

1 **Brief Communication: Inclusiveness in designing an early warning**
2 **system for flood resilience**

3 **Tahmina Yasmin¹, Kieran Khamis¹, Anthony Ross², Subir Sen³, Anita Sharma⁴, Debashish Sen⁴, Sumit**
4 **Sen³, Wouter Buytaert², David M. Hannah¹**

5 ¹School of Geography, Earth & Environmental Sciences, University of Birmingham, Birmingham, UK.

6 ²Department of Civil and Environmental Engineering, Imperial College London, London, UK

7 ³Centre of Excellence in Disaster Mitigation and Management, Indian Institute of Technology Roorkee, India.

8 ⁴People’s Science Institute, Dehradun, India.

9
10 **Correspondence:** Tahmina Yasmin (t.yasmin@bham.ac.uk) and David M. Hannah (d.m.hannah@bham.ac.uk)

11
12
13
14 **Abstract**

15
16 Floods remain a wicked-problem and are becoming more destructive with widespread ecological-social-and-
17 economic impacts. The problem is acute in mountainous-river-catchments where plausible-assumptions of risk-
18 behaviour to flood-exposure-and-vulnerability are crucial. Inclusive approaches are required to design suitable
19 flood early-warning-systems (EWS) with a focus on local social-and-governance context rather technology, as
20 is the case with existing practice. We assess potential approaches for facilitating inclusiveness in designing
21 EWS by integrating diverse-contexts and identifying preconditions and missing-links. We advocate the use of
22 a SMART-approach as a checklist for good-practice to facilitate bottom-up-initiatives that benefit the
23 community-at-risk by engaging them in every stage of the decision-making process.

24
25
26
27
28
29
30
31
32
33
34
35
36
37
38
39
40
41
42
43
44
45
46
47
48
49
50

51 **1 Introduction**

52 The theme for World Meteorological Day 2022 (March 23) was ‘Early Warning and Early Action –
53 Hydrometeorological and Climate Information for Disaster Risk Reduction’ which emphasises the vital
54 importance of information generation and sharing to minimize the risks from hydrometeorological extremes.
55 Further, the United Nations secretary-general announced a major initiative, to be delivered via COP 27 (UN
56 Climate Conference), for ‘everyone on Earth should be protected by early warning systems against extreme
57 weather and climate change within the next five years.’ These policy initiatives indicate the growing need for
58 new information and knowledge relating to risks arising directly from hazard but also from the complex
59 interactions with exposure and vulnerability (IPCC defined $\text{risk} = \text{hazard} \times \text{exposure} \times \text{vulnerability} \div \text{capacity}$
60 to cope, see details in Cardona et al., 2012). Although our understanding of hydrological extremes, such as
61 floods, has evolved in recent decades as we view them through the lens of hydro-complexity (Kosow et al.,
62 2022). However, floods remain a “wicked” problem and are becoming more destructive with ecological, social
63 and economic impacts (i.e., source of water pollution, damages to wastewater and irrigation system, excessive
64 erosion damaging riverbank settlements, see details in Kosow et al., 2022; Hannah et al., 2020). In mountainous
65 regions floods are becoming more unpredictable and destructive in response to increasing climatic extremes.
66 This is exacerbated by anthropogenic pressures which have severely modified formerly pristine, high altitude
67 river catchments. Furthermore, increased encroachment of riverbanks, dumping of solid and sewer waste and
68 rapid urbanisation has increased the proportion of low-income communities living in flood-prone areas (Mao et
69 al., 2018; Paul et al., 2018). The lack of adequate hydrometeorological monitoring networks or early warning
70 system in these regions causes undue damage to lives and property (Mountain-EVO, 2017; Pandeya et al., 2021).
71 Yet prediction of risks associated with floods is difficult to achieve in such data-scarce mountainous regions.

72
73 Indeed, the most recent report of the Intergovernmental Panel on Climate Change (IPCC, 2022) highlighted the
74 urgent need for investment in adaptation and resilience, particularly in developing regions which have been
75 historically underfunded but are already impacted by extreme weather events. A key requirement is to improve
76 early warning alerts of anticipated storms, heatwaves, floods and droughts. To generate such warning
77 information for floods, systematic development of monitoring networks that utilise appropriate technologies are
78 required. These systems should also consider social, cultural and political dimensions to identify context-
79 specific understanding on inequality and its impact on assessing vulnerabilities and exposure, so that the
80 warning system can ensure inclusiveness in responses following appropriate decision-making chains (Mao et
81 al., 2018; Acosta-Coll et al., 2018). Such an integrated and interconnected monitoring system requires science,
82 policy and local community-led approaches that can bring diverse stakeholders (i.e., gender, sex, age, socio-
83 economic status and physical abilities) together and generate knowledge to guide their decision to propose
84 solutions that fit the local context (Buytaert et al., 2018; Kosow et al., 2022; Roque et al., 2021; Zulkafli et al.,
85 2017). Despite this call for an inclusive approach for generating an early warning alert system, the existing flood
86 monitoring practices and designs are strongly technology-driven (i.e., information and communications
87 technology [ICT]) and focus less on converging with the local socio-cultural and governance context (Mao et
88 al., 2018; Westerhoff et al., 2021). There are still questions on how, where and at what level science, policy and
89 society may converge and facilitate bottom-up initiatives for decision-making and develop innovative solutions
90 to address challenges posed by floods.

