

Supplement of

Geospatial micro-estimates of slum populations in 129 Global South countries using machine learning and public data

Supplementary Figures S1-S4

Supplementary Tables S1-S5

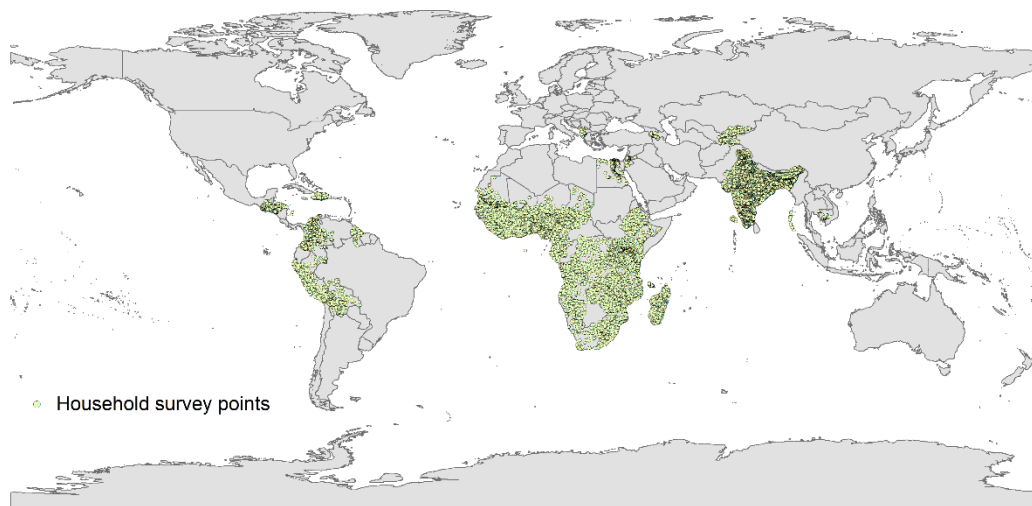


Figure S1 Locations of household surveys, referred to as clusters, collected from Demographic and Health Surveys conducted in 53 countries of the Global South. To protect respondent confidentiality, the GPS latitude/longitude coordinates of surveyed households within each cluster are randomly displaced.

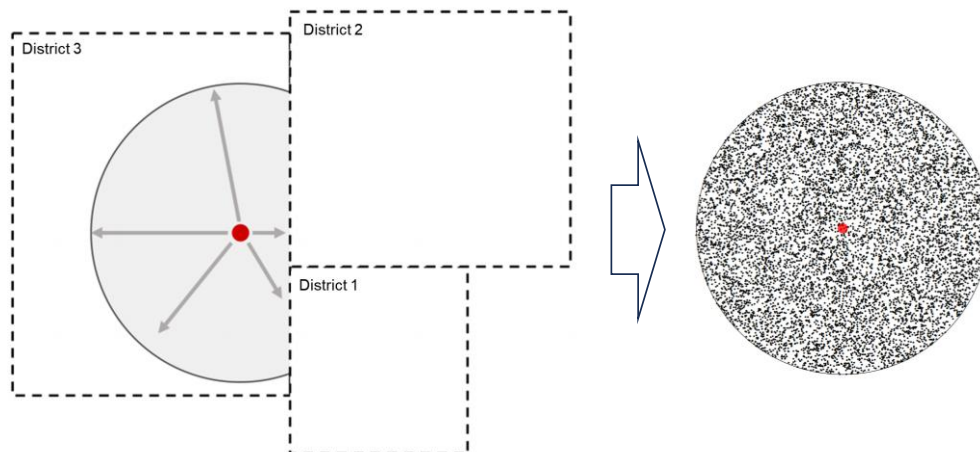


Figure S2 A illustration of GPS perturbation. The red point means the GPS coordinates after perturbation. Districts are at the administrative level 3.

