# Longitudinal research: Opportunities and Challenges

The benefits of panel data in an imperfect (data) world

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### Purpose

- 1. Explain why longitudinal data (rather than cross-section data) is so important to understanding the *dynamics* of poverty
- 2. Show how we can use these data to evaluate policy, using examples from Young Lives research
- 3. Outline plans for our third round of data collection later this year



# Young Lives panel data design



Longitudinal design..., 2 cohorts (8,000 born 2001-2) (4,000 born 1994-5)

... for 15 years
Collect information every 3-4

years
Two rounds of data already available

Different to national crosssection surveys (e.g. Census, Demographic Health Survey)



### Two existing rounds of panel data

- · Child, Household and Community level data
- Household data: similar to other cross-sectional datasets (e.g. LSMS) but with questions to the caregiver on psychosocial competencies, social capital
- · Detailed time use data for all family members
- Child-level information on anthropometrics and from the caregiver
- Child testing of cognitive achievement (language, maths)
- Directly asking the child! School & Work likes and dislikes, psychosocial
- Also have 2 rounds of qualitative data



### Third round of data 2009

- · Children are now 8 and 15
- Self-administered questionnaire (aged 15)
- More on educational history, psychosocial indicators, work life, social capital
- Focus on Social Protection: detailed modules on e.g. PSNP in Ethiopia, in order to evaluate programme impact
- Linking to existing data where possible school data in India and Peru, job card in India
- · School survey to complement existing data



### Benefits of Poverty Narratives

- Cross-section data can have errors of inclusion and exclusion for many reasons
- Repeated cross-section can tell us the proportion of poor each year
- Panel data can tell us whether it is the same people who are poor each year
- Allowing us to analyse poverty dynamics, and chronic or persistent poverty



# Example: Chronic poverty in Ethiopia

- Over ten years, poverty was measured five times
  - Poverty rate fell from 40% in 1994 to just over 20% in 2004
  - 18% of households were never poor during this time
  - The majority (45%) were poor once or twice
  - 30% poor 3-4 times
  - 7% always poor
- · Different policy implications



### Multidimensionality of poverty

- Early nutritional deprivation can have lasting negative impacts later on, in terms of:
  - grade completion
  - learning
  - psychosocial competencies (pride, sense of inclusion, agency, aspirations, etc.)

See Dercon (2008), Sanchez (2008), Le Thuc (2008)



# Stunting at age 7: its impact at age12

Comparing stunted versus non-stunted children

	Ethiopia	India	Peru	Vietnam
Grade deficit (in years)	-0.9*	-0.3*	-0.5*	-0.4*
Writing skills (deficit in %)	-18.1*	-7.0	-13.4*	-6.8
Reading skills (gap in %)	-15.6*	-2.5	-2.3	-5.4*
Self-esteem (based on shame) (gap in %)	-0.6	-3.0*	-10.3*	-2.4
Grade aspiration gap (in years)	-0.4*	-0.4	-0.4	-0.7*
(iii yours)				

Source Dercon(2008)



### Policy implications

Early childhood nutritional deprivations can have:

- permanent negative impacts on various short and long-term outcomes.
- social protection may work, but will it help improve psychosocial competencies?
- some groups of children worse than others exacerbating existing inequalities



### Who are the vulnerable children?

- 1 in 5 children in our sample of Ethiopian children aged 12 have lost a parent (200 children)
- 60% of ethnic minority children in Vietnam are stunted (19% of Kinh)
- Also less likely to enrol in school (even controlling for wealth differences)



### **Evaluating Policy**

Any role for longitudinal data?

"Randomisation allows us to answer a non-random set of key questions in development."

Key questions in development that cannot be randomised!

- · Impact of orphanhood
- · Impact of programmes targeted at the poorest
- · Food price rises
- Global recession



# Policy and Poverty

Longitudinal data helps addressing causal links In many cases we have "pre-programme" or baseline information

Compare with outcomes post implementation

- · Difference-in-difference of affected and non-affected groups over time
- A precaution against concluding easily that food aid makes people poorer, or that roads make people richer
- Examples YL Ethiopia workfare programme on child labour, India mid-day meal scheme



# Policy example: Safety net in Ethiopia

Round 3 survey will be able to evaluate the impact of the program using....

- Detailed questions on participation in the food/cash for work programmes
- Food security and nutrition questions
- Information on the younger sibling of our index child
- Research questions:

  - What is the impact of the programme on child outcomes?

    Do children whose parents work in the programme have to do more work
  - Are there any effects on education?



# 3. Linking to other databases

- YL is an example of how carefully planned longitudinal surveys can be used to merge with other data sources, expanding our database and issues investigated, while improving robustness of results.
  - Asking about the job card ID number to enable us to link to national data on the employment generation scheme in India (AP)
  - Use GPS code information to link with DHS and other national data.



# 4. Conclusions

- Longitudinal data has a unique place in helping us understand the causes and consequences of poverty
- Young Lives data is particularly unique as it can be linked to other available datasets
- The findings can make important new contributions to policy and research
- Should be combined with other analysis

