

Technical workshop on geospatial population estimation for selected countries in the Arab Region

Case Study: Filling census gaps in Afghanistan

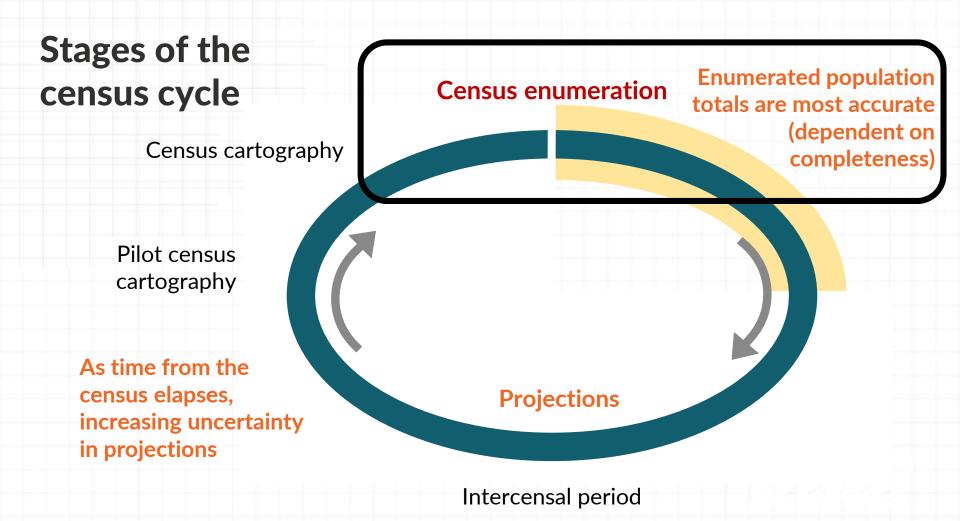
### WorldFop

High-resolution population mapping in Afghanistan 2017/18

#### Background

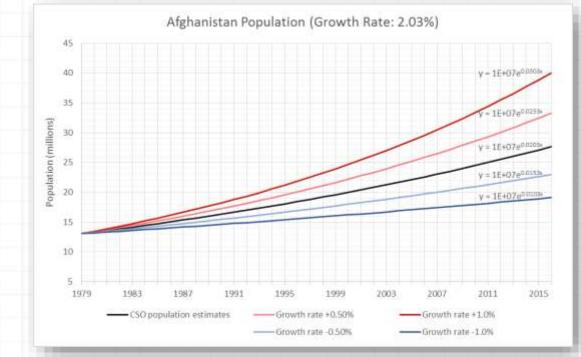
- Last census conducted in 1979; insecurity prevented a recent one; existing estimates based on 1979 baseline population;
- From 2011-2017 the CSO conducted a form of rolling census, the Socio-Demographic and Economic Survey (SDES) which included enumeration for 50% of households (the survey covered 12 of 34 provinces);
- □ Challenge: Lack of reliable current disaggregated population data at provincial, district level;
- President Ghani requested UN to assist CSO in estimating spatially disaggregated population data - Collaborative partnership including Government, UNFPA, WorldPop





### Existing CSO estimates

- Existing estimates based on 2.03% growth rate from 1979 baseline
- Significant differences with recently gathered SDES numbers (typically underestimates)
- Analyses of newer data suggest that recent growth rates are more likely to be 3% or higher



### Afghanistan: Project aims

- Estimate population counts at enumeration area (EA), district, province and national levels and associated measures of uncertainty;
- 100m x 100m gridded population estimates;
- Estimates of the age/sex structure of the population and associated measures of uncertainty
- Co-development with CSO and capacity strengthening







# How does the population estimation process work?

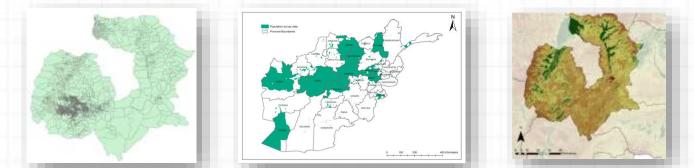
### Population estimation & mapping process

- 1. Enumerate population with ground-based surveys in defined areas
- 2. Predict population in un-surveyed areas
- 3. Spatial disaggregation of population estimates



### We want to....

- Estimate population density and total in each EA (with confidence intervals)
- We have population density and total data for some EAs, but not all
- We do have data on factors related to populations in all EAs from satellites and other geospatial data



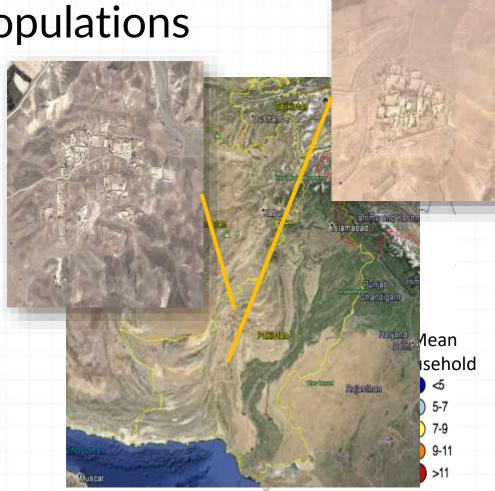
### Mapping populations

Mapping buildings/settlements is a valuable initial step

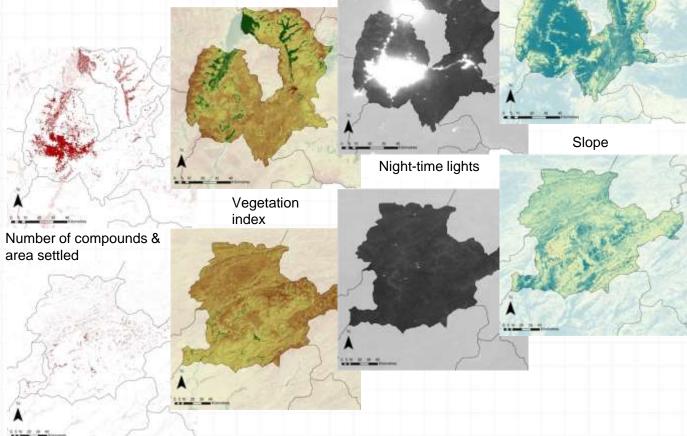
But, buildings do not directly translate to people

