

1 **Brief communication: Lessons Learned and Experiences Gained**
2 **from Building Up a Global Survey on Societal Resilience to**
3 **Changing Droughts**

4
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18
19 **Abstract**

20
21 This paper describes the process of creating a global survey of experts to evaluate drought
22 resilience indicators. The lessons learned include five main points: (1) the heterogeneity of
23 the conceptual background should be minimized prior to the construction of the survey; (2)
24 large numbers of indicators decrease the engagement of respondents through the survey,
25 ways to apportion indicators whilst maintaining reliability should be considered; (3) it is
26 necessary to design the survey to balance response rate and accuracy, (4) the survey
27 questions should have clear statements with a logical and flowing structure, (5) reaching
28 experts from different domain experience and regional representation is difficult, but crucial
29 to minimize biased results.

30
31 **Keywords:** drought resilience, indicators, expert elicitation, global survey

32 1 Introduction

33 The formulation of a global survey is a complex process that poses several challenges
34 both in its preparation (*a priori*), and evaluation of results (*a posteriori*) phases. In general,
35 studies focusing on surveys and expert elicitation address *a posteriori* challenges, such as
36 the data analysis tools used for samples of different sizes and compositions. However, *a*
37 *priori* challenges are rarely addressed and represent an important and defining step in the
38 process. For example, Baker et al. (2014) state that “while there is a rich literature on expert
39 elicitation approaches and protocols, there is less information available on the specifics of
40 how an elicitation is carried out”.

41 Harzing et al. (2013) have reviewed the issues faced in global surveys and identified
42 cultural and language differences, which may lead to different interpretations of questions
43 or loss of meaning, and varying response rates between countries as significant sources of
44 bias in global surveys. ProductLab (2023) also discusses the difficulties of global surveys
45 and provides best practices for their formulation. They also mention the challenges due to
46 cultural and language differences and finally recommend appropriate survey timing for all
47 countries. However, both studies focus on business and product development.

48 Therefore, our main motivation for writing this brief communication is due to the
49 scarcity of papers or other materials discussing the challenges of creating global surveys in
50 complex subjects where we face conceptual and definitional divergences - such as resilience.
51 We believe that the challenges and problems faced during the survey-building process are
52 often not discussed by the researchers, as doing so may weaken confidence in their final
53 results. However, it is important to face this fear and openly share difficulties encountered,
54 as this sharing of challenges can also lead to valuable new knowledge and insights gained.

55 In this study, we used a global survey to elicit experts' opinions on drought resilience
56 indicators. These indicators have been increasingly used in Decision Support Systems (DSS)
57 to reflect different socioeconomic, ecological, and technological conditions (WMO & GWP,
58 2016; Meza et al., 2019; Blauhut, 2020). Although numerous indicators for drought
59 resilience are found in the literature, certain aspects may make them unfeasible for
60 comparative analysis across global regions (Bachmair et al., 2016; Blauhut, 2020). The
61 absence of spatial and temporal data, variability of measurements in different regions, and
62 difficulty in understanding indicators can make it hard to select indicators to compose a
63 global drought resilience index (Blauhut, 2020). However, these aspects are usually
64 overlooked when rating the relevance of the indicators during surveys. For example, Meza
65 et al. (2019) have not incorporated these aspects in their global expert survey on drought
66 vulnerability indicators. Therefore, there is a need for a more in-depth analysis of the drought
67 resilience indicators to ensure their suitability for cross-regional comparisons.

68 Our focus was on agricultural drought resilience linked to systems of small farmers
69 for food production. By following the Sendai Framework for Disaster Risk Reduction
70 (DRR) 2015-2030 (UNDDR, 2015), we listed and screened indicators proposed in the
71 scientific literature for drought resilience related to food systems. The initial screening of
72 indicators provided the basis for the expert global survey to assess the relevance, the data
73 availability, and the shareholders' perception and understanding of these indicators in
74 different contexts.

75 Constructing the survey took about a year due to the challenges as presented in this
76 brief communication. We believe that it is important to discuss the process of formulating
77 the survey to prevent other researchers from encountering the same problems and improving
78 the use and interpretation of this method. The importance of preparing a global expert survey

79 for any generic field has also been discussed by Elangovan and Sundaravel (2021). We hope
80 to complement studies and suggestions for works in the resilience field.

