defined beforehand.

197

## 198 **3. Materials and Methods**

199 Here we present (i) how we built the database of indicator profiles on which we based our spatial  
200 aggregation analysis, (ii) the database itself, and (iii) the objective and method of the analysis.

201

### 202 **3.1 Origin of the database of indicator profiles**

203 The database of indicator profiles was developed by implementing steps 1 and 2 described above  
204 (Fig. 1) to address water management in the downstream area of the Aveyron watershed (southern  
205 France). This agricultural area experiences a structural water imbalance in the low-flow period,  
206 when low rainfall coincides with maximum water needs for agriculture. Water management is based  
207 on water restrictions that are implemented when river flow falls below a specific threshold. Since

208 national and European policies advocate for more structural management options (Erdlenbruch et  
209 al., 2013), we created a set of potential indicators to assess contrasting water management options  
210 and encourage public debate over future water management in the Aveyron watershed. Because the  
211 topic of agricultural water management is complex and prone to conflict in the area, undertaking the  
212 sustainability assessment of different scenarios challenges both our ability to represent the social-  
213 ecological processes at stake and create the conditions for dialogue between stakeholders.

214 This context drove our choices for developing indicators. Social mistrust and power asymmetries  
215 made collective meetings impossible from the outset and a top-down approach risky. Therefore, we  
216 opted for a bottom-up approach, based on card-sorting interviews with local and regional  
217 stakeholders (Allain et al., 2016), to construct a grid of criteria (Table 1) reflecting the plurality of  
218 stakeholder preoccupations. Then, we used these criteria to identify (step 1) and formalize (step 2)  
219 the indicator profiles.

220 Although stakeholders sometimes suggested useful indicators or “proto-indicators” (ideas for  
221 indicators, with a variable number of guidelines for their scale, representation, calculation, etc.),  
222 most indicators were identified through a series of 14 meetings with scientific or technical experts,  
223 most of whom worked in the region. We asked the experts to suggest potential indicators that  
224 corresponded to the criteria defined by stakeholders, were relevant at the landscape scale, and if  
225 possible, could be estimated quantitatively or qualitatively using model simulations, expertise, or by  
226 processing geographic information. Expert interviews included questions to facilitate formalization,  
227 such as at which spatial and temporal scales the indicator would be relevant.

228 Stakeholder and expert interviews led to identification of 156 proto-indicators. We specified  
229 complete profiles for 146 of them based on our own expertise. The remaining 10 reflected  
230 arguments stakeholders provided that neither we nor the experts we met with could transform into  
231 variables assessable in an *ex ante* manner.

232 Although the same individuals were sometimes interviewed twice, first as a stakeholder to identify  
233 criteria and then as a local expert to suggest indicators, we could not push the ideal of  
234 democratizing indicator development further. One reason was that local stakeholders, such as  
235 reservoir managers or farmers, use many indicators in their everyday life (e.g. the plant-growth  
236 stage for farmers to launch irrigation, the level of water in reservoirs for public agents to release  
237 water into rivers). Switching from the time and space scale of operational decisions to the scales of  
238 the landscape and multiple years is challenging to them. Also, we aimed at assessing scenarios and  
239 not real situations. We therefore had to identify and develop indicators not measurable here and  
240 now, hence disconnected from what practitioners are accustomed to handle. These barriers can be  
241 levered, but the time required to do so was prohibiting (to stakeholders and to us researchers).



242 Hence, for our case study, we, the experts we interviewed, and colleagues who helped specify and  
243 interpret indicator profiles served as indicator developers.

244

245

### 246 **3.2 Database of indicator profiles**

247 The database (see Appendix) contains 146 indicator profiles (123 based on spatially-explicit  
248 data), with details on the following attributes:

249 - Indicator definition

250 - Meaning: related criteria, justification for suggesting the indicator

251 - Estimation: estimation method (model simulation, expertise, calculation based on scenario  
252 characteristics), scale of raw data (e.g., field, farm, elementary watershed, river, entire  
253 landscape)

254 - Customization: type of representation preferred (e.g., simple value, graph, map, illustration,  
255 narrative), aggregation scale

256 - Evaluation scale: landscape or landscape + sub-landscape

257

### 258 **3.2 Analysis of spatial aggregation**

259 We analyzed the indicator profiles in the database with no additional input from stakeholders or  
260 experts. We focused on the attributes of 115 “horizontally” aggregated indicators (i.e., those that are  
261 based on spatially-explicit raw data provided for multiple locations and thus require aggregation  
262 during either indicator customization or use to become relevant at the landscape scale). Indicators  
263 referring to only a point location (e.g., a watershed outlet) were not considered horizontally  
264 aggregated because a single value at a single point cannot reflect any spatial heterogeneity.  
265 Horizontal aggregation differs from “vertical” aggregation, which yields composite indicators or  
266 overall sustainability scores (Allain et al., 2017).