91 In this commentary, we assess potential approaches for facilitating inclusiveness in the design of a flood early
92 warning system by integrating social, cultural and political aspects, and identify preconditions and missing links.

93 **2 Current approaches embedding inclusiveness in water and disaster research**

94 In water and disaster research several approaches are emerging to provide concepts, tools and framings that can
95 be used to support inclusiveness and disciplinary convergence for actionable knowledge production. The
96 concept of knowledge co-production has emerged from science-society interaction under the umbrella of
97 adaptive governance thinking where polycentric models and power relation received attention (see details in
98 Buytaert et al., 2018; Paul et al., 2018 and Zulkafli et al., 2017). Scholarly research has identified several
99 potential approaches to achieve knowledge co-production under the broader umbrella of the participatory action
100 research (PAR) including participatory modelling (Sterling et al., 2019), community-based participatory

101 approaches (Wallerstein et al., 2017), participatory scenario analysis (Birthisel et al., 2020; Lakhina et al., 2021;
102 Westerhoff et al., 2021), among others. More recently, citizen science has emerged with an emphasis on
103 “knowledge cocreation and co-generation” (i.e. the interactive processes across science, policy and
104 implementation to collaborate and to generate knowledge for supporting environmental decision-making see
105 further details in Buytaert et al., 2018) and new technologies, especially ICT, but limited focus on action and
106 development. In addition, citizen science focuses more on participation by volunteers, developing trust and
107 nurturing existing working relationships among involved actors towards knowledge co-production (Buytaert et
108 al., 2018; Zulkafli et al., 2017).

109
110 In the contemporary disaster research literature, knowledge co-production is advocated along with participatory
111 actions and transdisciplinary research, which laid the foundation for the participatory convergence concept to
112 translate research into practice (Lakhina et al., 2021; Peek et al., 2020; Roque et al., 2021). Peek et al. (2020)
113 define the participatory convergence research as ‘an approach to knowledge production and action that involves
114 diverse teams working together in novel ways—transcending disciplinary and organizational boundaries—to
115 address vexing social, economic, environmental, and technical challenges in an effort to reduce disaster losses
116 and promote collective well-being’ (pp. 2). While this research approach has been identified as one of the best
117 ten big ideas in funding allocation and research direction by the National Science Foundation of USA (2016),
118 there has been little exploration on the framing (i.e., methods and ethics) to apply this in practice (Westerhoff
119 et al., 2021). Indeed, scholars are focusing on more empirical exploration of convergence research to generate
120 ethics and methods that may deliver successful outcomes. For example, research attempting to address coping
121 with water extremes such as floods and droughts (Lakhina et al., 2021, Roque et al., 2021; Westerhoff et al.,
122 2021). Recently scholars have proposed ethics that have proven useful. For example, Lakhina et al., (2021)
123 proposed ‘convergence with CARE: collaboration, accountability, responsiveness and empowerment’ which
124 require community engagement and further highlight their perspective, questions and experiences while
125 disregarding traditional hierarchical approaches. However, much of hydrological research is focused on
126 improving scientific measurements and developing technological solutions. For example, improving model
127 uncertainty or the instruments and networks used to measure different facets of the hydrosphere (Beven et al.,
128 2020) while being useful for advancing the discipline result in solutions that are often difficult to disseminate
129 to local communities (Birthisel et al., 2020; Roque et al., 2021; Westerhoff et al., 2021). Earlier reviews indicate
130 many empirical investigations on how social context, such as culture, politics and economics have shaped water
131 knowledge and how and what interventions influence or shape communities’ respond differently (Roque et al.,
132 2021). This emphasises a need for future research to understand the underlying principles and ethics that would
133 facilitate bottom-up driven activities or active participation of engaged stakeholders for knowledge co-
134 production to responds and reshape convergence research methods.