81 **2 Methods for eliciting expert views and knowledge**

82 Mukherjee et al. (2017) identify six strategies that are best suited to the various stages
83 of the decision-making process and for eliciting different judgments: Interviews, Focus
84 Group Discussions (FGD), Nominal Group Techniques (NGT), Q methodology (Q), Delphi
85 technique, and Multi-criteria Decision Analysis (MCDA). An interview consists of an
86 information exchange between two or more individuals in which one of them aims to obtain
87 information, opinions, or beliefs from the other person. The FGD is a technique in which a
88 researcher gathers a group of people to discuss a given issue. Aside from the FGD, which
89 aims to draw on the participants' complex personal experiences, actions, beliefs, perceptions,
90 and attitudes, the NGT is an interactive group decision-making process primarily focused
91 on reaching a consensus. The Delphi technique is traditionally aimed at reaching consensus
92 through a group-based, anonymous, and iterative technique. The Q, on the other hand, is a
93 tool for understanding the primary viewpoints or opinions on an issue among a group of
94 significant players, in which respondents are asked to rank a set of items. Finally, the MCDA
95 assists decision-making by considering the benefits and disadvantages of several
96 possibilities for achieving a specific objective.

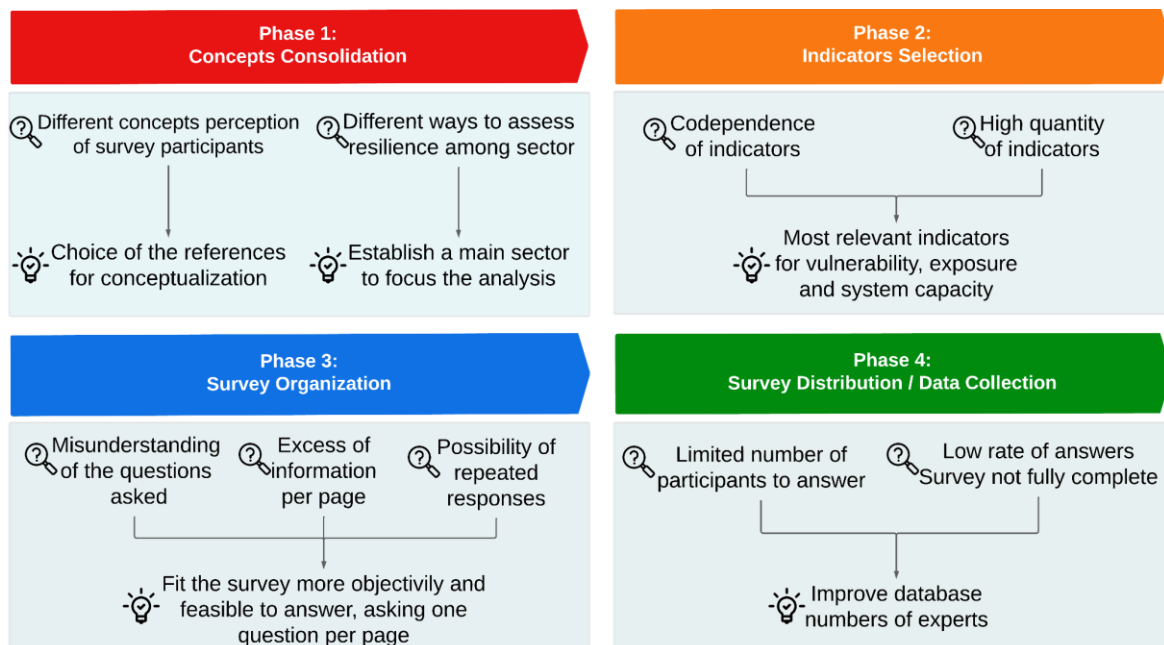
97 Each methodological approach has advantages and disadvantages. The interview, for
98 example, may be challenging to perform due to geographical proximity to the desired sample
99 group (Mukherjee et al., 2017). Another example of a challenge is that FGD is dependent
100 on participant engagement, giving researchers less control. There may be time restrictions
101 for the Q and NGT due to participant interpretation difficulties and insufficient time to reach
102 a consensus.

103 We chose the Delphi technique because it is a tool that can gather and assimilate a set
104 of experts' opinions across geographically diverse time zones on potentially complex
105 matters. The Delphi method has been applied to develop indices for desertification (Hai et
106 al., 2016) and water supply (Crisping et al., 2022), and previously used in global surveys
107 (Rastandeh et al., 2018). Nonetheless, the process of developing and conducting a global
108 survey based on the Delphi method at a global scale is not well documented for users and
109 requires further discussion.

