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Measuring vulnerability to poverty allowing for agricultural and non-agricultural risks: Evidence from Tanzanian household data

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Introduction

- Many studies have attempted to measure households' vulnerability to poverty: future probability of falling into poverty
- A household's vulnerability is usually calculated using its probability distribution for welfare indicators such as income and consumption
- However, it is sometimes difficult to obtain country-wide long-term panel data, particularly in developing countries
- Chaudhuri, Jalan and Suryahadi (CJS: 2002) proposed a method to estimate welfare distribution for each household using cross-sectional data
- The CJS method has widely been utilized owing to its less stringent data requirement
- The CJS method has been criticized, however, for its assumption that intertemporal welfare variation of a household can be approximated by crosssectional variation of household welfare

Objectives

- We propose a new method to measure vulnerability that can be implemented with cross-sectional data and capture the risk faced by each household more accurately
- Our goal is to obtain household-specific welfare distribution that allows for agricultural and non-agricultural risks, then calculate the vulnerability
- To show the validity of our method, we predict poverty status using both our method and the CJS method and compare accuracies of the predictions

Model

- Farmers choose the labor allocation between agriculture and non-agriculture to maximize the expected utility
- Both agricultural and non-agricultural incomes contain risk factors: climate risk and unemployment risk, respectively

$$\max_{L_{Ai},L_{Ni}} \quad \ln Y_i \left(L_{Ai} \left(L_{Ni} \right), X_i, S_i, r_i \right),$$
s.t.
$$L_{Ai} + L_{Ni} = \overline{L}_i$$

Y: household income, L: working hours, X: exogenous factors that affect income, S: asset holding, r: risk factors; A: agriculture, N: non agriculture

• This setting is consistent with Ito and Kurosaki (2009) who showed that households in India with a higher agricultural climate risk are more likely to allocate their labor to non-agricultural activities

Specification of the income variables:

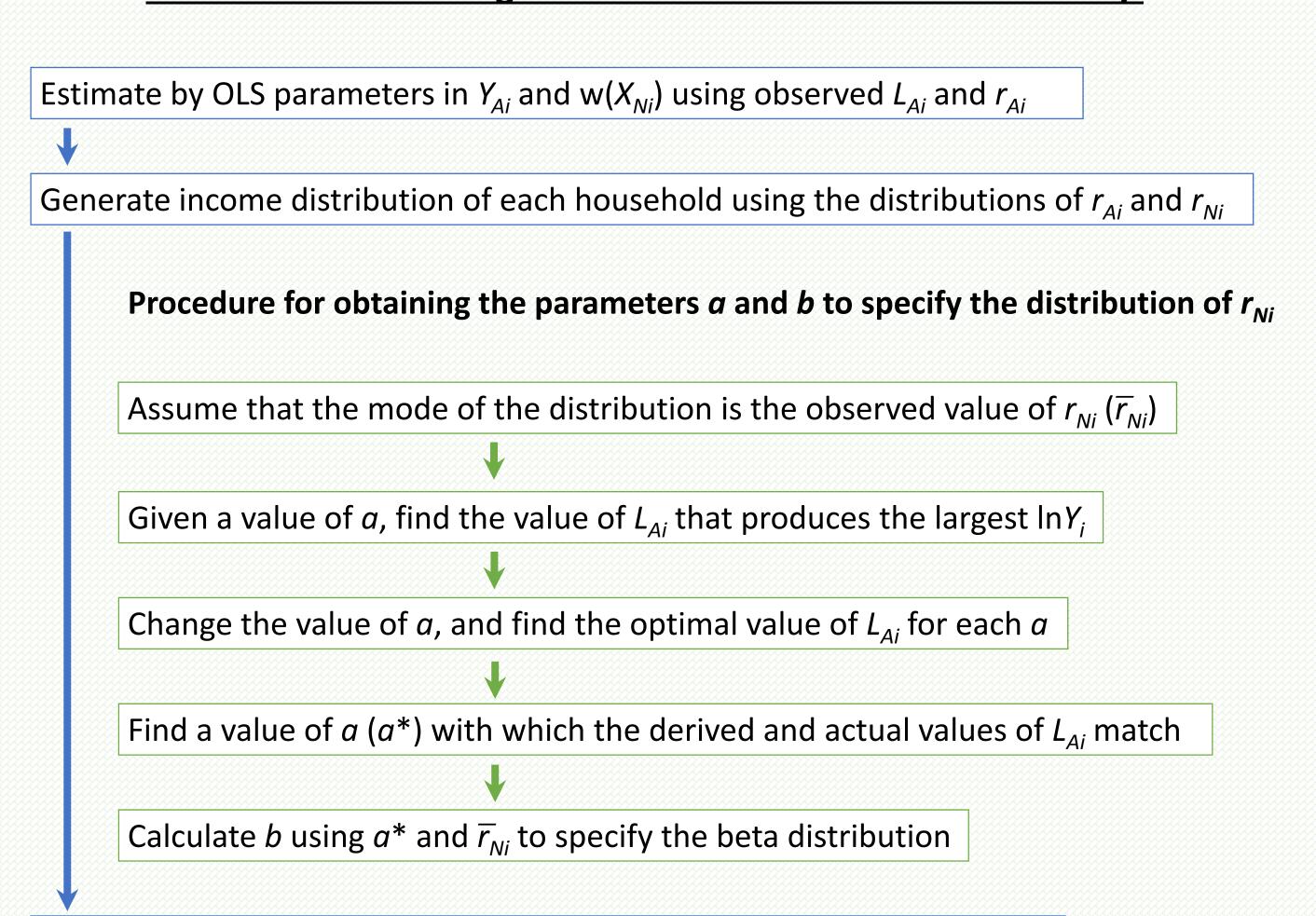
$$Y_i = Y_{Ai}(L_{Ai}, X_{Ai}, r_{Ai}) + Y_{Ni}(L_{Ni}, X_{Ni}, r_{Ni}) + S_i,$$
 $Y_{Ai} = \alpha + \beta L_{Ai} + X_{Ai} \gamma + \delta r_{Ai}, \quad r_{Ai} \sim \text{precipitation of recent 100 years,}$
 $Y_{Ni} = w(X_{Ni}) L_{Ni} r_{Ni}, \quad r_{Ni} \sim \text{Beta}(a,b)$