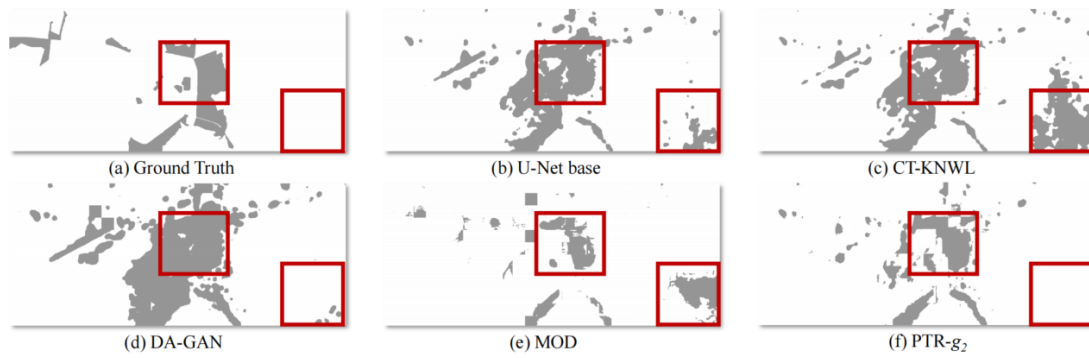


Figure S3 Examples of the uncertainty in morphological slum delineation using five image segmentation methods (b-f), compared to ground truth (a). sourced from Li et al., (2023). The existing slum population mapping frameworks heavily relied on slum geometry, exhibiting notable uncertainties.

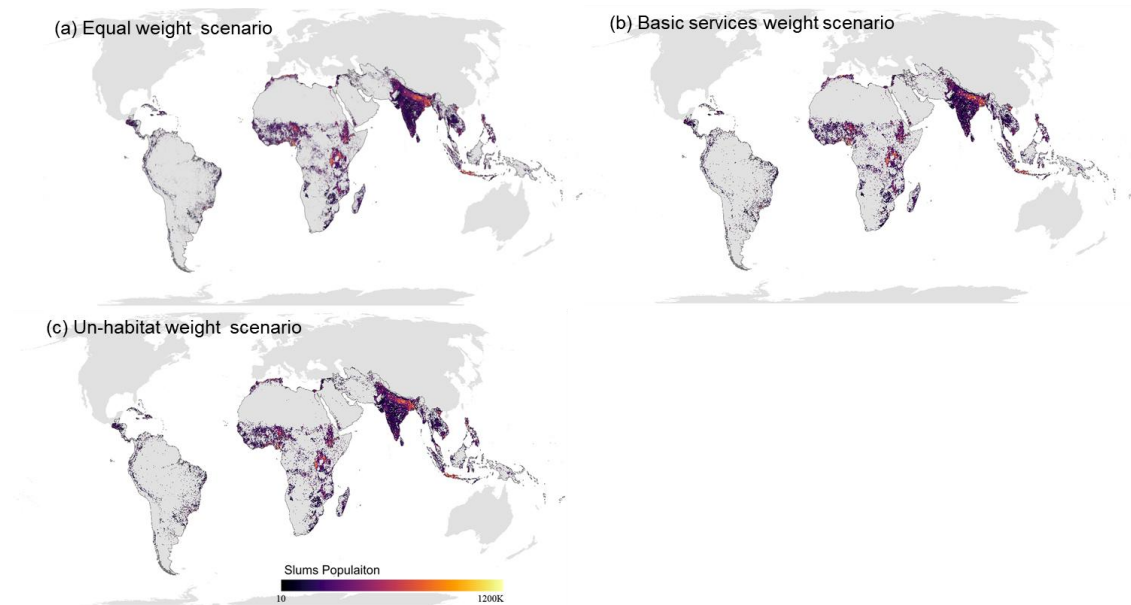


Figure S4 Maps of the slum population based on three different weighting scenarios of slum indicators. (a) equal weight scenario, (b) basic services weight scenario and (c) un-habitat weight scenario

Table S1 Household-based surveys for each country.