110 **3 Challenges in the Survey Planning**

111 The elaboration and consolidation process of the global survey was carried out in four
112 main phases: conceptualization (concept consolidation), indicators' selection, survey layout
113 organization, and distribution/data collection in survey execution (Figure 1). This section
114 discusses the challenges encountered in each phase and how the research team addressed
115 them using a collaborative approach. The four phases lasted 11 months, being the most time-
116 consuming part of the research so far. Additionally, it was a crucial part of the research since
117 the quality of the outcomes depended on the questions and the engagement of the responders.

118 Figure 1 – The phases of global survey elaboration and main steps



119

120 3.1 Phase 1: Concepts consolidation

121 The first challenge was related to the consolidation of the concepts frequently
 122 associated with drought resilience. We targeted experts from different fields, such as
 123 geophysics, engineering, economics, and social sciences who work and live in different
 124 countries. Thus, the concepts used in the Sendai Framework, such as drought, DRR,
 125 resilience, vulnerability, system capacity, and adaptation can be analyzed and perceived
 126 differently among participants.

127 Initially, we planned to ask the experts to classify the selected indicators into
 128 vulnerability or system capacity types based on the component in which they had the highest
 129 representation. However, due to the heterogeneity of expertise, backgrounds, and contexts,
 130 we realized that leaving the classifications open for a later consolidation would only
 131 propagate conceptual confusion instead of solving it. These conceptual divergences make it
 132 difficult to categorize the indicators in the resilience components and may affect the
 133 perception of their relevance to the respondent. This task was difficult even for the research
 134 group itself, which included researchers from different backgrounds and countries.

135 Therefore, we realized the importance of having a clearly defined *a priori* resilience model
136 to reduce conceptual confusion. For this purpose, we decided to adopt the Sendai Framework
137 (UNDRR, 2015), due to its global significance in developing public policies.

138 The goal of disaster risk management is to increase and strengthen resilience. The
139 UNDDR (2015) defines resilience as “the ability of a system or community to anticipate,
140 resist, prepare, respond to and recover from an event with multiple risks, with the least
141 possible harm to social, economic, and environmental well-being”. Several indices have
142 been proposed over the years to represent the level of resilience of a given system to a
143 disruptive event. In general, resilience assessment requires the identification of the risks in
144 the system due to disruptive events and the adoption of risk management policies to prevent
145 their occurrence or reduce their impacts along the system's chain, therefore it can be
146 represented by a function between risk and risk management (Eq. 1).

147 The risk can be represented by a function that correlates the probability of occurrence
148 of the disruptive event (H), the vulnerability of the system’s different components (V), and
149 their exposure to risk (E), so that vulnerability and exposure represent the potential impacts
150 on the system (Merz et al., 2014) (Eq. 2). Within the disaster risk management and risk-
151 oriented decision-making approach, the risk analysis stage is of fundamental importance and
152 a precursor to the decision-making process.

153 To evaluate the risk management stage, it is important to understand the type of the
154 proposed risk mitigation action, its temporal component, and the magnitude of the impacts
155 if the proposed action fails. According to these components, the actions can be correlated
156 with the different system capacities that help reduce the disaster risk and further impacts,
157 improving resilience, such as adaptive capacity (AC), coping capacity (CC), and
158 transformative capacity (TC) (Eq. 3).

159 $Resilience = f(Risk, Risk\ management) = Risk / Risk\ management$ (Eq.1)

160 $Risk = f(H, E, V) = H.E.V$ (Eq. 2)

161 $Risk\ management = f(system\ capacities) = \sum_{i=AC}^{CC} system\ capacities$ (Eq. 3)

162 3.2 Phase 2: Indicators selection

163 Droughts can have significant impacts on different economic and social sectors, and
164 likewise economic and social features will impact how drought is experienced. However,
165 assessing the drought resilience of each sector can be different. Initially, we focused on
166 agriculture, but we realized that vulnerability and system capacity to droughts can vary
167 significantly within this sector. Small farms produce a significant part of the world's food
168 production (Lowder et al., 2021), and they are more susceptible to climate change and
169 extreme events than commercial farms (Morton, 2007). Therefore, we prioritized the
170 selection of indicators related to small farms' drought system capacity and vulnerability. We
171 observed that prioritizing indicators specific to small farmers' drought system capacity and
172 vulnerability allows for tailored insights and interventions to address their unique needs.
173 However, such a specificity comes at the cost of broader applicability and requires more
174 intensive data collection and analysis. These observations highlighted a trade-off between
175 the targeted application effectiveness and the generalizability of a risk management index,
176 which is overlooked in the literature.

