

Simple Poverty Progress Indices for Bangladesh, Haiti, India, Mexico, Pakistan, and the Philippines

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Goal: Measure Clients' Poverty Status

1. Objective:

- Quantitative, expenditure-based**
- LSMS-type data**

2. Accurate (strong link with poverty)

3. Practical (accepted and used):

- Few, inexpensive-to-collect indicators**
- Simple enough to compute on paper, in the field, in real time (no software required)**

4. Applicable (for all, not just microfinance)

Goal: Give Managers a Decision Tool

1. Use score to:

- Target services (classify individuals)**
- Report poverty rates (e.g., for USAID)**
- Track changes in poverty over time**

2. ‘Practicality’ matters more than accuracy:

- For programs, not journals**
- Users resist change; if unused, why bother?**
- KISS (Keep Scoring Simple):**
 - Minimize ‘extra’ work**
 - Show users how scoring works**
 - Choose ‘common-sense’ indicators**
 - Adjust based on users’ feedback**

Approaches

1. Microfinance practitioners (housing indices):

- **Common-sense indicators and weights**
- **Easy-to-use, and well accepted**
- **Not linked to expenditure, unknown accuracy**

2. Academics (poverty map, proxy means test):

- **Regression with LSMS expenditure data**
- **Focus on complex statistics, ignore accuracy**
- **Scorecard rarely presented, let alone used**

3. Scoring Industry:

- **Use data, regression, *and* expertise**
- **Simple statistics (but not taught in school)**
- **Focus on results, thanks to profit incentive**

CASH-POR Housing Index

Indicator		Values		Points
1. Size of house?	Small	Medium	Big	
	0	2	4	
2. Structural condition?	Dilapidated	Average	Good	
	0	2	6	
3. Quality of walls?	Poor	Average	Good	
	0	2	6	
4. Quality of roof?	Thatch/leaves	Tin/Iron sheets	Permanent roof	
	0	2	6	
Source: "Overcoming the Obstacles of Identifying the Poorest Families", 2000, Simanowitz, Nkuna, and Kasim.				Total:

- **If score is 10, is the person poor?**
- **Are all people with scores of 0 poor?**
- **There are few thatch roofs on brick walls**

Egypt Regression (Datt & Jolliffe, 2005)

TABLE 2
BASIC AND AUGMENTED MODELS, LOG PER CAPITA CONSUMPTION—OLS, GOVERNORATE-LEVEL FIXED EFFECTS

Variable	Description	Rural Model (N = 1,326)				Urban Model (N = 1,122)			
		Basic Model		Augmented Model		Basic Model		Augmented Model	
		Coefficient	t-Ratio	Coefficient	t-Ratio	Coefficient	t-Ratio	Coefficient	t-Ratio
Upper	Upper Egypt	.147	2.91	.045	.43	-.145	-2.65	-.082	-1.82
Household characteristics:									
Hhsize	Household size	-.142	-8.83	-.156	-9.54	-.306	-12.51	-.245	-9.05
Hhsize2	Household size, squared	.004	5.23	.004	5.63	.015	9.98	.014	9.84
Hhage	Household head: age in years	.015	2.58	.012	1.97	.019	2.31	.010	3.91
Hhage2	Household head: age+ squared	.000	-1.95	.000	-1.69	.000	-1.65		
Femhead	Dummy: female-headed household	-.069	-1.77			-.114	-2.13	-.113	-2.11
Avgsch	Household average years of schooling	.041	10.51			.053	10.7	.054	10.73
Hhpedu	Household head's parent: primary school	.052	2.11	.068	2.61	.080	3.33	.079	3.27
Sppedu	Spouse's parent: primary school	.056	2.04						
Lowned	Log: owned cultivated land	.146	5.22	.253	6.98	.090	1.67		
Secsch	Distance to secondary school	.032	2.29						
H_post	Distance to hospital post/hospital	-.036	-3.06	-.022	-1.56				
Interaction effects:									
Hhsize × avgsh				.004	3.18				
Hhsize × sppedu				-.020	-2.47				
Hhsize × hhage								-.001	-2.38
Hhsize × lowned								.010	2.38
Hhage × avgsh				.000	2.82				
Hhage × sppedu				.005	4.25				
Hhage × lowned				-.001	-3.15			.001	1.92
Avgsh × lowned				-.006	-2.84				
Hhpedu × sppedu				-.041	-2.04				
Sppedu × h_post				.024	1.46				
Lowned × h_post				-.007	-1.52				
Intercept		4.876	23.88	5.196	23.79	5.352	15.35	5.246	23.68
R ²		.37		.38		.48		.48	

Note. Dummies for governorates and missing observations are suppressed from the output. The urban sample consists of 57 primary sampling units (PSUs); the rural sample has 68 PSUs. OLS = ordinary least squares.

