How Many People Live Near Protected Areas in Developing Countries? Estimates from Gridded Population Data (2000–2020)

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Abstract Protected areas (PAs) are the mainstream biodiversity conservation tool. While some research has examined the socioeconomic implications of conservation policies, the number of people affected by PAs remains unquantified. We estimate the population living within and near PAs in 2000 to 2020 in 76 countries classified as low-and lower-middle-income by the World bank in 2020. We use the World Database on Protected Areas and compute population estimates from two high-resolution gridded datase. Terrestrial PA coverage in developing countries increased from 9.3% in 2000 to 12.9% in 2020, while the share of the population living within 10 km of a PA rose from 16.4% to 21.8%. Excluding India, where PA expansion was minimal, coverage increased from 10.2% to 14.1%, and proximate populations from 26.5% to 33.5%. This scaling up of conservation's demographic outreach highlights the need for a more thorough assessment of its socioeconomic implications as efforts converge to expand PA coverage to 30% of terrestrial land by 2030.

1 Introduction

Protected areas (PAs) are a cornerstone of biodiversity conservation, primarily established for ecological purposes (Maxwell et al. 2020). However, their creation can have positive or negative effects on local populations (Kandel et al. 2022). PAs restrict access to natural resources (e.g., gathering, hunting, fishing, medicinal plants), limit land availability, and curb economic activities such as agriculture, livestock, and construction. Contoversely, they can provide benefits through compensation measures (e.g., local development projects, cash transfers), job creation (e.g., conservation-related employment, tourism), and ecosystem services (e.g., improved water resources, erosion control, fire prevention).

The impact of conservation on local well-being remains debated (Adams et al. 2004). Yet, no study has systematically quantified the population potentially affected by PAs. While some research has examined population density trends around PAs [joppa2009b; geldmann2019], none have assessed the demographic implications of PA creation or expansion.

The proportion of the population directly impacted by PAs is a critical issue for conservation planning, determining the distribution of benefits and burdens across social groups, shaping both local reception and long-term feasibility. PA expansion is thus not solely an ecological decision but a political one, affecting policy support from local elites, voters, and decision-makers, who balance conservation goals with social welfare mandates (Mangonnet, Kopas, and Urpelainen 2022). Recognizing the demographic extent of PA-affected populations is therefore crucial, as neglecting their socioeconomic realities risks generating resistance and conflict, undermining conservation effectiveness.

The rapid expansion of PAs makes this issue increasingly pressing. From 2000 to 2023, global terrestrial PA coverage expanded by 62%, reaching 17.6% of land area (UNEP-WCMC and IUCN 2024). This growth is set to accelerate with the COP15 commitment to protect 30% of terrestrial land by 2030. While crucial for biodiversity, these efforts also raise urgent questions about so-cioeconomic impacts. As more land falls under protection, more people will be affected, either through restricted resource access or conservation-driven economic opportunities. Quantifying how many people live near PAs is essential to designing policies that balance ecological and human well-being.

2 Methods

2.1 Study Area

The study focuses on countries classified by the World Bank as low-income and lower-middle-income economies in 2020, based on Gross National Income per capita, adjusted using the Atlas method (Vaggi 2017). We excluded South Sudan and Timor Leste, as they did not exist in 2000. The areas currently corresponding to these two countries were removed from the 2000 dataset.

We set a 10 km threshold to define the treatment group, a standard practice for assessing socioe-conomic impacts of protected areas in developing countries (Naidoo et al. 2019; Oldekop et al. 2016).

2.2 Data Sources

The national areas were selected using geoBoundaries (v.6.0.0), an open, standardized, and regularly updated dataset (Runfola et al. 2020).

The WDPA (February 2025 edition), managed by UNEP-WCMC in collaboration with IUCN, is the most comprehensive dataset on protected areas, updated monthly with submissions from national governments, non-governmental organizations, landowners, and local communities (Bingham et al. 2019; UNEP-WCMC and IUCN 2023). The status year field records the year in which a site was formally designated, allowing for historical analyses of PA expansion.

The Global Human Settlement Layer (GHSL) draws from Gridded Population of the World (G-PWv4) (Center for International Earth Science Information Network-CIESIN-Columbia University 2018). GHSL improves on this by integrating remote sensing and crowdsourced data, using regression models to refine 250m resolution estimates (Freire et al. 2016; Schiavina, Freire, and MacManus 2015). It has been assessed as the most reliable global gridded population dataset by

Chen et al. (2020), and we use GHSL population estimates as the default reference unless stated otherwise.

The second-best dataset according to Chen et al. (2020) is Worldpop, which applies a random forest algorithm trained on census data, land cover, infrastructure, elevation, and mobile phone records to produce high-resolution (100m) estimates (Stevens et al. 2015). It supports spatially explicit analyses of human presence near PAs. We use its 2000 and 2020 datasets (WorldPop and CIESIN 2020).

2.3 Analytical approach

We used Google Earth Engine (GEE), a cloud-based platform for large-scale environmental analysis (Gorelick et al. 2017), to process high-resolution population estimates within and around protected areas across 75 countries in 2000 and 2020. The computation, executed in iterative batches, handled billions of pixels over 100 hours.