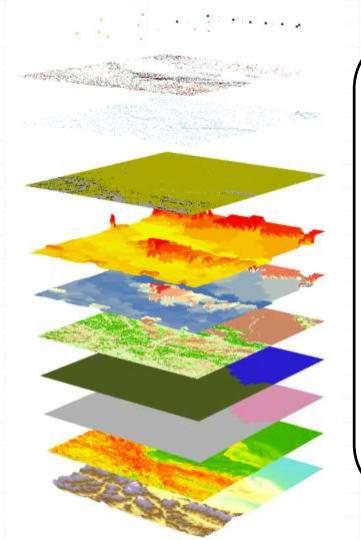
Multiple factors drive differences in distributions and demographics

Importance of *integrating multiple data sources* 

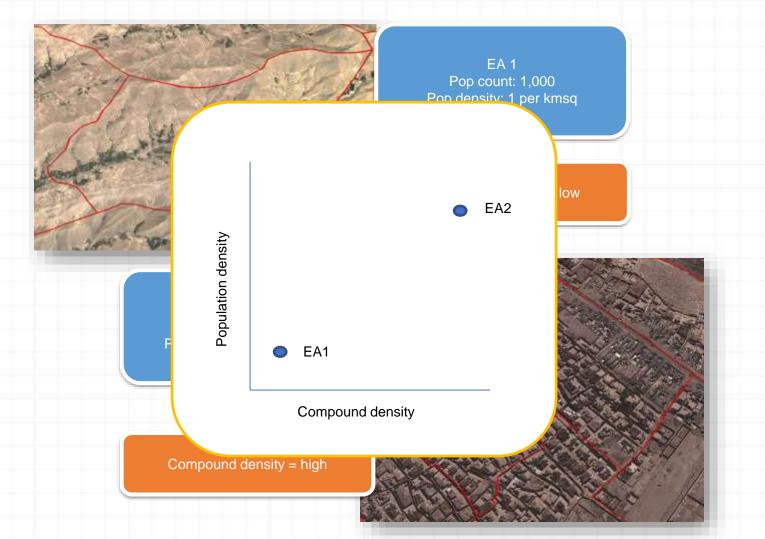


### Example geospatial datasets

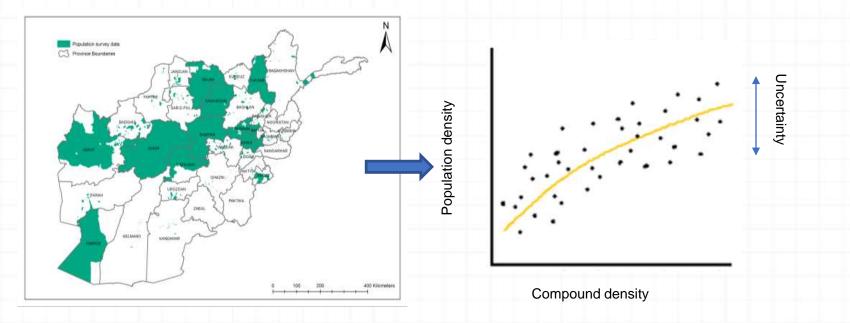


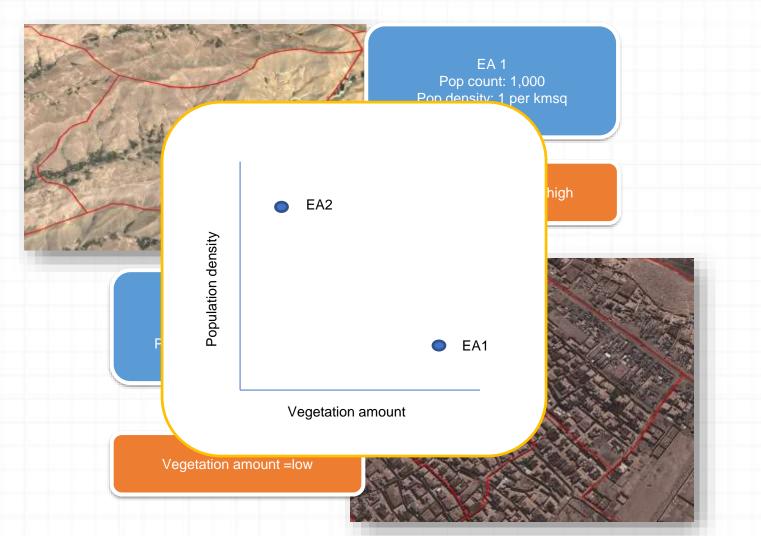


- >100 geospatial layers
- Sourced from across ministries, agencies and companies
  - Complete national coverage
  - Recent acquisition
  - Significant effort in checking, cleaning, harmonizing
- Processed to extract indicators likely to be predictive of population distributions and densities

