Modelling intrahousehold distribution to study gender inequality & individual poverty

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The MS in Economics & Development

- A great program!
- A honor to present some research to you
- A particular conjunction given the topic and the fact that this is the international day of women's rights

- Usual poverty measurement based on monetary concept at household level
 - ex: *per capita* consumption against a poverty line
 - at best, use of equivalence scales to correct for different needs and economies of scale
 - but no information on who gets what
- Evidence of some degrees of intrahousehold inequality
 - Start with nutrition data (ex: Haddad & Kanbur 1990)
 - Inequity may concern both children vs parents or gender inequality
- Ignoring it may often leads to policy mistargeting
 - Especially an **exclusion error** when poor individuals live in non-poor households
 - Broadly documented (Brown, Ravallion & van de Walle, 2019)

How to identify poor individuals in households?

Hard task. Here are a few approaches (from less to more similar to ours):

1-Vignettes

- Ask respondents how similar they are to several types of households
- How identity of decision-makers affect some child and household outcomes (Bernard, Doss, Hidrobo, Hoel & Kieran 2018)

2-Final say variables :

- Who decides about what on a 1-10 scale (subjective)
- Often available in surveys (Reggio 2011, Bergolo & Galvan 2018,...)
- In some contexts (ex: decide about expenses), it might reflect delegation more than power (Baland & Ziparo 2018)

3-Person-specific expenditure (usually health, education, nutrition, etc.) :

- Usually, provide a trace of intra-household inequality
- But specific goods that involve risk, uncertainty and dynamic behavior (ex: maybe one is deprived today because his household currently invests in his future) Personal expenses on health/education are hence difficult to interpret
- Sometimes, more instantaneous measures, for ex: individualized food (Hoddinott & Skoufias 2004; Brown, Calvi & Penglase 2021, etc)

4-Empowerment indices :

- Multidimensional deprivation indicators (IFPRI index, UNICEF indices, ...) based on education, health, personal income, survey questions on empowerment, etc.
- Cumulate the above material with the aim to define a person's living condition (but some of it is household-based: shelter, water, etc)

None of the above provides

- a comprehensive view of short-term individual control over resources
- and its implication for individual poverty

Moreover, fully individualized expenditure is rare and costly (cf. Bargain, Lacroix & Tiberti, 2021)

What to do ? -> attempts to model and identify resource allocation

- Early literature on collective models (from Chiappori 1988 to Bourguignon et al 2009)
 - Test basic rationality: **efficiency** (debatable!)
 - Identify the marginal sharing rule (how spouses share an extra \$)
- Interesting results on **distribution factors**
 - factors affecting balance of power only, Ex: divorce rules, sex ratio, etc
 - Maybe most interesting is natural experiments inducing a **pure bargaining effect**
 - For instance Lundberg, Pollak & Wales 1997, Ward-Batts 2008, Bradbury, 2004, etc.)
 - \rightarrow exogenous variation in spouses' control over unearned income
 - \rightarrow emphasize the role of income controls on individual consumption

- More recently, attempts to recover the full resource sharing process
- Browning, Chiappori, Lewbel (BCL, 2013)
 - More rigorous than equivalence scale (indifference scales)
 - Living with others implies: sharing resources and consuming jointly
 - Transparent approach to estimate the sharing function and economies of scales for each good
- Conceptually interesting but tedious implementation
 - Use panel data or repeated cross section for many years (price variation)
 - Identification comes from single individuals, used to estimate individual Engel curves for individuals in couples
 - \rightarrow assumption of preference stability across marital/demographic status), i.e. Rothbarth's flavor

- More tractable approaches
 - Lewbel & Pendakur (2008): require only one cross-section
 - economies of scale simply estimated as an average effect over all goods
- Extensions to households with children
 - Bargain & Donni (2012): use single data & the broad stability assumption as BCL
 - Seen as not adapted to developing countries (even though we did for Cote d'Ivoire: Bargain, Donni, Kwenda, 2015)
 - An approach that does not require singles: Dunbar, Lewbel, Pendakur (2013) but use specific functional form
 - & alternative identification assumptions (restrictive!) on preferences
 - & does not identify scale economies

- Notations:
 - Individuals of type: i = f, m, c
 - Household composition $n = (n_f, n_m, n_c)$ with n_i the number of type-*i* persons in the household
 - W_i : household **budget shares** for goods that can be assigned to type-*i* persons (ex: women's clothing)
 - w_i : budget share for that same good in the budget of a type-*i* person
 - *x* : household consumption
 - *p_i* : **resource share** (i.e. share of household consumption accruing to type-*i* person)
- Household budget shares can be written with basic structure (here for women):

$$W_f = n_f \times p_f \times w_f(p_f x)$$

Example: take a household with only $n_f = 1$ woman; she controls $p_f = 40\%$ of household resources; she dedicates $w_f = 20\%$ of her resources to female clothing; then the household dedicates $W_f = 8\%$ of its budget to female clothing.

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Super easy (compared to early models) !

And does not really require efficiency ... just to assume there is a sharing rule!

The difficulty remains to identify function p_i

What do we observe?