How Indicators Are Selected

'Practicality', not just accuracy

Regression w/LSMS-type data + 'expert': After stepwise, pick best predictors that are also:

- Common sense**
- Objective**
- Verifiable**
- Quick**
- Liable to change over time**
- Strongly linked with poverty**

Indicator Selection (cont.)

Exclude:

- **Expenditure amount, asset value**
- **Ratio, square, interaction, logarithm**
- **Continuous or subjective**

Include:

- **Presence/absence of assets**
- **Categories**
- **Objective**
- **Variety**
- **Liable to change over time**

Example Indicators

- **Number of children ages 0–14**
- **School attendance**
- **Type of floor**
- **Type of toilet**
- **Source of water**
- **Land ownership**
- **Cooking fuel**
- **TV ownership**
- **Radio ownership**
- **New shoes in past 6 months**

Point Selection (cont.)

- 1. Transform logit coefficients so that:**
 - All points are zero or positive integers**
 - Lowest score (most likely poor) is 0**
 - Highest score (least likely poor) is 100**
- 2. This transformation reduces accuracy a little but promotes acceptance**
- 3. Programs can download scorecard & use**
- 4. Field workers compute scores on paper, in field, in real time; no need for software**

Overview of PPIs

Country	Poverty line (person/day)	% poor	Survey	# HH	Indicators tested
Bangladesh	\$1 (PPP)	44	'00 HIES	7,440	>600
Haiti	\$1 (PPP)	56	'01 ECVH	7,168	>250
India	\$1 (PPP)	46	'03 SES	41,013	>400
Mexico	P31 rural, P45 urban	48	'02 ENIGH	17,167	>2,000
Pakistan	Rs25	40	'01 PIHS	15,503	>400
Philippines	P36	31	'01 APIS	38,014	>500

Bangladesh PPI

Indicator			Attributes			Points
1.	What type of latrine does the household use?			Kacha (temporary or permanent) or pit	Sanitary or water-seal	
			Open field	Pacca	Pacca	
			0	7	12	
2.	How many household members are 11 years old or younger?	4 or more	3	2	1	0
		0	7	12	17	26
3.	Does any household member work for a daily wage?			Yes	No	
				0	7	
4.	How many rooms does the house have (excluding ones used for business)?			1	2 or 3	4 or more
				0	3	9
5.	Do all children ages 6 to 17 attend school?			No	No children ages 6 to 17	Yes
				0	4	6
6.	Does the household own a television set?				No	Yes
					0	11
7.	How many hectares of cultivable land does the household own?	Less than 0.34	0.34 to 0.99	1 to 1.99	2 or more	
		0	3	4	9	
8.	What is the main construction material of the walls of the house?		Hemp/hay/bamboo or mud brick	C.I. sheet/wood	Brick/cement	
		0		5	7	
9.	Does the household own drawing room furniture?			No	Yes	
				0	9	
10.	Does the house have a separate kitchen?			No	Yes	
				0	4	
Source: Calculations by Microfinance Risk Management, L.L.C., based on 2000 HIES.						Total:

Haiti PPI

Indicator		Attributes				Points	
1.	How many people in the household are 14 years old or younger?	4 or more	3	2	1	0	
		0	3	8	11	19	
2.	Do all children of ages 6 to 14 attend school?			No	Yes	No children 6-14	
				0	3	3	
3.	Where does the household reside?				Not Port-a-Prince	Port-a-Prince	
					0	15	
4.	Does the household own a radio/cassette player?				No	Yes	
					0	7	
5.	What are the dwelling's floors made of?			Earth	Concrete or other	Ceramic or wood planks	
				0	4	12	
6.	In the past 12 months, did the household receive any money or gifts remitted from abroad?				No	Yes	
					0	7	
7.	Does any household member have salaried employment?				No	Yes	
					0	12	
8.	How many plots of agricultural land, forest land, pasture land, or gardens does the household use?		None	1	2 or 3	4 or more	
			2	0	5	11	
9.	What is the dwelling's roof made of?			Straw, palm leaves, or other	Iron	Concrete	
				0	4	9	
10.	Does the household own any pigs?				No	Yes	
					0	5	
Total:							

