Empirical Forecasting of Slow-Onset Disasters for Improved Emergency Response: An Application to Kenya's Arid Lands

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Introduction

The ability to forecast the onset, duration and severity of droughts, floods, and disease outbreaks with reasonable accuracy, especially in terms of their prospective human welfare effects, is critical to the design of timely and cost-effective early warnings and emergency response systems that can minimize the suffering of populations adversely affected by such relatively slow-onset events. As the consensus on climate change and its consequences grows, there is an increasing worry that the frequency of climate shocks will rise, with more frequent and serious humanitarian crises and ensuing demand for emergency response.

Given the finite resources allocated for emergency response initiatives, there is growing demand for the development of timely, rigorous, efficient and practical methods of emergency needs assessment. To contribute to this effort, the research reviewed in this brief develops an empirical forecasting model to predict the human impact of slow onset disasters for early warning and emergency needs assessment.

Geographic Focus: Kenya's Arid North

The research focuses on the arid lands of northern Kenya. Largely populated by nomadic pastoralists and particularly vulnerable to recurring shocks such as droughts and floods, the region is well-suited for the purposes of this study. As part of an effort to address the vulnerability of the region's population, the Arid Lands Resource Management Project (ALRMP) has been collecting data in various communities across Kenya's arid districts since 1996. Data collected include detailed household-level information on livestock such as herd sizes, mortality rates, lactation rates, and managed off-take rates, and child-specific nutritional data, specifically mid-upper arm circumference (MUAC) measures.

The ALRMP data are monthly, February 2002 to May 2005, for 54 communities across four districts (Baringo, Marsabit, Samburu and Turkana). The authors supplement the ALRMP data with a rich source of climate and forage availability data collected and produced by researchers of the USAID Global Livestock CRSP Livestock Early Warning System (LEWS) project and its successor, the Livestock Information Network and Knowledge System (LINKS) project. One desirable feature of these data is that they originate with remotely sensed and other data external to communities, so they are information imported into, rather than merely extracted from, the system under study. As changes in livestock fertility, mortality and productivity respond strongly to lagged changes in forage and water availability, use of these variables can improve the precision and forecast lead time of models to predict the human impact of the climate shocks that frequently confront pastoralist communities.

Severe Child Malnutrition and Food Crises

Emergency response to widespread acute food insecurity is largely conditioned by the degree and prevalence of gross malnutrition. Acute food insecurity is typically assessed based on the proportion of children whose anthropometric measure(s) of weight relative to height or age reflect widespread high levels of food stress and acute undernutrition, commonly known as "wasting". Mid-upper arm circumference (MUAC), a superior predictor of child mortality, is one such measure. A MUAC Z-score of less than -2 is widely regarded as an indicator of severe wasting. (A Z-score is a statistical measure of MUAC values relative to a universal reference population. A Z-score<-2 indicates a child more undernourished than 97.5% of children of similar age.) A food crisis might be objectively defined as occurring whenever twenty percent or more of children are severely wasted.

Using this definition of a food crisis, the authors developed a model to forecast child nutritional status (as given by MUAC) based on movements in key explanatory variables several months in advance. Herd dynamics variables (the size of herds, mortality rates, sales and slaughter rates), measures of food aid, as well as variables capturing rainfall and forage availability were used to predict the prevalence of MUAC, and consequently, the likelihood of a food crisis.

Early Warning: Forecasting Food Crises

Effective response to food crisis requires early warning of emergency conditions so as to mobilize resources. Two forecasting models are developed: a one-month forecast and a three-month forecast for the prevalence of severe wasting. While a one-month forecast will typically be more accurate, the short lead time leaves little leeway for food security managers to make effective use of the forecast. The longer forecast horizon afforded by a three-month forecast, however, comes at a cost of diminished accuracy. The inverse relationship between forecasting horizon and forecast precision forces an operational tradeoff between models. Different end-users will favor different characteristics and thus different models.

Figure 1 shows the monthly forecasts generated for January 2004 to May 2005 compared against subsequent, actual values. Figure 1a presents the forecasts superimposed on the full sample of actual values - the proportion of children with MUAC Z-scores<-2. The values are smoothed to highlight trends, especially the considerable variability in the prevalence of severe wasting. A food crisis existed during and following the 2000-1 drought, then re-emerged in 2003-4 in these northern Districts, then began reappearing last year. Figure 1b highlights forecast precision by zooming in on the forecasting period and plotting the unsmoothed actual values against the one- and three-month ahead predictions generated by the forecast model.

