

# Ex-ante Harmonization – Innovations in collection of welfare aggregates

The World Bank

The Poverty and Equity Global Practice

Global Solutions Group on Welfare Measurement and Statistical Capacity



**WORLD BANK GROUP**

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## Short consumption modules – Why?

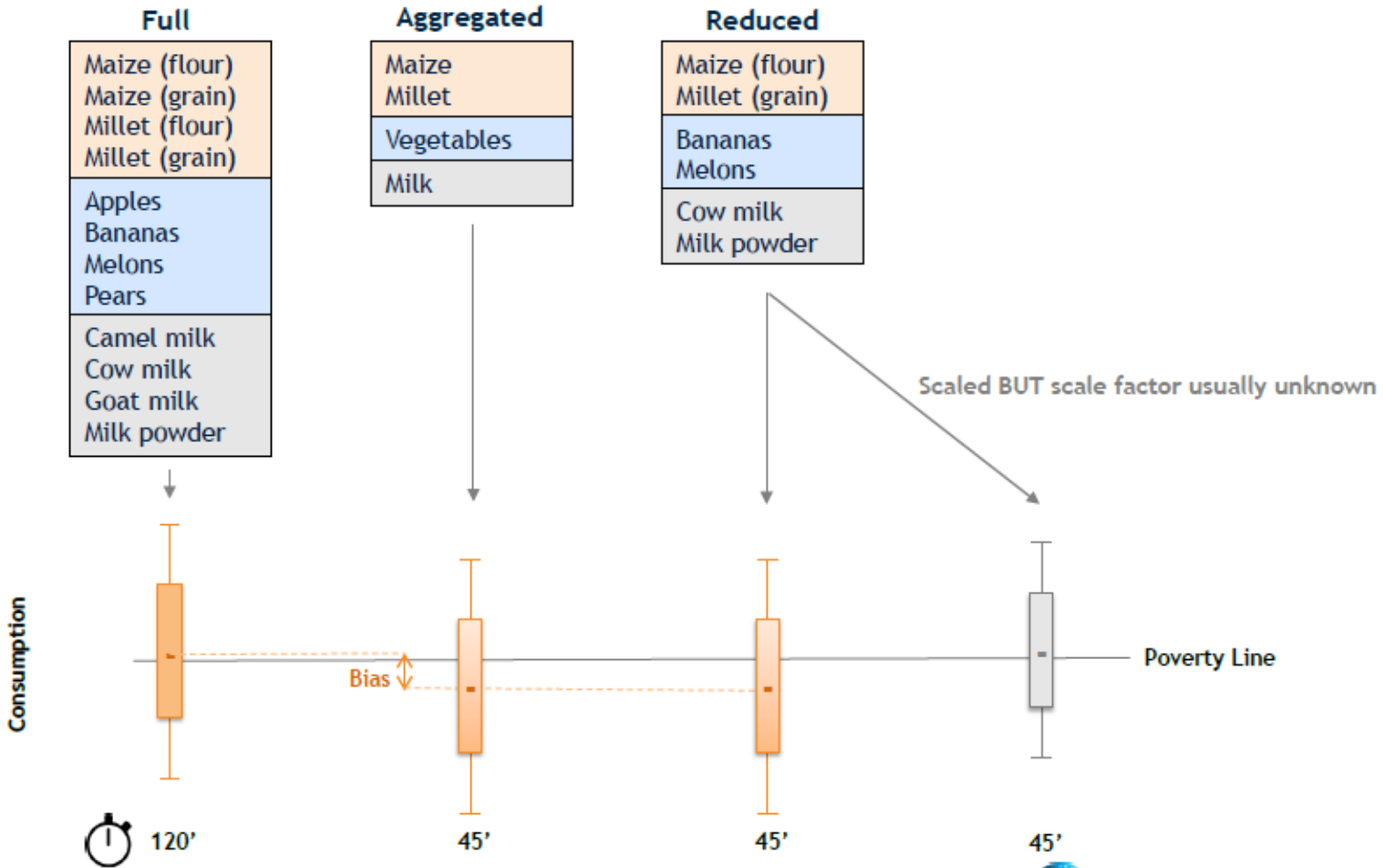
- In many cases it is not feasible to include consumption in one survey for every respondent in one survey
  - Security concerns
  - Cost
  - Time
  - Increases respondent fatigue
- A full aggregate is necessary for monitoring poverty, or program evaluation
- The need for an aggregate database which includes income and consumption
  - EUROMOD uses this for simulating direct and indirect tax policies