India PPI

Indicator			Values			Points	
1.	How many children aged 0 to 17 are in the household?	≥4	3	2	1	Zero	
		0	7	13	19	29	
2.	How many children aged 6 to 17 attend school?			There are no children	Not all attend	All attend	
				0	5	6	
3.	What is the household's primary energy source for cooking?			Firewood and chips, charcoal, dung cake, or no cooking arrangement	Electricity, coke, or coal	LPG, Kerosene, gobar gas, others, or unknown	
				0	4	11	
4.	Does the household own a television?				No	Yes	
					0	5	
5.	How many hectares of land does the household own?		Rural, ≤0.4 ha.	Rural, >0.4 and ≤1	Rural, >1 ha.	Urban, any amount	
			0	2	3	8	
6.	What is the principal occupation of the household?	Agricultural labourers	Operators and labourers, bricklayers, and construction workers	Cultivators, farmers, fishers, hunters, loggers, or unknown	Sales workers, service workers, and transport-equipment operators	Professional, technical, clerical, administrative, managerial, executive, and teachers	
		0	5	8	10	12	
7.	How many almirah/dressing tables does the household own?			0	1	2 or more	
				0	2	8	
8.	Does the household own a scooter or motorcycle?				No	Yes	
					0	11	
9.	Is the residence all pucca (burnt bricks, stone, cement, concrete, jackboard/cement-plastered reeds, timber, tiles, galvanised tin or asbestos cement sheets)?				Not all pucca	All pucca	
					0	5	
10.	Does the household own a pressure cooker or pressure pan?				No	Yes	
					0	7	
Source: Calculations based on Schedule 1.0 of the 59th Round (2003) of India's Socio-Economic Survey (NSSO, 2005).							Total:

Mexico PPI

Pregunta				Respuesta			Puntos
1.	¿Qué combustible utiliza para cocinar?			Leña	Gas		
				0	10		
2.	¿Esta vivienda cuenta con una regadera?			No	Sí		
				0	5		
3.	¿De qué material es la mayor parte de los pisos de la vivienda?			Cemento o Tierra	Madera, loseta o mosiaco		
				0	4	9	
4.	En los últimos tres meses, ¿Compró calzado para una persona de 17 años o más?			No	Sí		
				0	4		
5.	En los últimos tres meses, ¿Compró prendas de vestir para una persona de 17 años o más?			No	Sí		
				0	4		
6.	¿Cuántos miembros del hogar son de edades de 0 a 17 años?			Cuatro o más	Tres	Dos	
				0	10	13	
7.	¿Cuenta con teléfono fijo o teléfono celular?			No	Sí		
				0	6		
8.	¿Cuántos miembros del hogar son "obreros o empleados"?			Cero	Uno	Dos o mas	
				0	3	6	
9.	¿Cuenta con automóvil, camioneta, etc.?			No	Sí		
				0	5		
10.	¿Cuenta con un horno de microondas?			No	Sí		
				0	6		
11.	En la semana pasada, ¿Consumió en el hogar refrescos embotellados de cola o sabores?			No	Sí		
				0	5		
12.	En la semana pasada, ¿Consumió en el hogar carne de res o ternera en cualquier forma?			No	Sí		
				0	3		
13.	En la semana pasada, ¿Consumió en el hogar leche en cualquier forma?			No	Sí		
				0	2		
14.	El los últimos 3 meses, ¿Viajó alguien del hogar por colectivo pesero?			No	Sí		
				0	4		
15.	¿Cuenta con una batidora?			No	Sí		
				0	4		
Fuente: Cálculos basados en ENIGH 2002 por Microfinance Risk Management, L.L.C.						Total:	

