

## AF Method

Sabina Alkire, 2 October 2013, IDB

*Tabita, Kenya*

*Rabiya, India*

*Stéphanie, Madagascar*

*Agathe, Madagascar*

*Dalma, Kenya*

*Ann-Sophie, Kenya*

*Valérie, Madagascar*



[T]he job of a ‘measure’ or an ‘index’ is to distill what is particularly relevant for our purpose, and then to focus specifically on that. ... The central issues in devising an index relate to systematic assessment of importance. Measurement has to be integrated with evaluation. This is not an easy task.

–Amartya Sen (1989)

# Sources

- Alkire, S., Foster, J.E., 2011. “Counting and Multidimensional Poverty Measurement,” *Journal of Public Economics*
- See also Alkire, S., Foster, J.E., 2011. “Understandings and Misunderstandings of Multidimensional Poverty Measurement,” *Journal of Economic Inequality*

# Outline

- Motivation
- Multidimensional Data
- Identification
- Aggregation
- Examples

# Challenge

- A government would like to create an official multidimensional poverty indicator
- **Desiderata**
  - It must understandable and easy to describe
  - It must conform to “common sense” notions of poverty
  - It must be able to target the poor, track changes, and guide policy.
  - It must be technically solid
  - It must be operationally viable
  - It must be easily replicable
- **What would you advise?**

# Practical Steps

- **Select**

- Purpose of the index (monitor, target, etc)
- Unit of Analysis (indy, hh, cty)
- Dimensions
- Specific variables or indicators for each dimension
- Whether variables or dimensions should be aggregated with others or left independent
- Cutoff for each independent variable/dimension
- Value of deprivation for each variable/dimension
- **Identification method**
- **Aggregation method**

# This morning's focus:

- **Identification** – Dual cutoffs
- **Aggregation** – Adjusted FGT
- Purpose, Variables, Dimensional Cutoffs, Weights and all other steps – Assume given
- Sen (1976)

# Review: Unidimensional Poverty

**Variable** – income

**Identification** – poverty line

**Aggregation** – Foster-Greer-Thorbecke '84

**Example:**     **Incomes = (7,3,4,8)**     **Poverty line  $z = 5$**

Deprivation vector  $g^0 = (0,1,1,0)$

**Headcount ratio** =  $P_0 = \mu(g^0) = 2/4$

Normalized gap vector  $g^1 = (0, 2/5, 1/5, 0)$

**Poverty gap** =  $P_1 = \mu(g^1) = 3/20$

Squared gap vector  $g^2 = (0, 4/25, 1/25, 0)$

**FGT Measure** =  $P_2 = \mu(g^2) = 5/100$



# Multidimensional Data

Matrix of well-being scores for  $n$  persons in  $d$  domains

	Domains				
$y =$	13.1	14	4	1	Persons
	15.2	7	5	0	
	12.5	10	1	0	
	20	11	3	1	

# Multidimensional Data

Matrix of well-being scores for  $n$  persons in  $d$  domains

$$y = \begin{matrix} & \text{Domains} \\ \begin{matrix} \left[ \begin{array}{cccc} 13.1 & 14 & 4 & 1 \\ 15.2 & 7 & 5 & 0 \\ 12.5 & 10 & 1 & 0 \\ 20 & 11 & 3 & 1 \end{array} \right] & \text{Persons} \end{matrix} \\ z & (13 \quad 12 \quad 3 \quad 1) & \text{Cutoffs} \end{matrix}$$