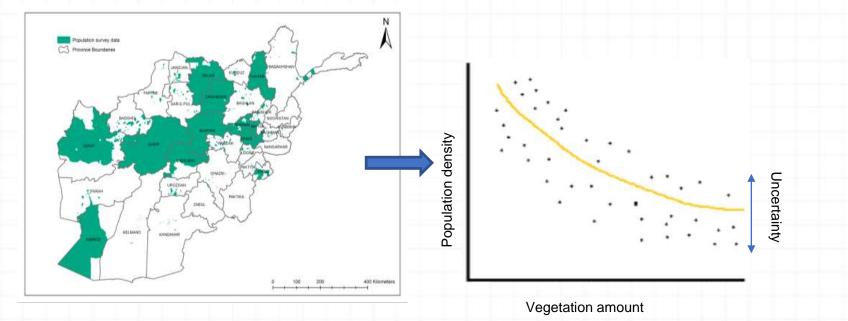


# What about other EAs where we have population data?





## What about other EAs where we have population data?





Distance from urban area = high

Compound density = low

Mean slope gradient = high

Nightlights brightness = low

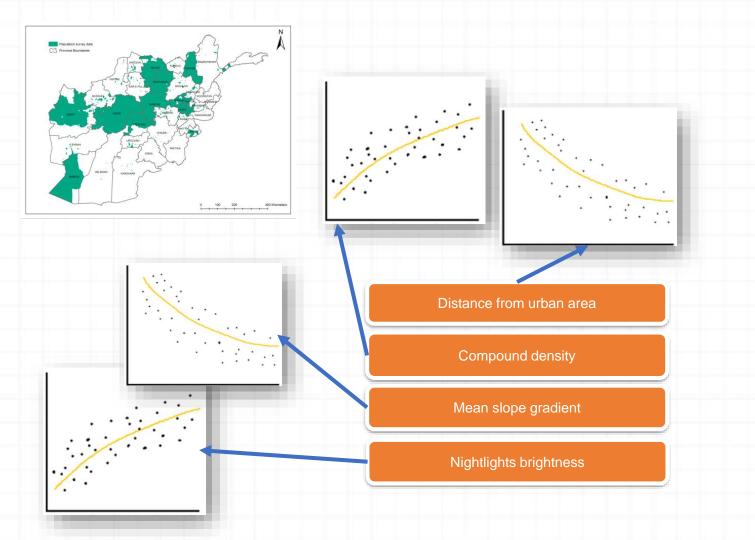
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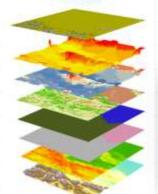


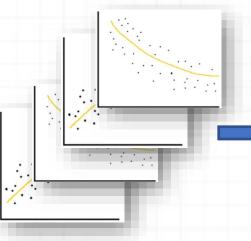




#### EA 3 Pop count: ? Pop density: ? per kmsq







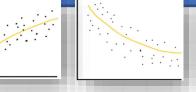
EA 3 Pop count: 350 (95% range = 300-400) Pop density: 10 per kmsq (95% range = 7-13)

### **Population estimation steps**

Assemble existing recent population data

Explore and test relationships between population data and geospatial datasets

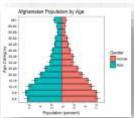
Select most predictive subset of geospatial datasets



Build model, make predictions of populations per EA in unsampled locations, cross-validation

Disaggregate to 100x100m grid cells

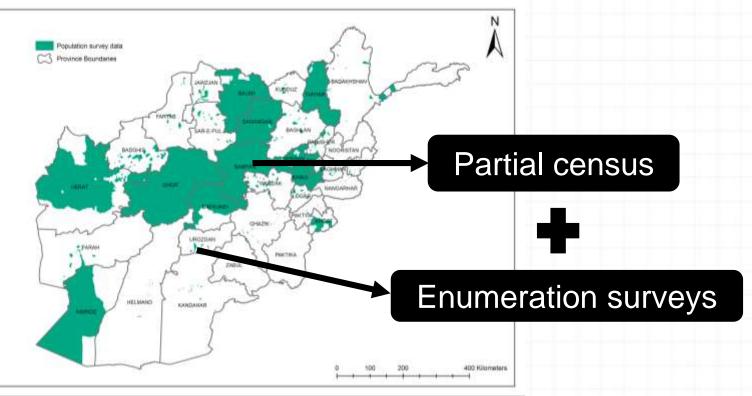
#### Repeat for age/sex group proportions



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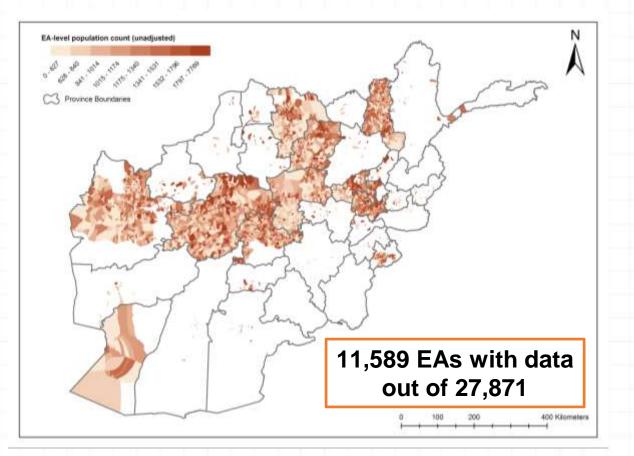
## Implementation for Afghanistan

### Existing recent data

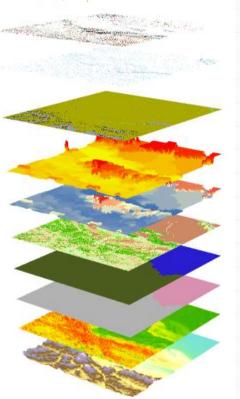


#### 11,589 EAs with data out of 27,871

### Existing recent data



### Finding the strongest predictors



Selected geospatial variables
Number of residential compounds
Compound density
Distance from edge of urban areas
Mean slope
Mean average distance between compounds
Minimum distance from a village
Maximum nightlight brightness
Fragmentation of settled area
Area of the EA
Vegetation amount