# Rapid consumption surveys

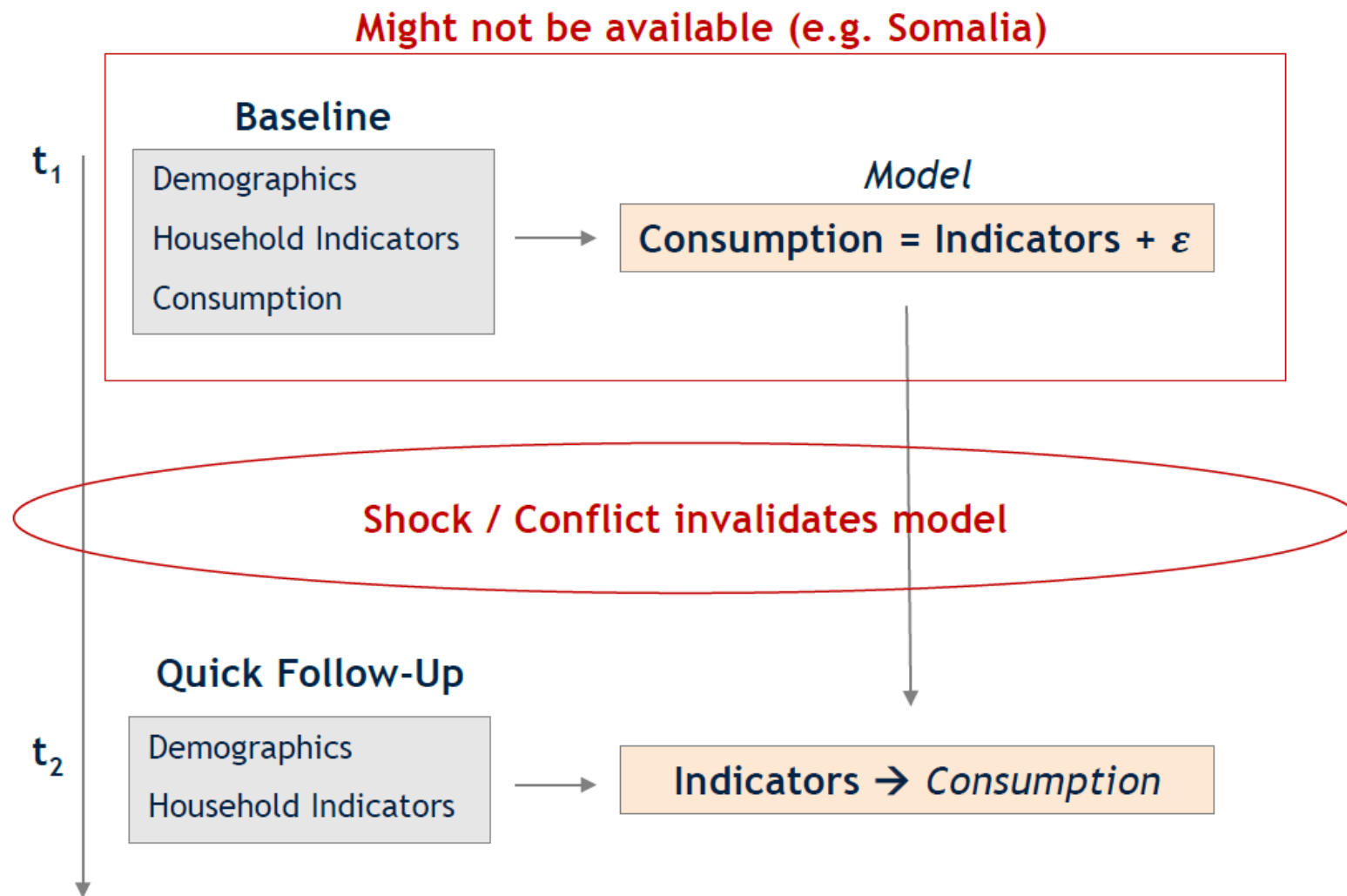
- Ask few items per household, but all items across households
- Use statistical methods to obtain full aggregate for all households
- Ex-ante simulations show virtually unbiased results at aggregated levels
  - Successfully implemented in Somalia, South Sudan, and Kenya

