

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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19
20 **Abstract**

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

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

33 1 Introduction

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

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

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

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

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

78 Due to the challenges presented in this brief communication, constructing the survey
79 took about a year. It is important to discuss the process of formulating the survey study to
80 prevent other researchers from encountering the same problems and improve the use and

81 interpretation of this method. Elangovan and Sundaravel (2021) have also discussed the
82 importance of preparing a global expert survey for any generic field. We hope to
83 complement studies and suggestions for works in the resilience field.

84 **2 Methods for eliciting expert views and knowledge**

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

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

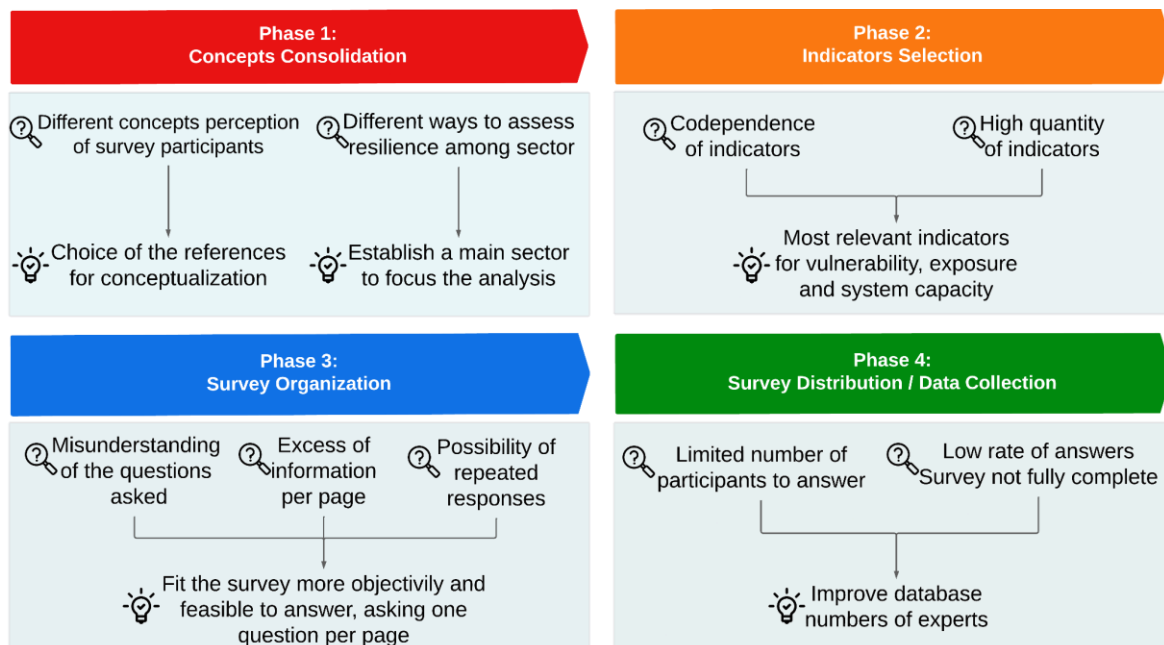
104 for the Q and NGT due to participant interpretation difficulties and insufficient time to reach
105 a consensus.

106 We chose the Delphi technique because it is a tool that can gather and assimilate a set
107 of experts' opinions across geographically diverse time zones on potentially complex
108 matters. The Delphi method has been applied to develop indices for desertification (Hai et
109 al., 2016) and water supply (Crisping et al., 2022), and has been previously used in global
110 surveys (Rastandeh et al., 2018). Nonetheless, the process of developing and conducting a
111 global survey based on the Delphi method at a global scale needs to be better documented
112 for users and requires further discussion.

113 **3 Challenges in the Survey Planning**

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

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



122

123 3.1 Phase 1: Concepts consolidation

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

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

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

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

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

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

162 $Resilience = f(Risk, Risk\ management) = Risk\ management / Risk$ (Eq.1)

163 $Risk = f(H, E, V) = H \times E \times V$ (Eq. 2)

164 $Risk\ management = f(system\ capacities) = AC + CC + TC$ (Eq. 3)

165 3.2 Phase 2: Indicators selection

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

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

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

189 There is a myriad of indicators for evaluating drought and its impact on agriculture.
190 Two issues were raised from this initial list: (1) There were too many correlated indicators
191 (e.g., Gini index and poverty rate). Including the codependent indicators would affect the
192 final index by unintentionally attributing a higher weight to this factor; (2) Including all the
193 136 indicators, the survey would become too extensive and exhaustive, which could affect
194 the response rate.

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

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

206 & GWP (2016). After a group discussion, we selected 33 indicators considering the
207 independent ratings of the research team.