135 136 **3 Processes and preconditions in early warning system development**

137 A synthesis of the literature on flood early warning systems was reviewed to develop a schematic representation
138 of an idealised framework for developing an inclusive early warning system (Figure 1) (for more details see
139 Acosta-Coll et al., 2018; Buytaert et al., 2018; Mashi et al., 2020; Paul et al., 2018; Zulkafli et al., 2017). The
140 foundation of this schematic representation (Figure 1) is adapted from the concept of knowledge co-generation
141 processes (Buytaert et al., 2018) and co-design framing for environmental decision-making processes in a
142 polycentric system (Zulkafli et al., 2017) and then applied with the key elements (i.e., risk knowledge; technical
143 monitoring and warning service; communication and dissemination of warnings and community response
144 capability (ISDR, 2020) identified by the World Meteorological Organization, International Strategy for
145 Disaster Reduction (ISDR). All these concepts, in general advocated participatory and citizen science approach
146 to become inclusive and generate actionable knowledge (Buytaert et al., 2018; ISDR, 2020; Paul et al., 2018;
147 WMO, 2020). The disaster risk equation provided by the IPCC ($risk = hazard \times exposure \times$
148 $vulnerability \div capacity\ to\ cope$) suggest that reduction in risk is dependent not only on efficient forecasting
149 of hazard, but also on the understanding of associated exposure, vulnerability and capacity to cope by the
150 exposed community. Therefore, in Figure 1, we present three interdependent steps, i.e., collate data on risk
151 generate data and models to facilitate forecasting and disseminate that is necessary to develop a system that not

152 only produce flood alerts, but also provide risks information through monitoring exposure, vulnerability and
153 capacity of the community-at-risk.

154

155 3.1 Mapping the risks through data collection and observation

156 In this step, it is crucial to collect as much information possible, to generate knowledge on the locality and the
157 community at risk to design a purposeful early warning system. The knowledge generated can also inform on
158 exposure, vulnerability and ability to cope if a disaster strikes and enables decision-makers to adjust or adapt
159 necessary precautionary measures to respond efficiently in a timely manner (Buytaert et al., 2018; Pandeya et
160 al., 2020). The required knowledge includes scientific measurements of the hydrological hazard, various context
161 of risks information (i.e., vulnerability and exposure mapping) across the socio, cultural and political domains
162 that contribute to the risk portfolio to be more intense and having long-term consequences (Mao et al., 2018).
163 In general, we found most studies generate information on risk through a baseline survey of exposure and
164 vulnerability analysis vis observation, interviews, focus group discussions, stakeholders' meetings. The data
165 focuses on a variety of aspects including historical analysis, geographical aspects, environmental, social,
166 economic and governance structures. All these are relevant, however, what is missing here is the lens through
167 which it is possible to explore the complexity of the risk portfolio determined through different angles of
168 exposure and vulnerability perceived by different stakeholders. Reaction to risks in terms of exposure and
169 vulnerability are dependent on the social, cultural and political stances of stakeholders, and thus is highly
170 variable (Mashi et al., 2020; Hermans et al., 2022). For instance, the communities that are living in flood
171 vulnerable areas might not have legal rights to do so therefore, they might decide to tolerate that risk due to fear
172 of eviction. Other stakeholders may be from state organisations which are not bound to provide services to this
173 illegal settlement and therefore, will not engage. People might not engage also as they already lost their trust on
174 the governance system (i.e., did not receive compensation for their previous flood damage, recurring failed
175 commitments from the political parties to reduce flood vulnerability). Previous research partly discussed these
176 complexities (e.g., Acosta-Coll et al., 2018; Hermans et al., 2022; Mashi et al., 2020) however, solutions to
177 these challenges are limited.

178

179 [Figure 1]

180

181 The citizen science approach, in such cases, recommend utilising social capital tools, such as building a
182 relationship with trust across stakeholders, identifying the people with leadership qualities or local champions
183 (i.e., community members or a social activist/government/non-government employee who have some form of
184 knowledge of flood risks and keen to learn about the early warning system) (Acosta-Coll et al., 2018; Mashi et
185 al., 2020). Previous research and project experiences in a similar context demonstrated conducting structured
186 dialogue through stakeholders' meetings, focus group discussions and forming of community groups (see
187 further details in Acosta-Coll et al., 2018; Mashi et. al., 2020). However, these interactions can lead to confusion
188 and unrealistic expectation relating to the monitoring system. Therefore, it is crucial to make plausible
189 assumptions of risk behaviour relevant to flood exposure and vulnerability that can feed into designing the early
190 warning system including having more focused conversation with the community at risks, specifying the aim
191 and expected outcome of the flood monitoring system.

192

193 3.2 Forecasting hazard risks and establish an alert system in real time

194 This step utilises information from the previous step to identify design specifications to build the early warning
195 system. For example, suitable sensor technology, identification of relevant variables (i.e., rainfall, water level),
196 suitable location(s) to install the components and transmit/receive data. In addition, decision-making on data
197 collection attributes, such as data transmission frequency, among others is critical because there will always be
198 a trade-off between lead time and the potential for an early warning to facilitate appropriate community
199 responses to reduce the likelihood of life. Thus, an understanding of what the optimal lead time in a certain
200 context should be is crucial. To enable any data processing activity, adequate monitoring of relevant variables
201 must be undertaken at the relevant spatial and temporal resolution or scale. This scale will vary depending on
202 the topographic complexity, landcover, geology and hydrodynamic properties of the catchment of interest