177 The list of indicators to be evaluated in the global survey was compiled from a
178 structured literature review. At the beginning of the process, we identified over 136
179 indicators that are frequently used in literature (Supplementary Material 1). We observed
180 that indicators related to the hazard component of the agricultural drought risk were already
181 well established and could also be easily obtained from global open databases, or even

182 remote sensing satellite data, through geoprocessing. For example, the Global Drought
183 Observatory¹ already monitors hazard indicators globally. Therefore, our focus on this
184 survey was to identify indicators related to risk impacts (vulnerability and exposure) and
185 risk management actions to increase resilience (adaptive, coping, and transformative
186 capacity).

187 There are a myriad of indicators for evaluating drought and its impact on agriculture.
188 Two issues were raised from this initial list: (1) There were too many correlated indicators
189 (e.g., Gini index and poverty rate). Including the codependent indicators would affect the
190 final index by unintentionally attributing a higher weight to this factor; (2) Including all the
191 136 indicators, the survey would become too extensive and exhaustive, which could affect
192 the response rate.

193 Therefore, narrowing the selection of the final list of indicators was made through
194 three steps. The first step was to remove hazard indicators, as previously discussed. In this
195 step, 31 indicators were removed. In the second step, we removed codependent indicators
196 from the list, keeping the ones with more availability and easy-to-access data. For example,
197 from the Gini index and poverty rate, we opted for the poverty rate, since it is a more direct
198 measurement and easier to get in different contexts. This process of eliminating codependent
199 indicators was made interactively in group discussion sessions with the members of our
200 research team. A total of 28 indicators were removed from consideration through this
201 process. The third step was reducing the total number of indicators to avoid the survey
202 becoming too extensive and exhaustive to answer. In this stage, each participant of the group
203 independently rated the relevance of the indicators, through a form available only for the

¹<https://edo.jrc.ec.europa.eu/gdo/php/index.php?id=2000>.

204 group, based on the seven questions given by WMO & GWP (2016). After a group
205 discussion, we selected 33 indicators based on these independent ratings.

206 In the next stage, we sought independent expert opinions concerning the indicators
207 chosen and the overall structure of the survey. External experts recommended three
208 additional indicators after the first pilot run of the survey. In the end, we had a list of 36
209 indicators (Table 1).

210 Additionally, during our internal group discussion sessions, one of the concerns was
211 that some indicators are very interesting and relevant, but they are challenging to obtain. In
212 this sense, we identified important complementary questions on data quality that are usually
213 not asked in the surveys (where all the relevant data are assumed to be equally accessible to
214 obtain and understand). We asked the experts to rate the usability of indicators in terms of:
215 relevancy, ease of understanding, accessibility, and objectivity (we included a definition of
216 each one at the beginning of the formulary).

217 The choice of these specific metrics came from Sweya et al. (2021) which identified
218 five essential attributes for the social resilience indicators of water supply systems:
219 affordability, availability, reliability, simplicity, and transparency. They found that data
220 availability, reliability, and affordability were the most limiting factors when selecting
221 indicators in Tanzania. In this sense, as the project focus was the Global South, our group
222 selected the three metrics adapted from Sweya et al. (2021) to be complementary to the
223 relevancy: (1) understanding – it was used to represent the simplicity; (2) accessibility – it
224 was used as a single attribute to account for affordability and availability; and (3) objectivity
225 - it was an additional attribute that we chose to evaluate how objective is the final measure
226 (since some of our social indicators are political measurements and may be subjective).