Pakistan PPI

Indicator		Attributes		Points
1.	Do all children of ages 6 to 17 attend school?	No	Yes	No children this age
		0	10	21
2.	In the past two weeks, did anyone in the household eat any chicken?	No	Yes	
		0	6	
3.	What is the household's main source of drinking water?	Hand pump	Any other source	
		0	3	
4.	In the past two weeks, did anyone in the household eat any curd or yoghurt?	No	Yes	
		0	5	
5.	Does the household own a refrigerator or freezer?	No	Yes	
		0	7	
6.	In the past month, did anyone in the household spend anything on telephone, telegraph, postal, fax, e-mail, internet, etc.?	No	Yes	
		0	8	
7.	In the past two weeks, did anyone in the household eat any mutton?	No	Yes	
		0	6	
8.	What type of toilet is used by the household?	All others	Flush connect to pit	Flush connected to public sewerage
		0	3	7
9.	In the past two weeks, did anyone in the household eat any Desi ghee?	No	Yes	
		0	10	
10.	In the past two weeks, did anyone in the household eat any tomatoes?	No	Yes	
		0	5	
11.	In the past year, did anyone in the household buy a Burka, Chadar, Ajrak, etc.?	No	Yes	
		0	6	
12.	In the past month, did anyone in the household use any shampoo?	No	Yes	
		0	4	
13.	In the past two weeks, did anyone in the household eat any eggs?	No	Yes	
		0	2	
14.	In the past month, did anyone in the household use any gas from a cylinder?	No	Yes	
		0	6	
15.	In the past two weeks, did anyone in the household eat any apples?	No	Yes	
		0	5	
Source: Calculations by Microfinance Risk Management, L.L.C. based on the 2001 PIHS.				Total:

Philippines PPI

Indicator		Values		Points
1. Does the family own a gas stove or gas range?		No	Yes	
		0	13	
2. How many people in the family are aged 0 to 17?	≥5	3 or 4	Zero	
	0	6	15	26
3. How many television sets does the family own?	Zero	1	≥2	
	0	9	20	
4. What are the house's outer walls made of?		Light (cogon, nipa, or sawali, bamboo, anahaw)	Strong (iron, aluminum, tile, concrete, brick, stone, wood, asbestos)	
		0	4	
5. Do any family members have salaried employment?		No	Yes	
		0	7	
6. How many radios does the family own?	Zero	1	≥2	
	0	3	12	
7. Does the family own a sala set?		No	Yes	
		0	8	
8. What is the house's roof made of?		Light (Salvaged, makeshift, cogon, nipa, or anahaw)	Strong (Galvanized iron, aluminum tile, concrete, brick, stone, or asbestos)	
		0	2	
9. What kind of toilet facility does the family have in the house?		None, open pit, closed pit, or other	Water sealed	
		0	3	
10. Do all children in the family of ages 6 to 11 go to school?	No	Yes	No children ages 6-11	
	0	2	4	
Source: Calculations based on the 2002 APIS.				Total: <input type="text"/>

Day-to-Day Use of PPI

- **Photocopy, go to field, ask/observe questions**
- **Circle responses, add up points**
- **Apply simple score-based policy, e.g.:**
 - **Score $\leq x$, qualify**
 - **Score $> y$, disqualify**
- **Put scorecard in client file, and perhaps record score and/or indicator values in database**
- **Reward field agents and branches based on:**
 - **Portfolio poverty rate**
 - **Changes in portfolio poverty rate**

Scores and Poverty Likelihoods

- **No score (not even 0 or 100) gives certainty of being poor or non-poor (always margin of error)**
- **Use survey data to convert scores to *poverty likelihoods* (probability poor for a given score)**
- **India example:**
 - **94 of 95 in survey w/scores 0–4 were poor, so poverty likelihood for 0–4 is $94 \div 95 = 98.8\%$**
 - **167 of 2,720 in survey w/scores 50–100 were poor, so poverty likelihood for 50–100 is $167 \div 2,720 = 6.1\%$**