# How does it work?

## Reducing the number of items under-estimates consumption



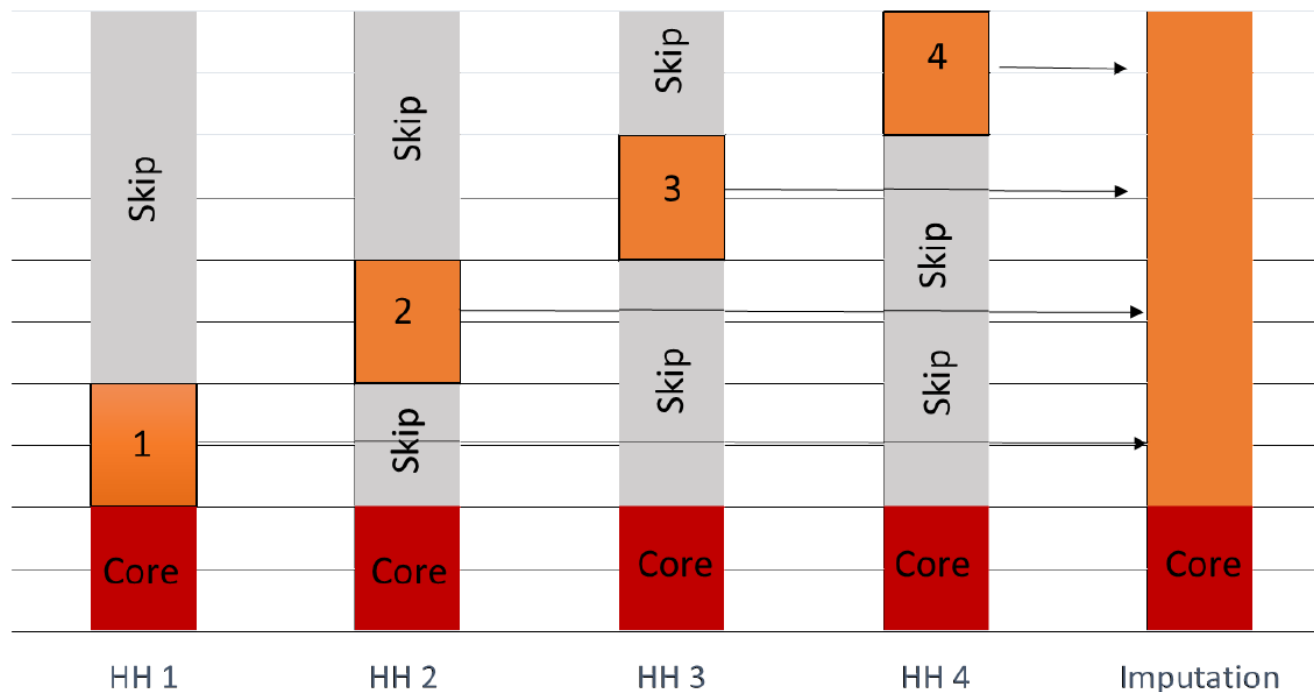
# Why not use cross survey imputation?



- When the source survey was conducted matters
- If the survey is too old, it is unlikely the parameters obtained may be applicable for valid imputation