267 We analyzed the 115 profiles according to who performs the spatial aggregation: developers  
268 (during customization) or users (during use). For the former, developers aggregate raw data to  
269 obtain a coarser resolution (partial aggregation) or a landscape-scale indicator (complete  
270 aggregation) (Fig. 2). Under partial aggregation, outputs remain heterogeneous. Partial aggregation  
271 can remain spatially explicit (e.g., zones) or yield non-spatial classifications (e.g., farm types).  
272 Complete aggregation results in a single value for the entire landscape. In user-led aggregation,  
273 users aggregate spatially heterogeneous information to assess or evaluate the entire landscape. User-  
274 led aggregations are therefore necessarily complete, although some indicators can be used to assess  
275 or evaluate landscape zones or classes in addition to the entire landscape.

276 We reflexively clarified reasons underlying developer-led and user-led aggregations, since we  
277 were involved in the identification and formalization processes for all indicators, most of the time as  
278 indicator developer interviewers but also as developers (see 3.1). Our aggregation choices were not  
279 discussed with the stakeholders expected to use them because the research was not part of a joint  
280 project in which stakeholders are clearly named. Also, as our experience showed, people get  
281 involved and interested at one moment and then, as the research progresses, other people participate  
282 in it. This prompted us to consider a wide range of potential users and favor, when possible,  
283 indicators prone to diverse interpretations.

284

## 285 **4. Results**

286 In presenting results for the 115 indicators, we use 15 indicators as examples to illustrate  
287 important points<sup>1</sup> (Table 2).

288

### 289 **4.1 Reasons underlying developer-led spatial aggregation**

#### 290 **4.1.1 Partial spatial aggregation**

291 The database of indicator profiles shows that only six indicators, associated with four criteria  
292 (“safety”, “biodiversity”, “adjustment potential”, and “maintaining natural capital”), should result  
293 from partial spatial aggregation. For each, raw data estimated at the field or field-islet scale should  
294 be aggregated according to the landscape’s hydrological zones (elementary watersheds) or water  
295 management zones (e.g., for water restrictions). Reasons underlying partial aggregation were  
296 similar to those for aggregation into classes, which indicator developers chose for 24 indicators (for  
297 the criteria “local identity”, “wealth and employment”, “long-term adaptability”, “adjustment  
298 potential”, and “equity” and “efficiency”). For these indicators, raw data at the field or field-islet  
299 scale should be aggregated by crop or farm type.

300 We identified three main reasons for partial aggregation:

- 301 - *to highlight differences* (between zones or classes). For instance, the indicator “area affected  
302 by water restrictions at key periods” estimated at the field scale should be aggregated at the

---

1 1 We often refer to characteristics of “indicators”, but the characteristics are those of profiles and  
2 not of the indicators themselves. Indicator profiles (mainly “definition”, “scale of raw data”,  
3 “aggregation scale”, “representation”, and “estimation method”) provide guidelines for spatial  
4 aggregation that will be usually strictly implemented to produce the indicators, because many  
5 calculations are based on long simulations. This constraint, which precludes a trial-and error  
6 approach, also explains the time and effort devoted to clarifying aggregation choices. In certain  
7 cases, an indicator profile allows for multiple aggregation scales, which can generate several  
8 indicators (one per scale). Consequently, although reasons underlying aggregation choices can be  
9 analyzed before indicators are produced, quantitative data (the number of indicators concerned)  
10 should be considered only indicative.

303 scale of water restriction zones. This would allow users to visualize whether certain zones  
304 are more affected than others, and in which zones farmers have less leeway to adjust crop  
305 management practices during the year. Likewise, raw data for “irrigation costs per ha”  
306 should be aggregated at the crop-type scale to compare crop types and thus provide an  
307 indication of how fairly costs are shared.

308 - *to make an indicator more relevant for the targeted criteria.* For instance, the indicator  
309 “variation in water stored in soils” estimated at the field scale could be used “as is” to  
310 describe the resilience of farms or crops to water stress; however, to produce an indicator  
311 informing the “maintaining natural capital” criterion, data should be aggregated at the  
312 elementary watershed scale.

