# Poverty Mapping in Uganda by integrating socio-economic, environmental and satellite data David Rogers<sup>1</sup>, Thomas Emwanu<sup>2</sup>, William Wint<sup>1</sup>, Francesca Pozzi<sup>3</sup> and Tim Robinson<sup>4</sup>



<sup>1</sup> University of Oxford <sup>2</sup> Uganda Bureau of Statistics <sup>3</sup> IGAD Livestock Policy Initiative, FAO <sup>4</sup> Pro-Poor Livestock Policy Initiative, FAO



### Abstract

In order to reduce poverty we must first describe, explain and predict its spatial distribution over large areas with as high a level of local accuracy as possible. Poverty maps are traditionally produced by exploiting links between census (wide area) and survey (smaller area coverage) data. The detailed relationships found within the survey data are extended to the census data that must share some predictor variables in common with the survey data. Both census and survey data tend to be socio-economic in nature; the mapping thus exploits the internal correlations within potentially strongly correlated data sets - one 'measure' of poverty is often correlated with another. Rather than look at the correlates of poverty, we should like to identify its causes. We suggest that poverty is multi-dimensional and that many of its dimensions are environmentally related; people are poor because they are unhealthy, or under-fed, or without access to fuel and water etc. Each of these is environmental in some way or other, and a correct approach to reducing poverty might be first to identify its (environmental) causes. We have attempted to do this with survey data from Uganda and environmental data derived from multi-temporal satellite imagery that measures land-surface conditions and processes (temperature, rainfall, vegetation growth etc.). The same satellite data have already been used to understand the distribution of farming systems throughout Africa and to predict the distribution and intensity of insect and tick carriers of a variety of diseases, and the incidence and prevalence of the diseases they transmit.

In this analysis therefore we examined to what extent satellite data (as a proxy for environmental conditions) are correlated with household survey data. Whilst correlation obviously does not automatically imply causation, we suggest an environmental approach is more likely to reveal causes than will the traditional approach of small area mapping using census and survey data. However, it is first necessary to establish the relative predictive accuracies of the traditional and environmental approaches. This poster goes some way towards reaching this objective.

## Methodology

The predicted (household income) and predictor (ground-based and satellite) variables are linked via discriminant analytical methods. This approach is shown diagrammatically for the simple case of two-group discrimination in the panel below.



The data and method define the discriminant axis (linear in this case) that distinguishes the two groups. Further observations can be assigned to one or other category depending on which side of the discriminant axis they lie. In the case of the Ugandan data, income was divided into ten income categories and the predictor variables that best discriminate these categories were selected in a step-wise inclusive manner that maximised the fit of the model to the data (using kappa, the index of agreement between predicted and observed categories). In this case, the discriminating axes between groups were no longer linear; otherwise the principles were the same as illustrated.

Models were constructed at a variety of different spatial resolutions, from 0.01 to 1.0 degree (approximately 1.1 km and 110 km at the equator). Both predictor and predicted variables were averaged at each spatial resolution before the models were constructed.

For more information contact

Tim Robinson

Pro-Poor Livestock Policy Facility (PPLPF), FAO Via delle Terme di Caracalla, 00100 Rome, Italy Email: Tim.Robinson@fao.org Input data

The poverty data are derived from breakdowns of food expenditure from the 2002-2003 Ugandan National Household Survey (UNHS), which covered 9,711 households in 973 communities (left-hand panel below). This survey was also designed to be integrated with the 2002 National Population and Housing Census for detailed poverty mapping using small area techniques. Ground-based predictors used in the analysis included human population density, livestock density (second

Ground-based predictors used in the analysis included human population density, livestock density (second panel below), 'distance' to markets (calculated as time to travel to towns of a certain size – third panel below), and the probability of occurrence of three important tsetse species.

Satellite-derived predictor variables included an elevation surface and data layers arising from temporal Fourier analysis (TFA) of data from the Advanced Very High Resolution Radiometer (AVHRR) on-board the NOAA series of oceanographic satellites. TFA outputs include separate images recording the mean, maximum and minimum of each satellite channel, the amplitude and phase (= timing) of annual, bi-annual and tri-annual cycles that collectively define habitat seasonality of land-surface and air temperatures, Cold Cloud Duration (a proxy for rainfall), Vapour Pressure Deficit and vegetation growth (Normalised Difference Vegetation Index or NDVI – right-hand panel below).



## Results

The maps below show the results of the predictions for a range of different spatial resolutions, along with a graph of the percentage explained variance ( $r^2$ ) at each spatial resolution. For comparison, the last map panel shows the results of applying the 'traditional' small area mapping approach to data from an earlier survey carried out in 1992, which estimated welfare at the county level. As is the case with the small area mapping approach, the accuracy of the environmental approach to poverty mapping increases as spatial resolution decreases. The environmental method explains more than 50% of the variance in the poverty data at a spatial resolution of about 0.20 to 0.30 degrees – a result comparable with that of the traditional methods.

NOTE: the classes of expenditure in the maps below are the same as those of the households in the figure above



### Conclusions

• Environmental data derived from ground surveys and satellites appear to be at least as good as census and survey data at describing the spatial distribution of poverty in Uganda.

The accuracies of both methods are scale-dependent. The environmental approach gives satisfactory results at spatial scales of c. 20km and above (r<sup>2</sup> of 50% or greater).
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• The environmental approach is more likely to identify the causes of poverty than is the traditional approach, and thus leads more naturally to appropriate interventions.

• It is likely that the ultimate causes of poverty vary locally; the environmental approach can establish the environmental correlates of these causes which may include soil fertility, agricultural production, health and the availability of fuel and water. Appropriate, targetted intervention can thus be designed once the causes are identified from their environmental correlates.