The spatial distribution of the resident population and the "Degree of Urbanisation" in the GHSL R2023

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Structure & Outline

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 The porting of the Degree of Urbanisation into the GHSL framework: GHS-SMOD and GHS-DUC R2023A



The population grid

What is

Why

How

The population grid: What is

- A population grid is a representation of the population distribution on a regular grid usually in raster format or vector format
 - Resident population
 - Temporary population (e.g. commuters)

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The population grid: Why

- Does not depend on the shape and size of census units
- Does not depend on the boundary of census units
- Grid cells have the same size and are stable over time
- Grids integrate easily with other data
- Grid cells can be assembled into custom zones



The population grid: How



• Census aggregation (bottom-up)







The GHSL population grid: GHS-POP R2023A

Input data

Workflow

Multi-temporal population estimates

Covariate selection

GHS-POP R2023A: Input data

- Resident population data (census) collected by the Center for Integrated Earth System Information (CIESIN, Columbia University, NY) for the production of the Gridded Population of the World (<u>GPW v4.11</u>):
 - 251 Countries
 - 13,415,003 polygons
 - Harmonised geometries
 - 2 Census epochs per country (2000-2010 on average, ranging between 1983 and 2015)
 - Adjusted growth rates computed between the two epochs
- UN World Population Prospects 2022 (WPP2022)
 - Countries population time series (to adjust country total annual population)
- UN World Urbanisation Prospects 2018 (WUP2018)
 - Urban Agglomerations population time series (increase the accuracy of estimates)
- GHSL built-up layers as covariates (Surface/Volume; TOT/NRES) [disaggregation top-down approach]



GHS-POP R2023A: Workflow

- Data Statistics
- Layer Pre-processing
- Multi-temporal population estimates (1975-2030 5yrs interval)
- Coastline extension
- Falsely unpopulated polygons & Population data anomalies
- Identification of the best covariate
- Population downscaling
- Quality check and assessment



GHS-POP R2023A: Multi-temporal population estimates (1)





GHS-POP R2023A: Multi-temporal population estimates (2)



Blue: estimated boundaries (SmartDissolve, *Schiavina et al. 2023, <u>http://dx.doi.org/10.1038/s41598-023-31253-z</u>) Green: 'city' boundaries in US, from UN data*



GHS-POP R2023A: Multi-temporal population estimates (3)

CALIBRATION SET						
Belgium	-2.5%	Greece	6.0%	Slovenia	<mark>-4</mark> 1.3%	
Bulgaria	-3 3.1%	Croatia	-18.2%	Slovakia	-1.1%	
Switzerland	3 <mark>0</mark> .4%	Ireland	0.7%	USA	<mark>-3</mark> 8.3%	
Czechia	-61.1%	Iceland	-17.0%	Argentina	- 1 4.5%	
Germany	-58.9%	Italy	8.5%	Nepal	-6.3%	
Denmark	0.8%	Liechtenstein	-2 9.5%	Georgia_C	-3 5.0%	
Estonia	-5 5.9%	Lithuania	6.3%	New Zealand	7.3%	
Spain	-3.5%	Luxembourg	-5 9.3%	Afghanistan	- 1 1.6%	
Finland	8.7%	Martinique	-13.5%	Azerbaijan	-1.5%	
France	-12.5%	Norway	<mark>-4</mark> 2.7%	Benin	0.0%	
United Kingdom	-2.0%	Poland	-6.4%	Botswana	0.0%	
French Guiana	- 6 0.6%	Réunion	-19.9%	Guatemala	-4.6%	
Guadeloupe	0.9%	Sweden	<mark>-4</mark> 9.9%	Myanmar	-16.6%	
VALILDATION SET						
Austria	-7 0.7%	Latvia	- <mark>1</mark> 8.3%	Portugal	-0.7%	
Cyprus	9.7%	Malta	2 <mark>1</mark> .5%	Romania	- 1 5.7%	
Hungary	<mark>-3</mark> 6.2%	Netherlands	<mark>-3</mark> 8.0%	Kenya	-4.9%	





GHS-POP R2023A: Covariate selection

- Built-up covariate tested:
 - Surface (GHS-BUILT-S), Volume (GHS-BUILT-V)
 - TOT built-up layer as target (noNRES)
 - TOT built-up full NRES layer (fullRES)
 - mixed TOT and NRES using weight (mixNRES)
 - mixed TOT and binarised NRES weight (binNRES)