w: wage function (determined outside of this model), Greek letters: parameters

Numerical analysis

- Because the above optimization problem cannot be solved analytically, we solve it numerically using the Monte Carlo approach
- We propose the following procedure to derive the income distribution of each household using cross-sectional data and then calculate its vulnerability

Procedure for deriving income distribution and vulnerability



Using the generated incomes, calculate each household's vulnerability

Application to Tanzanian household data

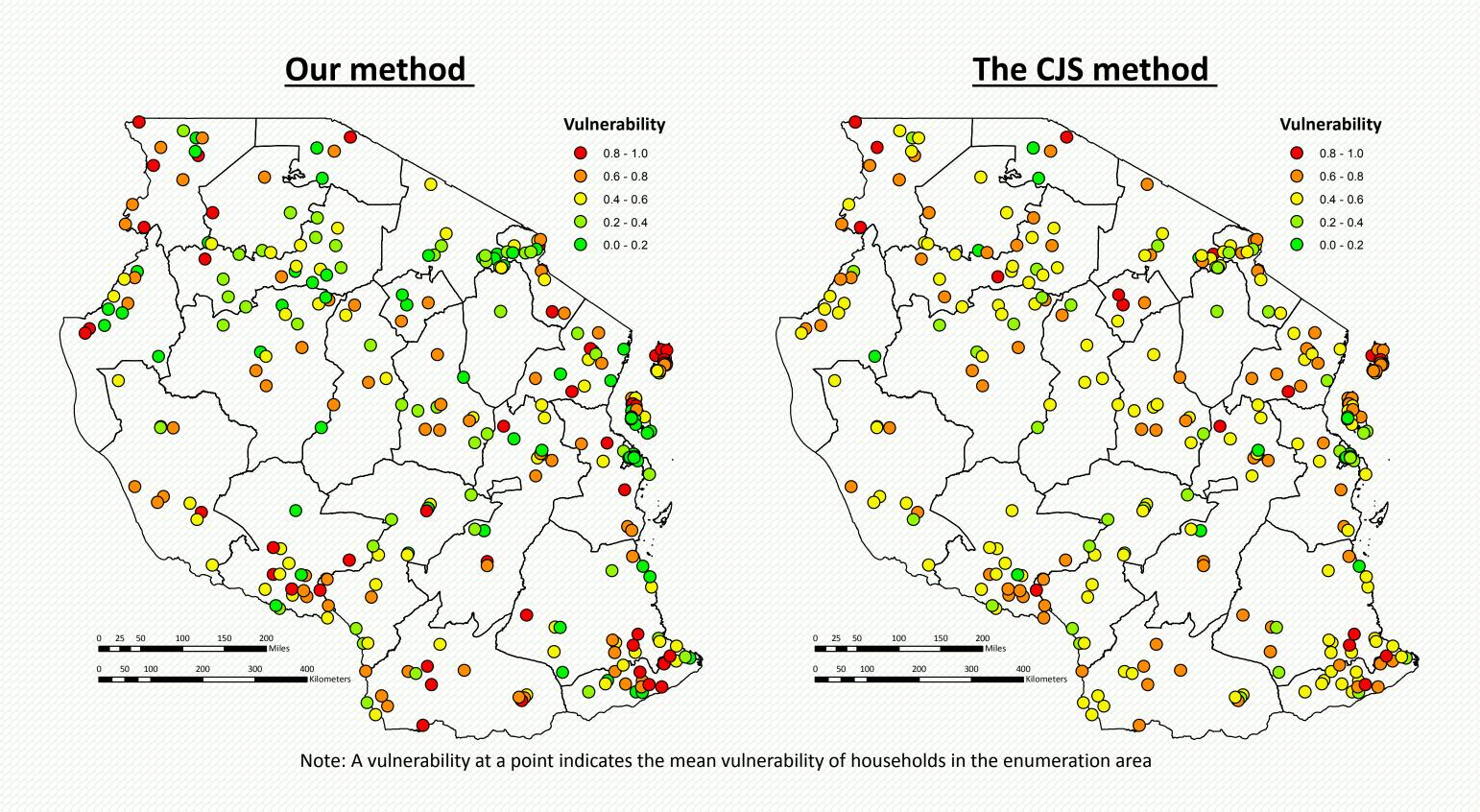
- We applied both our method and the CJS method to the Tanzanian LSMS data
- We calculated vulnerability of each household using the 2008-09 data and predicted its poverty status in 2010-11 based on this value
- We then compared the accuracies of prediction based on the realized income included in the 2010-11 data