203 (Lauden and Sponseller 2018). If historical data is limited (often the case with mountainous and logistically
204 challenging environments) a period of baseline data collection through the previous step is required to “get to
205 your catchment” before establishing a monitoring network. A range of analytical tools are available, including,
206 statistical modelling and simulation, to provide robust thresholds to trigger alert levels based on the collected
207 data. This forecasting step – i.e., predicting the likelihood of flood based on antecedent conditions - is a
208 challenge in data-scarce regions like the Himalaya where there may be significant uncertainty associated with
209 any alert/alarm thresholds due to insufficient training data (Mountain-EVO, 2017; Pandeya et al., 2019).
210 Therefore, many risk assumptions are involved in this step such as over-promising for a sensor-based alert
211 system and if the forecasts are not accurate, there may be a resentment in the community regarding the project.
212 This raises an important question relating to understanding the local context to get a good understanding on how
213 risk management happens and what this means for the design? Moreover, how and when to involve the
214 community (non-scientists) in the development process? Also, what is the purpose of involving the community
215 and other organisations and how will their involvement shape the design process? All these questions are
216 important for the emerging disaster risk management paradigm, where leading organisations (e.g. World
217 Meteorological Organisation (WMO) and other humanitarian agencies (i.e. International Federation of Red
218 Cross and Red Crescent Societies) are suggesting moving towards impact-based forecasting and anticipatory
219 humanitarian actions so that context specific risks could be identified and necessary relevant action plan could
220 develop on time (please see further details in report link 6).

221
222 Previous research has highlighted the importance of involving relevant state organisations, such as disaster
223 management departments or meteorological organisations at this stage (Acosta-Coll et al., 2018; Pandeya et al.,
224 2019). However, this can potentially lead to a divergence in terms of priorities; scientist and engineers are
225 generally focused on the success of the adopted technique and necessary data generation, while the state-led
226 organisations might focus on bureaucracy, policy, existing government beliefs and long-term operational plans
227 (e.g., maintenance and legacy costs). Therefore, engaging with the state departments at this stage can become
228 very difficult (Mashi et al., 2020), nonetheless from a design perspective, understanding both contexts are very
229 crucial for building a purposeful early warning system. The previous researcher recommended utilising a
230 bridging or boundary organisation that can act as a mediator and bridge the gap (Acosta-Coll et al., 2018; Mashi
231 et. al., 2020). Few projects involved local technological start-up companies or local research and development
232 organisations. However, there is limited exploration on the community engagement at this stage who struggle
233 to visualise such technical details in real-time application. Further, there are also missing on the crucial aspects
234 of what levels of technical details to share and which is the right time/phase to share with the community or the
235 state authority. This inadequate understanding to decide the right time or phase will risk of over-promising for
236 warning alert.

237 238 3.3 Communication and dissemination

239 After installation of the alert system, identification of the best possible modes of dissemination is critical to
240 further interact with the vulnerable communities and communicate the potential risks along with tentative
241 necessary actions to minimise the risks. While this has been the most critical part, it is also one of the most
242 interactive components in the entire scheme. New ICT technologies such as interactive dashboard visualisations,
243 give more flexibility in developing the visualisation to disseminate the EWS outputs in a way that can be easily
244 understood by the community is a major challenge (Mashi et al., 2020; Pandeya et al., 2019). Several questions
245 arise in this step including a strategy to ensure the alert levels reaches to all those who are at risk, the risk
246 information is easy to understand and there is a desired reaction to such information. Previous research
247 highlights different visualisation techniques to showcase alert levels such as text, colour coding, graphics, audio
248 mobile messages, and showcasing locational maps (Acosta-Coll et al., 2018; Pandeya et al., 2019). What may
249 be missing in this step is what would be the best possible methods to communicate with the community at risk
250 and understanding how they perceived and responded to such forms of alerts or warnings? Here, communication
251 not only with the communities but also with the responsible state authorities and how they are supporting or
252 engaged in with the decision-making processes to respond in a timely manner.

253

254 **4 A SMART way forward**

255 We believe that through this commentary we have raised critical questions and identified missing links in the
256 context of disaster resilience and the development of tools to improve preparedness and response. The most
257 important include i) the absence of diverse contextual risk angle and community reactions; ii) a lack of
258 community trust in government agencies and technology focused forecasting; iii) significant data limitations to
259 ensure effective EWS operation and impact-based forecasting; and iv) a lack of effective communication
260 strategies. All these points need deeper exploration to ensure inclusive EWS are developed in data-scarce
261 mountainous regions or geographic regions similar in context. We acknowledge that many countries are
262 currently implementing EWS focusing on active community participation (please see reports links 1-5)
263 however, solutions to address these missing links are limited and thus ensuring inclusiveness and impact
264 remained challenging. We have highlighted the need for multiple lenses to establish and explore the complexity
265 of the risk portfolio and thus understand the architecture of the engaged stakeholders and their behaviour. This
266 is essential to ensure actionable knowledge is generated and bottom-up initiatives are strengthened and the
267 capacity to respond is improved.