- Budget *x*
- Potential determinants z of sharing function $p_i(z)$
- Assignable consumption (female, male and child clothing), hence W_i for i = f, m, c

Bargain & Donni (2012):

• In families:

 $W_i = n_i \times p_i \times w_i(p_i x)$

 $W_i = w_i(x)$

• The intuition is that for singles:

• So we can estimate w_i () on singles, and assuming preference stability, recover p_i in families

• More info means more identification, i.e. recovering a summary parameter for scale economies

Dunbar, Lewbel & Pendakur (2013)

• Start from Piglog utility, which gives individual Engel curves of the form:

$$w_{i,n} = \alpha_{i,n} + \beta_{i,n} \log(p_i x)$$

for household of composition n, so that

$$W_{i,n} = \{p_{i,n}\alpha_{i,n} + p_{i,n}\beta_{i,n}\log(p_{i,n})\} + p_{i,n}\beta_{i,n}\log(x)$$

• Then, alternative identifying assumptions:

SAT: "Similarity across types": $\beta_{i,n} = \beta_i$ (individuals have same slope across demographic groups n for households with children only)

> identified but instable (using singles would bring more information)

SAP: "Similarity across persons": $\beta_{i,n} = \beta_n$ (men, women, kids have same slope within each demographic group)

> relatively strong assumption (but tested in a few studies)... what we use hereafter

• For each *n*, take the derivative for all three types:

$$\frac{\partial W_{f,n}}{\partial \log(x)} = n_f \times p_{f,n} \times \beta_{f,n} \\ \frac{\partial W_{m,n}}{\partial \log(x)} = n_m \times p_{m,n} \times \beta_{m,n} \\ \frac{\partial W_{c,n}}{\partial \log(x)} = n_c \times (1 - n_f p_{f,n} - n_m p_{m,n}) \times \beta_{c,n}$$

- SAP means that: $\beta_{i,n} = \beta_n$ for all i = f, m, c
- So, for each *n*, we have 3 equations and 3 unknowns $(p_{f,n}, p_{m,n} \text{ and } \beta_n) \rightarrow$ exact identification

Resource shares $p_{i,n}(z)$ depends on n and various factors z

- additional demographic factors, ex
 - proportion of boys among children \rightarrow gender discrimination
- distribution factors related to women's employment opportunities
 - or control over labor and nonlabor income, hence on **redistributive policies**
- distribution factors related to culture/norms
 - **Traditional norms** may influence women's right, child treatment, etc. (and may be accounted for in policy/targeting design: cultural 'tags').

→ applications:

- Individual incidence for policy making (Uruguay)
- Individual poverty & culture (Ghana, Malawi)
- Who contributes to child costs (UK)

3.1 Individual incidence for policy making

- Context: **women's financial power** expected to improve their condition and children's (Doss, 2006; Hoddinott and Haddad, 1995; Lundberg et al., 1997)
- Hence cash transfers often targeted at women

(e.g. Handa et al., 2009)

• CCT granted to women indeed show positive effects on child-related expenditures

(Attanasio and Lechene, 2002; Akresh et al., 2016; Benhassine et al., 2015; Bobonis, 2009; Haushofer and Shapiro, 2016; Armand et al., 2020)

- Encouraging, but little is know about how gender-targeted tax-benefit instruments affect intra-household resources sharing (also true for individual earnings)
 - Worse scenario: a gender-targeted transfer might be shared according to the "usual" sharing rule of a household (or worse: backlash)
 - Best scenario: she keeps it all for her and the kids

Reality probably in-between \rightarrow transfers might disproportionally benefit women, but **by how much?**

3.1 Individual incidence for policy making: natural experiments

- Rare possibilities to combine structural model and experiments
 - Households randomly receive a transfer via the wife
 - Ex. of PROGRESA: Tommasi (2019), Sokullu & Valente (2020) and De Rock, Potoms & Tommasi (2020)
- More frequent: combine structural model and natural experiments
 - DD approach in Borga & D'Ambrosio (2021)
 - Here **RDD**: Bargain & Colacce (2022): focus on the *Asignaciones Familiares-Plan de Equidad* in Uruguay (AFAM-PE), a gender-based targeting CCT program
 - Generate a **discontinuity** (used in Bergolo & Galvan 2018 who focus on final say data)
- Implementation
 - Encuesta Nacional de Gasto e Ingreso de los Hogares (ENGIH)
 - Score variable S and eligibility threshold \overline{S}
 - z will include smooth function of S and treatment variable (ITT): $T = 1(S > \overline{S})$

3.1 Individual incidence for policy making: natural experiments

Around 6 point increase in female+child score	Marginal effects on women+children's share					
		linear	quadratic	cubic	spline	
	Eligible	0.058 **	0.057 **	0.056 **	0.056 **	
		(0.026)	(0.026)	(0.026)	(0.026)	
Effect driven mainly by rural households	Eligible x urban	0.036	$0.045 \ *$	0.043	0.039	
		(0.028)	(0.028)	(0.028)	(0.028)	
	Eligible x rural	0.087 ***	0.090 ***	0.090 ***	0.088 ***	
		(0.033)	(0.032)	(0.031)	(0.032)	