227 Table 1. List of indicators evaluated in the survey

Indicator*	Description
1. Agriculture income dependence	Percentage of participation of crop and livestock production in the income of smallholder farming
2. Crop loss	Crop Damage & Sensitivity (Crop Loss)
3. Drought resistant crops	Cultivation of drought-resistant crops (%)
4. Crop varieties	Farmers use different crop varieties (%)
5. Protected area	Area protected and designated for the conservation of biodiversity (%)
6. Use of agricultural inputs	Use of Insecticides and pesticides (Use of agricultural inputs)
7. WUE	Crop water use efficiency (WUE)
8. Land degradation	Degree of land degradation and desertification*
9. Land rights	Land rights clearly defined (yes/no)
10. Drought management policies	Existence of drought management policies
11. Technical assistance	Technical assistance from local entities
12. Drought insurance	Farmers with crop, livestock or drought insurance (%)
13. Water use rights	Water use rights are clearly defined
14. Prediction system	Availability of drought prediction and warning systems or climatic predictions
15. Transportation network	Transportation network
16. Electricity	Access to electricity (Access to energy)
17. Conflict	Prevalence of conflict/insecurity
18. Sanitation condition	Population without access to (improved) sanitation (%)
19. Gender inequality	Gender inequality (categorical)
20. Rural population	Rural population (% of the total population)
21. Unemployment	Unemployment rate (and/or proportion of formal work)
22. Working-age population	Population ages 15-64 (% of the total population)
23. Displaced population	Percentage of the population displaced internally or transboundary
24. Drivers of migration	Presence of drivers of migration and displacement
25. Poverty	Poverty Rate
26. Food source reliability	Food source reliability and diversity
27. Participation in local policy	Public participation in local policy
28. Cooperatives or associations	Participation in farming cooperatives or associations
29. Employment in small farms	% of the population employed in small farms
30. Financing and credit	Access to financing and credit

	Indicator*	Description
31.	Water stress	Baseline water stress (ratio of withdrawals to renewable supply)
32.	Water quality	Water quality (categorical)
33.	Groundwater level	Groundwater level/sources
34.	Integrated policies	Integrated land and water management policies
35.	Retained renewable water	Percentage of retained renewable water
36.	Dam capacity	Total dam capacity

*The reference to each indicator is provided in Supplementary Material 1

228

229 3.3 Phase 3: Survey Organization

230 Another challenge was presenting the indicators and relevant information effectively
231 in an online survey instrument to make viewing, understanding, and comparing the
232 indicators as straightforward as possible. The survey design was made based on guidelines
233 for operationalizing the Delphi method (Hasson et al., 2000) and the suggestions made by
234 Elangovan and Sundaravel (2021). The last provided a template to validate the survey
235 instrument. However, they present a generic document, in which we still experienced
236 difficulties related to the resilience field study. Therefore, we have improved our survey
237 design based on the evaluation of different literature that used the Delphi method to access
238 resilience indicators (e.g., Alshehri et al., 2015; Ogah et al., 2021).

239 During the process of identifying the best layout, we tested different survey question
240 designs. We created several prototype surveys that varied in terms of question layout, types
241 of questions (such as Likert scales versus ranking), number of scales, and how the definition
242 of concepts was presented. To evaluate each prototype, we considered the ease of
243 understanding, cognitive load, and the time required to complete the survey. These survey
244 prototypes were modified and combined based on the user experience. After the first
245 consolidation of the survey design to be used, a pilot pre-test was carried out with a small

246 external group of experts who were asked for their opinions on the final design and
247 indicators. We used the same process to design the second stage of the survey, using the
248 Delphi method.

249 In the final selected design, each page of the survey refers to one specific attribute
250 and rates of importance that should be given to each indicator. This format was chosen
251 because it allows a comparison between the indicators when answering, reducing the
252 possibility of repeated responses for all indicators, and allowing a hierarchy between them
253 and greater fluidity in conducting the survey.

254 Each indicator could be rated on a three-point scale: “Low”, “Medium”, and “High”.
255 The definition of this point scale changes according to the metric that is being evaluated.
256 The category “Don’t know” was included to filter pseudo-opinions. On the last page of the
257 survey, we asked for some demographic information, like area of expertise, years of
258 experience, region of analysis, etc. The final format of the survey (Supplementary Material
259 2) was consolidated after all members of the group and the piloting phase group answered
260 the survey and did not provide any new inputs or suggestions. For the second stage of the
261 survey, we used the same layout, but we included the percentage of the first-phase
262 responders at each level of the scale for each indicator and each metric.

263 3.4 Phase 4: Survey Distribution/Data Collection

264 The last challenge was defining the experts to whom the survey should be sent. As
265 the purpose was to obtain the opinions of experts from different backgrounds and socio-
266 economic contexts, a list of experts was created from recently published papers on droughts
267 in the Web of Science and Scopus databases. The members of the group also shared the
268 survey in their networks.