We extracted protected area boundaries from the WDPA, excluding UNESCO-MAB Biosphere Reserves, which often lack legal protection (Hanson 2022; Coetzer, Witkowski, and Erasmus 2014). We retained only terrestrial PAs or the terrestrial portion of coastal PAs that intersected geoBoundaries and excluded purely offshore PAs. We generated binary PA presence layers at 100m resolution and created 10-km buffer zones to assess population proximity. We applied zonal statistics to WorldPop data, aggregating results at the national level for small and medium-sized countries (<1,000,000 km²) and at the subnational level for large ones to manage memory constraints.

To assess the consistency of GHSL and WorldPop population estimates, we compared values per country and year. Absolute differences were unsuitable means of comparison, as they are influenced by the estimate levels, which vary significantly across countries and depending on whether we consider PAs or surrounding areas. We used the Standardized Mean Difference (SMD), which standardizes the difference by accounting for variations in scale and dispersion. It is computed as:

$$SMD = \frac{\left| \boldsymbol{X}_{1}^{^{-}} - \boldsymbol{X}_{2}^{^{-}} \right|}{\sqrt{\frac{\sigma_{1}^{2} + \sigma_{2}^{2}}{2}}}$$

where X_1^- and X_1^- are the mean population estimates from respectively WorldPop and GHSL, and σ_1^2 and σ_2^2 are their variances.

We conducted the final analyses in R (R Core Team 2023), using a coherent framework for data processing (Wickham et al. 2019) and packages optimized for graphical visualisation (Wickham 2016; Slowikowski 2023; Wilke 2024) and tabular presentation (Iannone et al. 2023). The GEE JavaScript code, R code and output dataset are included in the replication package available online (Anonymized, 2025).

3 Results

3.1 Population

Table 1 summarizes the extent of PAs and population proximity to PAs in low- and lower-middle-income countries. It presents the percentage of national land within PAs and the percentage of population within 10 km of PAs for the years 2000 and 2020. Due to its large population and relatively low PA coverage, India has an overwhelming impact on the global average for developing countries. To account for this, we provide two totals: one including India and another excluding it.

Table 1: PA Coverage and Population Proximity (2000-2020)

Percentage of land and population within and near protected areas

Tercentage of land and population within and field protected areas												
	%	land witl	hin (WDI	PA)	%	populat	tion (GH	SL)	% population (WorldPop)			
	withi	n PAs	10km	of PAs	withi	n PAs	10km	of PAs	withi	n PAs	10km	of PAs
Country	2000	2020	2000	2020	2000	2020	2000	2020	2000	2020	2000	2020
Total	9.3	12.9	21.0	27.2	0.9	1.8	16.4	21.8	1.8	2.7	16.2	21.8
Total without India	10.2	14.1	22.9	29.5	1.4	2.8	26.5	33.5	2.8	4.2	26.1	33.6
Afghanistan	0.1	3.7	0.6	6.0	0.1	1.2	0.6	14.9	0.1	1.1	0.6	13.6
Angola	5.4	10.8	9.0	15.8	0.4	1.0	2.5	4.3	0.9	1.3	3.0	5.0
Bangladesh	4.0	4.5	9.8	17.8	0.3	0.5	5.0	18.8	0.6	1.0	4.6	18.3
Benin	23.7	23.7	54.9	54.9	1.7	1.8	25.7	25.7	5.2	5.7	25.8	26.5
Bhutan	38.5	51.1	72.0	80.5	4.4	4.5	55.2	56.1	11.2	11.6	56.2	58.4
Bolivia	20.9	23.3	33.4	38.4	22.2	31.0	61.3	68.8	21.1	25.8	58.3	63.0
Burkina Faso	14.2	16.6	33.3	38.8	3.3	4.0	31.0	42.7	6.1	7.8	31.5	45.8
Burundi	5.4	5.4	40.8	41.4	0.3	0.3	40.1	44.7	2.8	2.6	39.6	43.9
Cabo Verde	0.0	18.4	0.0	94.9	0.0	8.8	0.0	96.0	0.0	14.1	0.0	94.6
Cambodia	14.1	39.9	34.3	68.5	2.1	5.3	13.5	30.7	2.4	7.7	13.6	31.3
Cameroon	5.2	11.1	13.0	25.4	0.1	3.3	4.5	25.1	0.9	6.0	5.2	24.4
Central African Republic	14.0	14.0	21.8	21.9	0.6	0.6	19.8	21.3	2.0	1.8	22.2	23.1
Chad	11.4	13.9	14.7	18.4	3.8	3.8	11.8	13.6	6.5	7.7	11.7	16.8
Comoros	0.8	33.0	6.5	97.8	0.5	11.9	36.1	99.5	1.7	38.7	21.7	98.9
Congo, Dem. Rep.	9.0	14.9	14.9	23.0	2.5	6.1	23.3	28.1	3.9	7.3	22.7	24.5
Congo, Rep.	9.3	36.5	16.3	53.1	0.7	13.2	49.2	74.3	3.3	26.9	31.7	66.4
Côte d'Ivoire	21.5	21.7	69.5	70.0	5.1	5.6	69.7	71.6	10.0	9.3	71.4	72.6