268
269 Based on the above discussions of key questions, missing links and design needs, we propose the ‘**SMART**
270 convergence participatory research’ approach to support the EWS development phase and provide a checklist
271 of good practices. The SMART approach highlights crucial activity layers to incorporate into EWS
272 development which can help guide multi-disciplinary teams (e.g. disaster risk manager, hydrologist, engineer,
273 and social scientist) (Figure 2). This will enable to incorporate diverse disciplinary lenses (i.e., social science
274 and meteorological data) along with risks diversity identify by the community-at-risk (illegal settlement beside
275 riverbank or slums) which mentioned earlier as missing-link. This will support to expose vulnerability and risks
276 from different socio-cultural, institutional and scientific context. Following a **SMART** approach will ensure
277 inclusiveness by helping to identify and connect missing components and linkages when designing an EWS.

278
279 The first step, **S**, represents ‘**Shared understanding of the risks**’ ensuring all stakeholder engagements are diverse
280 and representative (irrespective to their gender, sex, age, socio-economic status and physical abilities) and a
281 wide range of data forms and collection methods are utilised, as stated in EWS step-1 (Figure 1). This knowledge
282 generated from the community will help the expert group to better understand context specific risks with more
283 focused exposure and vulnerability analysis. This further helps to identify common goals and anticipate damage
284 from the natural hazards and thus ensures impact through appropriate forecasting.

285
286 Secondly, **M** representing ‘**Monitoring of the risks**’ aligned closely with establishing alert system and
287 forecasting hazard information as stated in step-2 (Figure 1). This includes an intersection of generated
288 knowledge that will lead towards practicing collaborative activities, such as trust-building (which is key to
289 inclusive and impact-based forecasting), exchanging critical risk information to enrich data sets, feedbacks,
290 forming small groups for maintaining forecasting system.

291
292 Thirdly, **A**, building **Awareness** (i.e., training and capacity development activities to embed understanding of
293 real time weather and alert information) is critical for this approach and is a continuous process throughout the
294 development and utilisation of early warning system, in particular focus to EWS step 3 to support effective
295 communication and dissemination and will further also support legacy and sustainability of the warning system
296 into the local context.

297
298 Finally, **RT** indicating pre-planning **Response actions on Time** (i.e., comprehensive disaster management plan,
299 evacuation plan) based on the alert produced by the EWS and could be used to inform the effectiveness of the
300 overall EWS to minimize risks from the anticipated hazard. This will inform further the level of knowledge
301 produced through collaboration and how this can facilitate effective action by the community and responsible
302 agencies.

303
304 [Figure 2]

305
306
307
308
309
310
311
312
313
314
315
316
317
318
319
320
321
322
323
324
325
326
327
328
329
330
331
332
333
334
335
336
337

We advocate the use of this **SMART** approach to facilitate bottom-up initiatives for developing an inclusive and purposeful early warning system and to benefit the community-at-risk by engaging them every step of the way along with including other stakeholders at multiple scales of operations (i.e., scientific and policy actors). We advocate that the **SMART** convergence approach along with the dominant largely top-down initiatives will contribute to developing capacity and redefining adaptation and resilience in the face of more extreme water extremes (floods, droughts) and increased uncertainty under global change.

Figures (1& 2)

