

Perspectives on flood forecast-based early action and opportunities for Earth observations

Claire Nauman,^{a,b,*} Eric Anderson^{b,c}, Erin Coughlan de Perez^{b,d,e},
Andrew Kruczkiewicz^{b,d,e}, Shanna McClain,^f Amanda Markert^{b,c},
Robert Griffin^b,^{a,b} and Pablo Suarez^{d,g}

^aThe University of Alabama in Huntsville, Department of Atmospheric and Earth Science,
Huntsville, Alabama, United States

^bNASA/SERVIR Science Coordination Office, NASA Marshall Space Flight Center, Huntsville,
Alabama, United States

^cThe University of Alabama in Huntsville, Earth System Science Center, Huntsville, Alabama,
United States

^dRed Cross Red Crescent Climate Centre, The Hague, Netherlands

^eColumbia University, International Research Institute for Climate and Society,
The Earth Institute, New York, United States

^fNASA Headquarters, Washington, DC, United States

^gBoston University, Pardee Center for the Study of the Longer-Range Future, Boston,
Massachusetts, United States

Abstract. This paper seeks to identify opportunities to integrate Earth observations (EO) into flood forecast-based early action and propose future directions for research and collaboration between EO and humanitarian communities. Forecast-based early action (FbA) is an approach to shift disaster response toward anticipation to mitigate impacts to at-risk communities; however, timely and accurate information is needed in the development of data-based triggers and thresholds for action. Therefore, this paper considers the readiness of a wide range of EO for flood monitoring and forecasting in the design, operations, and evaluation phases of FbA. The most significant opportunities for EO to inform FbA efforts lie in the design and evaluation phases, as EO can aid in the development of impact-based triggers. The EO products most readily applicable include precipitation, streamflow estimates, and exposure mapping, and those requiring the greatest amount of further research include vulnerability and impact assessments. This paper identifies collaboration opportunities for the EO and humanitarian communities to create tailored products, such as overlays combining flood extents with exposure maps. Such collaboration opportunities can be fostered by open data sharing, data verification efforts, and incentives for supporting boundary organizations capable of enabling the use of EO for FbA. © *The Authors*. Published by SPIE under a Creative Commons Attribution 4.0 Unported License. Distribution or reproduction of this work in whole or in part requires full attribution of the original publication, including its DOI. [DOI: [10.1117/1.JRS.15.032002](https://doi.org/10.1117/1.JRS.15.032002)]

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1 Introduction

Floods cause significant damage and loss of lives worldwide, more so than any other type of natural disaster.¹ In an effort to minimize flood impact and the impact of all disasters, international bodies have come together to set goals and priorities for disaster risk reduction, as presented in the Hyogo Framework for Action and its successor, the Sendai Framework for Disaster Risk Reduction.^{2,3} The goals outlined in these frameworks have encouraged the development of risk reduction strategies by a wide array of sectors. One such approach is the humanitarian community's forecast-based early action (FbA), which endeavors one to enable action

*Address all correspondence to Claire Nauman, cmn0009@uah.edu

based on forecasts of extreme events. Forecast-based Financing for early action is gaining momentum as an approach to manage changing climate risks, including through United Nations Framework Convention on Climate Change (UNFCCC)-related processes and the Green Climate Fund.⁴

In recent years, many organizations have developed and strengthened FbA efforts for a variety of hazards, with flooding being one of the most common.⁵ Based on these initiatives, challenges in linking flood forecasts to action include bridging the gap between forecasts of physical phenomena and impact on lives and livelihoods, tailoring existing forecasts to local contexts and needs, and sharing information that is relevant and translatable across disciplines. Given these challenges and the promising growth of the field, we see an opportunity for the EO community to learn from and contribute to FbA efforts aimed at flood risk reduction.

In this paper, we provide a brief introduction to FbA, describe the readiness of Earth observations (EO) to inform flood FbA, and discuss challenges, opportunities, and recommendations for the EO community to better serve the needs of humanitarian programs implementing FbA.

2 Context

For the EO community to serve the needs of the FbA community, a general understanding of the FbA concept is needed. In this section, we provide an overview of the FbA framework, locations where FbA is currently implemented, and the core FbA concepts EO scientists should know.

2.1 Introduction to the Forecast-Based Early Action Concept

Despite efforts to shift focus toward risk reduction, in the last 20 years, ~12% of funding for disasters has been spent on reducing disaster risk, with the remaining 88% spent on emergency response, reconstruction, and rehabilitation.⁶ With such an imbalance in funding, the investments in risk reduction strategies must be highly effective. Studies have found that taking preventative action to avoid disaster losses can provide significant return on investment.⁷ Hallegatte⁸ describes how investing in early warning systems (EWS) can yield returns via avoided losses and increased economic production in weather-sensitive sectors. Another study estimated that flood early action could save US\$34 per dollar invested in flood prone districts in Nepal.⁹ Statistics such as these have strengthened financial support from governments and donor agencies for the development of EWS, with the hopes that these systems might make it possible for people to act before an extreme event occurs, saving lives, and reducing impact. However, often-times, action is not taken in response to warnings from such EWS,^{10,11} and an early warning has no effect without early action.¹² FbA systems are meant to change this, enabling people to take pre-emptive action based on a forecast, even under conditions of uncertainty.^{13,14}

FbA is an approach utilized in humanitarian communities to trigger early action when a severe weather or climate forecast reflects a high likelihood of critical impact to lives and livelihoods in a targeted area. The intention is to anticipate disasters and mitigate their impact to reduce human suffering and loss. FbA combines meteorological forecast information with risk analysis to identify where and when early actions should be implemented. To identify the spatial and temporal areas of action, FbA relies on a complex system of variables to prioritize risk and impact.