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

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

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

229 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

230

231 3.3 Phase 3: Survey Organization

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

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

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

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

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

265 3.4 Phase 4: Survey Distribution/Data Collection

266 The final challenge involved identifying and recruiting the experts to send the
267 survey. To obtain the opinions of experts from different backgrounds and socio-economic
268 contexts, a list of experts was created from recently published papers on droughts in the Web
269 of Science and Scopus databases. The group members also shared the survey in their
270 networks. As a result of the disproportionate amount of research conducted in countries and
271 regions in the Global North due to economic factors, scientific databases have a bias toward

272 the Global North in terms of institutional affiliation. Therefore, it is important to address and
273 remedy this issue in the recruitment process. After this initial data collection, a distribution
274 analysis was carried out about continents and countries to assess whether there was a need
275 to complement any specific region.

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

283 The survey was approved by the Institutional Review Board (IRB) of Penn State
284 University for Human Subjects Protection (IRB # STUDY00021208), and a consent form
285 was provided to all the participants before starting the survey. We customized the research
286 consent form to align with the legal and ethical standards of the participant's country as much
287 as possible. For example, the survey presented a different consent form that accurately
288 reflects the customized considerations of the European Union. After the survey concluded,
289 we received responses from 326 experts from 46 countries, with 120 complete responses.
290 The data obtained from the survey and their *a posteriori* analysis are presented in Sass et al.
291 (2023). For the second stage of the survey (as required by the Delphi method), we obtained
292 32 respondents from 21 countries.

293 **4 Lessons Learned and Recommendations**

294 In this study, a great effort was made to understand how to equalize regional issues
295 during the construction of an international survey aiming at identifying indicators to

296 compose a global index to evaluate resilience to agricultural droughts in the context of small
 297 farms for food production. The challenges encountered *a priori* in the application of the
 298 method (e.g., construction of questions and engagement of participants in the process) are
 299 not explained and discussed in length in the academic literature despite being crucial for the
 300 quality of the data obtained. In Table 2, we summarize our processes for designing such a
 301 survey, highlight the main challenges, and present suggestions for working around them.

302 Table 2. Summary of challenges, lessons, and recommendations for building a global
 303 survey

| Survey phase | Challenges | Lessons learned | Suggestion |
|--|--|---|--|
| Phase 1 – Concepts consolidation | <ul style="list-style-type: none"> - Resilience is a slippery concept. - Conceptual divergence between expertise, backgrounds, context, and frameworks. | <ul style="list-style-type: none"> - Need to consolidate the resilience concepts and framework used before starting the survey construction. - Do not ask the respondents to classify the indicators into the resilience components. This would only propagate conceptual confusion, instead of solving it. | <ul style="list-style-type: none"> - Define an a priori resilience model to reduce conceptual confusion. - Define the main concepts of your survey. |
| Phase 2 – Indicator selection | <ul style="list-style-type: none"> - High number of resilience indicators in literature. - 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. | <ul style="list-style-type: none"> - Hazard indicators are well-established and well-assessed. - 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. | <ul style="list-style-type: none"> - Narrow down the list of indicators according to the purpose of the study. Use at most 40 indicators. - 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. - Include qualitative metrics besides relevance: ease of understanding, accessibility, and objectivity. |
| Phase 3 – Survey organization | <ul style="list-style-type: none"> - Presenting the indicators and all relevant information effectively in an online instrument. | <ul style="list-style-type: none"> - It is easier to compare indicators when they are presented all together. When the indicators are presented on separate pages, the | <ul style="list-style-type: none"> - Use a three-point scale: “Low”, “Medium”, and “High” and include "Don't know" to filter pseudo-opinions. |

| Survey phase | Challenges | Lessons learned | Suggestion |
|--|---|---|--|
| | | <ul style="list-style-type: none"> respondents lose a sense of comparison, and they can provide the same ratings to all of them (usually as "High"). - More than a three-point scale can confuse responses. | <ul style="list-style-type: none"> - Each metric should be questioned on each page, presenting all the indicators to be rated to allow comparison between them. - 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 | <ul style="list-style-type: none"> - Defining the experts to whom the survey should be sent. | <ul style="list-style-type: none"> - Bias to Global North representation. - Difficult to have access to databases of other shareholders than the academy. | <ul style="list-style-type: none"> - 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. |

304

305 Next, we present and discuss the five main points to be considered when conducting
306 reliable and representative research on a global scale.

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

309 To deal with this challenge in the construction of a global indicator, we suggest
310 choosing an internationally relevant and well-consolidated resilience framework (in this
311 case, the Sendai Framework due to its relevance in public policies), rigidly adopting the
312 presented settings. Additionally, to account for differences in local contexts, in addition to
313 the relevance of each indicator, we utilized complementary attributes, such as ease of
314 understanding, accessibility, objectivity, and temporal consistency.

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

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

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

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

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

335 Respondent engagement from the beginning to the end of the survey is crucial to
336 maintaining consistent results for all questions. Therefore, the format of the applied survey
337 is important. The survey should facilitate a quick and clear comparison of the main

338 components being evaluated—in this case, the indicators. To minimize cognitive load,
339 questions regarding different attributes should be organized into separate sections. Response
340 time should be 15 minutes.

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

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

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

359 **5 Conflict of Interest**

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

362 **6 Author Contributions**

363 All authors contributed to the development and execution of the global survey which
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