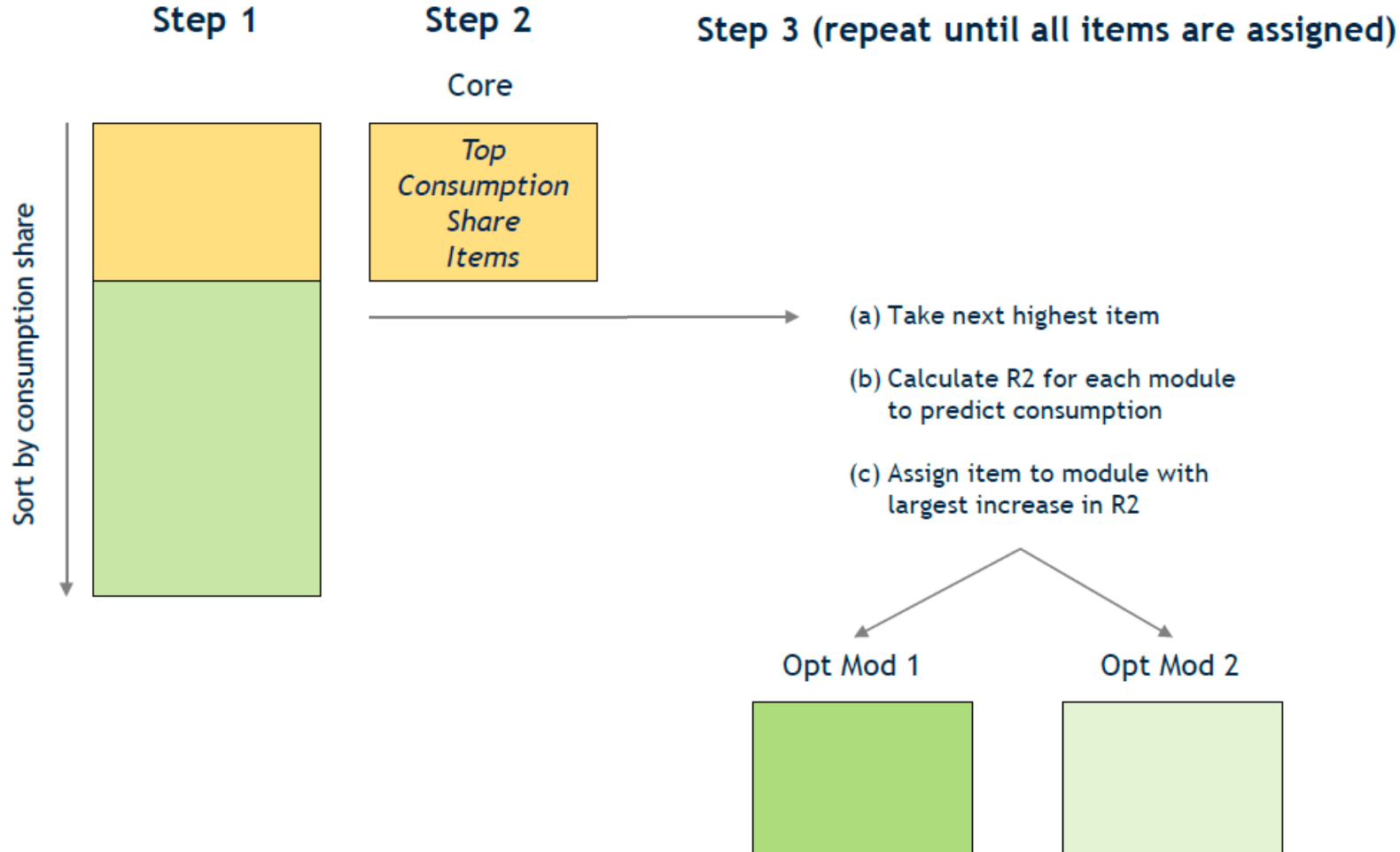
# The basics

Modules and total consumption



- All households are asked the core questions, and one additional module
- Ask a few items per household, but all items across households
- An imputed complete aggregate is obtained for every household
- Different methods are available for imputing (OLS, Tobit, MICE, averages and medians)