# *z* vector = Deprivation Cutoffs

- **Schooling:** “How many years of schooling have you completed?”
  - **6 or more (bold is non-poor)**
  - 1-5 years (non-bold is poor)
- **Drinking Water:** “What is the main water source for drinking for this household?”
  - **9. Piped Water**
  - **8. Well/Pump (electric, hand)**
  - 7. Well Water
  - 6. Spring Water / Rain Water / River/Creek Water / Pond/Fishpond
  - 5. Other
- **Sanitation:** “Where do the majority of householders go to the toilet?”
  - **11. Own toilet with septic tank**
  - **10. Own toilet without septic tank**
  - 9. Shared toilet
  - 8. Public toilet
  - 7. Creek/river/ditch (without toilet)
  - 6. Yard/field (without toilet)
  - 5. Sewer
  - 4. Pond/fishpond
  - 3. Animal stable
  - 2. Sea/lake
  - 1. Other

# Deprivation Matrix

Replace entries: 1 if deprived, 0 if not deprived

$$\begin{array}{rcc} & \text{Domains} & \\ y = & \begin{bmatrix} 13.1 & 14 & 4 & 1 \\ 15.2 & \underline{7} & 5 & \underline{0} \\ \underline{12.5} & \underline{10} & \underline{1} & \underline{0} \\ 20 & \underline{11} & 3 & 1 \end{bmatrix} & \text{Persons} \\ z & (13 \quad 12 \quad 3 \quad 1) & \text{Cutoffs} \end{array}$$

# Deprivation Matrix

Replace entries: 1 if deprived, 0 if not deprived

$$g^0 = \begin{matrix} & \text{Domains} & & & \\ & & & & \text{Persons} \\ \begin{matrix} z \\ (13 \quad 12 \quad 3 \quad 1) \end{matrix} & \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 \end{bmatrix} & & & \text{Cutoffs} \end{matrix}$$

# Identification

$$g^0 = \begin{matrix} & \text{Domains} \\ \begin{matrix} \text{Persons} \\ \text{Persons} \\ \text{Persons} \\ \text{Persons} \end{matrix} & \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 \end{bmatrix} \end{matrix}$$

Matrix of deprivations

# Identification – Counting Deprivations

$$\mathbf{g}^0 = \begin{array}{cccc|c} & \text{Domains} & & & c \\ \left[ \begin{array}{cccc} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 \end{array} \right] & \dots & & & \\ & & & & \text{Persons} \end{array}$$

# Identification – Counting Deprivations

Q/ Who is poor?

$$g^0 = \begin{array}{c} \text{Domains} \\ \left[ \begin{array}{cccc} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 \end{array} \right] \begin{array}{c} \dots\dots\dots \\ \dots\dots\dots \end{array} \\ \text{Persons} \end{array} \quad \begin{array}{c} c \\ 0 \\ 2 \\ 4 \\ 1 \end{array}$$



# Identification – Union Approach

Q/ Who is poor?

A1/ Poor if deprived in any dimension  $c_i \geq 1$

	Domains				$c$	
$\alpha_0$	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>	Persons
	<b>0</b>	<b>1</b>	<b>0</b>	<b>1</b>	<b>2</b>	
	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>4</b>	
	<b>0</b>	<b>1</b>	<b>0</b>	<b>0</b>	<b>1</b>	

# Identification – Union Approach

Q/ Who is poor?

A1/ Poor if deprived in any dimension  $c_i \geq 1$

	Domains	$c$
Persons	<b>0</b> <b>0</b> <b>0</b> <b>0</b>	<b>0</b>
	<b>0</b> <b>1</b> <b>0</b> <b>1</b>	<b><u>2</u></b>
	<b>1</b> <b>1</b> <b>1</b> <b>1</b>	<b><u>4</u></b>
	<b>0</b> <b>1</b> <b>0</b> <b>0</b>	<b><u>1</u></b>

## Observations

Union approach often predicts high numbers.

Charavarty et al '98, Tsui '02, Bourguignon & Chakravarty 2003 etc use the union approach

# Identification – Intersection Approach

Q/ Who is poor?

A2/ Poor if deprived in all dimensions  $c_i = d$

	Domains				$c$	
$\alpha_0$	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>	Persons
	<b>0</b>	<b>1</b>	<b>0</b>	<b>1</b>	<b>2</b>	
	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>4</b>	
	<b>0</b>	<b>1</b>	<b>0</b>	<b>0</b>	<b>1</b>	

# Identification – Intersection Approach

Q/ Who is poor?