Source: Analysis based on WDPA, WorldPop & GHSL data (2000-2020)

Table 1: PA Coverage and Population Proximity (2000-2020)

Percentage of land and population within and near protected areas

	%	land with	nin (WDI	PA)	% population (GHSL)				% population (WorldPop)			
	within PAs		10km of PAs		within PAs		10km of PAs		within PAs		10km of PAs	
Country	2000	2020	2000	2020	2000	2020	2000	2020	2000	2020	2000	2020
Djibouti	0.0	1.3	0.0	8.3	0.0	0.0	0.0	1.4	0.0	0.8	0.0	7.8
Egypt, Arab Rep.	4.3	11.2	6.5	15.7	0.1	0.3	12.6	13.4	0.4	0.5	11.2	11.3
El Salvador	0.1	8.6	8.2	80.7	0.0	2.9	4.2	90.2	0.0	3.8	4.2	90.3
Eritrea	0.0	0.0	0.0	1.3	0.0	0.0	0.0	0.9	0.0	0.0	0.0	1.6
Eswatini	4.1	4.4	28.3	38.4	0.2	0.1	34.1	41.8	1.6	1.5	30.7	39.5
Ethiopia	10.4	17.0	17.6	27.2	3.2	7.0	9.8	17.4	3.8	7.7	9.8	17.8
Gambia, The	5.1	8.0	32.2	61.8	0.2	1.0	57.2	79.2	2.8	8.8	55.8	78.9
Ghana	14.7	14.9	56.9	57.2	-1.2	-0.2	68.2	69.0	6.2	5.0	69.1	68.9
Guinea	7.9	23.7	35.9	52.3	2.6	14.9	46.7	57.1	4.0	15.4	47.3	59.0
Guinea-Bissau	18.0	24.4	40.8	49.9	2.8	5.8	16.7	20.7	5.8	9.4	20.9	25.3
Haiti	2.2	7.4	16.6	46.0	0.9	3.3	10.4	50.8	1.0	3.2	10.8	54.4
Honduras	20.4	23.0	54.6	63.7	4.0	5.6	67.5	76.3	8.2	9.2	65.5	74.1
India	0.2	0.5	1.1	2.4	0.0	0.0	0.5	1.1	0.1	0.1	0.5	1.1
Indonesia	11.5	11.9	31.5	33.9	0.2	0.5	32.1	34.6	1.7	2.2	31.7	34.6
Kenya	10.8	12.7	32.4	38.2	2.2	2.7	57.0	59.4	4.0	4.2	59.4	60.7
Kiribati	0.0	4.6	0.0	4.6	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Korea, Dem. Rep.	0.0	0.0	0.0	1.1	0.0	0.0	0.0	0.4	0.0	0.0	0.0	0.4
Kyrgyz Republic	6.4	6.7	21.6	23.3	0.6	0.6	17.0	16.7	1.2	1.1	15.1	14.7
Lao PDR	8.9	15.9	23.1	40.0	1.4	3.2	19.2	31.0	1.9	3.8	18.7	28.6
Lesotho	2.0	23.5	19.0	40.1	0.3	8.6	10.9	27.2	0.8	9.9	11.2	25.8
Liberia	1.7	3.7	6.6	12.3	0.0	0.9	3.7	10.0	0.2	1.8	3.5	10.2
Madagascar	4.3	12.8	14.0	35.8	0.2	4.8	7.7	41.7	1.6	8.0	8.4	40.2
Malawi	18.0	18.7	61.2	63.2	1.9	2.0	63.3	63.9	2.2	2.5	62.6	62.4
Mali	3.3	7.5	9.1	15.9	3.4	12.1	36.7	50.7	4.5	11.9	36.2	52.0
Mauritania	0.6	0.6	1.0	1.0	0.4	0.1	1.2	0.8	0.4	0.1	1.1	0.7
Micronesia, Fed. Sts.	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
Moldova	9.1	9.1	86.1	86.4	1.4	1.6	91.5	91.5	6.3	6.0	83.2	82.1

Source: Analysis based on WDPA, WorldPop & GHSL data (2000-2020)

Table 1: PA Coverage and Population Proximity (2000-2020)