### Training and testing

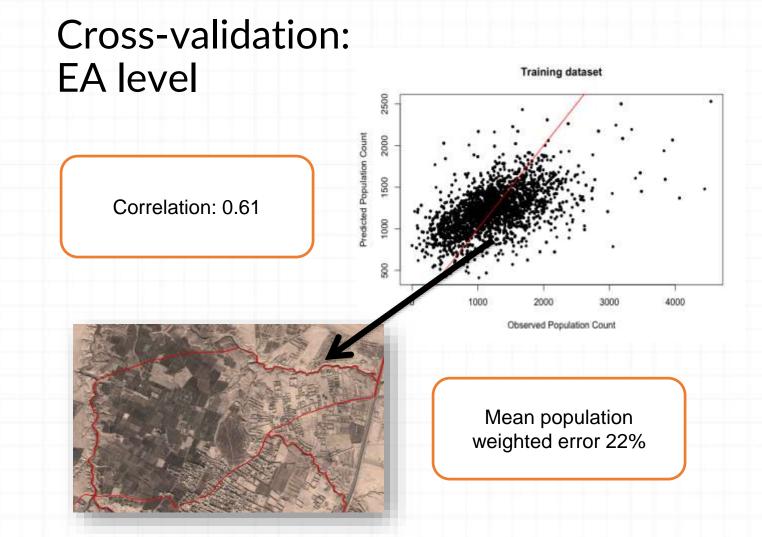
- Multiple model types and forms tested
- Selected the type and set of covariates that optimised predictive ability

Bayesian hierarchical spatial model:

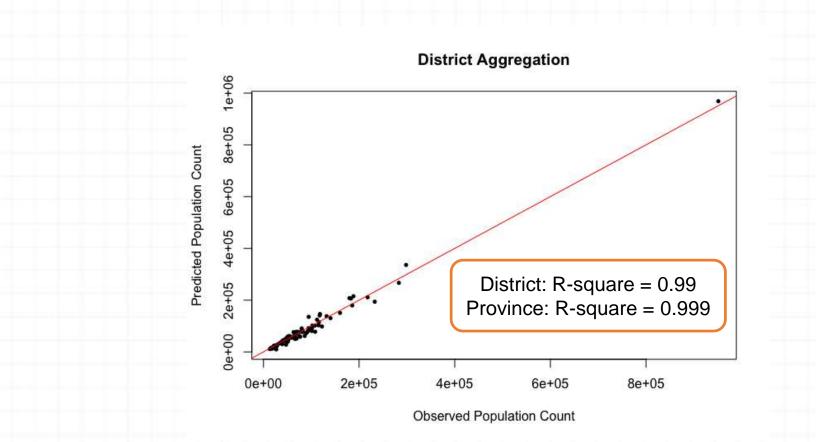
-multiple model forms -accounts for spatial relationships -measures of uncertainty

- 10-fold cross-validation
- 11,589 EAs with data
- Split randomly into 10 groups
- Further subdivided into training and testing sets
  - Mean observed vs prediction stats

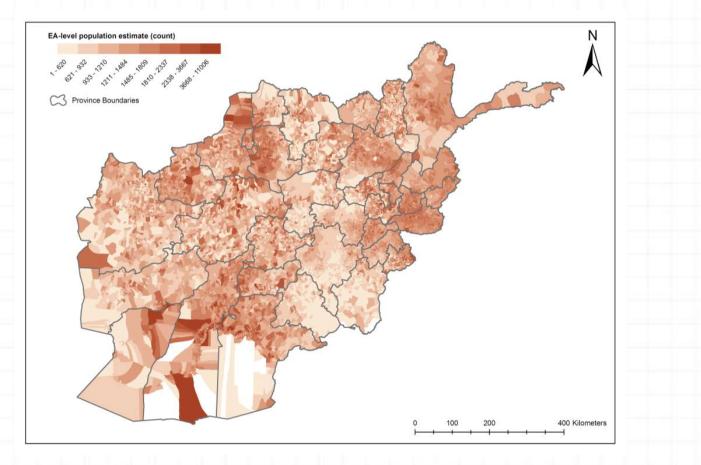




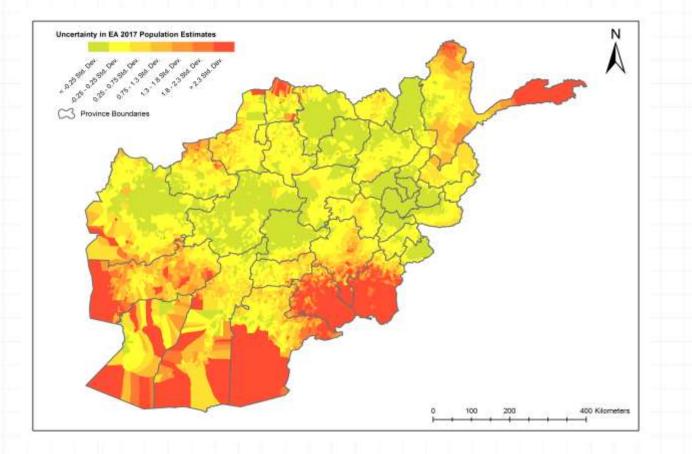
### Accuracy assessment: District and Province level

