

Development goal denominators

- All MDGs and SDGs are based on ensuring a certain percentage of the population has access to specific services or resources, or achieves a certain level of social, economic, or physical health.
- Improved understanding of *sub-national geographic variation and inequity* in health status, wealth, and access to resources within countries is increasingly recognized as central to meeting development goals.
- Requires a consistent, comparable and regularly updated understanding of not only how many people live in a country, but where the people are, and who they are.

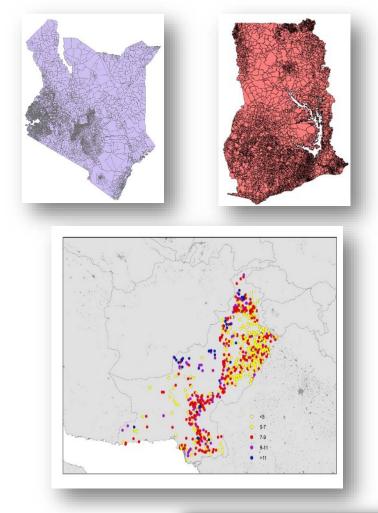


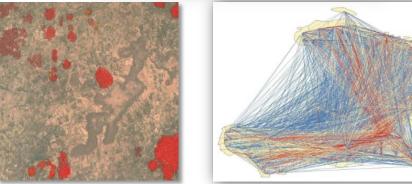
WorldPop

Improve the spatial demographic evidence base for low and middle income countries

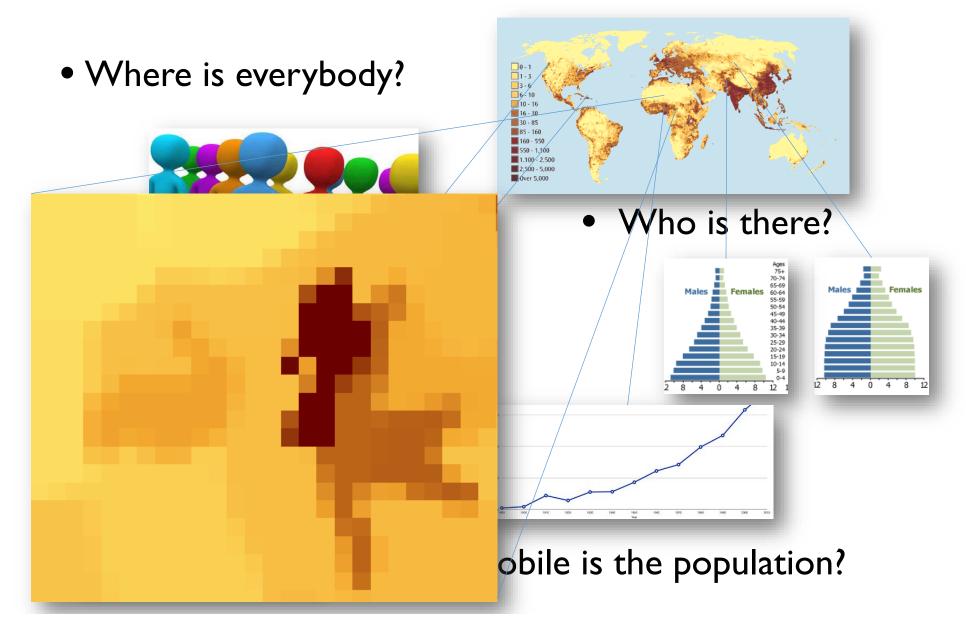
- Work with statistical agencies, health depts to obtain finest detail, boundary-matched census data
- Develop scalable methods and models for integrating ancillary datasources to complement and fill data gaps in census
- Integrate new technologies
- Freely available high resolution data and fully documented, peer-reviewed methods







What do we mean by spatial demographics here?