Percentage of land and population within and near protected areas

	Percentage of land and population within and hear protected areas											
	% land within (WDPA)				%	populat	ion (GH	SL)	% population (WorldPop)			
	withi	n PAs	10km	of PAs	withi	n PAs	10km	of PAs	withi	n PAs	10km	of PAs
Country	2000	2020	2000	2020	2000	2020	2000	2020	2000	2020	2000	2020
Mongolia	14.8	19.7	23.4	32.1	1.7	2.2	36.2	54.4	5.5	6.1	33.1	55.5
Morocco	0.6	2.2	2.6	11.9	0.1	1.0	6.3	20.4	0.2	1.5	5.8	19.5
Mozambique	21.1	29.4	33.2	42.9	3.6	8.8	11.5	28.8	4.5	9.8	11.9	28.2
Myanmar	1.6	6.6	7.2	18.1	0.2	0.6	7.7	12.0	0.4	1.1	8.0	12.7
Nepal	18.9	23.5	35.2	42.8	2.6	5.5	24.4	35.8	2.9	5.6	21.2	41.6
Nicaragua	14.0	21.5	42.2	55.8	2.5	5.4	65.0	70.3	5.5	9.3	65.0	70.5
Niger	7.4	17.0	9.5	20.7	2.1	5.0	5.0	10.3	2.7	5.3	6.0	10.7
Nigeria	12.0	13.3	57.6	59.4	1.5	2.2	51.0	51.7	4.4	4.6	53.0	54.0
Pakistan	7.2	7.7	15.2	17.1	1.3	1.4	11.3	11.4	1.7	1.7	10.8	11.1
Papua New Guinea	2.9	3.7	8.6	11.5	1.2	1.2	11.1	11.8	1.5	1.7	11.6	12.7
Philippines	8.7	15.9	30.3	54.6	3.5	4.7	35.5	57.1	3.6	5.2	36.3	58.3
Rwanda	8.7	9.3	29.7	41.3	0.0	0.2	21.0	30.7	0.3	0.7	21.5	31.5
Senegal	25.5	26.5	60.1	61.4	3.1	4.1	39.5	62.1	7.6	9.4	40.8	63.4
Sierra Leone	9.4	13.0	50.3	60.1	2.5	5.6	59.2	68.5	5.8	9.1	58.9	68.7
Solomon Islands	1.0	1.5	2.1	11.6	5.2	5.7	4.9	3.6	0.1	0.2	2.4	8.1
Somalia	0.0	0.0	0.1	0.2	0.0	0.0	0.2	0.2	0.0	0.0	0.1	0.1
Sudan	1.4	1.4	1.8	2.0	0.1	0.1	0.2	0.2	0.4	0.3	0.6	0.5
Syrian Arab Re- public	0.0	0.1	0.2	0.9	0.0	0.0	0.1	0.6	0.0	0.0	0.1	0.6
São Tomé and Príncipe	0.0	32.5	0.0	93.8	0.0	0.1	0.0	49.5	0.0	9.1	0.0	58.2
Tajikistan	18.7	21.8	35.1	40.9	0.6	0.7	13.3	15.7	1.5	1.7	12.5	15.2
Tanzania	36.4	39.9	65.8	69.1	4.9	6.1	60.5	66.0	8.6	9.7	61.3	67.5
Togo	12.2	12.8	62.8	66.4	3.8	4.4	49.9	65.3	5.9	6.3	49.9	66.5
Tunisia	5.1	8.0	18.1	33.5	0.1	0.3	26.1	49.7	0.8	1.5	27.6	58.5
Uganda	13.2	15.2	72.0	74.2	2.3	2.8	85.6	86.4	6.7	7.3	86.0	87.0
Ukraine	18.1	18.1	82.5	82.5	1.2	1.2	89.5	89.9	9.6	9.2	85.6	86.4
Uzbekistan	2.9	7.9	8.1	17.0	0.1	0.3	2.4	10.0	0.9	1.1	4.9	10.6

Source: Analysis based on WDPA, WorldPop & GHSL data (2000-2020)

Table 1: PA Coverage and Population Proximity (2000-2020)

Percentage of land and population within and near protected areas

	%	land witl	hin (WDI	PA)	%	% population (GHSL)				% population (WorldPop)			
	within PAs		10km	0km of PAs wi		nin PAs 10kr		10km of PAs		within PAs		of PAs	
Country	2000	2020	2000	2020	2000	2020	2000	2020	2000	2020	2000	2020	
Vanuatu	3.9	3.9	37.0	37.0	0.0	-0.1	-5.8	-11.2	0.6	0.7	48.1	55.2	
Viet Nam	1.9	7.6	14.9	40.1	0.5	1.1	9.1	20.4	0.9	2.0	10.4	21.9	
Yemen, Rep.	1.0	1.1	1.5	2.6	1.0	2.0	3.1	10.4	1.0	2.0	3.2	10.4	
Zambia	38.1	38.9	71.3	72.1	10.5	10.6	76.4	75.6	17.4	16.2	74.2	72.7	
Zimbabwe	17.8	28.2	38.1	49.8	1.6	6.8	31.6	39.8	4.4	9.1	32.0	42.8	
West Bank and Gaza	NA	NA	NA	NA	0.5	0.4	60.9	62.5	NA	NA	NA	NA	

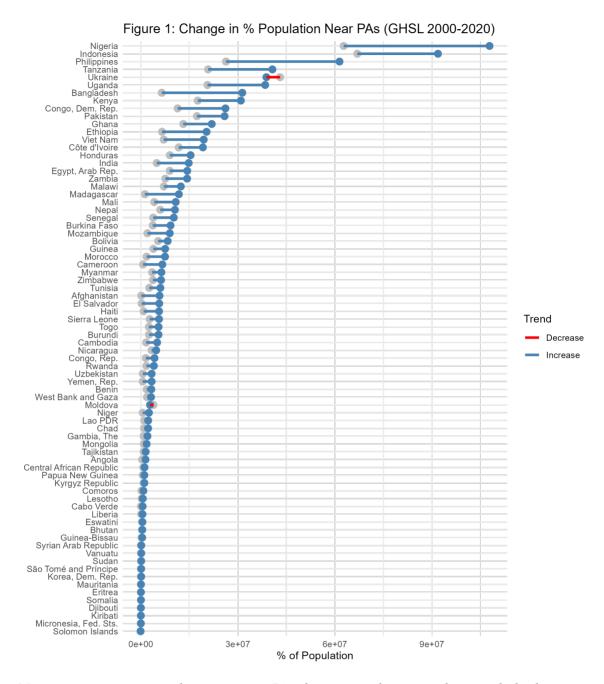
Source: Analysis based on WDPA, WorldPop & GHSL data (2000-2020)