India: Poverty Likelihoods

Score	Poverty likelihood for people with score in range (%)	% of people <=score who are poor	% of people >score who are non-poor
0-4	98.7	98.7	54.1
5-9	97.0	97.5	55.2
10-14	95.3	96.2	57.7
15-19	90.9	93.6	61.9
20-24	80.7	88.8	66.9
25-29	72.5	83.9	72.9
30-34	61.7	78.2	80.4
35-39	42.5	71.5	85.9
40-44	32.9	66.4	90.5
45-49	21.1	61.5	93.8
50-54	13.3	57.5	96.1
55-59	10.0	54.3	98.4
60-64	3.3	51.7	99.2
65-69	1.5	49.7	99.5
70-74	0.6	48.3	99.6
75-79	0.1	47.3	99.4
80-84	0.6	46.7	99.5
85-89	0.7	46.5	100.0
90-94	0.0	46.4	100.0
95-100	0.0	46.4	N/A
Total:	46.4		

Surveyed cases weighted to represent all India.

Source: Calculations by Microfinance Risk Management, L.L.C. based on Schedule 1.0 of the 59th Round (2003) of India's Socio-Economic Survey by the National Sample-Survey Organisation

Setting Cut-Off for Targeting

- **Set by programs, not consultant nor USAID:**
 - **Based on values and mission**
 - **Balance ‘benefit’ of covering poor versus ‘cost’ of leaking to non-poor**
 - **PPI makes explicit the leakage to non-poor that is always inevitably there**
- **India: If target cut-off score is 24, 88% will be poor (if random sample of population)**
- **Choice of target cut-off score does *not* define the poverty line**

Portfolio Poverty Rates

The share of all clients who are poor is the average of their poverty likelihoods.

Indian example, portfolio of 3 clients, 1/1/06

Score		
Client	1/1/06	Poverty likelihood (%)
A	20	80.7
B	30	61.7
C	40	32.9
Average:		58.4

USAID will mandate reporting this.

Accuracy

- **Test method: ‘Out-of-sample bootstrap’:**
 - **Use part of data to build PPI**
 - **Test 10,000 samples from other part**
- **Portfolio poverty rates (90% confidence):**
 - **Bangladesh: +/- 1.5 percentage points**
 - **India: +/- 1.7 percentage points**
 - **Haiti: +/- 1.9 percentage points**
- **Individual poverty likelihoods (90% confidence) are generally about +/- 5 to 10 percentage points**

Poverty Progress (or Regress)

Change (*not impact*) in portfolio poverty rates.

Indian example, 3 clients, 1/1/06 to 1/1/07

Client	<u>Score</u>		<u>Poverty likelihood (%)</u>	
	1/1/06	1/1/07	1/1/06	1/1/07
A	20	25	80.7	72.5
B	30	35	61.7	42.5
C	40	60	32.9	3.3
Average:			58.4	39.4

- **19.0% = 58.4 – 39.4 of all clients left poverty**
- **32.5% = 19.0 ÷ 58.4 of poor clients left poverty**

Scoring Policy

Programs set their own policy cut-offs based on:

- **Their mission and values**
- **Benefits/costs of classification outcomes**

		<u>Poverty segment</u>	
		<u>Poor</u>	<u>Non-poor</u>
<u>True poverty status</u>	<u>Poor</u>	<u>A.</u> Truly poor correctly classified in poor segment	<u>B.</u> Truly poor incorrectly classified in non-poor segment
	<u>Non-poor</u>	<u>C.</u> Truly non-poor incorrectly classified in poor segment	<u>D.</u> Truly non-poor correctly classified in non-poor segment

India: Classification Outcomes

	A.	B.	C.	D.
	Truly poor	Truly poor	Truly non-poor	Truly non-poor
	correctly classified	incorrectly classified	incorrectly classified	correctly classified
Score	in poor segment	in non-poor segment	in poor segment	in non-poor segment
0-4	94	4,523	1	5,339
5-9	298	4,320	7	5,333
10-14	719	3,899	28	5,312
15-19	1,385	3,233	95	5,245
20-24	2,103	2,515	266	5,074
25-29	2,840	1,778	547	4,793
30-34	3,559	1,059	994	4,347
35-39	4,005	613	1,598	3,742
40-44	4,285	333	2,169	3,171
45-49	4,450	167	2,787	2,553
50-54	4,537	81	3,351	1,989
55-59	4,594	24	3,863	1,477
60-64	4,609	9	4,299	1,041
65-69	4,615	3	4,667	673
70-74	4,616	1	4,933	407
75-79	4,617	1	5,140	200
80-84	4,617	0	5,266	75
85-89	4,618	0	5,321	19
90-94	4,618	0	5,337	4
95-100	4,618	0	5,340	0
Total:	4,618		5,340	

Surveyed cases weighted to represent all India, in units of 100,000 people.