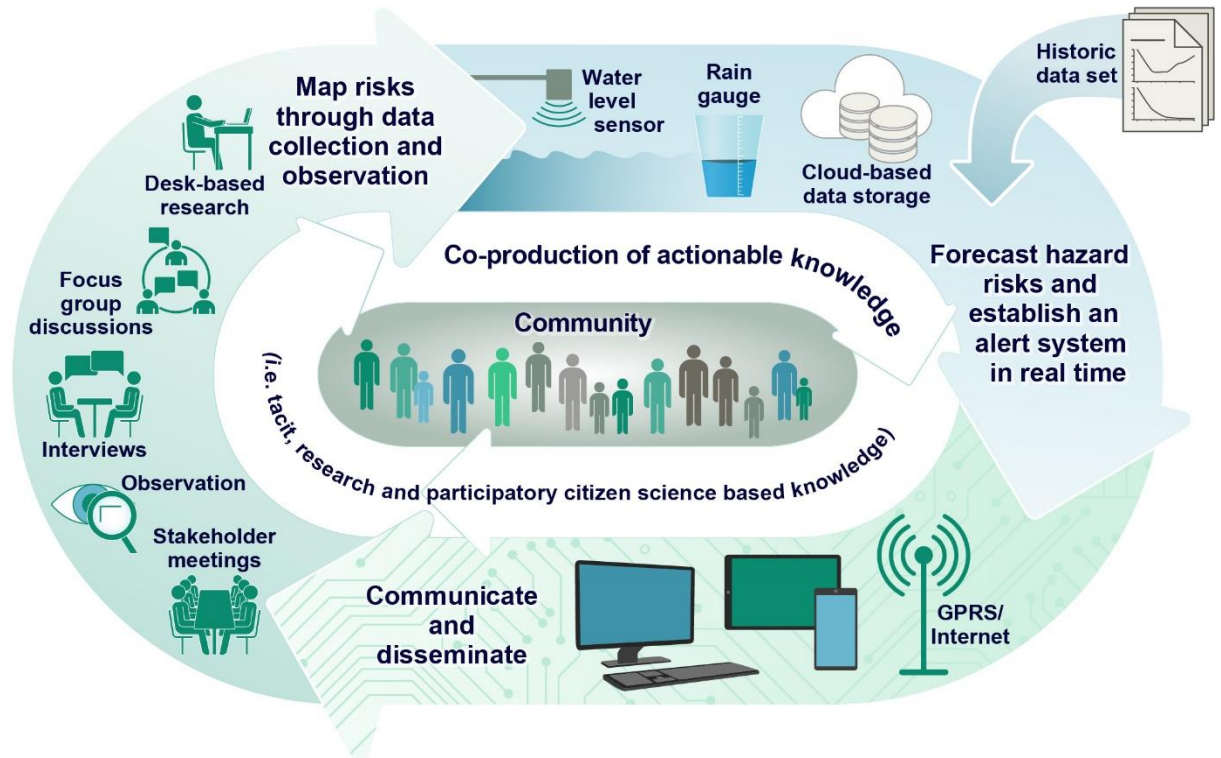
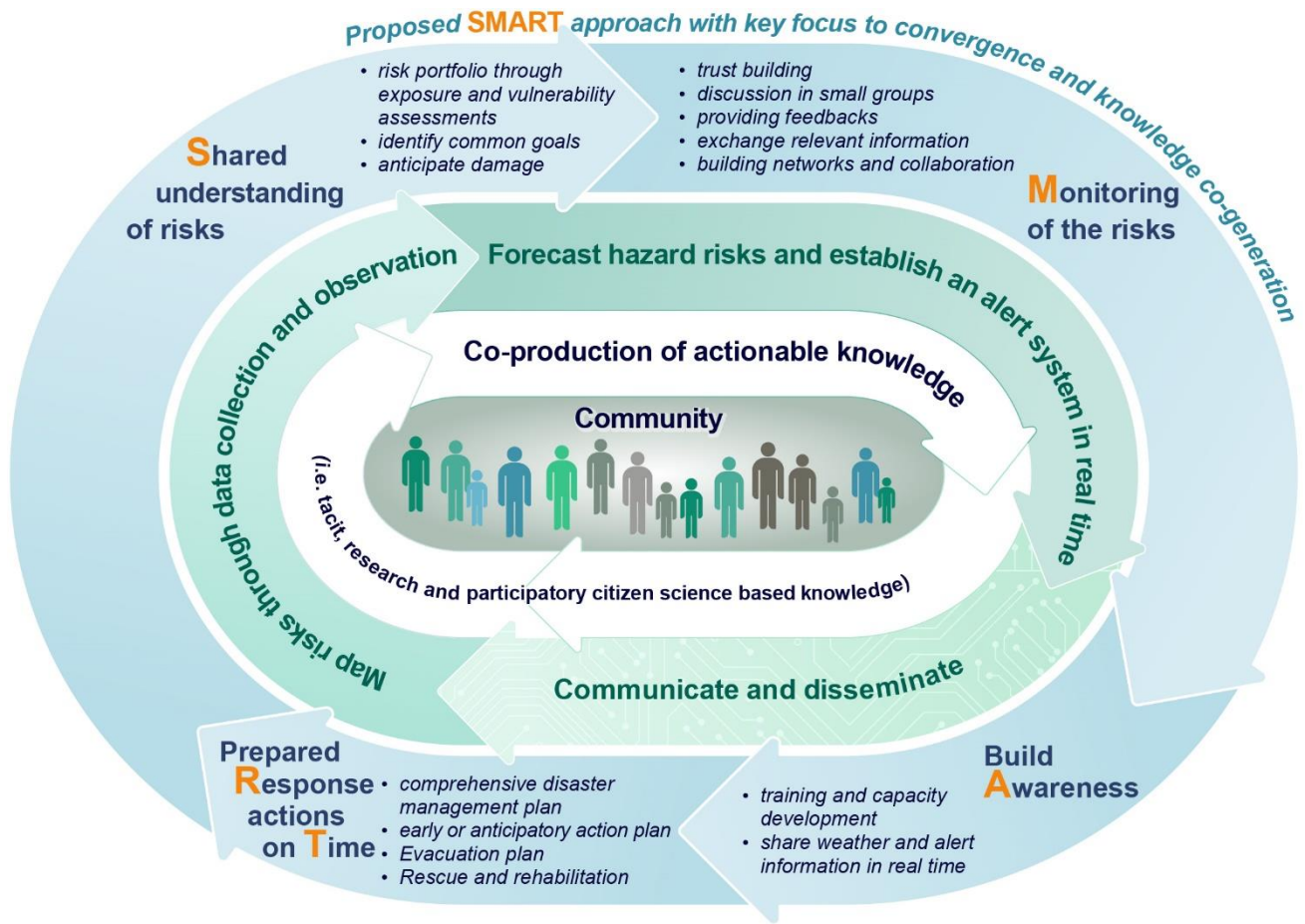


Figure 1: An idealised scenario for developing a monitoring and alert system to provide an early warning of potentially life/livelihood threatening natural hazards.