Terrestrial PA coverage in developing countries increased from 9.3 % in 2000 (3.4 million km2) to 12.9 % in 2020 (4.7 million km2), while the share of the population living within 10 km of a PA according to GHSL rose from 16.4 % (446.5 million people) to 21.8 % (844.4 million people) over the same period. This global average is largely driven by India, due to its demographic size and almost null PA expansion. For the 75 other countries, PA coverage increased from 10.2 % in 2000 (3.4 million km2) to 14.1 % in 2020 (4.7 million km2), while the share of the population living within 10 km of a PA according to GHSL rose from 26.5 % (441.6 million people) to 33.5 % (829.6 million people).

The expansion of PA coverage was contrasted across countries. Some countries saw substantial PA expansion (e.g., Bhutan, Cambodia), while others experienced minimal growth (e.g., India, Ghana, Côte d'Ivoire). Countries with already high PA coverage in 2000 saw only marginal increases (e.g., Zambia, Tanzania).

3.2 Shifts in Shares of Population Near Protected Areas (2000-2020)

Figure 1 illustrates country-level changes in the percentage of population residing within 10 km of protected areas between 2000 and 2020. The points represent the population share near PAs in both years, while the connecting lines indicate whether the trend is increasing (blue) or decreasing (red).



Most countries experienced an increase in PA-adjacent populations, with particularly sharp rises in Comoros, Cabo Verde, El Salvador, Tunisia, São Tomé and Príncipe, Haiti, and Madagascar, where PA expansion and population growth intensified proximity. In contrast, Ukraine and Moldova saw slight declines due to urban demographic shifts despite stable PA coverage.

3.3 Growth of PA Coverage and Population near PAs between 2000 and 2020

Figure 2 presents changes in PA land coverage and the population share living within 10 km of PAs. The diagonal green line represents equal growth in both metrics. Countries above the line experienced faster growth in PA-adjacent populations than in PA coverage.

1% PA coverage in 2000 НТІ Growth in Population Share Within 10km of PAs CMR MDG BGD LBR MOZ KHM LSO TUN VNM UZB NER MMR GIN MLI EGY 2.5 5.0 10.0 Growth in Land Share Covered by PAs ETH MNG PHL Equal PA & Population Growth LAO GMB RWA ZWE GNB COD BD IDN NIC

Figure 2: Evolution of PA Land Coverage and Nearby Population Share (GHSL 2000-2020) Subsample of 58 countries classified as Low- or Low-Middle Income in 2020, with at least 1% PA coverage in 2000

Country codes: AFG: Afghanistan, AGO: Angola, BDI: Burundi, BEN: Benin, BFA: Burkina Faso, BGD: Bangladesh, BOL: Bolivia, BTN: Bhutan, CAF: Central African Republic, CIV: Côte d'Ivoire, CMR: Cameroon, COD: Congo, Dem. Rep., COG: Congo, Rep., COM: Comoros, CPV: Cabo Verde, DJI: Djibouti, EGY: Egypt, Arab Rep., ERI: Eritrea, ETH: Ethiopia, FSM: Micronesia, Fed. Sts., GHA: Ghana, GIN: Guinea, GMB: Gambia, The, GNB: Guinea-Bissau, HND: Honduras, HTI: Haiti, IDN: Indonesia, IND: India, KEN: Kenya, KGZ: Kyrgyz Republic, KHM: Cambodia, KIR: Kiribati, LAO: Lao PDR, LBR: Liberia, LSO: Lesotho, MAR: Morocco, MDA: Moldova, MDG: Madagascar, MLI: Mali, MMR: Myanmar, MNG: Mongolia, MOZ: Mozambique MRT: Mauritania, MWI: Malawi, NER: Niger, NGA: Nigeria, NIC: Nicaragua, NPL: Nepal, PAK: Pakistan, PHL: Philippines, PNG: Papua New Guinea, PRK: Korea, Dem. Rep., PSE: West Bank and Gaza, RWA: Rwanda, SDN: Sudan, SEN: Senegal, SLB: Solomon Islands, SLE: Sierra Leone, SLV: El Salvador, SOM: Somalia, STP: São Tomé and Príncipe, SWZ: Eswatini, SYR: Syrian Arab Republic, TCD: Chad, TGO: Togo, TJK: Tajikistan, TUN: Tunisia, TZA: Tanzania, UGA: Uganda, UKR: Ukraine, UZB: Uzbekistan, VNM: Viet Nam, VUT: Vanuatu, YEM: Yemen, Rep., ZMB: Zambia, ZWE: Zimbahwa

Most countries fall below the diagonal, where PA spatial expansion outpaced the growth of demographic share of adjacent populations. Lesotho in particular, exhibit large PA expansions with

relatively small shifts in population share, indicating that new PAs were primarily established in low-density areas. This reflect the classical trend of PAs to be typically established "high and far", where pressure for land conversion is lower (Joppa and Pfaff 2009).