Number	Country name	Code	Year	Counts		
				Total	Urban	Rural
1	Angola	AO	2015	625	345	280
2	Burkina Faso	BF	2010	573	176	397
3	Benin	BJ	2017	555	251	304
4	Burundi	BU	2016	554	106	448
5	D. R. of Congo	CD	2013	536	161	375
6	Côte d'Ivoire	CI	2012	350	160	190
7	Cameroon	CM	2018	430	237	193
8	Egypt	EG	2018	1836	977	859
9	Ethiopia	ET	2016	643	202	441
10	Gabon	GA	2012	336	183	153
11	Ghana	GH	2014	427	216	211
12	The Gambia	GM	2019	280	173	107
13	Guinea	GN	2018	401	138	263
14	Kenya	KE	2014	1594	617	977
15	Comoros	KM	2012	252	108	144
16	Liberia	LB	2019	325	120	205
17	Lesotho	LS	2014	399	118	281
18	Madagascar	MD	2021	650	164	486
19	Mali	ML	2018	345	104	241
20	Mauritania	MR	2020	1200	565	635
21	Malawi	MW	2015	850	173	677
22	Mozambique	MZ	2011	610	256	354
23	Nigeria	NG	2018	1389	576	813
24	Niger	NI	2012	476	130	346
25	Namibia	NM	2013	550	267	283
26	Rwanda	RW	2019	500	112	388
27	Sierra Leone	SL	2019	576	214	362
28	Senegal	SN	2019	214	84	130
29	Chad	TD	2014	624	163	461
30	Togo	TG	2013	330	128	202
31	U. R. Tanzania	TZ	2015	608	180	428
32	Uganda	UG	2016	696	162	534
33	South Africa	ZA	2017	746	464	282
34	Zambia	ZM	2018	545	198	347
35	Zimbabwe	ZW	2015	400	166	234
36	Bangladesh	BD	2018	672	226	446
37	India	IA	2020	30170	7780	22390
38	Jordan	JO	2017	970	769	201

39	Cambodia	KH	2014	611	188	423
40	Myanmar	MM	2015	441	122	319
41	Nepal	NP	2016	383	243	140
42	Philippines	PH	2017	1250	444	806
43	Pakistan	PK	2017	561	282	279
44	Tajikistan	TJ	2017	366	166	200
45	Timor Leste	TL	2016	455	129	326
46	Bolivia	BO	2008	998	593	405
47	Colombia	CO	2010	4868	3401	1467
48	Dominican R.	DR	2013	524	364	160
49	Guatemala	GU	2015	853	381	472
50	Guyana	GY	2009	312	87	225
51	Honduras	HN	2011	1128	494	634
52	Haiti	HT	2016	450	152	298
53	Peru	PE	2009	1131	692	439

Table S2 Definitions of the 10 slum sub-indicators and their corresponding criteria.