In many FbA systems, anticipatory or early action protocols detail the roles and responsibilities for action. These protocols take into account the forecast probability, the type of hazard, the magnitude of the event predicted, the early actions that may be taken, the cost and effects associated with the action or inaction, and the organizations required to facilitate such activities.¹³ Each component has associated questions and challenges to consider.¹⁵ For example, how should one determine critical magnitudes of an extreme event that should trigger action? What type of impacts might different magnitudes cause? Once that magnitude or danger level is identified, what is the appropriate level of probability for a certain action to be justified? Does the probability of the forecast reflect reality, or does the forecast predict higher or lower probabilities than actually occur?

As interest in the concept of FbA grows, so does the number of FbA programs and actors involved. Figure 1 shows the locations of FbA programs implemented by the International

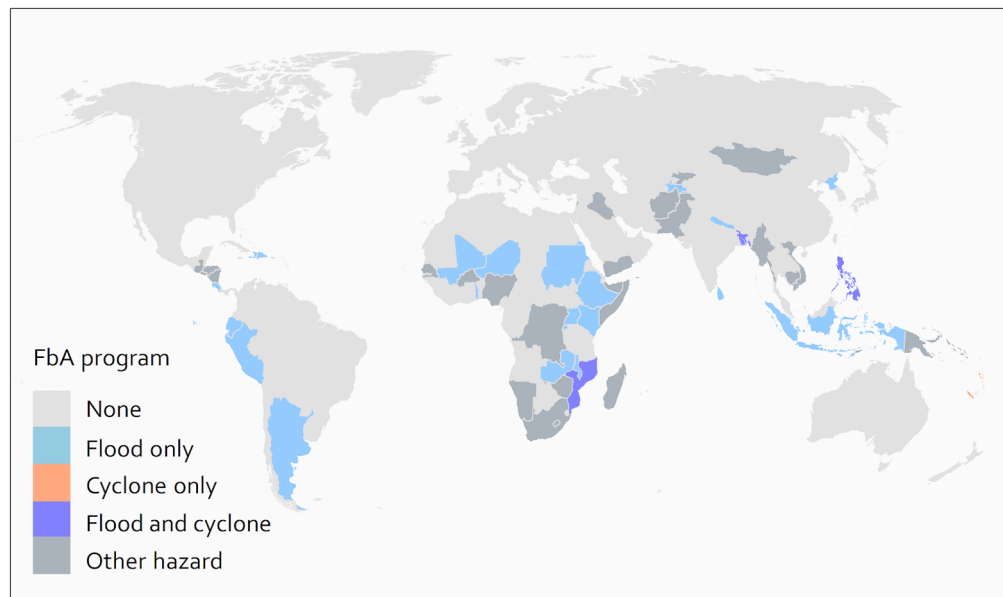


Fig. 1 Countries with FbA programs and their associated hazard focus implemented by the International Federation of Red Cross and Red Crescent Societies, Start Network, World Food Program, and Food and Agriculture Organization.^{16,17}

Federation of Red Cross and Red Crescent Societies, Start Network, World Food Program, and Food and Agriculture Organization.^{16,17} Most of these programs focus on droughts or floods, with a few other hazard types interspersed, such as cold waves in Peru, volcanoes in Ecuador, dzud in Mongolia, or heat waves in Vietnam. Of the 53 countries with active FbA programs, 19 have a flood or cyclone focus. Many of these flood FbA programs use probabilistic or deterministic forecasts, whereas others use real-time monitoring data, historical data, or a combination of forecasts and real-time data.⁵

2.2 Determining Triggers

Most FbA programs take one of two main approaches to determine when to act: pre-defined triggers or real-time forecast-informed decision-making.⁵ We focus on pre-defined triggers here, because they represent a novel aspect of FbA, distinguishing FbA from typical EWS. Triggers are defined by the danger level or magnitude of a forecasted event (e.g., 100 mm of rain over 24 h) and forecast probability (e.g., 80% chance of the event occurring). In more recent implementations of FbA, triggers are being defined in terms of impact, rather than hazard, following guidelines proposed for impact-based forecasting.^{18,19} For example, an impact-based trigger would instigate action if more than 30% of houses in a community are forecasted to flood, which could happen from several different hazards, such as storm surge or rainfall-induced flooding. A trigger determined solely based on a hazard (e.g., a given river height) is defined by relating those historic river heights to their documented impacts but does not include potential impacts of related hazards.