In contrast, countries in the top left quarter of Figure 2 experienced a much larger increase in the percentage of their population near PAs than in PA coverage, indicating that new PAs or expansions occurred in relatively high-density areas. Such pattern is expected in countries with predominantly rural populations, such as Madagascar, Bangladesh, and Mozambique, where PA establishment is more likely to affect human settlements In other cases, such as Haiti, Cameroon, and Liberia, the trend appears to result from deliberate conservation choices to designate PAs in more densely inhabited regions, rather than in remote areas. The zoomed-in portion highlights countries with smaller relative changes, where PA coverage and nearby population less than doubled between 2000 and 2020.

3.4 Share of PA Coverage and Population Near PAs in 2020

Figures 3 displays the relationship between the percentage of national land designated as protected areas and the percentage of the population residing within 10 km of these areas in 2020. Each point represents a country, with country codes labeled on the graph and a corresponding key provided below. The size of the points reflects the total national population.

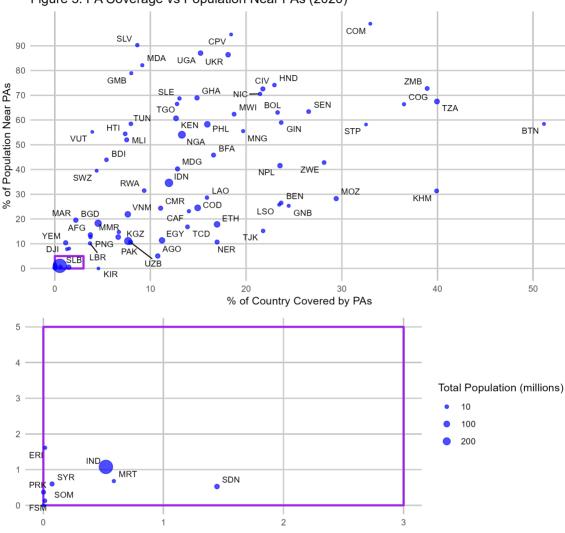


Figure 3: PA Coverage vs Population Near PAs (2020)

Country codes: AFG: Afghanistan, AGO: Angola, BDI: Burundi, BEN: Benin, BFA: Burkina Faso, BGD: Bangladesh, BOL: Bolivia, BTN: Bhutan, CAF: Central African Republic, CIV: Côte d'Ivoire, CMR: Cameroon, COD: Congo, Dem. Rep., COG: Congo, Rep., COM: Comoros, CPV: Cabo Verde, DJI: Djibouti, EGY: Egypt, Arab Rep., ERI: Eritrea, ETH: Ethiopia, FSM: Micronesia, Fed. Sts., GHA: Ghana, GIN: Guinea, GMB: Gambia, The, GNB: Guinea-Bissau, HND: Honduras, HTI: Haiti, IDN: Indonesia, IND: India, KEN: Kenya, KGZ: Kyrgyz Republic, KHM: Cambodia, KIR: Kiribati, LAO: Lao PDR, LBR: Liberia, LSO: Lesotho, MAR: Morocco, MDA: Moldova, MDG: Madagascar, MLI: Mali, MMR: Myanmar, MNG: Mongolia, MOZ: Mozambique MRT: Mauritania, MWI: Malawi, NER: Niger, NGA: Nigeria, NIC: Nicaragua, NPL: Nepal, PAK: Pakistan, PHL: Philippines, PNG: Papua New Guinea, PRK: Korea, Dem. Rep., PSE: West Bank and Gaza, RWA: Rwanda, SDN: Sudan, SEN: Senegal, SLB: Solomon Islands, SLE: Sierra Leone, SLV: El Salvador, SOM: Somalia, STP: São Tomé and Príncipe, SWZ: Eswatini, SYR: Syrian Arab Republic, TCD: Chad, TGO: Togo, TJK: Tajikistan, TUN: Tunisia, TZA: Tanzania, UGA: Uganda, UKR: Ukraine, UZB: Uzbekistan, VNM: Viet Nam, VUT: Vanuatu, YEM: Yemen, Rep., ZMB: Zambia, ZWE: Zimbabwa

fig-alt="A scatter plot with, for 2020, the percent of spatial PA coverage as X and the percent of population within 10 km of PAs."}

Figure 3 highlights considerable variation across countries in both PA coverage and the proportion of people living near these areas. Some countries, such as Bhutan and Tanzania, have extensive PA networks. A cluster of countries with low PA coverage and limited nearby population presence can also be observed. Notably India falls in this section, and it has such a large population that it weighs heavily on the global averages drawn from these national statistics.

3.5 Consistency between sources

Table 2 presents the SMDs between GHSL and WorldPop estimates for populations residing within and near protected areas in 2000 and 2020, with higher values indicating greater discrepancies.