	SOURCE Z	ONES	REF DATA			
Country	Aggregation	Units	Units	Pop 2020		
JPN	10	5,016	501,600	126,146,099		
BEL	7	25	32,140	11,485,814		
FIN	32	385	352,885	5,424,562		
KEN	admin4	2,713	7,125	53,771,300		
AUS	SA1	61,819	368,286	25,418,422		
COL	Municipio	1,122	551,028	46,925,773		
TOTAL			1,813,064	269,171,970		

	RMSE						%TAA									
	Surface				Volume			Surface				Volume				
Country	noNRES	fullRES	mixNRES	binNRES	noNRES	fullRES	mixNRES	binNRES	noNRES	fullRES	mixNRES	binNRES	noNRES	fullRES	mixNRES	binNRES
JPN	477.2	433.6	434.3	435.3	456.0	381.3	381.9	382.7	0.813	0.825	0.825	0.824	0.838	0.859	0.858	0.858
BEL	531.5	405.2	407.7	411.3	511.3	311.9	314.1	316.7	0.765	0.814	0.813	0.811	0.772	0.852	0.851	0.850
FIN	109.4	87.3	87.6	88.1	126.1	79.9	80.4	81.0	0.661	0.716	0.715	0.714	0.622	0.723	0.722	0.720
KEN	3847.8	3811.1	3811.8	3812.8	3780.1	3725.0	3725.5	3726.1	0.9095	0.9103	0.9102	0.9102	0.9087	0.9102	0.9102	0.9102
AUS	59.6	57.975	58.006	58.050	60.7	58.9	58.9	58.9	0.800	0.8080	0.8078	0.808	0.793	0.801	0.801	0.801
COL	317.9	270.4	271.3	272.6	316.0	248.8	249.4	250.2	0.468	0.473	0.473	0.473	0.453	0.460	0.460	0.539
TOTAL	400.0	369.3	369.8	370.6	390.5	342.4	342.8	343.3	0.766	0.776	0.776	0.776	0.774	0.7909	0.7907	0.8847



GHS-BUILT-V TOT and NRES 100m



GHS-BUILT-V NRES 2020 100m, in m3

20 km

100000

0

GHS-BUILT-V 2020 100m, in m3

100000

Census data

GHS-LAND Water

Thailand admin2 census 2010, in p/km2

0 - 300
300 - 600
600 - 1,250
1,250 - 2,500
2,500 - 5,000
5,000 - 10,000
10,000 - 15,000
> 15,000

5

0

1()

20 km

Phra Nakhon Si Ayutthaya

Census data and WUP2018 data

Samut Sakhon

Suphan

Rup

Pathum Thani

GHS-LAND

Water

GHS-SDATA WUP2018

 $\times \times$

Thailand admin2 census 2010, in p/km2

1() 20 km

0 - 300 300 - 600 600 - 1,250 1,250 - 2,500 2,500 - 5,000 5,000 - 10,000 10,000 - 15,000 > 15,000

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Nonthaburi

Krung Thep (Bangkok)

Samut Prakan

Chachoengsao

Lingin Bunnission

Estimated population data

GHS-LAND Water

Thailand population estimate 2020, in p/km2

20 km

0 - 300 300 - 600 600 - 1,250 1,250 - 2,500 2,500 - 5,000 5,000 - 10,000 10,000 - 15,000 > 15,000

10

5

Population grid 100m

GHS-LAND

Water

GHS-POP 2020 100m

0

0 - 300

300 - 600

600 - 1,250

1,250 - 2,500

2,500 - 5,000

5,000 - 10,000

10,000 - 15,000

> 15,000

Thailand admin2 census 2010

0 5 10 20 km

Thailand admin2 census 2010

Population grid 1km

GHS-LAND

Water

GHS-POP 2020 1km

0

0 - 300

300 - 600

600 - 1,250

1,250 - 2,500

2,500 - 5,000

5,000 - 10,000

10,000 - 15,000

0 > 5 1 (20 km)

> 15,000

Thailand admin2 census 2010

The Degree of Urbanisation

What is Why How

The Degree of Urbanisation: What is

- The Degree of Urbanisation (DEGURBA) is a population based methodology recommended by UN to classify the urban/rural continuum
 - Harmonised people-centric definition for international statistical comparison (SDG)