A2/ Poor if deprived in all dimensions  $c_i = d$

	Domains	$c$	
$\sigma_0$	$\begin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} & \mathbf{1} \\ \mathbf{1} & \mathbf{1} & \mathbf{1} & \mathbf{1} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} & \mathbf{0} \end{bmatrix}$	$\begin{matrix} \mathbf{0} \\ \mathbf{2} \\ \mathbf{4} \\ \mathbf{1} \end{matrix}$	Persons

## Observations

Demanding requirement (especially if  $d$  large)

Often identifies a very narrow slice of population

Atkinson 2003 first to apply these terms.

# Identification – Dual Cutoff Approach

Q/ Who is poor?

A/ Fix cutoff  $k$ , identify as poor if  $c_i \geq k$

	Domains				$c$	
$\alpha_0$	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>	Persons
	<b>0</b>	<b>1</b>	<b>0</b>	<b>1</b>	<b>2</b>	
	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>4</b>	
	<b>0</b>	<b>1</b>	<b>0</b>	<b>0</b>	<b>1</b>	

# Identification – Dual Cutoff Approach

Q/ Who is poor?

A/ Fix cutoff  $k$ , identify as poor if  $c_i \geq k$  (Ex:  $k = 2$ )

	Domains				$c$	
$\alpha_0$	0	0	0	0	0	Persons
	0	1	0	1	<u>2</u>	
	1	1	1	1	<u>4</u>	
	0	1	0	0	1	

# Identification – Dual Cutoff Approach

Q/ Who is poor?

A/ Fix cutoff  $k$ , identify as poor if  $c_i \geq k$  (Ex:  $k = 2$ )

$c_i$	Domains	$c$																	
	<table style="border-collapse: collapse; margin: 0 auto;"> <tr><td style="padding: 5px 10px;"><b>0</b></td><td style="padding: 5px 10px;"><b>0</b></td><td style="padding: 5px 10px;"><b>0</b></td><td style="padding: 5px 10px;"><b>0</b></td></tr> <tr><td style="padding: 5px 10px;"><b>0</b></td><td style="padding: 5px 10px;"><b>1</b></td><td style="padding: 5px 10px;"><b>0</b></td><td style="padding: 5px 10px;"><b>1</b></td></tr> <tr><td style="padding: 5px 10px;"><b>1</b></td><td style="padding: 5px 10px;"><b>1</b></td><td style="padding: 5px 10px;"><b>1</b></td><td style="padding: 5px 10px;"><b>1</b></td></tr> <tr><td style="padding: 5px 10px;"><b>0</b></td><td style="padding: 5px 10px;"><b>1</b></td><td style="padding: 5px 10px;"><b>0</b></td><td style="padding: 5px 10px;"><b>0</b></td></tr> </table>	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>	<b>1</b>	<b>0</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b>0</b>	<b>1</b>	<b>0</b>	<b>0</b>	<b>0</b>	
<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>																
<b>0</b>	<b>1</b>	<b>0</b>	<b>1</b>																
<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>																
<b>0</b>	<b>1</b>	<b>0</b>	<b>0</b>																
		<b><u>2</u></b>	Persons																
		<b><u>4</u></b>																	
		<b>1</b>																	

Note

Includes both union ( $k = 1$ ) and intersection ( $k = d$ )

# Identification – The problem empirically

$k =$	H
Union 1	91.2%
2	75.5%
3	54.4%
4	33.3%
5	16.5%
6	6.3%
7	1.5%
8	0.2%
9	0.0%
Inters. 10	0.0%

## Poverty in India for 10 dimensions:

91% of population would be targeted using union,  
0% using intersection  
Need something in the middle.