313 - *to fit the scale of a process.* For instance, since the indicator “pollution from plant protection  
314 practices” describes water pollution caused by agricultural chemicals, raw data should be  
315 aggregated at the elementary watershed scale. Aggregation scales could vary to reflect other  
316 processes, such as pollution of food products or soils. This reason appears linked to the use  
317 of pressure indicators, which are often needed in *ex ante* assessments since direct  
318 measurements and model simulations are not always possible.

319 Our study yielded one reason specific to aggregation into classes: to *increase the intelligibility* of  
320 indicators. Because the study area has many fields and farms, data at these two scales become  
321 unintelligible when considering the entire landscape. Classifying thus renders fine-resolution data  
322 tractable.

323

#### 324 **4.1.2 Complete spatial aggregation**

325 The database of indicator profiles contains 79 indicators (among the 11 criteria) that should be  
326 completely spatially aggregated. Two reasons identified for partial aggregation also exist for  
327 complete aggregation: relevance for the targeted criteria and fit to the process scale. The former is  
328 illustrated by the indicator “volume of rainwater returned to the environment”, which qualifies the  
329 biophysical system (the water system, in the “maintaining natural capital” criterion). Since it does  
330 not seek to indicate the extent to which each water user manages water capital, it should be  
331 aggregated at the landscape scale. The second reason (fit to the process scale) is illustrated by the  
332 indicator “irrigation capacity of all farms”. Although its raw data, at the farm scale, can be  
333 meaningful at this scale, the experts who suggested this indicator referred to the potential to expand  
334 irrigation to additional fields and reallocate water within and among farms. It was also meant to  
335 address how quickly agriculture can respond to shocks (e.g., a rise in maize price), which justifies  
336 aggregation at the landscape scale.

337 We identified three additional reasons for complete aggregation:

- 338 - *easily comparing management options* by providing a single value. For instance, the  
339 indicator “impact of water use restrictions on agricultural yields” (“adjustment capacity”  
340 criterion) should be aggregated at the landscape scale for two reasons: (1) information  
341 would be too complex at smaller scales (e.g., field or farm) and (2) aggregation by farm type  
342 or crop type would mask a potential imbalance among the number of each type. Thus, each  
343 field’s production should be aggregated to calculate a landscape-scale value. To make this  
344 choice, developers assumed a commensuration process, i.e. to consider all crops and all  
345 farms equivalent.
- 346 - *addressing uncertainties*. An example related to model and data uncertainties is the indicator  
347 “use rate of reservoirs”, based on model predictions. Since data for individual water  
348 withdrawals were unavailable, accuracy of the model (MAELIA, Gaudou et al., 2013) could  
349 be verified only by comparing its predictions with observed data for river flows and  
350 cumulative water withdrawals. Also, allocating a withdrawal point to a field was based on  
351 simplified decision rules that are uncertain at the field scale but robust at the landscape  
352 scale. Consequently, the indicator could be used only to quantify cumulative use of  
353 reservoirs and not use of each reservoir separately. Other uncertainties arise from a lack of  
354 knowledge. The indicator “semi-natural elements contributing to water purification” is a  
355 good example. It was aggregated at the landscape scale because we simply do not know  
356 enough about water purification processes. It is incorrect to assume that the degree of water  
357 purification increases linearly as the area of forest or grassland increases. Thus, aggregating  
358 this indicator at the landscape scale seems a more accurate approach than not doing so.
- 359 - *prompting expression of social rationality*. The indicator “match between water storage  
360 capacity and irrigation needs” (“efficiency” criterion) seeks to reflect how farmers  
361 collectively manage the available infrastructure. At a disaggregated scale, it would indicate  
362 whether individual farmers use their reservoirs efficiently or not. This is why complete  
363 aggregation was preferred.

364

#### 365 **4.2 Reasons underlying user-led aggregation**

366 The 115 horizontally aggregated indicators can be split in two groups after customization:

- 367 - 71 indicators that already exist at the scale of the evaluation and therefore do not require  
368 users to aggregate. Most are at the landscape scale (to assess/evaluate the landscape), but  
369 some are also at smaller scales and can be used to evaluate/assess both the landscape and  
370 sub-landscapes.

371 - 44 indicators require users to aggregate. Indicators in this group are at sub-landscape scales  
372 (sometimes also at the landscape scale) but are meant to be used to assess/evaluate only the  
373 entire landscape.