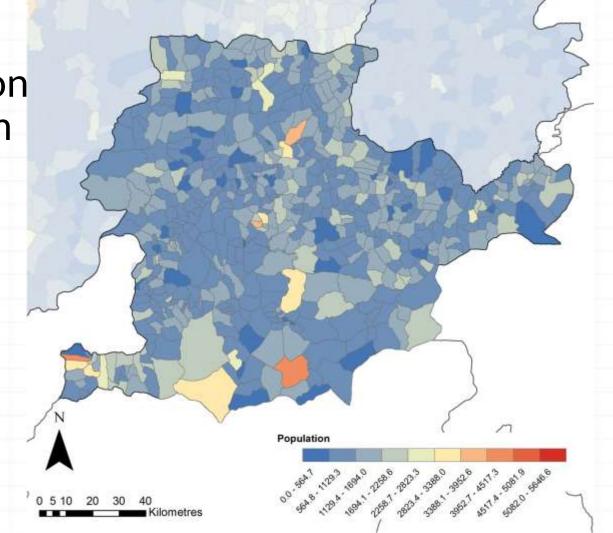


### **Outputs: EA counts and estimates**

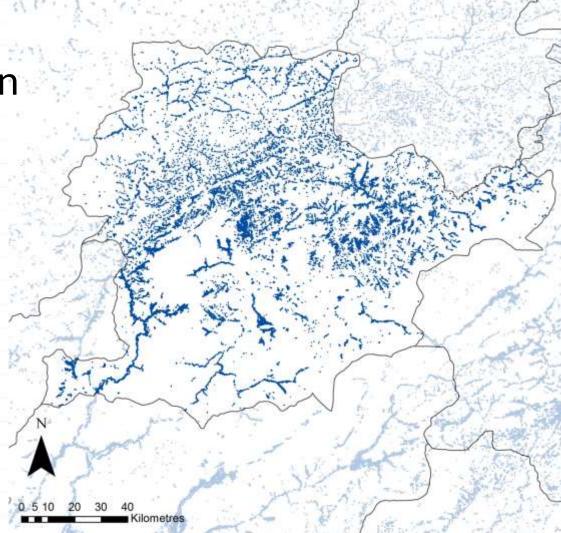


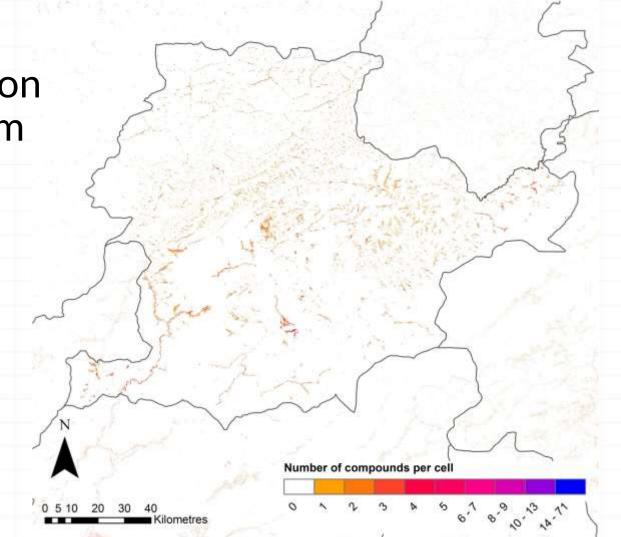
### **Outputs: EA level uncertainties**

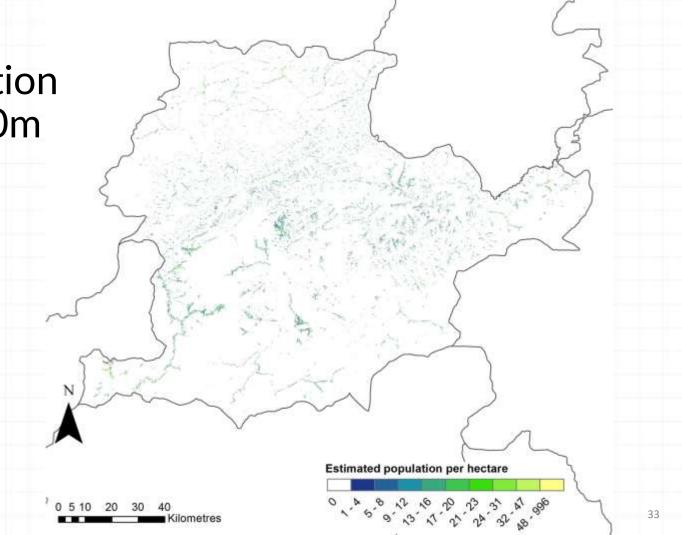


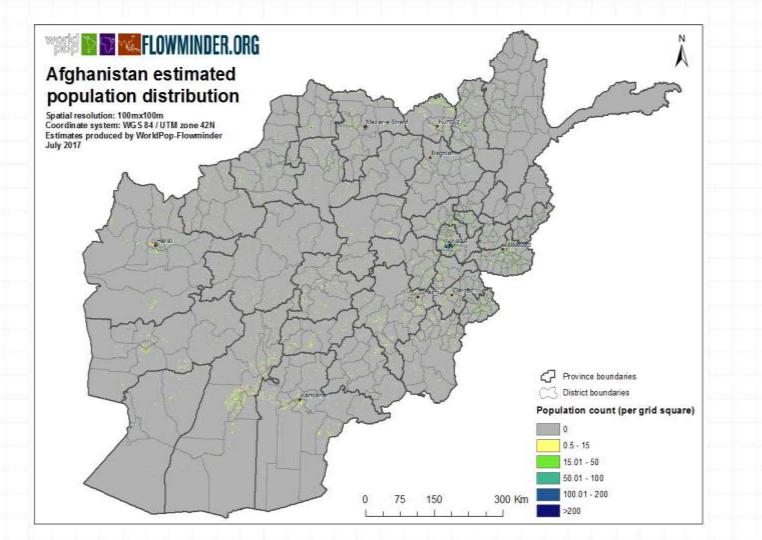


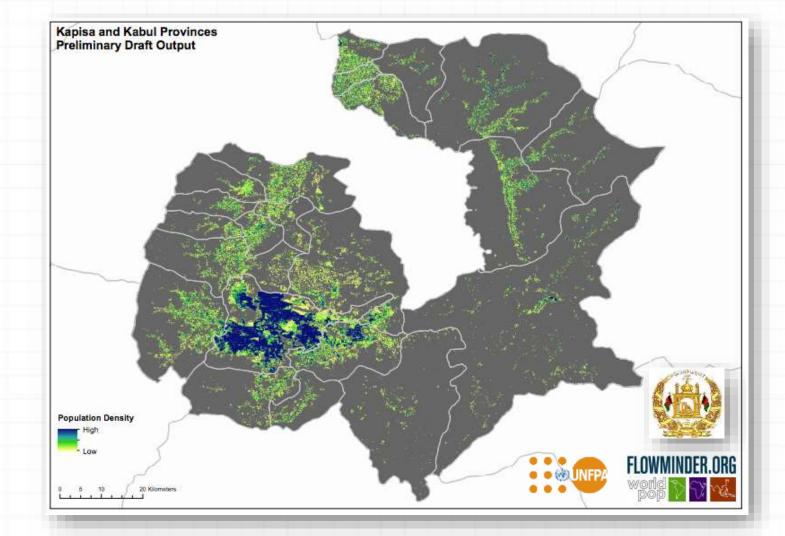
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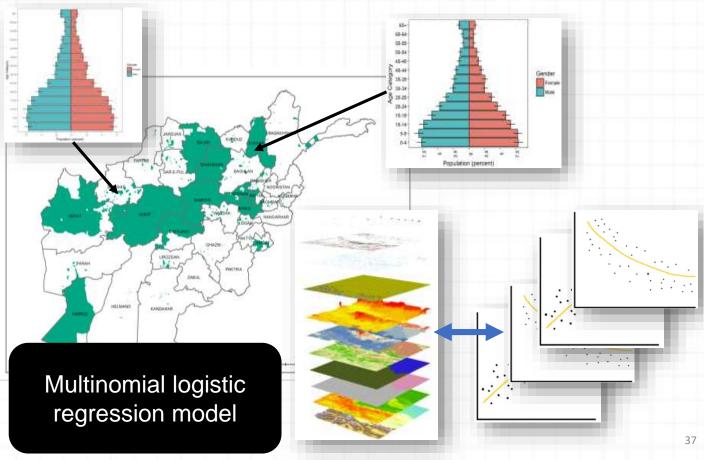