*(Alkire and Seth 2009)*



# Create Censored Deprivation Matrix

Censor data of nonpoor

$$\mathbf{c}_{\text{nonpoor}} = \begin{array}{c} \text{Domains} \\ \left[ \begin{array}{cccc} \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} & \mathbf{1} \\ \mathbf{1} & \mathbf{1} & \mathbf{1} & \mathbf{1} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} & \mathbf{0} \end{array} \right] \end{array} \begin{array}{c} c \\ \mathbf{0} \\ \underline{\mathbf{2}} \\ \underline{\mathbf{4}} \\ \mathbf{1} \end{array} \quad \text{Persons}$$

# Censored Deprivation Matrix

Censor data of nonpoor

$$g^0(k) = \begin{array}{c} \text{Domains} \\ \left[ \begin{array}{cccc} \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} & \mathbf{1} \\ \mathbf{1} & \mathbf{1} & \mathbf{1} & \mathbf{1} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \end{array} \right] \end{array} \quad \begin{array}{c} c(k) \\ \mathbf{0} \\ \underline{\mathbf{2}} \\ \underline{\mathbf{4}} \\ \mathbf{0} \end{array} \quad \text{Persons}$$

# Aggregation – Headcount Ratio

$$g^0(k) = \begin{array}{c} \text{Domains} \\ \left[ \begin{array}{cccc} \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} & \mathbf{1} \\ \mathbf{1} & \mathbf{1} & \mathbf{1} & \mathbf{1} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \end{array} \right] \end{array} \begin{array}{c} c(k) \\ \mathbf{0} \\ \underline{\mathbf{2}} \\ \underline{\mathbf{4}} \\ \mathbf{0} \end{array} \quad \text{Persons}$$

# Aggregation – Headcount Ratio

$$g^0(k) = \begin{array}{c} \text{Domains} \\ \left[ \begin{array}{cccc} \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} & \mathbf{1} \\ \mathbf{1} & \mathbf{1} & \mathbf{1} & \mathbf{1} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \end{array} \right] \end{array} \quad \begin{array}{c} c(k) \\ \mathbf{0} \\ \underline{\mathbf{2}} \\ \underline{\mathbf{4}} \\ \mathbf{0} \end{array} \quad \text{Persons}$$

Two poor persons out of four: **H = 1/2**

# Aggregation – Intensity (A)

Need to augment information

deprivation shares among poor

$$g^0(k) = \begin{array}{c} \text{Domains} \\ \left[ \begin{array}{cccc} \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{1} & \mathbf{0} & \mathbf{1} \\ \mathbf{1} & \mathbf{1} & \mathbf{1} & \mathbf{1} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} \end{array} \right] \end{array} \quad \begin{array}{cc} c(k) & c(k)/d \\ \mathbf{0} & \\ \underline{\mathbf{2}} & \mathbf{2 / 4} \\ \underline{\mathbf{4}} & \mathbf{4 / 4} \\ \mathbf{0} & \end{array} \quad \text{Persons}$$

# Aggregation – Intensity (A)

Need to augment information

deprivation shares among poor

	Domains	$c(k)$	$c(k)/d$	
$g^0(k) =$	<b>0</b> <b>0</b> <b>0</b> <b>0</b>	<b>0</b>		Persons
	<b>0</b> <b>1</b> <b>0</b> <b>1</b>	<u><b>2</b></u>	<b>2 / 4</b>	
	<b>1</b> <b>1</b> <b>1</b> <b>1</b>	<u><b>4</b></u>	<b>4 / 4</b>	
	<b>0</b> <b>0</b> <b>0</b> <b>0</b>	<b>0</b>		