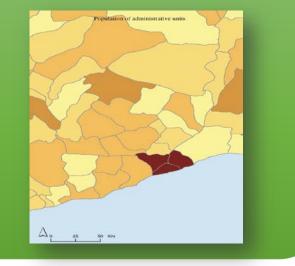
Modelling population distributions



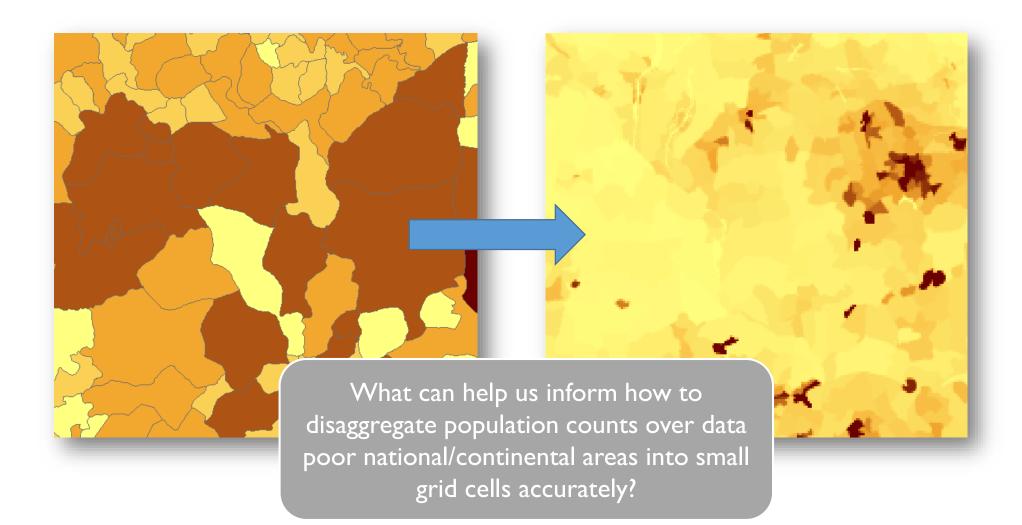
Bottom-up: where census data are outdated/unreliable, integration of high resolution settlement mapping with survey data



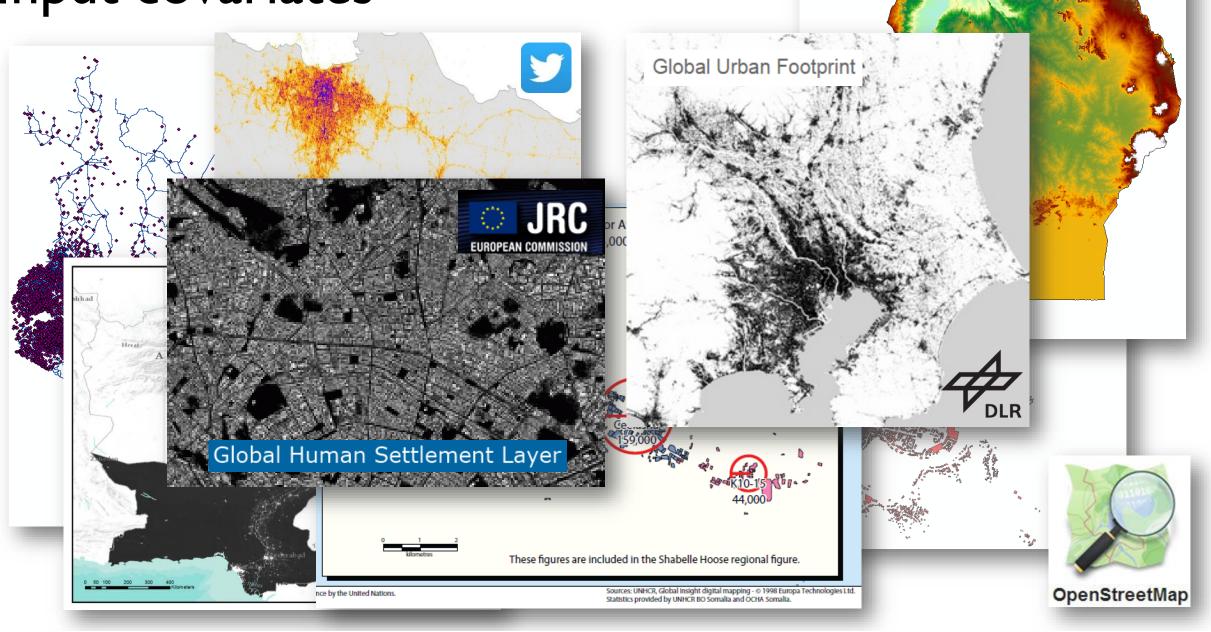
Top-down: disaggregation of admin-unit based census/official estimate counts