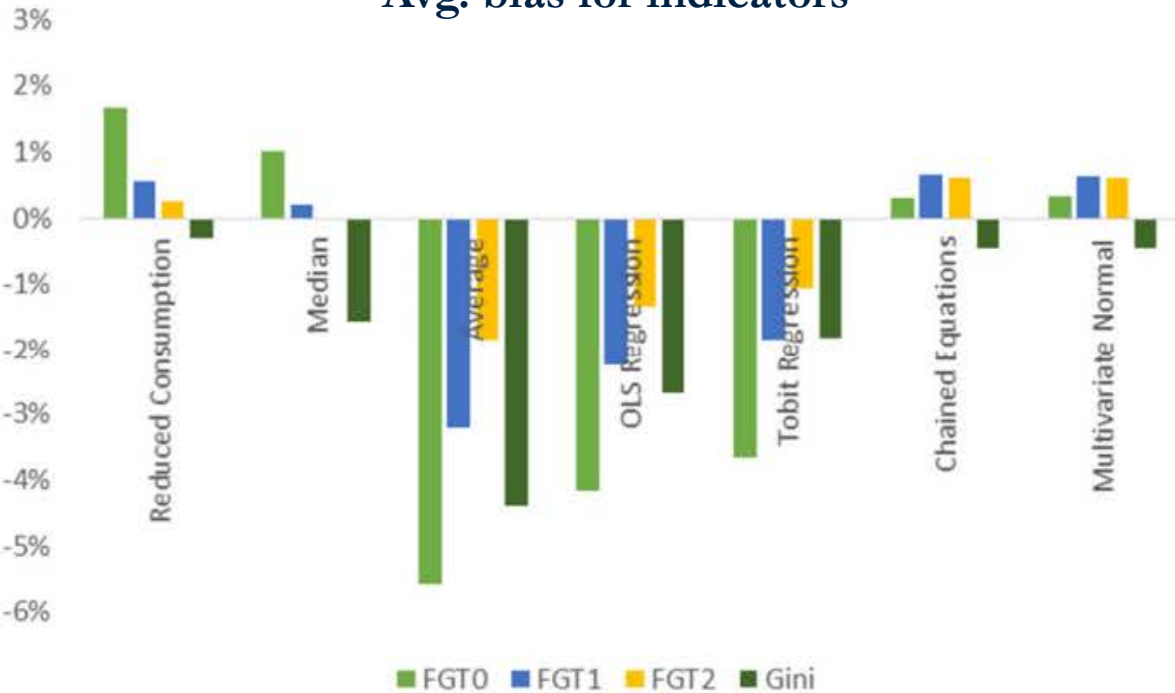
# The basics



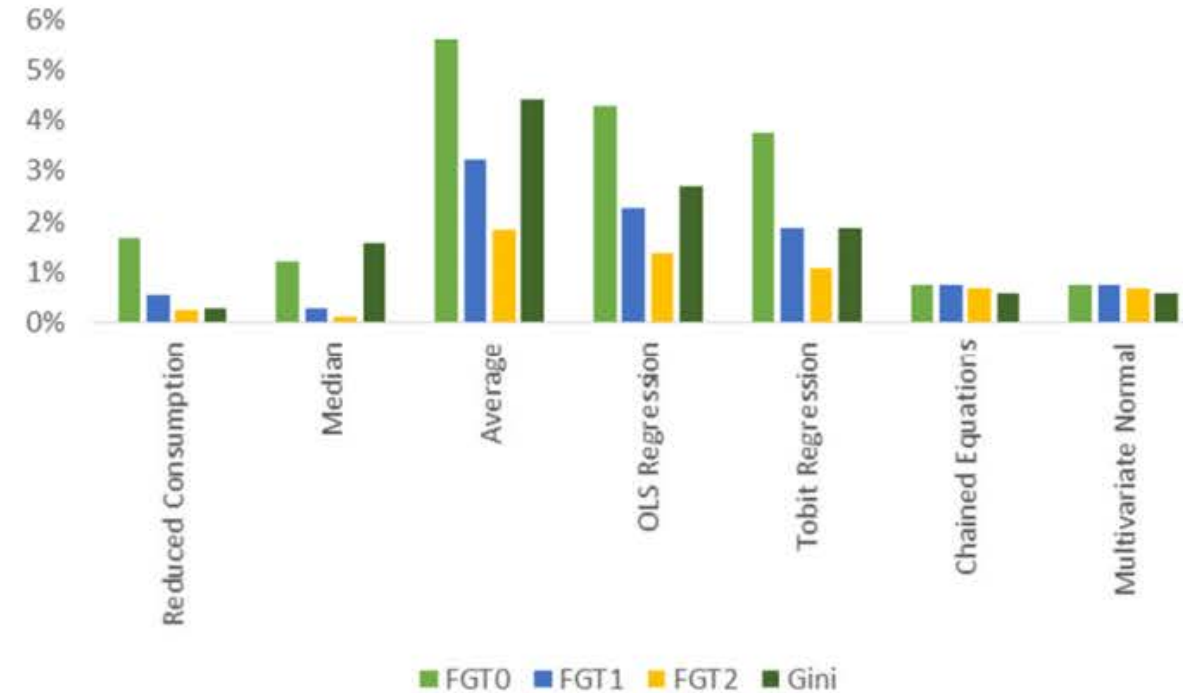
- Module partitions are done on previous consumption data for a similar country
- This improves estimation precision, but does not introduce bias if previous data is poorly correlated with current consumption

# How well does it work?

Avg. bias for indicators



Avg. std. error by indicator



- Relative bias is avg. error
- Relative std. error is the standard deviation of the relative error
- MI methods work best



## An alternative?

### Common approaches for imputation rely on:

- Collecting data from samples which are representative of the underlying population of interest
  - Use welfare correlates on a sub-sample to estimate poverty for all the sample
  - Correlates are usually demographic and socio-economic characteristics

### However:

- Household consumption patterns can be a great predictor of overall consumption levels
  - A limited set of questions on whether or not a household has consumed certain COICOP level 4 item can greatly assist in obtaining a comparable consumption aggregates

## How ? – Selecting the candidate questions on consumption items

- COICOP 4 questions are selected depending on how well these predict overall consumption
- The pool of possible questions is filtered by requiring a proportion of households reporting consumption between 0.10 and 0.90
- Correlates are chosen based on overall model fit (adjusted R<sup>2</sup> and F-stat), as well as the significance of the correlate
  - All correlates with high variance inflation factors are removed
  - Since we are simulating welfare onto a different sample, multicollinearity and overfitting are issues we want to avoid