A = average deprivation share among poor = 3/4

# Aggregation – Adjusted Headcount Ratio

Adjusted Headcount Ratio =  $M_0$  = HA

	Domains	$c(k)$	$c(k)/d$				
$g^0(k) =$	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>			
$g^0(k) =$	<b>0</b>	<b>1</b>	<b>0</b>	<b>1</b>	<b><u>2</u></b>	<b>2 / 4</b>	
$g^0(k) =$	<b>1</b>	<b>1</b>	<b>1</b>	<b>1</b>	<b><u>4</u></b>	<b>4 / 4</b>	
$g^0(k) =$	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>	<b>0</b>		Persons

A = average deprivation share among poor = 3/4

# Aggregation – Adjusted Headcount Ratio

$$\text{Adjusted Headcount Ratio} = M_0 = HA = \mu(g^0(\mathbf{k}))$$

	Domains	$c(k)$	$c(k)/d$	
$g^0(k) =$	$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$	$0$	$2 / 4$	Persons
		$\underline{4}$	$4 / 4$	
		$0$		

$A =$  average deprivation share among poor  $= 3/4$



# Aggregation – Adjusted Headcount Ratio

$$\text{Adjusted Headcount Ratio} = M_0 = HA = \mu(g^0(\mathbf{k})) = 6/16 = .375$$

	Domains	$c(k)$	$c(k)/d$	
$g^0(k) =$	$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$	$\mathbf{0}$	$\mathbf{2 / 4}$	Persons
		$\mathbf{4}$	$\mathbf{4 / 4}$	
		$\mathbf{0}$		

$A$  = average deprivation share among poor = 3/4

# Aggregation – Adjusted Headcount Ratio

$$\text{Adjusted Headcount Ratio} = M_0 = \text{HA} = \mu(g^0(\mathbf{k})) = 7/16 = 0.44$$

	Domains	$c(k)$	$c(k)/d$	
$g^0(k) =$	$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$	$0$	$3 \dots 3/4$	Persons
	$\dots$	$4 \dots 4/4$		
	$0$	$0$		
	$0$	$0$		

A = average deprivation share among poor = 3/4

Note: if person 2 has an additional deprivation,  $M_0$  rises

Satisfies dimensional monotonicity

# Aggregation: Adjusted FGT Family

Adjusted FGT is  $M_\alpha = \mu(\mathbf{g}^\alpha(\boldsymbol{\tau}))$  for  $\alpha \geq 0$

Domains

$$\mathbf{g}^\alpha(k) = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0.42^\alpha & 0 & 1^\alpha \\ 0.04^\alpha & 0.17^\alpha & 0.67^\alpha & 1^\alpha \\ 0 & 0 & 0 & 0 \end{bmatrix} \text{Persons}$$

**Theorem 1** For any given weighting vector and cutoffs, the methodology  $M_{ka} = (\rho_k, M_\alpha)$  satisfies: decomposability, replication invariance, symmetry, poverty and deprivation focus, weak and dimensional monotonicity, nontriviality, normalisation, and weak rearrangement for  $\alpha \geq 0$ ; monotonicity for  $\alpha > 0$ ; and weak transfer for  $\alpha \geq 1$ .

# Extension – General Weights

Modifying for weights: identification and aggregation  
(technically weights need not be the same, but conceptually probably should be)

- Use the  $g_0$  or  $g_1$  matrix
- Choose relative weights for each dimension  $w_d$
- Apply the weights (sum =  $d$ ) to the matrix
- $c_k$  now reflects the *weighted sum* of the dimensions.
- Set cutoff  $k$  across the weighted sum.
- Censor data as before to create  $g_0(k)$  or  $g_1(k)$
- Measures are *still* the mean of the matrix.

# Example: Weights

$$\mathbf{g}_0 = \begin{matrix} & \text{Domains} \\ \text{Persons} & \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 \end{bmatrix} \end{matrix}$$

Matrix of deprivations

Weighting vector  $\omega = (.5 \ 2 \ 1 \ .5)$

# Example: Weights

$$g^0 = \begin{matrix} & \text{Domains} & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ \text{Persons} & & & \end{matrix} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & .5 \\ .5 & 2 & 1 & .5 \\ 0 & 2 & 0 & 0 \end{bmatrix}$$

Matrix of deprivations

Weighting vector  $\omega = (.5 \ 2 \ 1 \ .5)$

# Weighted Deprivation Matrix

Note that we use the same notation as for the deprivation matrix on purpose.