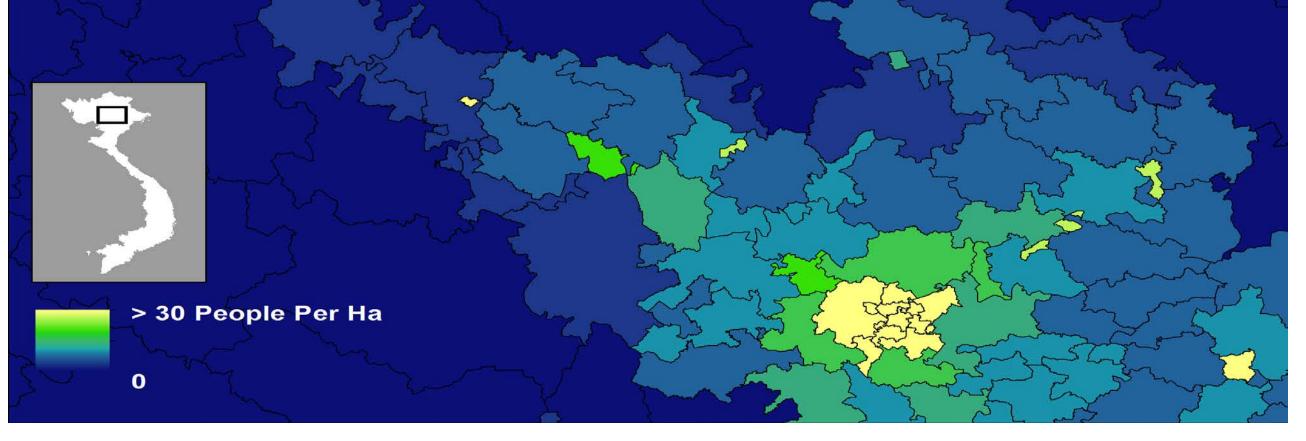
Top-down: How can we go from aggregated counts to gridded surfaces?