Table 2: Standard Mean Differences Between GHSL and Worldpop Estimates

	2000	2020
Population within PAs	0.29	0.20
Population within 10km of PAs	0.09	0.08

There is no universal interpretation of SMD values for comparing different population datasets. Cohen's rule of thumb suggests an SMD below 0.2 represents a negligible difference, while Austin proposes a more conservative threshold of 0.1 for comparability (Cohen 1988; Austin 2009). Applying this principle, GHSL and WorldPop differ significantly for PA-internal population estimates but align well for populations within 10 km of PAs. GHSL is considered more reliable for high-and low-density settings (Chen et al. 2020).

Figure 4 provides a disaggregated comparison of GHSL and WorldPop estimates. Each segment represents the difference between the two sources for a given country, with black and blue points denoting GHSL and WorldPop, respectively.

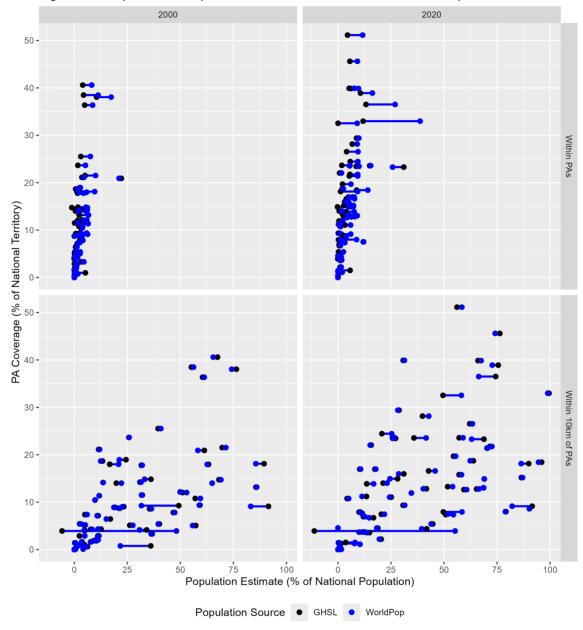


Figure 4: Comparison of Population Estimates from GHSL and WorldPop

For most countries, GHSL and WorldPop estimates align closely, with overlapping black and blue dots on Figure 4. However, some countries exhibit large discrepancies. Table S1 in the appendix details the 15 countries where GHSL and WorldPop estimates differ by more than 5 percentage points and by more than 10%. Notably, WorldPop estimates significantly exceed GHSL in countries such as Comoros, the Republic of Congo, and São Tomé and Príncipe, while GHSL reports higher population shares near PAs than WorldPop in countries such as Ukraine and Zambia. These variations may stem from differences in population allocation methodologies or underlying census data.

4 Discussion

4.1 A Relevant Parameter to Frame the Reflection on Conservation

This study has relatively modest theoretical claims and does not aim to establish a causal analysis for protected area expansion. Instead, we emphasize the value of a "mere description" (Gerring 2012) quantifying the magnitude of the population directly affected by protected areas in the developing world. To our knowledge, no other study has systematically estimated the number of people impacted by area-based conservation policies. Given the scale of the phenomenon described in this paper, addressing conservation challenges increasingly requires integrating socioeconomic considerations.

The demographic dynamics around protected areas have been a subject of debate in the literature. Wittemyer et al. (2008) reported accelerated population growth near existing PAs, suggesting an attraction effect, while Joppa, Loarie, and Pimm (2009) argued that such cases were isolated and that population growth near PAs was not disproportionately high overall. Our analysis aligns with the latter, as we find no evidence of increased neighboring population growth in areas where PA coverage has remained stable. Instead, our results indicate that the observed rise in populations near PAs in some countries is primarily driven by the expansion of PA coverage itself. In the cases where population growth outpaces PA expansion, this can be attributed to the selection of denser areas for new PAs. Additionally, a geometric effect may also play a role: as smaller and more irregularly shaped PAs are added to enhance connectivity (Saura et al. 2019) or accommodate political and ecological constraints, the relative extent of surrounding buffer zones increases compared to the total protected area (Schauman et al. 2024).

4.2 The Least Inaccurate Method Available

Our estimation of the share of people living near protected areas is derived from GHSL and World-Pop, and the reliability of our analysis depends on their accuracy. Several studies have attempted to assess Worldpop's accuracy, but they all face a key challenge: census data are only available at the administrative unit level, and no alternative dataset provides fine-grained "ground truth" population estimates at low resolution. Consequently, most assessments compare gridded datasets aggregated to administrative units with official census counts or evaluate the spatial consistency across gridded population datasets. The most comprehensive study is Chen et al. (2020), which evaluated four global gridded datasets – WorldPop, HYDE (Goldewijk et al. 2017), GPWv4, and GHSL – across four countries with diverse population distribution patterns. GPWv4 and GHSL were closer to census totals, while WorldPop had larger deviations but the highest spatial consistency with other datasets, suggesting its population allocation model aligns well with relative distribution patterns.