Indicators		Improved	Unimproved
1	House made of finished materials (roof, wall and floor)	Finished roof materials typically include metal, wood, cement fiber, ceramic tiles, cement and roofing shingles	Natural or rudimentary roof materials typically include thatch, palm leaf, sod, rustic matting, bamboo, wood planks and cardboard.
		Finished wall materials typically include cement, stone with lime or cement, burnt bricks, cement blocks, covered adobe and wood planks or shingles	Natural or rudimentary wall materials typically include cane, palm, dirt, bamboo or stone with mud, uncovered adobe, plywood, cardboard and reused wood planks.
		Finished floor materials include parquet or polished wood, vinyl or asphalt strips, ceramic tiles, cement and carpet	Natural or rudimentary floor materials include earth, sand, dung, wood planks, palm and bamboo
2	House with a sufficient living room	no more than 3 people sharing one sleeping room	more than 3 people sharing one sleeping room
3	Household using safely managed drinking water	Piped water into dwelling Piped water to yard/plot Public tap or standpipe Tubewell or borehole Protected dug well Protected spring Rainwater	Unprotected spring Unprotected dug well Tanker-truck Cart with small tank/drum Surface water Bottled water
4	Household with access to water availability for continuous two weeks	Yes, not interrupted for a full day	No, interrupted for a full day or more
5	Household with access to drinking water located within a round trip of 30 minutes	<30, on premises	>30, don't know
6	Household using safely managed sanitation toilets	Flush toilet Piped sewer system Septic tank Flush/pour flush to pit latrine Ventilated improved pit latrine (VIP) Pit latrine with slab Composting toilet	Flush/pour flush to elsewhere Pit latrine without slab Bucket Hanging toilet or hanging latrine Shared sanitation No facilities or bush or field
7	Household using safely managed hand-washing facility with water.	Observed, fixed place	Not observed: no permission to see not observed: not in dwelling
8	Household using safely managed hand-washing facility with soap or detergent	Yes	No

9	Household with access to electricity	Yes	No
10	Household with clean cooking fuels	Electricity, Natural gas, Biogas, LPG	Cardboard/paper charcoal other petroleum/kerosene wood straw/shrubs/grass

Table S3 Regional models categorized by income levels and geographical locations.

	Regional models	Countries with available training labels	Countries with unavailable training labels
Africa	Low income& west of Africa countries Model	Burkina Faso; Chad; Democratic Republic of the Congo; The Gambia; Liberia; Mali; Niger; Sierra Leone; Togo	Central African Republic; Guinea-Bissau
	Low income& east of Africa countries Model	Burundi; Ethiopia; Madagascar; Malawi; Mozambique; Rwanda; Uganda	Eritrea; Somalia; Sudan
	Lower-middle income& west of Africa countries Model	Angola; Benin; Cameroon; Cote d'Ivoire; Ghana; Guinea; Mauritania; Nigeria; Senegal	Algeria; Cabo Verde; The Republic of the Congo; Morocco; Sao Tome and Principe; Tunisia; Western Sahara
	Lower-middle income& East of Africa countries Model	Comoros; Egypt; Kenya; Lesotho; the United Republic of Tanzania; Zambia; Zimbabwe	Djibouti; Swaziland (Eswatini)
	Upper-middle income countries Model	Gabon; Namibia; South Africa	Gabon; Namibia; South Africa;
Asia	Lower-middle income& west of Asia countries Model	Bangladesh; India; Jordan; Nepal; Pakistan; Tajikistan	Afghanistan; Bahrain; Bhutan; Iran; Iraq; Israel; Kuwait; Lebanon; Maldives; Oman; Qatar; Saudi Arabia; Sri Lanka; Syria; Turkmenistan; United Arab Emirates; West Bank; Yemen
	Lower-middle income& east of Asia countries Model	Cambodia; Myanmar (Burma); Philippines; Timor-Leste	Brunei; Fiji; Indonesia; Laos; Malaysia; Marshall Islands; Federated States of Micronesia; Nauru; Papua New Guinea; Singapore; Solomon Islands; Thailand; Vanuatu; Vietnam
Latin America	Lower-middle- and low income Model	Guyana; Colombia; Dominican Republic; Guatemala; Peru	Antigua and Barbuda; Barbados; Chile; Panama; St. Kitts and Nevis; The Bahamas; Trinidad and Tobago; Uruguay; French Guiana; Argentina; Belize; Brazil; Costa Rica; Cuba; Dominica; Ecuador; El Salvador; Grenada; Jamaica; Paraguay; St. Lucia; St. Vincent and the Grenadines; Suriname
	High- and upper middle- income Model	Bolivia; Haiti; Honduras	Marshall Islands; Federated States of Micronesia

Table S4 Optimal Hyperparameters of XGBoost Models for all three weighting scenarios.