Several methods exist to design triggers for FbA systems.^{11,14,15,20–22} Several approaches draw on the value of information arguments, which seek to move beyond using typical forecast verification metrics to quantify skill, toward assigning economic value to observations and forecasts.^{23–28} To assess the financial costs and benefits of any potential trigger, designers build contingency tables such as the one shown in Table 1, building on the choice molecule approach described by Suarez et al.²⁹ Together, the false alarm ratio (FAR) [FAR = false positives/(true positives + false positives)] and the total number of times an action will be triggered will determine the expected cost of the FbA program over time. The number of expected “worthy actions” directly conveys FbA program benefit via avoided losses. “actions in vain” may pose reputational risks but not all false positives may be considered total losses—some actions in vain may

Table 1 FbA outcome scenarios.

	Extreme event	No extreme event
Action	Worthy action (true positive)	Action in vain (false positive)
No action	Fail to act (false negative)	Worthy inaction (true negative)

Note: adapted from Lopez.²²

have secondary benefits (e.g., awareness campaigns on hygiene may help prevent the spread of disease; first aid kits can serve any type of unexpected accident).

2.3 FbA Phases

This paper considers three phases of FbA programs: design, operations, and evaluation. The design phase includes the development of an impact-based trigger and results in protocols outlining anticipatory or early actions. A key element of the design phase is to explore questions around both climate risk and vulnerability. For example, there is a need to explore who is vulnerable to the flood hazard and where they are located, to consider the skill of available forecasts, and to identify the probability of occurrence of flood events of different magnitudes (linked to the so-called return period, a problematic term for risk communication instances—especially given climate change and physical interventions in river basins). The operations phase is when real-time monitoring and forecasting occur, and early actions take place according to protocol. The evaluation phase includes steps to monitor and track indicators to learn from and evaluate whether the program met its objectives to reduce overall flood impact in cost-effective ways [akin to Monitoring, Evaluation, and Learning (MEL) in development parlance]. Evaluation can occur during and after the design and operations phase.

3 Readiness of Earth Observation to Inform Flood Forecast-based Action

The next sections consider the abilities and limitations of EO in all three phases of FbA. Further, the ability of EO to characterize physical hazard versus exposure, vulnerability, and impact of floods is also examined. Many thorough reviews exist on methods and uncertainties of flood modeling, the integration of EO into flood modeling, and the state of operational use of flood models (e.g., Refs. 30–34). The role of EO builds on an understanding that flood forecasts can be generated from hydrologic and/or hydraulic models that use combinations of *in situ*, remotely sensed, and modeled inputs. These inputs can be static (e.g., elevation, land cover, and river channel morphology) and dynamic (e.g., precipitation, evapotranspiration, soil moisture, snow cover, and snow water equivalent). Flood monitoring with EO often refers to historical or near real-time observation of flood extent with both active and passive sensors. To measure or forecast flood exposure, vulnerability, and impact, EO data and techniques must be combined with dynamic socioeconomic data and further multidisciplinary perspectives. This poses nearly insurmountable challenges for accuracy in an operational sense, but careful incorporation of EO in the design phase may position the satellite remote sensing community to more meaningfully inform impact-based triggers.

3.1 Current Examples of EO in the Operational Phase of FbA Programs

The main variables currently used to trigger action for floods are forecasted river height and streamflow. EO-based precipitation observations and numerical weather prediction (NWP) products are often used as inputs to hydrologic models that provide streamflow or river height-based triggers (e.g., Refs. 18 and 35), as evidenced in Bangladesh, Zambia, and Peru, among other countries.³⁶ In this way, EOs are also the basis of NWP-derived triggers for cyclones (e.g., wind speed) in the Philippines and El Niño-related hazards (e.g., temperature and precipitation) in

Peru.^{37,38} Early action protocols such as those in Bangladesh and Peru are based on global and national flood forecasting systems. Often, global systems with longer outlooks (e.g., the Global Flood Awareness System) trigger a first phase of initial preparedness actions and are followed by national forecasts with shorter outlooks (e.g., Bangladesh Flood Forecasting and Warning Center) that can trigger a second phase of actions like evacuation and cash assistance.^{36,39}

We should point out that even the lowest latency, near-real time EO-based flood monitoring products, by definition, provide less lead time for early actions, especially if consulted without any additional historical context related to expected impact. These flood monitoring products might be most useful in the anticipation of secondary flood impacts, such as disease outbreaks. Therefore, the design and evaluation phases are reviewed to uncover EO's greatest untapped potential for FbA.

3.2 Future of EO in Designing and Evaluating FbA Programs

For EO science and data to more meaningfully serve FbA goals of reducing costs of responding to floods and reducing flood impacts, the EO community must play a stronger role in the design phase. Table 2 reflects the readiness of EO inputs in the design and evaluation phases. This was conducted by weighing scientific and organizational factors in assigning readiness and considering the NASA Application Readiness Level (ARL) metric.⁴⁰ Readiness factors include the ability of EO science and technology to characterize the specific flood-related phenomenon, and the extent to which EO has been used for basic research (low ARL) versus approved for decision-making environments (high ARL). In Table 2, a key is provided for interpreting the readiness ratings shown in Table 2. To aid discussion, the flood-related variables are ordered from fundamental (i.e., precipitation and soil moisture), intermediate (i.e., river height, streamflow, flood extent, and depth), to complex (i.e., exposure, vulnerability, and impact).