374 Several reasons explain why these 44 indicators should not be completely aggregated during  
375 customization and why users must perform at least some of the aggregation:

376 - *Visualizing variability or distribution patterns*, allowing users to interpret it/them. This  
377 applies mainly to indicators related to the “equity” criterion, such as the indicator “farms  
378 concerned with increase/decrease in water costs”. This variability can be expressed spatially  
379 (e.g., in maps) but also statistically (e.g., in boxplots) to help users interpret it.

380 - *Promoting expression of diverse (and possibly conflicting) interpretations*. Indicator  
381 developers can choose to let users perform the commensurations that the latter judge  
382 relevant or to use only some of the information produced. For instance, the indicator “two-  
383 year flood flow” (“safety” criterion) should be provided for the outlet of each sub-  
384 watershed. Users can consider the indicator value at the most downstream outlet in the  
385 watershed to compare water management options “in general” for the entire watershed, or  
386 consider the value for the river where they consider safety concerns greater.

387 - *Contextualizing information* provided by the indicator. For instance, the indicator “impacts  
388 on the number of recreational water activities” should be provided for each site of activity,  
389 even though this diversity could be expressed more simply by the number of potential  
390 activities. Presenting the indicator in a spatially-explicit way supports users’ learning  
391 capacities (i.e., they can understand better the source of its values) and expression of  
392 expertise (i.e., they can appreciate a value better when they know what it refers to). This  
393 reason often accompanies the previous reason of diverse interpretations.

394 - *Acknowledging a lack of knowledge*. In this case, the indicator is provided in a spatially  
395 disaggregated form, which makes interpretation difficult, nay, impossible, but developers  
396 still convey to users that, despite being weak for landscape-scale assessments, the indicator  
397 matters. A change in its value reflects a change for a specific location or context, but not  
398 enough knowledge is present to infer impact at the landscape scale. For instance, the  
399 indicator “nitrate pressure” should be provided at the elementary watershed scale but not for  
400 the entire landscape. Nutrient flows are complex processes, which makes it difficult to  
401 convert pressure on water quality into a quantified impact. Another example, the indicator  
402 “changes in gross margin generated by each type of agricultural production”, considered a  
403 proxy for employment and wealth, should be provided for each type of production. This  
404 indicator lacked complete spatial aggregation because the influence of gross margin on  
405 employment and wealth remains little understood.

406

## 407 5. Discussion

### 408 5.1 Many reasons underlie aggregation choices

409 Our analysis empirically illustrates how descriptive and normative claims together influence  
410 indicator development and more specifically aggregation choices. We showed that customization,  
411 which modifies the information provided by an indicator and its potential uses, is shaped by  
412 developers' subjectivities. For instance, by providing users with an indicator in a spatially  
413 disaggregated form, developers can decide to widen interpretations of the indicator or can  
414 complicate indicator use by making it difficult to interpret. Similarly, by aggregating, developers  
415 can limit potential uses of an indicator, for instance by prompting the expression of one type of  
416 rationality only, e.g. social rationality over individual interests.

417 Although researchers who study customization of spatial indicators and spatial aggregation  
418 problems acknowledge that these processes are non-neutral and affect indicator uses (Janssen et al.,  
419 2005; Malczewski and Rinner, 2015; Uran and Janssen, 2003; Walz, 2000), they generally justify or  
420 question customization choices in terms of "functionality" (Uran and Janssen, 2003). Accordingly,  
421 customization choices result from a trade-off between ease of use and "scientific soundness". These  
422 choices could be optimized according to customization rules (e.g., a maximum number of colors, a  
423 spatial resolution that allows patterns to be perceived quickly).

424 Among the reasons for spatial aggregation highlighted in our analysis, most of them can be  
425 understood in terms of functionality. These reasons include relevance for the criteria, fit to process  
426 scale, addressing uncertainty, increased intelligibility, contextualization, increasing the visibility of  
427 variability and easily comparing management options. However, the desire to make an indicator  
428 more functional does not always completely underlie them. Developers can also intend for users to  
429 focus on differences within the landscape by highlighting its heterogeneity or on differences among  
430 management options by providing a single quantitative value for each option. Furthermore, other  
431 reasons escape the logic of functionality (i.e. increasing indicator's ease of use and ability to  
432 describe univocally and accurately a management option for a given criterion), such as promoting  
433 expression of diverse interpretations or prompting social rationality. The former reflects a desire to  
434 yield multiple indicator uses rather than the only "right" use. The latter reason shows a desire to  
435 promote specific indicator uses agreeing with developers' view of the problem.