Where

- $g_{ij}^0 = w_j$  if  $x_{ij} < z_j$  (deprived)
- $g_{ij}^0 = 0$  if  $x_{ij} \geq z_j$  (non-deprived)
- Or equivalently:

$$g_{ij}^0 = w_j \left( \frac{z_j - x_{ij}}{z_j} \right)^0$$

$$g^0 = \begin{bmatrix} g_{11}^0 & \dots & g_{1d}^0 \\ g_{21}^0 & \dots & g_{2d}^0 \\ \dots & & \dots \\ g_{n1}^0 & \dots & g_{nd}^0 \end{bmatrix}$$

$$Z = (z_1, z_2, \dots, z_d)$$
$$W = (w_1, w_2, \dots, w_d)$$

# AF Method: Decompositions

By Population Subgroup

$M_0$  Poverty

H Headcount

A Intensity

Post-identification: By Dimension

Censored Headcount (column vector)

Percentage Contribution (weighted)

**All draw on censored matrix**



# Informal Glossary of Terms

**Deprivation:** if  $y_{id} < z$  person  $i$  is **deprived** in  $y_d$

**Poverty:** if  $c_i \leq k$  person  $i$  is poor.

**Deprivation cutoffs:** the  $z$  cutoffs for each dimension

**Poverty cutoff:** the overall cutoff  $k$

**Dimension:** for AF – a column in the matrix having its own deprivation cutoff (sometimes called an ‘indicator’)

**Joint distribution:** showing the simultaneous or coupled deprivations a person/hh has

# Deprivation Count Vector

Where the 'deprivation count' or score for each person is the sum of her weighted deprivations

- $c_i = g_{i1} + \dots + g_{id}$

$$c = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$$

# Identify the poor

Given a poverty cut-off  $k$ , we compare the deprivation count with the  $k$  cutoff and then censor the deprivations of those who were not identified as poor.

$$\begin{aligned} \rho_k(x_i; z) &= 1 && \text{if } c_i \geq k && \text{poor} \\ \rho_k(x_i; z) &= 0 && \text{if } c_i < k && \text{non-poor} \end{aligned}$$

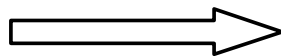
$$c = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$$

# Key point: Deprivation and Censored Matrix

$g^0(k)$  needed for associated partial indices

Deprivation Matrix

$$g^0 = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 \\ 2 \\ 4 \\ 1 \end{bmatrix}$$



Censored Deprivation Matrix,  $k=2$

$$g^0(k) = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 \\ 2 \\ 4 \\ 0 \end{bmatrix}$$

# Headcount Ratio of MD Poverty

It is the proportion of people who have been identified as poor. Thus:

$$H = \frac{\sum_{i=1}^n \rho_k(x_i; z)}{n} = \frac{q}{n}$$

Where  $q$  is the number of poor people.

Headcount Ratio is sometimes called the *incidence* of poverty, or the poverty rate.

# Intensity (or breadth) of MD Poverty

- It is the average proportion of deprivations in which the poor are deprived.

$$A = \frac{\sum_{i=1}^n c_i(k)}{dq}$$

Note that it is simple to compute:

- 1) You compute the proportion of total deprivations each poor person has ( $c_i(k)/d$ ). Note we need to use the censored deprivation count vector, ie: we ignore the deprivations of the non-poor.
- 2) You take the average of those proportions (that's why we divide by  $q$ , the number of the poor)

# Multidimensional Poverty: $M_0$

(Adjusted Headcount Ratio)

- It is the product of incidence and intensity.

$$M_0 = H * A$$

- Or equivalently, it is the mean of the censored (weighted) deprivation matrix:

$$M_0 = \mu(g^0(k)) = \frac{\sum_{i=1}^n \sum_{j=1}^d g_{ij}^0(k)}{nd}$$

# How do we interpret $M_0$ ?

- $M_0$  Interpretation

$M_0$  is the mean of the weighted censored deprivation matrix.