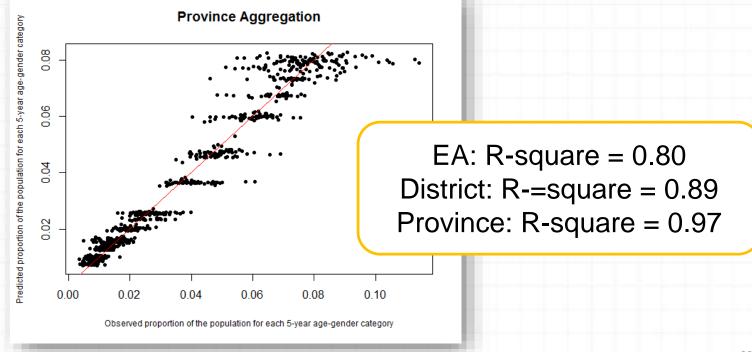


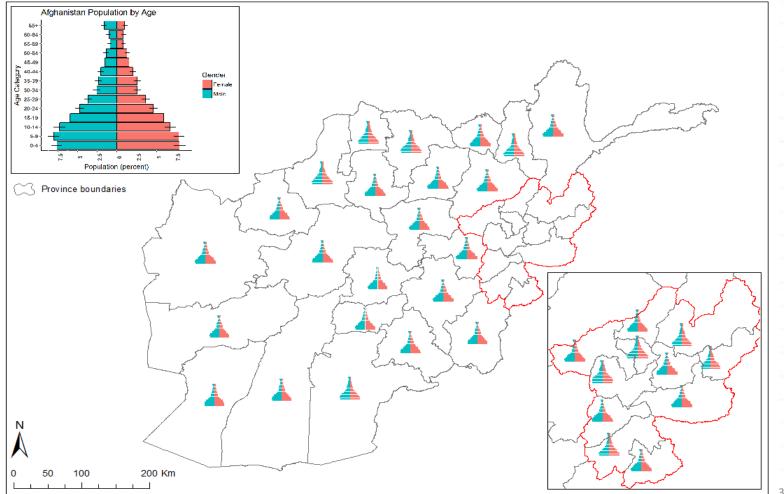
# Modelling population pyramids

## Estimating age/sex structures



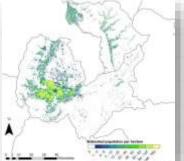
### Age/sex structure cross-validation





# Key considerations

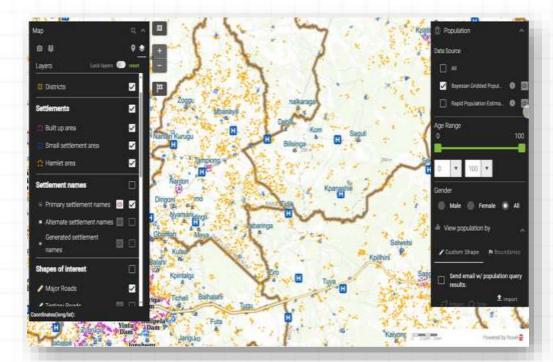
- Uncertainties are substantially higher in unsampled locations
- Uncertainties are higher for predictions made at fine spatial scales
- Mobile populations are not explicitly accounted for
- Estimates can be improved further with new survey and geospatial data





# Usage since 2017

- Child vaccination
  microplanning
- Sample frame for designing new household surveys
- Health financing
- Modelling updates with WHO planned for 2024/25

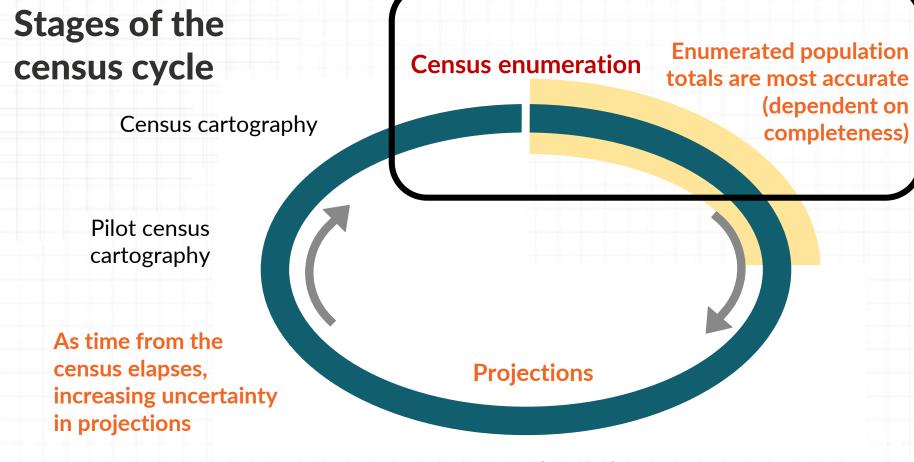


GeoPode polio vaccination portal



## Brief overview: Similar examples

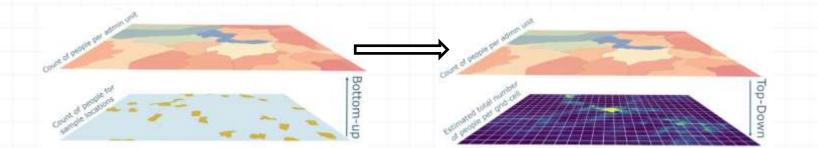
# **WorldFop**



Intercensal period

## Census under-enumeration

- Unable to enumerate in hard-to-access regions
- Bottom-up estimation for under-enumerated units, followed by top-down disaggregation to grid squares