436 As a result, when indicators are expected to foster deliberation among multiple actors, spatial  
437 aggregation cannot be viewed only through the lens of functionality. Functionality primarily makes  
438 an indicator accessible to those with different skills in a way that is consistent with the  
439 system/problem described. It therefore agrees with a descriptive view of indicators as tools to  
440 measure progress towards sustainability. However, when confrontation of value systems is

441 considered another role of indicators (Bell and Morse, 2004; Frame and O'Connor, 2011),  
442 customization can influence the degree of leeway available for interpreting indicators.  
443 Consequently, customization is one way to increase and/or decrease the number of potential value  
444 judgments.

445

## 446 **5.2 What should drive aggregation choices**

447 By making customization choices, indicator developers influence the size of the “interpretation  
448 space” left to users. Developers tend to decrease the space for what they consider indicator  
449 “misuses” and to increase the space for what they consider “legitimate uses” (but they can also  
450 consider that providing a large amount of space is intrinsically good). The issue then arises of  
451 whether certain aggregation choices are more legitimate than others. In other words, what should  
452 guide aggregation choices besides functionality? We provide two responses that are related to the  
453 collective nature of the problem and the commensuration issue inherent in all information  
454 aggregation processes.

455 Environmental appraisal methods can be considered value-articulating institutions (as defined by  
456 Vatn (2009)) because they address problems that are collective in nature and that promote  
457 expression of different types of rationality. For instance, for managing common goods, methods that  
458 promote expression of social rationality rather than the confrontation of individual interests should  
459 prevail (e.g., the forum rather than the market) (Vatn, 2009). Fine-resolution, spatially-explicit  
460 indicators can encourage expression of individual interests in certain cases since they can, for  
461 instance, allow users to consider impacts on an individual’s private property. Customization can be  
462 a useful way to switch from individual to social rationality. For instance, partial aggregation could  
463 blur individual information or show values in a boxplot to transform spatially-explicit heterogeneity  
464 into statistical heterogeneity.

465 However, aggregation, whether spatial or not, implies commensuration, i.e. comparing different  
466 objects according to a single measurement unit (Espeland and Stevens, 1998; Martinez-Alier et al.,  
467 1998). In spatial aggregation, spatial entities with different values for an indicator are summed or  
468 averaged into a single value for the entire area considered. We posit that indicator developers should  
469 first address the commensuration issue by considering the extent to which different spatial units are  
470 commensurable. It may not be appropriate to make performances of farms on alluvial plains  
471 commensurate with those of farms in the hills. Some may advocate that performances can be  
472 calculated relative to a location-based potential and that it is acceptable to provide an aggregated  
473 value. In contexts with multiple actors, however, the relevance of spatial aggregation as a  
474 commensuration process should be framed in normative terms (i.e., whether aggregation helps or  
475 hinders confrontation of value judgments). If we consider indicators as “channels for bridging

476 realities and meanings” (Abbot and Guijt, 1998), this indicator would probably be more relevant in  
477 a spatially disaggregated form, even though it would make comparisons among management  
478 options more complex.

479 Other experiences with the uses of maps (which are by essence spatially-disaggregated  
480 representations) in governance contexts support our argument that the commensurability of  
481 viewpoints should drive spatial aggregation choices. The use of maps as deliberation support tools  
482 (Caron and Cheylan, 2005; Lardon and Piveteau, 2005; Rinner, 2006) show that they can mediate  
483 the expression and confrontation of different viewpoints. Similarly, the emergence of participatory  
484 or collaborative mapping (Goosen et al., 2007; Jankowski, 2009) strengthens the claim that  
485 spatialization is a powerful tool for empowering people, even those who are marginalized. Finally,  
486 studies in the field of landscape aesthetics have shown that although landscape metrics are  
487 correlated with visual preferences of the landscape, stakeholder groups do not value the same  
488 aspects of the landscape (Dramstad et al., 2006; Howley, 2011), which argues in favor of using  
489 landscape visualizations when users are not known beforehand. These studies reinforce the idea that  
490 spatial representations and other disaggregated forms can help make evaluations more deliberative,  
491 under certain conditions. This claim differs greatly from the use of horizontally and vertically  
492 aggregated forms (i.e., numbers) in most decision-making tools, including sustainability  
493 assessments.