3.1 Individual incidence for policy making: natural experiments

Bargain & Colacce (2022): female+child consumption share

Bergolo & Galvan (2018): « Who decides on food expenses: woman decides"



- Surge of research on **culture and ethnic norms** in economics (Baland et al., 2020, Nunn, 2020, Bau and Fernandez, 2022, Giuliano, 2020)
- Often with gender focus and role of crucial dimensions such as education (Dessy, Tiberti, Zoundi, 2022)
- Possible that culture explain a substantial part of within-household inequity
 - if this is the case, policy targeting (individualized PMT) could be improved by 'cultural tags'
 - i.e. observable demographics/traits associated with prevalence of poverty among specific persons
- Some evidence on women's resource shares
 - Dowry practice (Calvi and Keskar, 2021) in India
 - Ancestral post-marriage residence norms
 - → Aminjonov, Bargain, Colacce, & Tiberti (2023) on Ghana and Malawi

- We combine micro data
 - Ghana Living Standards Survey 2016/2017
 - Malawi Integrated Household Survey 2016/2017

with ethnic- or language-specific ancestral norm (Murdok's Ethnographic Atlas)

• We estimate the model with a Patrilocal dummy in z :

Marginal effects on per-woman's resource share								
	Ghana		Malawi					
Household type:	children, women and men	women and men	children, women and men	women and men				
Patrilocal (=1)	-0.026 *** (0.004)	-0.018 *** (0.005)	-0.032 *** (0.007)	-0.041 *** (0.010)				
in % of women's share:	-11%	-6%	-11%	-11%				
% of patrilocality	0.675	0.595	0.170	0.169				
Ν	6204	1552	7462	967				

- Rare evidence of gender-age discrimination in terms of intra-hh resource allocation (Calvi, 2020)
- So we also estimate age-heterogeneous effects
- -> growing influence on decision-making with life experience



- Informal check : for working women, use the question on who control the wife's earned income
- overall trend is compelling of women's authority strengthening with age, especially in a context where women have more responsibility due to kinship traditions



Implication for poverty



Implication for poverty

Similar results for India regarding dowry practice (Calvi & Keskar, 2021)

3.3 Who contributes to child cost?

- Extended model:
 - BCL 2013 with children \rightarrow Bargain, Donni, Hentati (2022)
 - Using price variation over many years
 - This way, identifying 'price effects' that correspond to scale economies of different goods (ex: consume the car 50% of the time with your wife \rightarrow 'actual' price divided by 1.5)
- Evidence from the UK
 - UK Family Expenditure Survey over the period 1978-2007
 - Clothing is used as an exclusive good
- Identification:
 - Sharing rule **depends on total resources**
 - Scale economies are identified
 - Contribution of each parent to child resources!
 - .. and assessment over time

3.3 Who contributes to child cost?

Results:

- children command from 14% (1 child) to 11% (families of 3) per child
- resource shares relatively balanced between spouses when there are no children
- however, women's contribution to children is around 40% larger than men's
 - makes their share lower than their husband's in large families.
- women's shares increase slightly, but significantly, with total expenditure
 - contributes to raise child resources since women are bigger providers

3.3 Who contributes to child cost?

With increased living standard **over time** :

- non-negligible redistribution from men to women
- progress in education levels plays a role in the reduction of intrahousehold inequality over time

Total, between- and within-household decomposition of trends in the variance of log expenditure. Shaded area represents 95% confidence bounds.

Conclusions

- Applications are motivating, and the overall approach seems promising
- Yet, empirically, it looks that
 - we are mainly able to capture the effect of some determinants (culture, policy)
 - slightly less to estimate the full shares in a precise enough way
- More validation needed
 - Comparing actual and estimated shares (Bangladesh : Bargain, Lacroix & Tiberti, 2021)
 → next for rich countries but accounting for scale economies (WTP)...
 - Also against nutrition data (Brown, Calvi, & Penglase, 2021)
 - **Comparing gradient of the shares** (traces with the UK experiment?)
- More creativity on exclusive/assignable goods
 - experiments on how parents decide on child vs adult goods

(Boutin & Filipkowki, 2022, Dizon-Ross & Jayachandran, 2022, Lichand & Thibaud 2020, Cherchye, Chiappori, de Rock, Ringdal & Vermeulen, 2021)

Thanks a lot

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Discussion

- Implementation hinges on somewhat strong assumptions
 - Here SAP (Dunbar, Lewbel & Pendakur): reducing heterogeneity ..at the source of the collective model)
 - Recent attempts to make things more tractable via linearization of the model (Lechene, Pendakur & Wolf, 2022): simplification of a simplification (!) and sharing rule determinants not recoverable
 - Bargain & Donni : requires singles & broad preference stability, but possible to use all the variation in demographic groups ?
- Exclusive goods for identification: not many choices
 - Clothing : one of the rare assignable goods in standard surveys
 - Pass tests (Bargain, Lacroix & Tiberti, 2021) but may vary across cultural contexts
 - Need for creative views regarding assignable goods
- And possibly need for better identification strategies that work in various contexts
 - Also confront the different approaches

(for that, we need to observe the sharing rule + a large dataset)