Also noted are the differences between forecasting and monitoring variables that EO can capture. Monitoring variables such as observed flood extent and depth can inform the design and evaluation phases, but it is unclear to what extent EO-based monitoring adds value to the operational phase of FbA, especially given wide ranges of flood duration and magnitude of impact. Generally, only forecast variables have roles in the operations phase. One notable exception may arise if a monitoring variable that characterizes antecedent conditions (e.g., soil moisture) is determined to add value to predicting flood impact when coupled with other forecast data.

The most significant opportunities for EO to enhance FbA programs and to gain applications experience through FbA lie in the design phase. To develop impact-based triggers, research is ongoing to determine the extent to which fundamental variables such as precipitation and soil moisture can play a role, whether used independently or as combined indices to serve as proxies or precursory signals of flood impact (e.g., Refs. 20 and 41–45). River height and streamflow (e.g., Refs. 46–50) are the most common trigger variables used in flood FbA programs, but by themselves do not describe flood impact. Intermediate variables such as flood extent and depth come closer and are traditionally derived through simulations of hydrologic and hydraulic models. Other studies demonstrate that EO can be used to construct historic catalogs that relate historic streamflow or river height to resulting flood extent and onset date (e.g., Refs. 41 and 51), depth (e.g., Ref. 52), and exposure (e.g., Refs. 53–56). To determine triggers based on impact and to operationalize impact-based forecasts for early action, it may not be necessary to produce forecasts of depth, exposure, vulnerability, or impact. Rather, in the design phase, triggers can be defined based on the impact associated with more readily available flood variables (e.g., river height or streamflow). Again, depth, exposure, and vulnerability play critical roles in determining expected levels of flood impact, but these complex variables may not need to be forecasted in real time to define and trigger early actions to reduce impacts.

While the entire range of flood variables captured by EO is poised to support the design phase, the evaluation phase likely demands perspectives from the more complex variables. To be clear, the evaluation phase of FbA assesses whether the FbA program achieved its intended goals of reducing flood impacts using MEL tools, not the accuracy of a flood forecast. This is why the readiness of “n/a” has been assigned to the fundamental and intermediate variables in the evaluation phase. Traditionally, program evaluations may involve expensive household and

Table 2 Readiness of EO inputs for FbA.

(a) Relating EO readiness to calls for action			
Symbol	Readiness of EO in FbA phase (ARL range) ^a	Interpreting EO for FbA Readiness as “calls to action”	
*	Low (1 to 2)	More multidisciplinary, fundamental, and applied research must be funded, ideally by multiple agencies	
**	Medium (3 to 5)	Some fundamental research and significant applied research are required	
***	High (6 to 7)	Some applied research remains; EO scientists and humanitarian organizations should be poised to co-develop better solutions	
****	Very high (8 to 9)	Connections may already exist; if not, these should be ready to incorporate into early action protocols	
(b) Readiness of EO inputs for flood phenomena by FbA phase			
FbA phase	Flood-related phenomena	EO inputs ^b	Readiness
DESIGN Development of impact-based trigger	Precipitation	Satellite precipitation products, quantitative precipitation forecast	****
	Soil moisture	All above inputs + typical variables in a Land Surface Model (LSM)	**
	River height	All above inputs + potentially including altimetry and reservoir operations into hydrologic model	***
	Streamflow	All above inputs	****
	Flood extent	All above inputs + optical and radar satellite imagery	***
	Flood depth	All above inputs + digital elevation model (DEM)	***
	Exposure	All above inputs + nighttime lights, populated areas/density, agricultural areas/crop type, building footprints, and infrastructure	***
	Vulnerability	All above inputs + additional socioeconomic analysis	*
	Impact	All above inputs + additional socioeconomic and cost analysis	*
EVALUATION Program MEL	Precipitation, soil moisture, river height, and streamflow		n/a
	Flood extent	Optical and radar satellite imagery	****
	Flood depth	All above inputs + DEM	***
	Exposure	All above inputs + nighttime lights, populated areas/density, agricultural areas/crop type, building footprints, infrastructure	***
	Vulnerability	All above inputs + additional socioeconomic analysis	*
	Impact	All above inputs + additional socioeconomic and cost analysis	*

^abased on the ability of EO to characterize the phenomenon and EO’s demonstrated current use, roughly mapped to NASA ARL

^bEO inputs, whether direct inputs or derived from/with satellite inputs

community surveys. EO-derived flood extent and depth estimates could add more extensive perspectives of FbA program areas. EO information could target locations for surveys or complement existing surveys. If the findings in the evaluation point to poorly performing triggers during operations, a more localized accuracy assessment of fundamental variables (i.e., a traditional, scientific validation exercise) might help diagnose failures or breakdowns in impact-based early actions. Ultimately, findings from programmatic evaluations should feedback into research and design, and hopefully improvements in operations. For instance, a monitored flood extent map during or immediately after a flood event may serve to adjust thresholds to better capture forecast flood impact. Annual redesigns may also benefit from the perspectives of EO. Lessons point to the potential value of EO to annually update maps, and resulting flood inundation forecasts, in regions with expected changes in vulnerability (e.g., vulnerable island populations).