Model			colsample_ bytree	gamma	learning_ rate	max_depth	n_estimators	subsample	
Low income	West Africa	of	Scenario1	0.6	0	0.1	3	65	0.8
			Scenario2	0.6	0	0.05	3	100	1
			Scenario3	0.6	0	0.1	3	81	1
	East Africa	of	Scenario1	1	0.2	0.05	5	117	0.8
			Scenario2	0.6	0	0.1	3	63	1
			Scenario3	0.6	0	0.1	3	55	1
Lower middle income	West Africa	of	Scenario1	0.6	0	0.1	3	94	0.8
			Scenario2	0.6	0.1	0.1	3	92	1
			Scenario3	0.6	0	0.1	3	110	1
	East Africa	of	Scenario1	0.6	0.1	0.1	3	56	0.8
			Scenario2	1.0	0	0.1	3	55	0.8
			Scenario3	1	0.2	0.1	3	63	0.8
Upper middle income	Asia		Scenario1	1	0.3	0.1	3	48	1
			Scenario2	0.6	0	0.1	3	54	1
			Scenario3	1	0.3	0.05	3	219	0.8
Lower middle income	West Asia	of	Scenario1	0.8	0	0.03	7	228	0.8
			Scenario2	0.8	0.1	0.05	6	138	0.8
			Scenario3	0.8	0.2	0.05	5	142	1
Lower middle income	East of Asia		Scenario1	0.6	0	0.1	3	69	1
			Scenario2	0.8	0	0.1	3	57	1
			Scenario3	1	0.2	0.1	3	97	1
Lower- middle and low income	Latin American		Scenario1	0.8	0	0.1	3	55	1
			Scenario2	1	0	0.1	3	51	1
			Scenario3	0.6	0.3	0.1	6	50	1
High and upper middle income	Latin American		Scenario1	0.8	0.2	0.1	5	93	0.8
			Scenario2	1	0	0.1	3	102	0.8
			Scenario3	0.8	0	0.1	3	84	1

Table S5 The country-specific model's performance measured by the root mean square error (RMSE) and R squared (R^2)

Country	RMSE	R^2	Country	RMSE	R^2
AO	6.79	0.93	LS	3.91	0.95
BD	4.68	0.83	MD	6.62	0.87
BF	5.62	0.90	ML	5.25	0.90
BJ	6.38	0.86	MM	4.50	0.87
BO	6.47	0.93	MR	7.64	0.91
BU	4.63	0.87	MW	3.95	0.89
CD	3.42	0.94	MZ	5.98	0.90
CI	7.59	0.85	NG	5.53	0.90
CM	4.49	0.95	NI	7.25	0.90
CO	5.98	0.88	NM	8.14	0.86
DR	3.19	0.82	NP	5.36	0.81
EG	1.38	0.92	PE	5.75	0.91
ET	4.10	0.97	PH	3.07	0.90
GA	3.51	0.93	PK	7.40	0.78
GH	6.51	0.77	RW	5.87	0.82
GM	2.88	0.95	SL	6.21	0.82
GN	5.26	0.91	SN	7.95	0.78
GU	4.85	0.85	TD	4.59	0.91
GY	4.35	0.92	TG	5.63	0.91
HN	5.80	0.77	TJ	3.68	0.81
HT	5.71	0.84	TL	6.43	0.84
IA	4.38	0.86	TZ	7.19	0.85
JO	2.55	0.79	UG	4.45	0.93
KE	5.98	0.89	ZA	4.45	0.93
KH	4.83	0.94	ZM	9.08	0.79
KM	3.52	0.81	ZW	4.61	0.95
LB	2.99	0.92			

References:

Li, Z., Xie, Y., Jia, X., Stuart, K., Delaire, C. and Skakun, S., 2023. Point-to-region co-learning for poverty mapping at high resolution using satellite imagery. In *Proceedings of the AAAI Conference on Artificial Intelligence*, 37(12), 14321-14328.