338 Figure 2:



339

Figure 2: A SMART convergence research approach to ensure inclusiveness in designing monitoring and alert system to provide early warning information to minimize disaster risks.

340

341

342

343

344

345

346

347

348

349

350

351

352

353 **Authors contribution**

354 TY and DMH prepared the manuscript with contributions from all co-authors.

355

356 **Acknowledgement**

357 The research was funded by the Natural Environment Research Council (UKRI NERC), Research grant
358 reference no: NERC COP26 A&R Project Scoping Call -2021COPA&R31Hannah.

359

360 **Useful links and report links**

361

- 362 1. Mountain-EVO (2017) project paper and reports are available at: paramo.cc.ic.ac.uk/espa/
- 363 2. International Strategy for Disaster Reduction (ISDR). Emerging Challenges for Early Warning Systems in context of
364 Climate Change and Urbanization. Available online:
365 http://www.preventionweb.net/files/15689_ewscontextofccandurbanization.pdf
- 366 3. Guidelines on Early Warning Systems and Application of Nowcasting and Warning Operations (2010) by World
367 Meteorological Organization (p. 25) and can be access through https://library.wmo.int/doc_num.php?explnum_id=9456
- 368 4. Explained: Why India's Early Warning Systems For Floods And Cyclones Fall Short (indiaspend.com)
- 369 5. Community-Based Flood Early-Warning system-India: A collaborative project by the International Centre for Integrated
370 Mountain Development (ICIMOD), Aranyak and Sustainable Eco Engineering (SEE) [https://unfccc.int/climate-action/un-
371 global-climate-action-awards/winning-projects/activity-database/community-based-flood-early-warning-system-
372 india?gclid=Cj0KCCQjw--2aBhD5ARIsALiRlWBy8J63opnqOTpqi_9ciM3IONeEat2vk2S1bNk88d-
373 IfxpVYIpld1MaAkpeEALw_wcB](https://unfccc.int/climate-action/un-global-climate-action-awards/winning-projects/activity-database/community-based-flood-early-warning-system-india?gclid=Cj0KCCQjw--2aBhD5ARIsALiRlWBy8J63opnqOTpqi_9ciM3IONeEat2vk2S1bNk88d-IfxpVYIpld1MaAkpeEALw_wcB)
- 374 6. The Future of Forecasts: Impact-Based Forecasting for Early Action: a joint report by the Red Cross Red Crescent and
375 the UK Met Office, <https://www.anticipation-hub.org/download/file-58>;
- 376 7. World Meteorological Organization (WMO) Guidelines on Multi-hazard Impact-based Forecast and Warning Services
377 https://library.wmo.int/?lvl=notice_display&id=21994#.YvN5LnbMKUk.

378

379 **References**

- 380 1. Acosta-Coll, M., Ballester-Merelo, F., Martinez-Peiró, M., & la Hoz-Franco, D. (2018). Real-time
381 early warning system design for pluvial flash floods—A review. *Sensors*, 18(7), 2255.
382 doi:10.3390/s18072255.
- 383 2. Beven, K., Asadullah, A., Bates, P., Blyth, E., Chappell, N., Child, S., ... & Wagener, T. (2020).
384 Developing observational methods to drive future hydrological science: Can we make a start as a
385 community?. *Hydrological Processes*, 34(3), 868-873. <https://doi.org/10.1002/hyp.13622>.
- 386 3. Birthisel, S. K., Eastman, B. A., Soucy, A. R., Paul, M., Clements, R. S., White, A., & Dittmer, K. M.
387 (2020). Convergence, continuity, and community: a framework for enabling emerging leaders to build
388 climate solutions in agriculture, forestry, and aquaculture. *Climatic Change*, 162(4), 2181–2195.
- 389 4. Buytaert, W., Ochoa-Tocachi, B. F., Hannah, D. M., Clark, J., & Dewulf, A. (2018). Co-generating
390 knowledge on ecosystem services and the role of new technologies. In *Ecosystem Services and
391 Poverty Alleviation* (pp. 174-188). Routledge.
- 392 5. Cardona, O. D., Van Aalst, M. K., Birkmann, J., Fordham, M., Mc Gregor, G., Rosa, P., ... &
393 Thomalla, F. (2012). Determinants of risk: exposure and vulnerability. In *Managing the risks of
394 extreme events and disasters to advance climate change adaptation: special report of the
395 intergovernmental panel on climate change* (pp. 65-108). Cambridge University Press.
396 10.1017/CBO9781139177245.005.
- 397 6. Hannah, D. M., Lynch, I., Mao, F., Miller, J. D., Young, S. L., & Krause, S. (2020). Water and
398 sanitation for all in a pandemic. *Nature Sustainability*, 3(10), 773–775.
399 <https://doi.org/10.1038/s41893-020-0593-7>
- 400 7. Hermans, T.D., Šakić Trogrlić, R., van den Homberg, M.J., Bailon, H., Sarku, R. and Mosurska, A.,
401 (2022). Exploring the integration of local and scientific knowledge in early warning systems for
402 disaster risk reduction: a review. *Natural Hazards*, pp.1-28.