4 Further Serving Humanitarian Flood Early Action

This section now considers a roadmap for collaboration. Researchers and practitioners in the EO community have a responsibility to demonstrate benefit to society; collaborating with the humanitarian community on FbA may be one way to do so. Some EO inputs are very mature, whereas others require further tailoring to be relevant to the different phases of FbA. This section reflects the challenges and opportunities for joint research and action between the EO and humanitarian communities, with the FbF research roadmap in mind.⁵⁷

4.1 Tailoring Forecasts to Humanitarian Needs

The EO community can serve humanitarian needs in the FbA design phase by tailoring existing forecasts so that stronger connections can be formed between EO-based information and flood impact: both smart, forecast-based decisions and simple, decision-based forecasts are needed.⁵⁸ In this context, forecast tailoring means more than model calibration and validation, and includes how to present and interpret forecasts and their uncertainties for defining impact-based triggers. The aim is to allow access to forecast outputs in such a way that they can be applied and interpreted with greatest flexibility, and whenever possible, to curate forecast variables that have good predictability and high correlations with flood impacts in the particular region of interest.

For example, tailoring precipitation forecasts for a given region, could include testing different combinations of rainfall variables such as rainfall intensity and duration to determine, which are the most correlated with flood impact.^{20,41} If a streamflow or river height forecast is available, tailoring could mean linking flood extents or depths associated with certain streamflows or river heights, as discussed in Sec. 3.2, and overlaying those extents/depths with exposure and vulnerability information to estimate potential impact. Tailoring could also mean respecting the limitations of EO and incorporating contextual local knowledge through household and community surveys. For example, if certain sections of the river are known to overflow their banks more often, this information can be used to interpret where potential impacts from a given streamflow are likely to occur and guide actions to reinforce levees or clear drainage canals.⁵⁹

To reach an FbA audience, the EO and flood forecasting communities could change how they report and present flood forecast verifications and accuracy assessments, beyond those called for by many in the hydrology community (e.g., Ref. 34). Further improvements could involve co-developing new and common performance metrics relevant to each early action and its required lead time, all of which could be disaggregated at local scales (e.g., How skilled is a flood forecast in identifying conditions that should trigger the distribution of water purification tablets in different communities across the program area?).⁶⁰ To facilitate this, flood modelers could communicate their accuracy assessments for extreme events for individual gauges and over the full range of forecast lead times.

Results of an FbA program evaluation may also open opportunities for further tailoring. For instance, within one FbA program area, an evaluation may uncover disparities among community benefits of early actions intended to reduce flood costs or impacts. Such findings may orient efforts to further tailor forecasts and redefine triggers to capture differing impacts within a program area, similar to the approaches and challenges faced in defining unit areas or zones by

the index insurance sector.^{61–64} In this scenario, the operational phase would need sufficient resources to take more locally specific actions. In another instance, an EO-informed evaluation may help assess the reach of the early action or intervention. For example, EO may show that a flood caused much wider impacts, but only a small section of the river was evacuated.

4.2 Challenges Beyond the Forecast

In approaching research on FbA through the lens of EOs, it is important to note that many of the challenges with translating forecasts into action are actually not related to the forecast itself. Certainly, work is needed in improving forecasts, understanding which forecasts to use, and tailoring them for specific applications so that it is easier for decision-makers to take action. However, much of the difficulty has to do with funding, political considerations, and coordination and information sharing across organizations—areas beyond the scope of the scientific community.⁵⁹ Since taking action on a forecast requires multiple groups to come together, it is important that no one group work in isolation. Scientists working with EOs and forecasts must work together with those who will use their forecasts and become aware of the many difficulties they face, even if they are beyond the scientific scope.

Trust in forecast sources remains a challenge in designing and operating flood FbA programs. Global monitoring and forecast systems play an important role, especially in monitoring ungauged and transboundary basins and forecasting at longer lead times, but it is critical that the EO community contributes to strengthening national and regional flood early warning systems. Doing so will not only improve our understanding of the water cycle and how to produce impact-based forecasts, but it will also build trust and confidence in the local agencies responsible for risk reduction.

4.3 Roadmap Moving Forward

This paper identifies clear opportunities for EO science and data to inform different phases of the FbA life cycle. The below roadmap reflects on key thematic elements that can enable the opportunities identified, including collaboration and partnership, consideration of future EO missions, transparency and no-regrets measures, and the role of “boundary organizations” to facilitate communication, translation, and mediation across the current EO-FbA divide. Collectively, these provide greater recognition and support toward this collaborative vision and hope for success.

Groups within the applied EO community and the humanitarian community both aim to reduce flood impacts. Both must also gather evidence of their contributions to risk reduction. Collaboration on FbA would allow each community to support each other’s scientific and humanitarian missions and in gathering evidence of individual and shared successes. By collaborating more deeply in the design and evaluation phases, the EO community can enhance the humanitarian community’s ability to take effective early action to reduce flood losses. In turn, the humanitarian community can provide additional evidence from the field on the use and remaining limitations of EO science and technology.