269 As a result of the disproportionate amount of research conducted in countries and
270 regions in the Global North due to economic factors, scientific databases have a bias toward
271 the Global North, in terms of institutional affiliation. Therefore, it is important to address
272 and remedy this issue in the recruitment process. After this initial data collection, a
273 distribution analysis was carried out about continents and countries to assess whether there
274 was a need to complement any specific region.

275 Despite the attempts to assemble the greatest diversity of experts' backgrounds on
276 drought resilience analysis, the study had a limitation in that it had a large concentration of
277 responses coming from academic experts (approximately 80%). This was due to the
278 difficulty in accessing the information of other practitioners and stakeholders, since there is
279 no unified database, as is the case with Scopus and Web of Science for researchers. For
280 future surveys, we recommend trying to reach out to existing policy and practitioner
281 networks around drought to reach other types of stakeholders.

282 The survey was approved by the Institutional Review Board (IRB) of Penn State
283 University for Human Subjects Protection (IRB # STUDY00021208), and a consent form
284 was provided to all the participants before starting the survey. We customized the research
285 consent form to align with the legal and ethical standards of the participant's country as much
286 as possible. For example, the survey presented a different consent form that accurately
287 reflects the customized considerations of the European Union. After the survey concluded,
288 we received responses from 326 experts from 46 countries, with 120 complete responses.
289 The data obtained from the survey and their *a posteriori* analysis are presented in Sass et al.
290 (2023). For the second stage of the survey (as required by the Delphi method), we obtained
291 32 respondents from 21 countries.

292 **4 Lessons Learned and Recommendations**

293 In this study, a great effort was made to understand how to equalize regional issues
 294 during the construction of a global survey aiming at identifying indicators to compose a
 295 global index to evaluate resilience to agricultural droughts in the context of small farms for
 296 food production. The challenges encountered *a priori* in the application of the method (e.g.,
 297 construction of questions and engagement of participants in the process) are not explained
 298 and discussed in length in the academic literature despite being crucial for the quality of the
 299 data obtained. In Table 2, we summarize our processes for designing such a survey, highlight
 300 the main challenges, and present suggestions for working around them.

301 Table 2. Summary of challenges, lessons and suggestions found on building a global
 302 survey

Survey phase	Challenges	Lessons learned	Suggestion
Phase 1 – Concepts consolidation	Resilience is a slippery concept. Conceptual divergence between expertise, backgrounds, context, and frameworks.	Need to consolidate the resilience concepts and framework used before starting the survey construction.	Define an a priori resilience model to reduce conceptual confusion.
	Consolidation of concepts.	Do not ask the respondents to classify the indicators into the resilience components. This would only propagate conceptual confusion, instead of solving it.	Define the main concepts of your survey
Phase 2 – Indicator selection	High number of resilience indicators in literature.	Hazard indicators are well- established and well-assessed.	Narrow down the list of indicators according to the purpose of the study. Use at most 40 indicators.
	Too many indicators make the survey too extensive and exhaustive, which affects the response rate, including the number of respondents who start the survey but do not complete it.	Many codependent indicators. Some indicators have a high relevance rate, but they are not easy to obtain or are not objective or easy to understand, which may affect their final use as a global indicator.	Remove hazard or secondary indicators, and remove codependent indicators (remaining with the easiest to access and direct measurement). Perform a first assessment of indicators by the internal group and select the most relevant. Use the pilot phase to validate chosen indicators by external experts.

Survey phase	Challenges	Lessons learned	Suggestion
			Include qualitative metrics besides relevance: ease of understanding, accessibility, and objectivity.
Phase 3 – Survey organization	Presenting the indicators and all relevant information effectively in an online instrument.	It is easier to compare indicators when they are presented all together. When the indicators are presented on separate pages, the respondents lose a sense of comparison, and they can provide the same ratings to all of them (usually as "High").	Use a three-point scale: "Low", "Medium", and "High" and include "Don't know" to filter pseudo-opinions. Each metric should be questioned on each page, presenting all the indicators to be rated to allow comparison between them.
		More than a three-point scale can cause confusion in responses.	The completion of the survey should not exceed 15 minutes, to prevent a decrease in the response rate to the final questions.
Phase 4 – Survey Distribution/ Data Collection	Defining the experts to whom the survey should be sent.	Bias to Global North representation. Difficult to have access to databases of other shareholders than the academy.	A list of experts can be created from authors of recently published papers in the Web of Science and Scopus databases. Evaluate the geographical coverage of the list and complement the list with specific contacts from underrepresented regions. To reach out to existing policy and practitioner networks around drought to reach other types of stakeholders.