494 When considering the normative role of indicators, aggregation choices can increase the ability of  
495 users to interpret indicators according to their values and knowledge while moving beyond  
496 individual interests. More generally, although identifying indicator sets is a necessary information  
497 compression (Giampietro et al., 2006), indicator customization can help to reintroduce plurality and  
498 social rationality into the debate. Finally, there is no contradiction between the normative role and  
499 descriptive role of indicators, but there is a tension between the need for consistency (e.g., in the  
500 final assessment/evaluation scale(s), in the “common” nature of the problem) and the need to  
501 represent diversity (e.g., spatial heterogeneity, plurality of individual values).

502 Therefore, when developing indicators, we recommend developers to investigate the following  
503 questions:

- 504 1. What are the scales that are consistent with the processes at stake, the assessment  
505 criteria, the level of commensurability between the different spatial units, the nature of  
506 the problem tackled and the model used? Answering this “consistency” question bounds  
507 the possibilities for aggregation choices. If more than one answer is possible, then other  
508 questions arise about the representation of diversity.



- 509 2. Would it enrich the debate or the analysis to visualize heterogeneity or contrasts? (if so,  
510 prefer disaggregated forms ; if not, prefer the most aggregated forms, which are  
511 generally more easy to handle)
- 512 3. Which forms would help users with the elaboration of their value judgment and make  
513 comparisons possible? Between this question and the previous one, a tension might  
514 exist. In such cases, developing two indicators, one in an aggregated form and one in a  
515 disaggregated form, could allow their concurrent use and stimulate discussions among  
516 users and social learning<sup>2</sup>.
- 517 4. If a value judgment appears difficult to access, is the information provided still useful, at  
518 least to understand the limitations of the indicator? (if not, the indicator should probably  
519 be discarded).

520

## 521 6. Conclusion

522 Using indicators to assess sustainability within a governance context requires producing them  
523 without knowing who will use them and how. This context also entails considering the ability of  
524 indicators to foster the expression and confrontation of multiple viewpoints (i.e., their normative  
525 quality) and not only their ability to accurately describe situations or processes (i.e., their  
526 descriptive quality). Involvement of multiple actors and iterations between the steps of  
527 identification, development, and use of indicators is therefore preferred. It is not always possible to  
528 perform iterations, however, for instance due to a lack of time, data or knowledge (from developers  
529 or potential users) or to the cost of developing certain indicators. In these cases, indicator  
530 developers have greater leeway in shaping indicators and therefore the descriptions and judgments  
531 the indicators will generate.

532 Indicator development, a stepwise process of coding information, is rarely described in the  
533 literature on sustainability assessment. We clarified the processing steps required to develop  
534 indicators: formalization, estimation, and customization. We used a case study on water  
535 management to investigate spatial aggregation, which is a specific customization process common  
536 to most landscape-scale or regional sustainability assessments.

537 Spatial aggregation, the process of changing fine-resolution data into coarser-resolution data to  
538 derive meaningful information, partly has a descriptive function. Accordingly, choices are made to  
539 increase indicators' ease of use, which generally leads towards aggregated forms (ultimately a

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11 2 Although in this article we do not report on the use of indicators but on their development, our experience showed how  
12 stimulating it can be to provide stakeholders with two indicators resulting from different aggregation choices. For  
13 example, the two indicators presented in Fig. 2, which are based on the same raw data, were used in evaluation workshops  
14 with stakeholders. Depending on the indicator used, the participants did not formulate the same value judgments and not  
15 with the same facility (it was easier with complete aggregation). Also, the diversity of value judgments among  
16 stakeholders was much higher and discussions more intensive with the partially aggregated indicator.

540 single value), and scientific soundness, which may lead towards aggregation or not. However,  
541 normative considerations can disrupt aggregation choices. These normative claims argue mainly for  
542 disaggregated forms of indicators because aggregation then becomes “user-led”, promoting  
543 expression of social incommensurabilities. Indicators can be aggregated, however, to limit  
544 expression of users’ individual interests.

545 When descriptive and normative claims are considered together, aggregation choices become  
546 more complex and leave more room for developer subjectivity. Completely aggregated forms  
547 (single values) do not guarantee that an indicator will be “good for describing” and “good for  
548 debating”. Leaving indicators in disaggregated forms, however, is not necessarily the best solution  
549 either, even for sustainability problems with a structuring spatial dimension. Spatial aggregation  
550 choices illustrate tension between the need for consistency and the need to represent diversity. In  
551 this respect, the spectrum of indicator forms (e.g., differing degrees of aggregation, different  
552 shapes) merits attention in the practice and theory of sustainability assessments.

553

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560

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