The applied sciences community will share immediate joint successes with the humanitarian community by tailoring EO inputs that are most “ready” to serve needs of impact-based early action (Table 2). The EO inputs that we classified as less ready for FbA point to opportunities for additional fundamental and interdisciplinary research. The humanitarian community has posed challenges that can justify future research and applications of new satellite missions and is likely to provide evidence of societal benefit from such missions (e.g., Ref. 65). For instance, the NASA-Indian Space Research Organization-synthetic aperture radar mission will allow us to significantly augment EO-based historic flood catalogs that can be integrated into the FbA design phase. Additionally, the FbA and EO communities can investigate the potential for satellite altimetry measurements to complement triggers designed with *in situ* stream gauge data and streamflow forecasts, especially in ungauged basins. Collaborators may ask which flood early actions, and where could data products from the forthcoming Surface Water Ocean Topography (SWOT) satellite mission inform with its significantly greater coverage and ability to capture finer scale transboundary river height and streamflow monitoring data to complement existing flood EWS than offered by previous EO (Fig. 2).

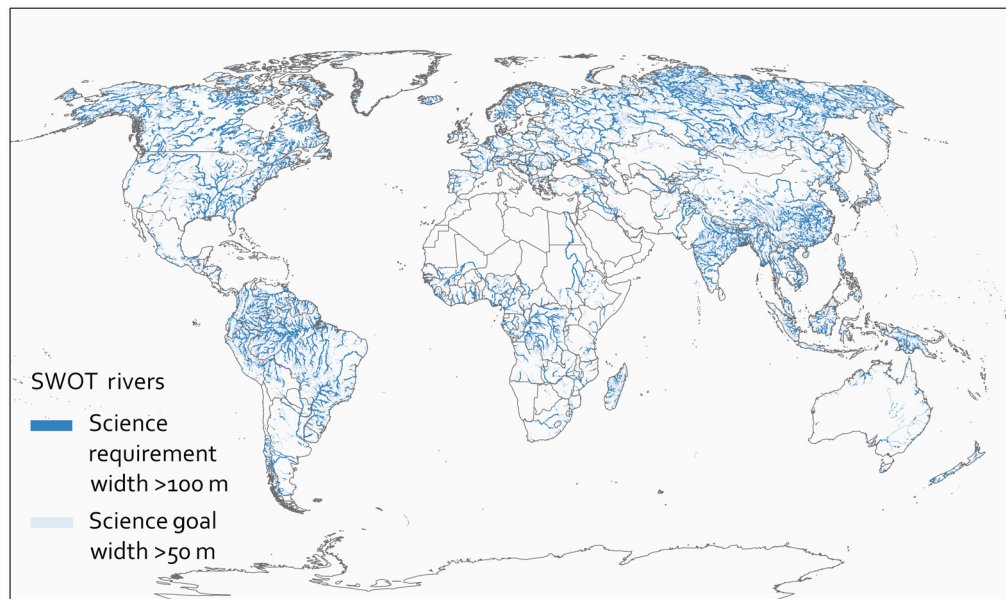


Fig. 2 SWOT will monitor significantly more rivers (derived from Ref. 66) than current and past satellite-based altimeters. The satellite mission is required to measure every river >100-m wide, and scientists aim to measure every river >50-m wide,⁶⁷ offering much more data to inform different phases of flood FbA.

Joint success in flood risk reduction requires more open sharing of information from both sides. This could mean presenting the skill of EO-informed forecasts in different ways (e.g., per basin, how skill changes with lead time). Open sharing also means more documentation and transparency in early action protocols, particularly on the early warning inputs: the original model or forecast sources (and if available, a reference to a publication describing the forecast model setup), the thresholds with units for each unique impact-based trigger, the geographic area that each early action covers, and the lead time required for each action. As anticipatory or early action protocols aim to trigger action based on reducing impact, the scientific community would learn from more detailed justifications that describe how triggers have been linked with local and regional impact; in turn, this knowledge would position EO scientists to collaborate more meaningfully in determining the most appropriate monitoring or forecasting inputs to inform impact-based early warning and early action. In fact, impact-based triggers assume an understanding of vulnerability to floods, which is often highly uncertain. More significant collaborations between the physical flood forecasting communities and the vulnerability mapping communities would result in scientific discoveries and more rigorously designed impact-based triggers, however with increased collaboration, there is a need to increase identification of roles and responsibilities, as well as incentive structures, for each scientists, decision makers, and intermediaries.^{68,69}

The early actions in Table 3 are only a few of the dozens of starting points for EO scientists and practitioners to collaborate with FbA designers, operators, and evaluators. Each early action should seed conversations about the readiness of EO to enhance existing inputs to reduce flood impacts. For each action, collaborators should discuss forecast lead time and uncertainty, the need or ability to design spatially disaggregated triggers, how each trigger for these actions is tied to impact, and more. The sample of early actions in Table 3 span lead times from general preparedness years in advance to months, weeks, days, and hours. The lead time column should orient EO scientists in considering the appropriateness of historic monitoring variables, seasonal forecasts, and near real-time forecasting variables, and the actions they can inform. Finally, the EO community should understand that some actions are no regrets such as awareness campaigns and general risk reduction actions. Other actions that affect livelihoods, such as evacuations, are highly regrettable if a false alarm occurs. We provide a more comprehensive table that can spark collaborative conversations toward realizing collective successes in flood impact reduction, in the appendix.