303

304 (1) There are different concepts related to resilience, especially about vulnerability and
305 system capacity, which can be very context-dependent.

306 To deal with this challenge in the construction of a global indicator, we suggest
307 choosing an internationally relevant and well-consolidated resilience framework (in this
308 case, the Sendai Framework due to its relevance in public policies), rigidly adopting the
309 presented settings. Additionally, to account for differences in local contexts, in addition to

310 the relevance of each indicator, we utilized complementary attributes, such as ease of
311 understanding, accessibility, objectivity, and temporal consistency.

312 (2) There are many indicators in the literature. Surveys containing all the indicators become
313 tiresome to answer, decreasing the engagement, response rate, and quality of the answers
314 obtained.

315 In our experience, including more than 40 indicators already significantly reduced
316 engagement and consistency in responses. Thus, the choice of the final and reduced list of
317 indicators should be based on the objective of the research, and the system evaluated, with
318 only the priority indicators being chosen for representativeness in different local contexts of
319 risks.

320 (3) It is important to identify the best survey design that clarifies questions and definitions,
321 to reduce misunderstanding and divergent answers across different contexts (expertise
322 and region-wise).

323 Before making the survey available to the experts and practitioners, it was essential
324 to study its face and conceptual validity by our internal research team and externally by a
325 smaller group of experts during a pilot phase. Face validation refers to whether the
326 participants can interpret the survey items according to their intended meaning. The
327 conceptual validity ensures that survey items accurately represent the theoretical concept
328 that they are intended to represent. These validation processes will help to identify and
329 correct poorly prepared items and ill-defined concepts to ensure the quality of the survey
330 responses. Providing conceptual definitions of the scales can improve the face validity of
331 surveys.

332 (4) The survey design must be clean and flow well between questions

333 The engagement of respondents from the beginning to the end of the survey is of
334 great importance to maintain consistent results for all questions. Therefore, the format of the
335 applied survey is important. The survey should allow quick and explicit comparison between
336 the main components evaluated (in our case, the indicators), and questions about different
337 attributes should be separated into different sections. Response time should preferably be at
338 most 15 min.

339 (5) It can be difficult to list participants from different areas of knowledge, professional
340 experience, and regions/countries. The small number of respondents for each area affects
341 the significance of the analysis *a posteriori*.

342 There is a lack of databases for practitioners and stakeholders other than experts, which
343 makes it difficult to gather names of other actors usually involved in decision-making
344 processes. Suggestions to obtain a more diverse participant base, including public and
345 private sectors and international organizations, include creating their buy-in and support to
346 share the survey with their members and employees. Developing collaborations with
347 international agencies involved in dealing with disasters, especially droughts (e.g. IDMP,
348 UNCCD, WMO, FAO) may help with their engagement and participation in the survey.
349 Moreover, even in academic databases, there is still a great bias for international research to
350 be centered on countries of the Global North, in terms of institutional affiliation. Since the
351 countries of the Global South are generally the ones with the greatest difficulty in coping
352 with the risks of droughts, studies of indicators benefit a lot by taking into account their
353 perspectives.

354 By sharing our experience in the process of constructing a global survey, we hope to
355 help other researchers by pointing out the key difficulties one may encounter and the
356 measures we followed to address them.

357 **5 Conflict of Interest**

358 *The authors declare that the research was conducted in the absence of any commercial or*
359 *financial relationships that could be construed as a potential conflict of interest.*

360 **6 Author Contributions**

361 All authors contributed to the development and execution of the global survey which
362 is part of the Management of Disaster Risk and Societal Resilience (MADIS) project, funded
363 by the Belmont Forum². MM, MB, KS, and AN contributed to this manuscript by writing,
364 reading, and reviewing. AK, NO, EM, GS, PS, and MJ contributed to the manuscript revision
365 and reading.

366 **7 Funding**

367 This study was funded by Grants #2019/23393-4 and #2022/15054-8, São Paulo
368 Research Foundation (FAPESP). The work of the USA authors is sponsored by the National
369 Science Foundation (NSF) Grant (2039506). The work of the UK author is funded by the
370 Engineering and Physical Sciences Research Council (EPSRC, EP/V006592/1, UK). Any
371 opinions and findings expressed in this material are of the authors and do not necessarily
372 reflect the views of the NSF.

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