Figure 1: Fraction of Children with MUAC Z-score<-2: One and Three Month Forecasts 1a) Full Sample Period 1b) Forecast Period 0.34 Actual (Smoothed) Actual One-Month Forecas 0.32 One-Month Forecast Three-Month Fore Three-Month Foreca Food-Crisis Threshold 0.30 Food-Crisis Threshold with MUAC Z-Score <-2 0.3 0.28 with MUAC Z-Score 0.26 0.30 Food Crisis Food Crisis 0.24 0.2 0.22 0.20 0.15 0.16 Jan02 Jan04 Apr04 Oct04 Jan05 Apr05

Three key points emerge from the figures. First, recall that a food crisis is defined as the state where 20% or more of children exhibit MUAC Z-scores <-2, reflecting severe wasting. Thus defined, the sample sites regularly experienced food crisis over the January 2000 and May 2005 period, but crisis was not constant. Second, the forecasts trace the actual values quite well, and seem to improve with time, as additional data improves forecast precision. Third, while the one-month forecasts outperform the three-month forecasts, the differences are not substantial.

Making Practical Use of the Forecasts

How can policy makers use such forecasts of food crisis to make critical emergency response decisions? Furthermore, how confident can policy makers be in decisions they make based on the model's recommendation? To answer these questions, the authors construct a plausible scenario in which the model's forecasts can be used, and offer a measure of forecasting performance.

Consider the case of an organization tasked with responding to food crises. This sort of forecast model can provide early warning of a food crisis at a particular site in the form of predictions of specific levels of severe wasting. The organization can decide on the minimum likelihood of food crisis required before they deploy a costly aid package. This policy decision will depend on a range of variables, including the availability of funds to support emergency response, the operational costs of serving the affected population, logistical considerations regarding access, etc. Once such a threshold likelihood is chosen, the organization might initiate response if the forecast of food crisis reaches or exceeds the pre-determined minimum threshold.

To test forecast performance operationalized in such a way, a minimum response threshold was arbitrarily set at 66% (that is, the organization deploys aid whenever the forecast predicts a 66% or greater likelihood of food crisis). Defining emergency response when there is actually a food crisis, or no response where there is no food crisis, as 'correct' decisions, one can then calibrate forecasting performance by calculating the

fraction of correct decisions generated by this decision rule in combination with the forecast model.

The results, presented in Table 1, are quite striking. Decisions based on forecasts are likely to be correct more that 75% of the time - quite an impressive forecast performance. Moreover, the fairly small depreciation in performance as the forecasting horizon increases shows that these models can be used fairly accurately to provide a reasonable, three-month early warning to help with emergency response to mitigate the consequences of impending crisis.

Table 1:Model Performance in Generating Correct Decision for Famine Response

Fraction of Correct Decisions	
One Month	Three Month
0.786	0.756

Conclusions and Policy Implications

Based on data collected from primarily pastoralist communities selected across four districts in Kenya's arid north, the authors have developed an empirical forecasting model that can predict, with reasonable accuracy and at least three months in advance, the expected human impact of slow onset shocks such as drought. Information on herd composition and herd management, climate and forage availability and food aid flows enable reasonably accurate three-month-ahead forecasting of child nutritional status, specifically severe wasting reflect in very low MUAC levels, with impressive precision. Longer lead forecasts may also be feasible and warrant investigation.

These forecasts were generated from a relatively small subset of variables that ALRMP regularly collects, augmented by data collected routinely by LEWS/LINKS. These data are not overly restrictive or costly to collect. Limiting data collection to these set of variables, collected consistently through time, might offer a cost-effective way to provide effective early warning to policymakers and emergency response professionals. The precision of these predictions appears sufficiently high that delays in acting on this information due to concerns over forecast accuracy should be limited. However, there remains work to be done to establish how best to communicate this information in as clear and timely a fashion as possible to appropriate audiences.

The authors recommend that the model be adapted as an effective famine early warning tool. As the model can be easily and regularly updated with new information that should continuously increase its forecast performance, a premium should be placed on developing standardized collection procedures and failsafe methods for entering, identifying and storing the necessary data. Such a forecasting model could prove an invaluable tool for early warning and emergency response to food crises.