**Thus, it gives the proportion of weighted deprivations that the poor experience in a society of all the total potential deprivations that the society could experience.**



# How do we decompose $M_0$ by Indicators?

There are two useful but distinct indicators to look at:

- 1) Censored headcount ratios
- 2) Contributions by indicators and dimensions.

# Censored Headcount Ratios

- Censored headcount ratios are the % of people who are poor and deprived in a certain indicator.
- Careful! Censored headcounts are not the % of the poor deprived in a certain indicator.
- Raw headcount ratios are the % of people who are deprived in a certain indicator.

# Censored Headcount Ratios

- They are the mean of each column of the (weighted) censored deprivation matrix divided by the indicator's weight.

$$H_j^C = \frac{\sum_{i=1}^n g_{ij}^0(k)}{w_j n}$$

- $M_0$  is the weighted sum of the censored headcount ratios.

$$M_0 = \sum_{j=1}^d \left( \frac{w_j}{d} \right) H_j^C$$

# Raw Headcount Ratios

- These are the deprivation rates by dimension, ie. the proportion of people who are deprived in that dimension.
- It is simply the mean of each column of the [uncensored] deprivation matrix:

$$H_j = (g_{1j}^0 + g_{2j}^0 + \dots + g_{nj}^0) / n$$

# Contribution by Indicator and Dimension

- It is the proportion of total poverty which arises from a particular deprivation.
- Recall from the previous slide:

$$M_0 = \sum_{j=1}^d \left( \frac{w_j}{d} \right) H_j^C$$

- Thus, the contribution of indicator  $j$  to overall poverty is given by:

$$C_j = \frac{(w_j / d) H_j^C}{M_0}$$

# Contribution by Indicator and Dimension

- Note: The sum of the contributions of all  $d$  indicators needs to add up to 1 (or 100%).
- The percentage contribution includes the **weight** on each indicator, and is useful when the weights are not equal.
- If there are more than one indicators in a dimension, the dimensional contribution is simply the sum of the indicators' contribution.

# How do we decompose $M_0$ by population subgroups?

Visually...

Peo	Ye. Ed	Child Attend	Nutr	Mor	Elec	Wat	Sani	Floor	Cook. Fuel	Assets	Depr. Count
1	1.67	1.67	0	0	0	0.56	0	0.56	0	0	4.44
2	0	0	1.67	1.67	0	0	0	0	0	0	3.33
3	0	0	0	0	0.56	0.56	0.56	0.56	0.56	0.56	3.33
4	1.67	0	0	1.67	0.56	0.00	0.56	0	0.56	0	5
5	0	0	0	0	0	0	0	0	0	0	0

**GROUP A** ↑

↓ **GROUP B**

# Decomposition by Population Subgroups

If the entire population  $X$  (of size  $n$ ) is divided into two subgroups  $X_1$  (of size  $n_1$ ) and  $X_2$  (of size  $n_2$ ), then overall  $M_0$  is the weighted sum of  $M_0$  in each subgroup:

$$M_0(X; z) = \left( \frac{n_1}{n} \right) M_0(X_1; z) + \left( \frac{n_2}{n} \right) M_0(X_2; z)$$

Thus, the contribution of subgroup  $i$  to overall poverty is

$$C_{Gi} = \frac{(n_i / n) M_0(X_i; z)}{M_0(X; z)}$$



# Decomposition by Population Subgroups

- Note that the sum of the contributions of all groups needs to add up to 1 (or 100%).
- You take the mean of the matrices of each population subgroup individually.
- The population-weighted mean of subgroups is equal to the national  $M_0$ .