Results

- Our method produced a vulnerability index with higher accuracy and fewer false-positive cases than that obtained by the CJS method
- Compared to the CJS method that assumes normally distributed vulnerability, our method gave more "deterministic" evaluations, i.e. index values close to zero and one
- The results were robust to the specification of the utility function and the location of the poverty line

		Our method			The CJS method		
		Vul (vul >.5)	Non-vul (vul <.5)	Total	Vul (vul >.5)	Non-vul (vul <.5)	Total
Actual	Poor	520	239	759	564	195	759
	Non-poor	257	486	743	355	385	740
	Total	777	725	1,502	919	580	1,499
Accuracy				0.670			0.633
Score (pov)				0.691			0.635
Score	e (non-pov)	nemannumumaniumachumatumumumunanananantiiniili katulisiinakkin		0.645		umaenomaenomaenubianasikoinaenomaenuenuenuenomaenoenaenaenaen	0.522
Score	e (total)			0.668			0.580
XXXXXX	er var stad oca v sens de etak sad oca v sal stad oca v sens de etak sed stad stad stad stad stad stad oca v s Oca var stad oca v sens de etak stad oca v sal stad oca v sens de etak sed stad stad stad stad stad oca v sed o	a oran suan tuno cara san suan sua suan sun sua suan suan su	ari sarona kiang dorawi saronan sari sari sa tsari kiang ari sari sari sari sari sari kang daki kari sari sari Sari	OR BETTS BETTS AT THE OLD THE SEPTEMBERS AND COME THE STORE CARE THE SEPTEMBERS AND THE CARE THE SEPTEMBERS AND CARE THE SEPTE	DELAT GART KATO BAT KATO BADELAT KATO BAT KADO BERKESADO BERKESADO BAT KATO BAT KATO BAT KATO BAT KATO BAT KAT Baran baran kato bat kato badelat kato bat kado bat kato	DERR KLANDERT KRONDERERKTING DER KLANDERFORMEN EINE DER KLANDERFORME KLANDERFORMEN.	JAKEN BETTAD DAN PARTIDA PARTIDA DAN BETTAD DAN PARTIDA PARTIDA PARTIDA PARTIDA PARTIDA PARTIDA PARTIDA PARTIDA

Note: Score (pov) = $(1/I)\sum_{i}^{I} vulnerability_{ki}$, $i \in \text{poverty}$, $k \in \{\text{Our method}, \text{The CJS method}\}$, Score (non-pov) = $(1/J)\sum_{j}^{J} (1-vulnerability_{kj})$, $j \in \text{non-poverty}$, Score (total) = $(1/(I+J))(\sum_{i}^{J} vulnerability_{ki} + \sum_{i}^{J} (1-vulnerability_{kj}))$



Conclusion

 Considering both agricultural and non-agricultural risks in an explicit manner can improve the accuracy of vulnerability index, even when only crosssectional data are available

References

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