Table 3 Example flood early actions (see full table in the appendix).

Early actions	Lead time	Reference
Work with communities to reduce risk through concrete actions (e.g., reforestation and reinforcement of houses)	General preparedness years in advance	12
Build shelters; build latrines ^a	Three months	37
Support storage of seeds and food items; bag vulnerable items and move storage facilities to higher ground	Three months to one week	5, 37
Preposition buckets and chlorine tablets; preposition hygiene kits	one month to one week	5, 37
Distribute biodegradable trash bags, jerry cans, soap, buckets, chlorine tablets, and hygiene kits to vulnerable households	Nine to seven days	5, 11, 37
Evacuate	Days to hours	12

^aDesigned in preparedness for El Niño, and related to floods

The international applied sciences community and the humanitarian community often target the same users and beneficiaries but from different orientations and vernaculars. Applied EO scientists often work with hydro-meteorological agencies to develop capacity to characterize flood hazards and their impacts. Likewise, humanitarian organizations implementing forecast-based early action programs rely on the same hydrometeorological agencies for actionable forecasts. Meaningful collaboration requires boundary organizations (referring both to persons and to organizations) in both camps who can empathize and understand others' unique challenges and opportunities, and who can communicate each other's goals, metrics, and challenges. Since the task goes beyond mere translation (i.e., tasking a person with replacing words from one language to another), we embrace the term "boundary organization": persons and organizational arrangements with functions that enable interactions across the EO-FbA boundary, including communication, translation, and mediation. Building on lessons outlined by Cash,⁷⁰ boundary organizations can act as intermediaries between the arenas of science and decisions, with specialized roles within relevant organizations for managing the boundary; clear lines of responsibility and accountability to distinct social arenas on opposite sides of the boundary; and a forum in which information can be co-produced by actors from different sides of the boundary. Boundary organizations play key roles in connecting early action challenges to tailored information, helping them move from data to decisions. Program managers must value boundary organizations and understand their struggles when faced with conflicting institutional cultures and disparate metrics of success. New performance indicators should celebrate the roles and unique capabilities that boundary organizations develop to achieve more together. In addition to the typical science metrics of papers published and research dollars generated, new applied science indicators could look to the development and humanitarian sector for new performance metrics. For instance, the performance of scientists working on FbA could be measured by number of programs supported for early warning/early action, number of non-scientists engaged or trained, service on early action protocol development committees, number of value-added EO products integrated into early action and risk reduction in any phase of FbA, or testimonials from humanitarian organizations, among others.

5 Conclusions

As FbA continues to gain momentum and EO science advances, new ways to connect these communities will emerge, not only for anticipating and addressing flood impacts as discussed here, but for other shocks triggered by events as diverse as droughts, El Niño, heat waves, cyclones, volcanic eruptions, cold waves, forest fires, and others. Future work could consider how the EO community can best serve humanitarian FbA in these other areas as well. The EO community

should see great opportunities to advance its fundamental science and applications in response to the FbA community's challenge to design impact-based triggers. Flood early warning systems are still many years away from producing forecasts of impact (advances are underway, e.g., Refs. 71 and 72), but this should not discourage the design of impact-based triggers for early action today. There is an immediate opportunity to collaborate in the FbA design phase so that the more readily available flood variables such as precipitation and streamflow are tied to on-the-ground impact. In the longer term, research and development can help to realize the value for FbA of the intermediate to complex variables such as extent and exposure. To best serve the needs of the humanitarian community, EO scientists can work to tailor their forecasts, understand the challenges beyond the scientific scope, and share information as clearly and openly as possible. As EO scientists and humanitarians work together, boundary organizations that can engage with the two communities will play key roles in facilitating the conversations needed to help turn science into action and ultimately reduce flood impacts.

6 Appendix

Table 4 expands on Table 3. It provides examples of flood early actions that can be used to seed discussions between collaborating EO scientists and practitioners and FbA designers, operators, and evaluators.

Table 4 Example flood early actions.

Early action	Reference	Hazards	Country	Lead time
Train volunteers from Red Cross branches	37	Floods and El Niño	Peru	Years/general preparedness at the beginning of the season
Conduct vulnerability and capacity assessments	37	Floods and El Niño	Peru	Years/general preparedness at the beginning of the season
Create community risk maps	37	Floods	Peru	Years/general preparedness at the beginning of the season
Key actors confirm standard operating procedures	37	Floods and El Niño	Peru	Years/general preparedness at the beginning of the season
Continually update risk maps and identify changing vulnerable groups	12	Floods	—	Years
Recruit additional volunteers	12	Floods	—	Years
Establish new areas of work	12	Floods	—	Years
Work with communities to reduce risk through concrete actions (e.g., reforestation and reinforcement of houses)	12	Floods	—	Years
Provide drinking water for families	5	El Niño	Peru	Seasonal
Assist families in healthcare; raise awareness on water and hygiene	5	El Niño	Peru	Seasonal
Raise awareness with families on hygiene and community health	5 and 37	Floods and El Niño	Peru	Seasonal to three months

Table 4 (Continued).