# Monte Carlo Simulation

Sample with aggregate

$$\hat{\beta}_{GLS} \sim N(\hat{\beta}_{GLS}, \text{Var}(\hat{\beta}_{GLS}))$$

$$\tilde{e}_{ch} \sim N(0, \hat{\sigma}_{e,ch}^2)$$

$$\tilde{\eta}_c \sim N(0, \hat{\sigma}_{\eta}^2)$$

- The goal is to simulate a sufficiently large number of vectors of welfare to allow for reliable estimates of poverty (usually 100)
- From the first stage parameters it is possible to take random drawings from the assumed distributions
  - Alternatively it is possible to get bootstrapped samples of the survey data to yield all parameters needed for simulating welfare vectors



- Take the drawn parameters and apply these to the  $X$  matrix of characteristics in the sample w/o welfare, and simulate the residuals
- This yields  $R$  simulated vectors in the sample data
- From these vectors we get  $R$  poverty rates

Sample w/o welfare

$$\tilde{Y}_{ch} = \exp(X\tilde{\beta}_{GLS} + \tilde{\eta}_c + \tilde{e}_{ch})$$

# How good are our results?

- Household consumption patterns can be a great predictor of overall consumption level
- A limited set of 35 questions if a household has consumed certain COICOP level 4 items can go a long way towards obtaining a comparable consumption aggregate

**Table 1: ELL simulation of adult equivalent consumption for Greece**

	Sim (1)	Sim (2)	Sub sample 2 sim (1)	Sub sample 2 sim (2)
<i>Model details:</i>				
Observations	5,857	5,857	2,860	2,860
Regressors	35	35	35	35
Adjusted R2	0.67	0.67	0.66	0.66
Max. VIF	3.63	3.63	3.59	3.59
F-Stat	345.80	345.80	249.40	161.51
Het. Adj. R2		2.E-03		1.E-03
Het. F-Stat		6.50		2.52