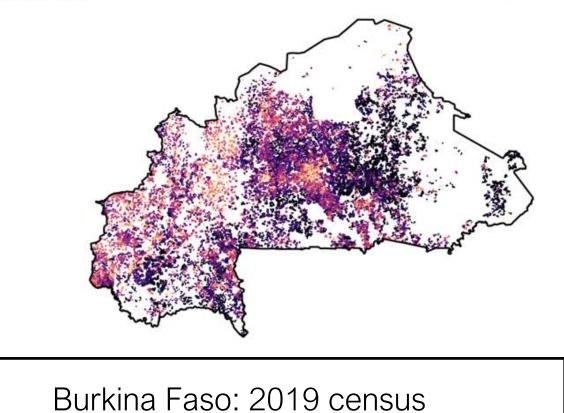


### 5e RGPH Burkina Faso

### **Population Density**

Low

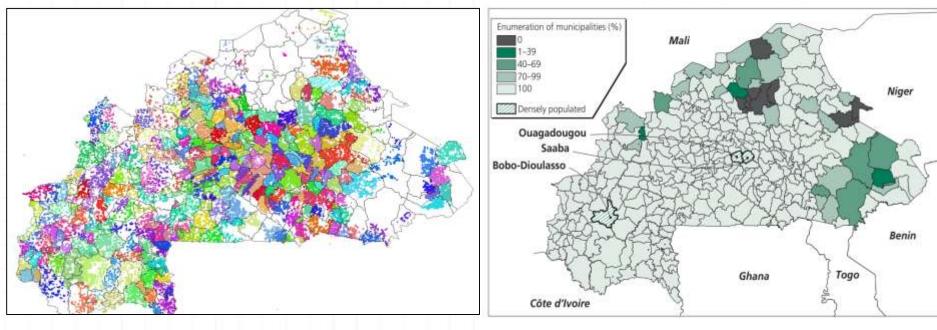
High



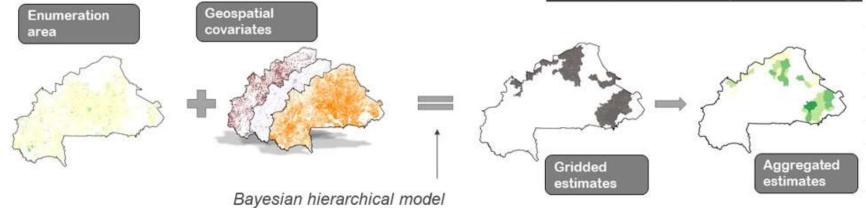


### Census 2019 database: Non-enumerated EAs

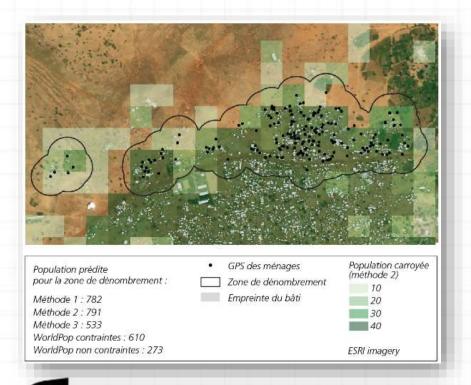
### Figure 1. Coverage of the 351 municipalities (communes) in the 2019 census of Burkina Faso



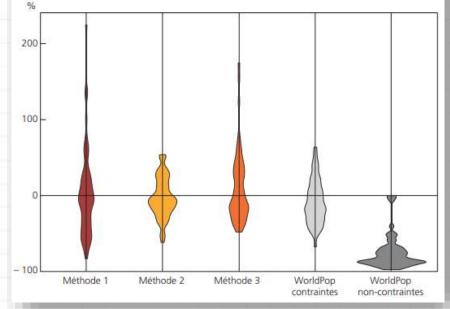
### 1. Bottom-up modelling



## Post-census validation



A) Distributions des erreurs de prédiction (%) Sur les 50 zones de dénombrement test



https://wopr.worldpop.org/?/Population;

Darin et al (2022) Population



Cinquième Recensement Général de la Population et de l'Habitation du **Burkina Faso** 

### Résultats **Préliminaires**







#### DEFINITION

La Recessiverent Général de la Population et de l'Hapitation est une opération s'envergure nationale qui Vise a fourner des priormations d'ordre.

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#### OBJECTIES

- Disposer d'informations détaitées et exhaustives sur la 1- Feffectif gobai de la population, sa structure par seve population at Thabitation. - Depend of une repartition par units administrative

RESULTATS

PRELIMINAIRES

#### **RÉSULTATS ATTENDUS**

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RAPPORT

PRELIMINAIRE



RESULTATS DU 5E RGPH

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### SE RGPH **Tawchargamant**

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16 novembre au 15 décembre 2019