In principle, household surveys such as the Demographic and Health Surveys (DHS) could complement or validate these gridded estimates. naidoo2019 used DHS data from 34 developing countries (2001–2011) to compare households near and far from PAs in assessing conservation impacts. While their objective was not to estimate the proportion of the population near PAs, their supplementary materials report the number of surveyed households located in proximity to PAs. Direct comparisons between our estimates and household survey data are not straightforward, but a

rough approximation using data extracted from Naidoo et al. (2019)'s appendix provides a useful reference point. Summing the total number of surveyed households and the number located within 10 km of a PA, we estimate that approximately 22% of households in the DHS sample reside within 10 km buffers, which is broadly in line with our estimates. However, several factors hinder direct comparison. 1. DHS estimates require inverse probability weighting to be populationrepresentative, which cannot be applied using Naidoo et al. (2019)'s appendix data. 2. DHS GPS coordinates represent the centroid of the enumeration area, not individual household locations. 3. Additionally, DHS relies on a two-stage cluster sampling approach, meaning that the survey covers only a limited number of sampled locations (typically a few hundred per country), reducing spatial representativity at a fine scale. 4. DHS also reports household counts, not individual population estimates, meaning that differences in household size could affect comparability. 5. DHS geolocations are intentionally displaced, with rural clusters randomly shifted by up to 10 km, introducing uncertainty in proximity assessments (Skiles et al. 2013). Processing to the raw DHS data would allow to tackle the issues 1., 4., and update this analysis with more surveys. However, obtaining DHS data also involves a vetting procedure with a USAID contractor and, at the time of writing, the foreign assistance programs have been indefinitely suspended and it is unknown when or if it will be resumed.

5 Conclusion

Our analysis provides an assessment of the population living within or in the immediate surroundings of protected areas following their expansion in low- and lower-middle-income countries. Between 2000 and 2020, PA coverage in these countries grew from 9.3% to 12.9%, while the share of people living within 10 km of a PA increased from 16.4% to 21.8%. This global trend is strongly influenced by India, where limited PA expansion and a large population shape the overall figures. Excluding India, PA coverage rose from 10.2% to 14.1%, and the proportion of the population near PAs increased from 26.5% to 33.5%. These figures highlight the extent to which conservation policies affect human populations and the need to address their socio-economic implications. Balancing ecological goals with the well-being and rights of nearby communities will become ever more critical as governments pursue accelerated PA expansion to meet global 30% targets.

Although our data and method have inherent limitations, it relies on the most consistent and scalable approach currently available for estimating demographic exposure to conservation policies. Future improvements in population mapping and survey integration could refine these estimates, but we believe that our findings offer a useful first step in quantifying the human dimensions of PA expansion at a global level.

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Appendix

Table S1: Largest Differences Between GHSL and WorldPop Estimates

Absolute difference of more then 5 percentage points (pp) and relative difference of more than 10%

Country	Year	Area	WorldPop Estimate	GHSL Estimate	Absolute Difference	Relative Difference
Comoros	2020	Within PAs	38.7%	11.9%	26.9 pp	226.4%
Congo, Rep.	2000	Within 10km of PAs	31.7%	49.2%	17.6 pp	35.7%
Comoros	2000	Within 10km of PAs	21.7%	36.1%	14.4 pp	39.8%
Congo, Rep.	2020	Within PAs	26.9%	13.2%	13.7 pp	103.6%
Moldova	2020	Within 10km of PAs	82.1%	91.5%	9.4 pp	10.3%
São Tomé and Príncipe	2020	Within PAs	9.1%	0.1%	9.0 pp	17,093.6%
Tunisia	2020	Within 10km of PAs	58.5%	49.7%	8.8 pp	17.7%
São Tomé and Príncipe	2020	Within 10km of PAs	58.2%	49.5%	8.7 pp	17.5%
Ukraine	2000	Within PAs	9.6%	1.2%	8.4 pp	670.2%
Ukraine	2020	Within PAs	9.2%	1.2%	8.0 pp	660.6%
Congo, Rep.	2020	Within 10km of PAs	66.4%	74.3%	8.0 pp	10.7%
Gambia, The	2020	Within PAs	8.8%	1.0%	7.8 pp	765.0%
Bhutan	2020	Within PAs	11.6%	4.5%	7.1 pp	156.9%
Zambia	2000	Within PAs	17.4%	10.5%	6.9 pp	65.3%
Bhutan	2000	Within PAs	11.2%	4.4%	6.9 pp	156.5%
Djibouti	2020	Within 10km of PAs	7.8%	1.4%	6.4 pp	446.9%
Nepal	2020	Within 10km of PAs	41.6%	35.8%	5.7 pp	16.0%
Zambia	2020	Within PAs	16.2%	10.6%	5.6 pp	53.0%
Solomon Is- lands	2020	Within PAs	0.2%	5.7%	5.5 pp	95.8%
Senegal	2020	Within PAs	9.4%	4.1%	5.4 pp	131.3%
Cabo Verde	2020	Within PAs	14.1%	8.8%	5.3 pp	60.3%
Bolivia	2020	Within PAs	25.8%	31.0%	5.3 pp	16.9%

Table S1: Largest Differences Between GHSL and WorldPop Estimates

Absolute difference of more than 5 percentage points (pp) and relative difference of more than 10%

Country	Year	Area	WorldPop Estimate	GHSL Estimate	Absolute Difference	Relative Difference
Solomon Is- lands	2000	Within PAs	0.1%	5.2%	5.1 pp	97.7%