Source: Greek 2014 HBS, own estimates

**Table 2: Simulation exercise for Greece**

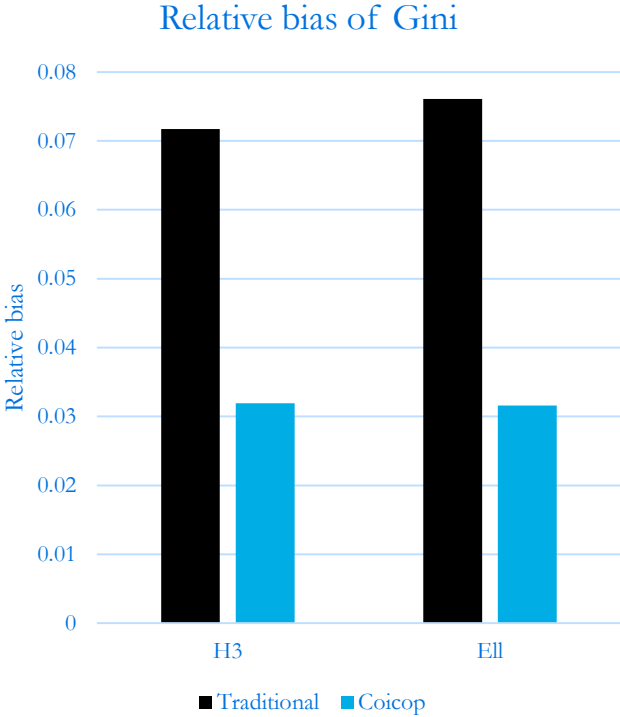
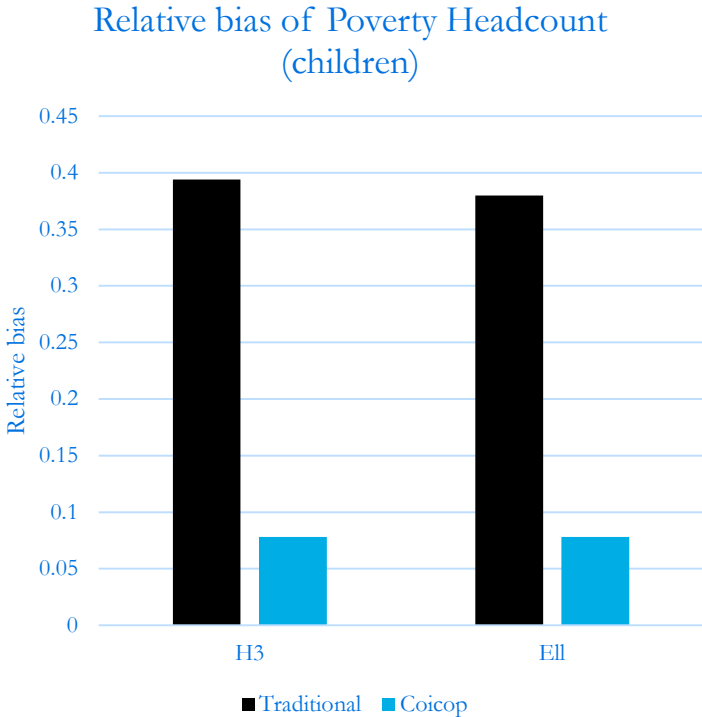
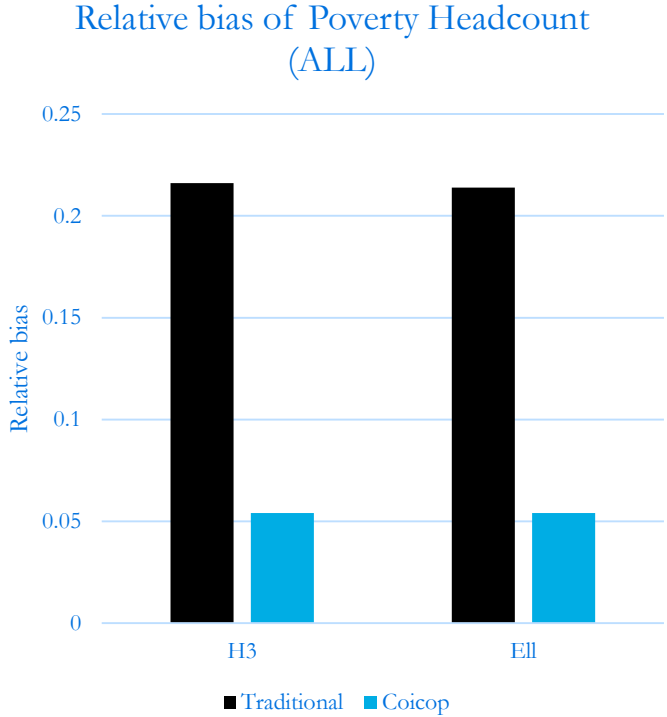
	Direct estimate	Full sample Sim (1)	Full sample Sim (2)	Sub-sample (50%) Sim (2)	Sub-sample (50%) Sim (2)
<b>Children</b>					
<i>Headcount</i>	21.5	25.1	20.3	25.1	20.0
<i>Gap</i>	4.1	8.4	5.2	8.3	5.2
<i>Severity</i>	1.4	3.9	1.9	3.9	2.0
<b>Elderly</b>					
<i>Headcount</i>	27.3	31.9	26.8	32.3	27.4
<i>Gap</i>	6.0	10.6	6.5	10.7	6.6
<i>Severity</i>	2.1	4.9	2.3	4.9	2.3
<b>National</b>					
<i>Headcount</i>	20.9	25.2	19.9	25.5	20.2
<i>Gap</i>	4.8	8.3	4.9	8.4	5.0
<i>Severity</i>	1.8	3.8	1.8	3.8	1.8
<b>Gini</b>	34.6	42.7	34.7	42.7	34.6
<b>GE 0</b>	20.0	31.5	19.8	31.6	19.7
<b>GE1 (Theil)</b>	21.1	32.1	20.3	32.0	20.2
<b>GE2</b>	28.2	45.7	25.6	45.3	25.2

Source: Greek HBS 2014, Own estimates.

Note: Relative line (60% of median adult equivalent consumption). Sim 1: Only consumption dummies, and nat log. of household size. Sim 2: Only consumption dummies, heteroskedasticity modeled with share children and elderly. Simulations done using ELL methodology. Full sample (5,888 hh).