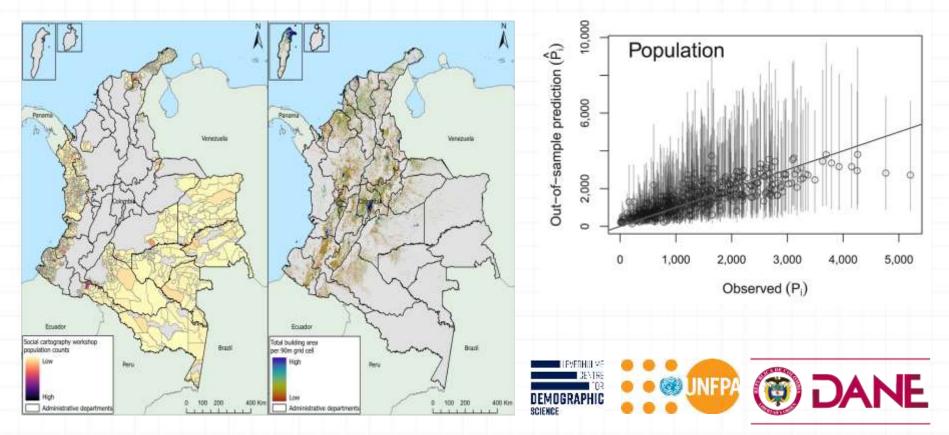
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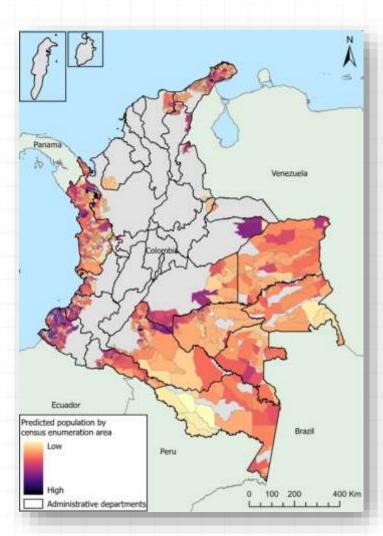
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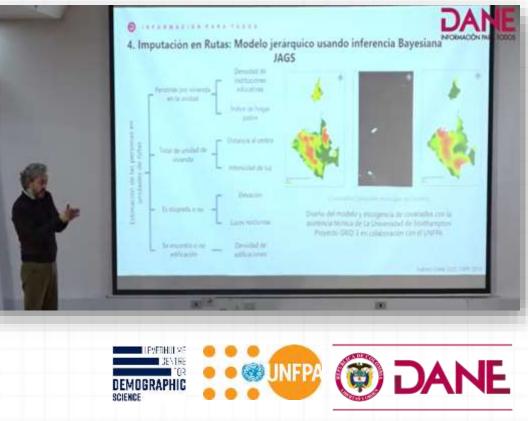
### Filling census gaps using satellite and "social cartography": Colombia



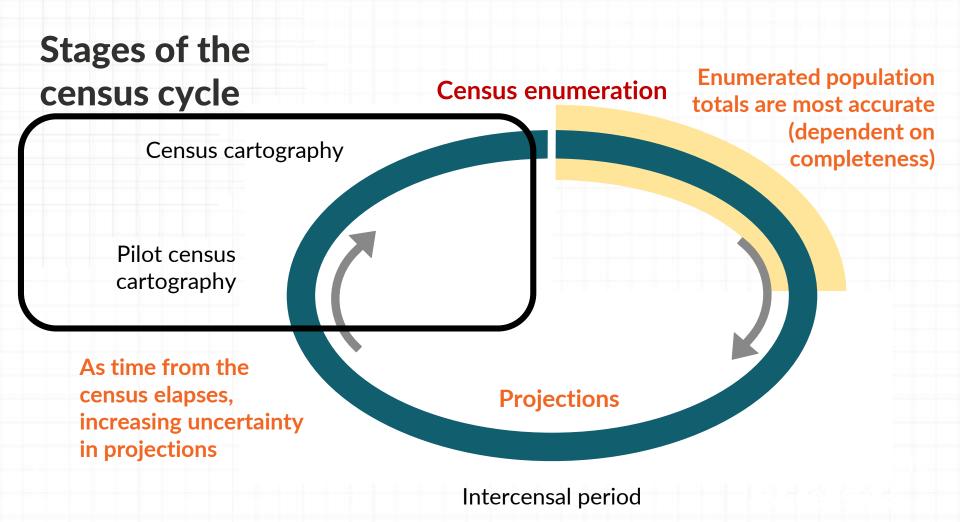
Sanchez-Cespedes et al (2023) Population Studies



# Filling census gaps using satellite and social cartography: Colombia

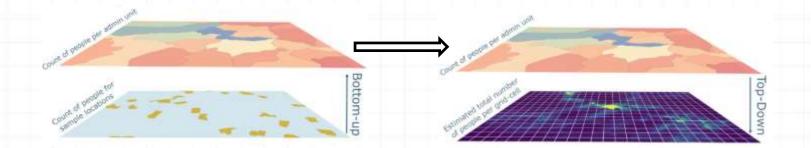


Sanchez-Cespedes et al (2023) Population Studies



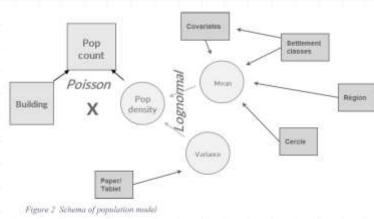
### Census process support: Use of census cartography

- Population counts are collected in census preparation phases e.g. cartography – but can often be incomplete
- Value in having full coverage of recent estimated population for planning the full census
- Bottom-up estimation for under-enumerated units, followed by top-down disaggregation to grid squares



### Census cartography: Mali

- <u>Bottom-up model</u> fitted using the population counts of the 25,466 enumeration areas fully covered by the census cartography
- <u>Top-down model</u> used to map enumeration area total to 100x100m grid square estimates



Census cartography coverage Enumeration areas [25552] Complete [25466] Incomplete (86) Gridded population 0.5666 269.4 538.3 1075 Figure 1 Census cartography coverage https://wopr.worldpop.org/?/Population



### WorldPop and INSTAT (2022) 10.5258/SOTON/WP00745

## Questions?

To think about (as you listen to case studies):

-Do the case studies presented have similarities to your own situations?

*-If you see a need for it in your country, what types of population data might be available to support bottom-up estimation modelling?*