Early action	Reference	Hazards	Country	Lead time
Establish early warning committees, and equip and train community brigades	5 and 37	Floods and El Niño	Peru	Seasonal to three months
Support storage of seeds and food items	5 and 37	El Niño	Peru	Seasonal to three months
Strengthen and protect homes at risk of collapse for families	5 and 37	El Niño	Peru	Seasonal to one month
Build shelters	37	El Niño	Peru	Three months
Revisit contingency plans	12	Floods	—	Months
Replenish stocks	12	Floods	—	Months
Inform communities about enhanced risk and what to do if the risk materializes (e.g., clear drains)	12	Floods	—	Months
Active chlorine production installations	37	Floods	Peru	One month
Distribute first aid kits	37	Floods and El Niño	Peru	Lead time
Provide communities with temporary housing for families	5	El Niño	Peru	Lead time
Install water collection system at critical points	37	Floods	Peru	Lead time
Build temporary platforms for water bladders (in case of forecast of high impact weather)	37	Floods	Peru	Lead time
Preposition hygiene kits	37	El Niño	Peru	Lead time
Fumigate communities	37	El Niño	Peru	Lead time
Preposition buckets and chlorine tablets	5, 37	El Niño	Peru	One month to One week
Alert volunteers and communities	12	Floods	—	Weeks
Meet with other response agencies to enable better coordination	12	Floods	—	Weeks
Closely monitor rainfall forecasts	12	Floods	—	Weeks
Bag vulnerable items and move storage facilities to higher ground	11	Floods	Uganda	Two to one weeks
Distribute biodegradable trash bags	37	Floods	Peru	Nine days
Coordinate water distribution	37	Floods	Peru	Nine days
Distribute buckets, chlorine tablets, and hygiene kits	5 and 37	Floods and El Niño	Peru	Nine to seven days
Install water reservoirs and temporary water tanks	5	El Niño	Peru	One week
Early evacuation	5	El Niño	Peru	One week
Distribute jerry cans to vulnerable households	11	Floods	Uganda	One week

Table 4 (Continued).

Early action	Reference	Hazards	Country	Lead time
Distribute soap and a 30-day supply of chlorine tablets to vulnerable households	11	Floods	Uganda	One week
Dig trenches around homes to divert water	11	Floods	Uganda	One week
Prepare for evacuation	12	Floods	—	Days
Mobilize volunteers	12	Floods	—	Days
Get warnings and instructions out to communities at risk	12	Floods	—	Lead time
Storm proof houses (cyclone/hurricane/typhoon)	12	Cyclone	—	Lead time
Check emergency kits for batteries, torch, food supplies, and radio (cyclone/hurricane/typhoon)	12	Cyclone	—	Lead time
Inform brigades and community leaders about evacuation alert	37	Floods	Peru	Two days
Evacuate	12	Floods and cyclone	—	Days to hours

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Claire Nauman is a recent graduate of the Earth system science master's program at the University of Alabama in Huntsville, where she served as the NASA/SERVIR Mekong Graduate Research Assistant. She received her BS degree in environmental engineering from the University of Notre Dame. She now works as a water resource engineer at SLR Consulting.

Eric Anderson is a research scientist with the University of Alabama in Huntsville Earth System Science Center, serving as the NASA/SERVIR associate chief scientist and Disasters Theme Lead. He holds an MS degree in Earth system science from the University of Alabama in Huntsville.

Erin Coughlan de Perez is the manager of the Climate Science Team at the Red Cross Red Crescent Climate Centre and an adjunct associate research scientist at Columbia University. She holds a PhD in climate science focused on disaster risk management.

Andrew Kruczkiewicz is a senior associate researcher at the Earth Institute, International Research Institute for Climate and Society, and faculty lecturer at the Columbia University Climate School. He is science advisor at the Red Cross Red Crescent Climate Centre. He holds an MA degree from Columbia University's Department of Earth and Environmental Sciences.

Shanna McClain serves a dual role as manager of the NASA Earth Sciences Division Global Partnerships Program and program advisor on risk reduction and resilience to NASA's Applied Sciences Program. She is a visiting scholar at the Environmental Law Institute. She received her PhD in environmental science and policy from Southern Illinois University.

Amanda Markert is a research scientist at the University of Alabama in Huntsville Earth System Science Center, serving as the NASA/SERVIR Mekong Regional Science Coordination Lead. She received her BS degree in meteorology from Virginia Tech and an MS degree in Earth system science from the University of Alabama in Huntsville.

Robert Griffin is an associate dean of the College of Science, an associate professor in the Department of Atmospheric and Earth Science, and a co-director for the Laboratory for Applied Science at the University of Alabama in Huntsville. He received his PhD in anthropology from Pennsylvania State University.

Pablo Suarez is the associate director for research and innovation at the Red Cross Red Crescent Climate Centre, as well as visiting research fellow at Boston University, and artist in residence at the National University of Singapore. He received his PhD in geography from Boston University.