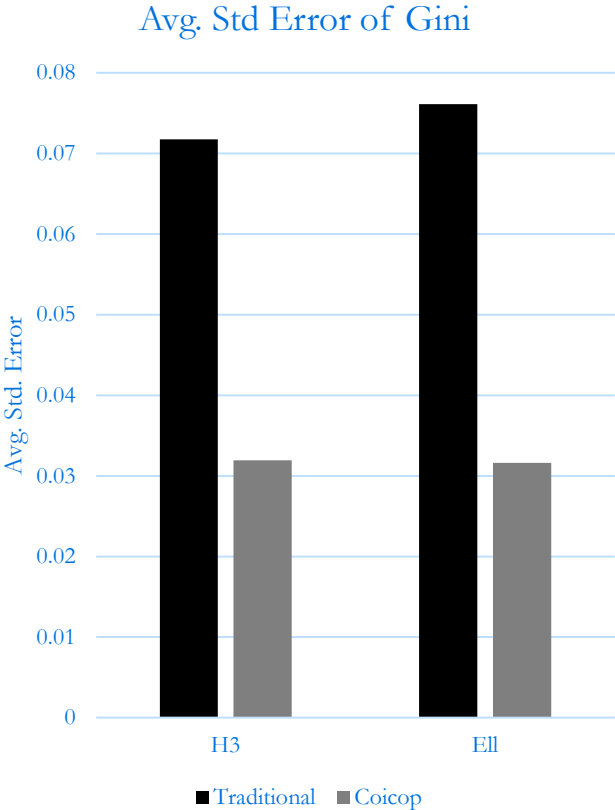
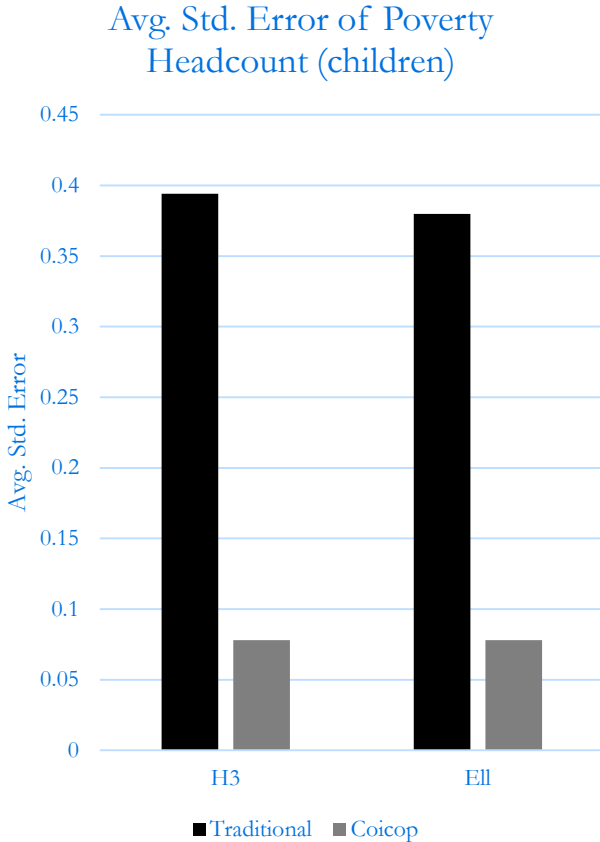
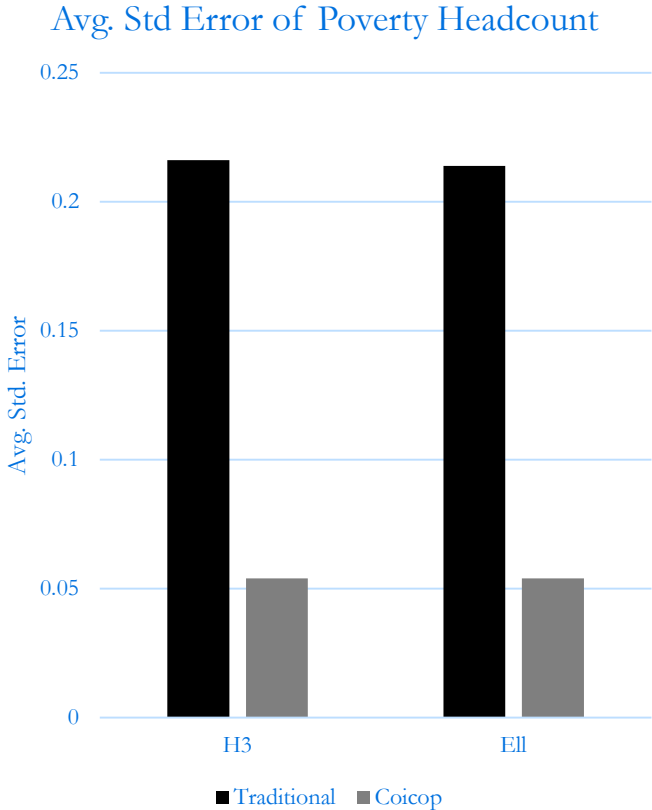
# How does it compare to other approaches?

- It does considerably better than imputations which use demographics and socio-economic characteristics



# How does it compare to other approaches?

- Also yields considerably lower standard average errors



## Next Steps & Conclusions

- Short consumption modules show plenty of promise
  - Possible to estimate consumption and poverty
  - Accurately simulate distribution sensitive parameters
  - Accurately simulate estimates for sub-groups of the population
- Not only valid, but a cost effective approach for obtaining consumption aggregates