- 403 8. Kosow, H., Kirschke, S., Borchardt, D., Cullmann, J., Guillaume, J. H. A., Hannah, D. M., Schaub, S.,
404 & Tosun, J. (2022). Scenarios of water extremes: Framing ways forward for wicked problems.
405 *Hydrological Processes*, 36(2), e14492. <https://doi.org/10.1002/hyp.14492>
- 406 9. Lakhina, S. J., Sutley, E. J., & Wilson, J. (2021). “How do we actually do convergence” for disaster
407 resilience? Cases from Australia and the United States. *International Journal of Disaster Risk Science*,
408 12, 1–13.
- 409 10. Laudon, H., & Sponseller, R. A. (2018). How landscape organization and scale shape catchment
410 hydrology and biogeochemistry: Insights from a long-term catchment study. *Wiley Interdisciplinary*
411 *Reviews: Water*, 5(2), e1265. <https://doi.org/10.1002/wat2.1265>.
- 412 11. Mao F, Clark J, Buytaert W, Krause S, Hannah DM. Water sensor network applications: Time to
413 move beyond the technical? *Hydrological Processes*. 2018; 32:2612–2615.
414 <https://doi.org/10.1002/hyp.13179>
- 415 12. Mashi, S. A., Inkani, A. I., Obaro, O., & Asanarimam, A. S. (2020). Community perception, response
416 and adaptation strategies towards flood risk in a traditional African city. *Natural Hazards*, 103(2),
417 1727-1759.
- 418 13. Paul, J. D., Buytaert, W., Allen, S., Ballesteros-Cánovas, J. A., Bhusal, J., Cieslik, K., ... & Supper, R.
419 (2018). Citizen science for hydrological risk reduction and resilience building. *Wiley Interdisciplinary*
420 *Reviews: Water*, 5(1), e1262. doi: 10.1002/wat2.1262
- 421 14. Pandeya, B., Uprety, M., Paul, J. D., Sharma, R. R., Dugar, S., & Buytaert, W. (2021). Mitigating
422 flood risk using low-cost sensors and citizen science: A proof-of-concept study from western
423 Nepal. *Journal of Flood Risk Management*, 14(1), e12675.
- 424 15. Peek, L., Tobin, J., Adams, R. M., Wu, H., & Mathews, M. C. (2020). A framework for convergence
425 research in the hazards and disaster field: The natural hazards engineering research infrastructure
426 CONVERGE facility. *Frontiers in Built Environment*, 6, 110.
- 427 16. Roque, A., Wutich, A., Quimby, B., Porter, S., Zheng, M., Hossain, M. J., & Brewis, A. Participatory
428 approaches in water research: A review. *Wiley Interdisciplinary Reviews: Water*, e1577.
- 429 17. Sterling, E. J., Zellner, M., Jenni, K., Leong, K. M., Glynn, P. D., BenDor, T. K., Bommel, T. K.,
430 Hubacek, K., Jetter, A. J., Jordan, R., Schmitt Olabisi, L., Paolisso, M., & Gray, S. (2019). Try, try
431 again: Lessons learned from success and failure in participatory modelling. *Elementa Science of the*
432 *Anthropocene*, 7(1), 9. <https://doi.org/10.1525/elementa.347>
- 433 18. Wallerstein, N., Duran, B., & Oetzel, J. G. (2017). In M. Minkler (Ed.), *Community-based*
434 *participatory research for health: Advancing social and health equity*. John Wiley & Sons.
- 435 19. Westerhoff, P., Wutich, A., & Carlson, C. (2021). Value propositions provide a roadmap for
436 convergent research on environmental topics. *Environmental Science & Technology*, 55(20), 13579-
437 13582.
- 438 20. Zulkafli, Z., Perez, K., Vitolo, C., Buytaert, W., Karpouzoglou, T., Dewulf, A., ... & Shaheed, S.
439 (2017). User-driven design of decision support systems for polycentric environmental resources
440 management. *Environmental Modelling & Software*, 88, 58-73.

441