Total Net Benefit: Indian Example

Policy cut-off: If ≤ 34 , count as 'poor'

Classification	People		Net benefit
A. Poor correct	3,559	x	+3
B. Poor incorrect	1,059	x	-2
C. Non-poor incorrect	994	x	-2
D. Non-poor correct	4,347	x	+1
Total net benefit:			10,918

Policy cut-off: If ≤ 39 , count as 'poor'

Classification	People		Net benefit
A. Poor correct	4,005	x	+3
B. Poor incorrect	613	x	-2
C. Non-poor incorrect	1,598	x	-2
D. Non-poor correct	3,742	x	+1
Total net benefit:			11,335

Setting Cut-Offs

- **Cut-offs represent trade-offs between:**
 - 1. Undercoverage (poor classified as non-poor)**
 - 2. Leakage (non-poor classified as poor)**
- **Being intentional and explicit about these inevitable trade-offs is unusual but healthy**
- **Programs (not Congress) decide how policy depends on poverty. But programs still do not:**
 - **Choose poverty lines**
 - **Manipulate portfolio poverty rates**

Accuracy

- 1. Estimated portfolio poverty rate:**
 - Bias, consistency, mean-squared error**
 - Test w/data not used to build scorecard**
- 2. Classification of individuals (concentrating poor in low scores, non-poor in high scores)**
 - Best: Net benefit at specific cut-offs**
 - # poor/non-poor classified right/wrong, by score**
 - Global 'c' (area under ROC curve)**

No scorecard is perfect (or even close).

Appropriate measure depends on scorecard purpose.

Compared w/others, GFUSA scorecards are accurate.

How Many Indicators?

GFUSA tested 5, 10, and 15

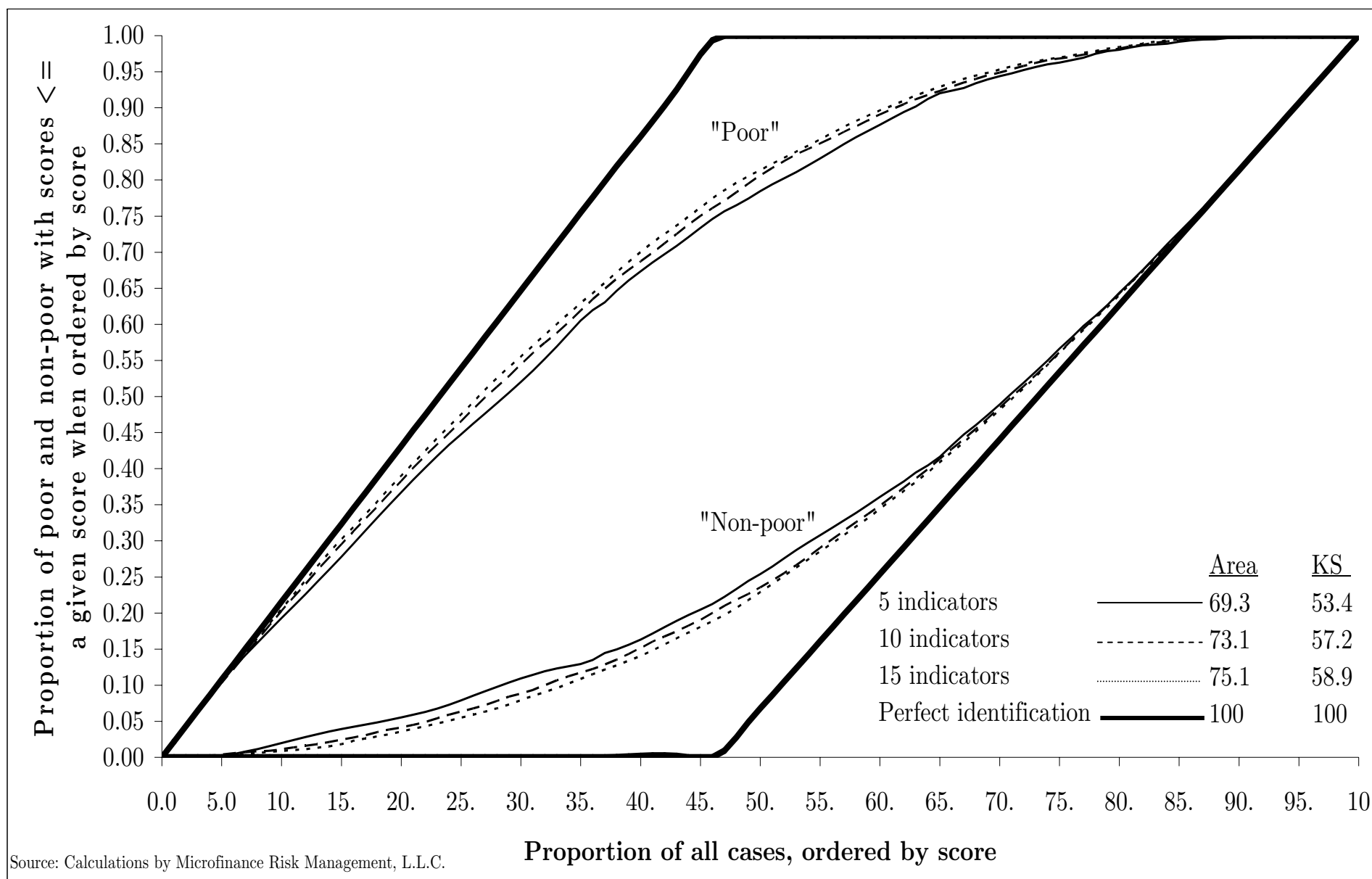
Costs of more indicators:

- **Longer interview**
- **Improves accuracy only a little ('flat max')**

Benefits of more indicators:

- **More sensitive to small changes over time**
- **Less sensitive to inaccurate data**
- **Smaller margin of error**

ROC for India: 5, 10, 15 indicators



What Regression Technique?

- 1. IRIS: Estimate expenditure**
- 2. GFUSA: Estimate poverty likelihood**
- 3. Hentschel: Expenditure, convert to likelihood**

'Flat max':

- Diminishing returns to sophistication**
- Regardless of methods or weights, most reasonable scorecards w/many indicators have similar accuracy**
- Additional indicators add little new info.**
- Poverty is relatively easy to score**

Wainer (1976): 'It don't make no nevermind.'

Comparison of Techniques

Aspect:	IRIS	Hentschel	GFUSA
Product of estimation:	Expenditure	Expenditure converted to likelihood	Poverty likelihood
Statistical model	OLS	OLS	Logit
Objective link of scores to data:	Yes	Yes	Yes
Considers distance from poverty line:	Yes	Yes	No
Steps in computing score:	2	2	1
Risk of overfitting:	High	Medium	Low
Complexity:	Most	Medium	Least
Accuracy, portfolio poverty rate:	Biased	Good	Unbiased
Accuracy, classification of individuals:	Highest	High	Good
Indicator criterion:	Stepwise R^2	Stepwise R^2	Stepwise 'c'/ROC and 'expert'

Objectivity

Objective: Scores linked to poverty via data

All GFUSA scorecards are objective: Poverty likelihood = share of surveyed HH with score who were poor by an expenditure poverty line

How to ‘certify’ scorecards?

- Not how indicators and weights are selected (data vs. ‘expert’, fancy vs. simple)**
- What matters is linking scores (regardless of how derived) to survey data**
- Basel II uses a similar criterion to qualify scorecards for credit-risk assessment**

Summary

- **GFUSA's PPIs are simple, easy-to-use, inexpensive, accurate, and objective**
- **Estimate likelihood that a person is poor:**
 - **Use policy cut-offs for targeting**
 - **Take average to get portfolio poverty rate**
 - **Track over time for progress out of poverty**
- **Put 'practicality' before accuracy (KISS):**
 - **One page, few indicators, simple weights**
 - **Field workers can compute scores on paper in real time (no software required)**
- **Useful for *any* program, not just microfinance**

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