European

• Based on 1km population grid in equal area projection

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Classification grid (Stage I) and classification of territorial units (Stage II)

The Degree of Urbanisation: Why

- Half the countries designate urban and rural areas, which cannot be replicated
- Different definitions hamper the possibility of statistics comparisons (SDG)

Definition relies on: Population size and/or density indicators A combination of indicators including population size or density Other indicators than population size or density No statistical definition reported

3,000 Km

The Degree of Urbanisation: How (Stage I, 1)



- Each grid cell is classified according to a set of rules applied on a 1km equal era population grid:
- Population density
 of each cell
- Contiguity & distance between cells to form clusters of cells
- **Population size** of the clusters



The Degree of Urbanisation: How (Stage I, 2)

1st Level 2nd Level





 See the <u>online</u> <u>manual</u> for details on the classes



Icons credits: City by Muhajir ila Robbi from the Noun Project, Town by Alice Design from the Noun Project, Farm by Andi Nur Abdillah from the Noun Project

The Degree of Urbanisation: How (Stage II, 1)

- Territorial units are classified based on
 - Settlement grid classification
 - Population grid
- Makes possible to disaggregate existing local statistics in the 3 or 7 classes

Cities	≥ 50% pop. in urban centres
Towns and suburbs	≥ 50% pop. in urban clusters and not classified as city
Rural area	> 50% pop. in rural grid cells



The Degree of Urbanisation: How (Stage II, 2)

1st Level 2nd Level





 See the <u>online</u> <u>manual</u> for details on the classes



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The porting of the Degree of Urbanisation into the GHSL framework: GHS-SMOD and GHS-DUC R2023A

GHS-SMOD and GHS-DUC R2023A

- The GHS-SMOD is the settlement classification grid produced by applying the Degree of Urbanisation Stage I to each epoch (1975-2030, 5yrs interval) of the GHS-POP 1-km
- The GHS-DUC is the Global Administrative boundaries layer (GADM, <u>www.gadm.org</u>) classified according to the Degree of Urbanisation Stage II:
 - GHS-SMOD as settlement classification grid
 - GHS-POP 100m as population counts
 - This is only a exemplificative exercise to show how the Degree of Urbanisation Stage II can be applied; there are no statistics associated to the GADM geometry to be disaggregated and produce indicators by Degree of Urbanisation and GADM geometry is quite coarse for some countries



GHS-SMOD 2020 1km

Water

Very low density rural

Low density rural

Rural cluster

Suburban or peri-urban

Semi-dense urban cluster

20 km

Dense urban cluster

Urban centre

GHS-SMOD 2020 1km

Water

Very low density rural Low density rural

Rural cluster

Suburban or peri-urban

Semi-dense urban cluster

20 km 🐂

Dense urban cluster

Urban centre

10



GHS-SMOD 2020 1km

Water

Very low density rural

Low density rural

Rural cluster

Suburban or peri-urban

Semi-dense urban cluster

20 km

Dense urban cluster

Urban centre

1()



GHS-SMOD 2020 1km

Water

Very low density rural

Low density rural

Rural cluster

Suburban or peri-urban

Semi-dense urban cluster

20 km

Dense urban cluster

Urban centre

1()

GHS-SMOD 2020 1km

Water

Very low density rural

Low density rural

Rural cluster

Suburban or peri-urban

Semi-dense urban cluster

20 km

Dense urban cluster

Urban centre

Commission

GHS-SMOD 2020 1km

Water

Very low density rural

Low density rural

Rural cluster

Suburban or peri-urban

Semi-dense urban cluster

20 km

Dense urban cluster

Urban centre

Classification of territorial units level 2

GHS-LAND Water

GHS-DUC

Water

Mostly Uninhabited Area

Dispersed Rural Area

Village

Citv

Suburban or Peri-urban Area

20 km

Semi-dense Town

Dense Town

1()

European

GHS-SMOD R2023A: Global Results

Between 1975 and 2020: urban centre pop almost x3 urban cluster pop almost x2



The global urban share: 69% (32%+37%) in 1975 80% (45%+35%) in 2020



Here more charts derived from these data, chart source OurWorldInData.org



GHS-SMOD R2023A: Regional Results

Africa has the largest increment in urban centre: from 96M to 550M (x5.7)



South America has the largest share of urban centre population: 57%



Here more charts derived from these data, chart source OurWorldInData.org



Thank you for your attention



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