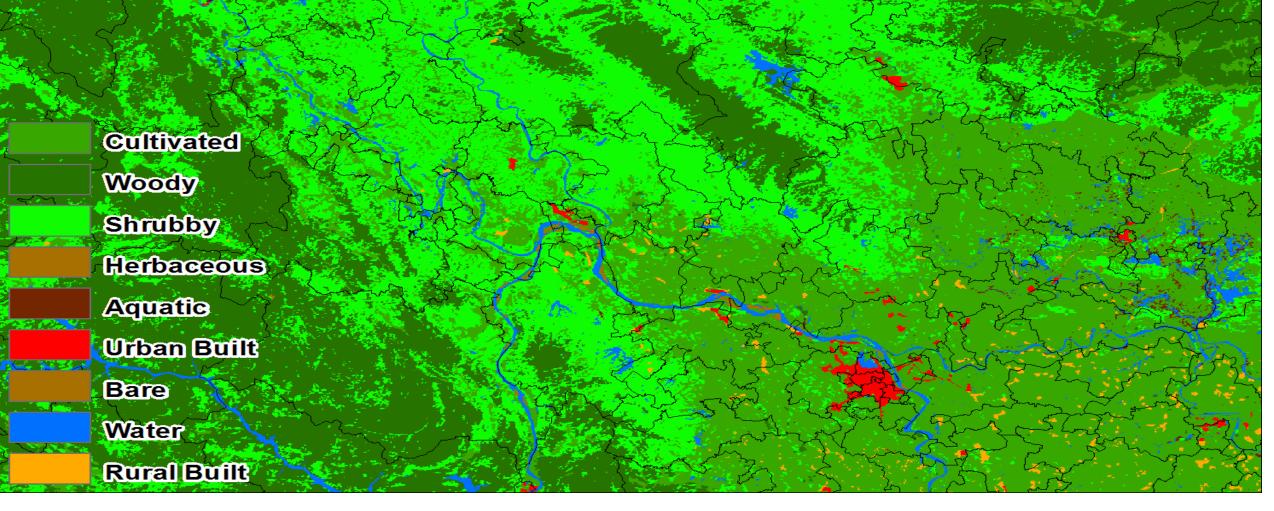
Input covariates



Population Density Vietnam Administrative Unit Level 002



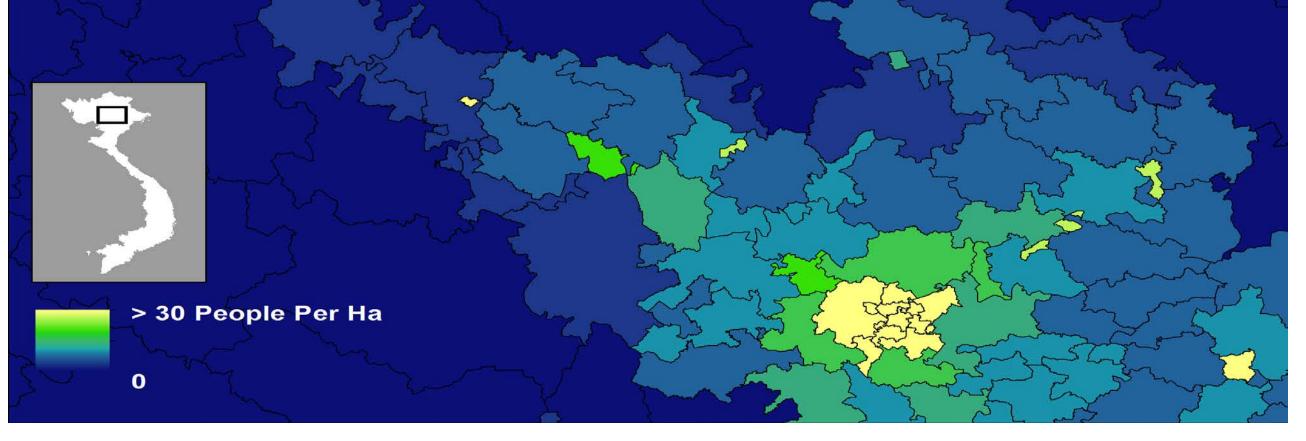
Vietnam MDA GeoCover Land Cover (30 m Pixels)



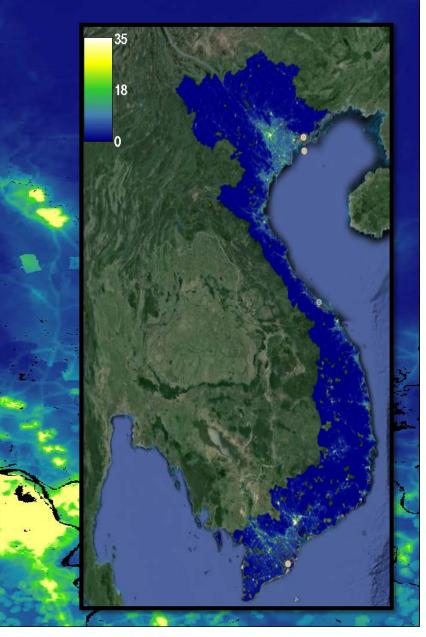
NOAA Suomi VIIRS-derived Lights at Night 2012 for Vietnam

Session 6. Data disaggregation and utilization challenges: Andrew Tatem (University of Southampton) – Experience of WorldPop in mapping population numbers, demographics and behaviours

Population Density Vietnam Administrative Unit Level 002

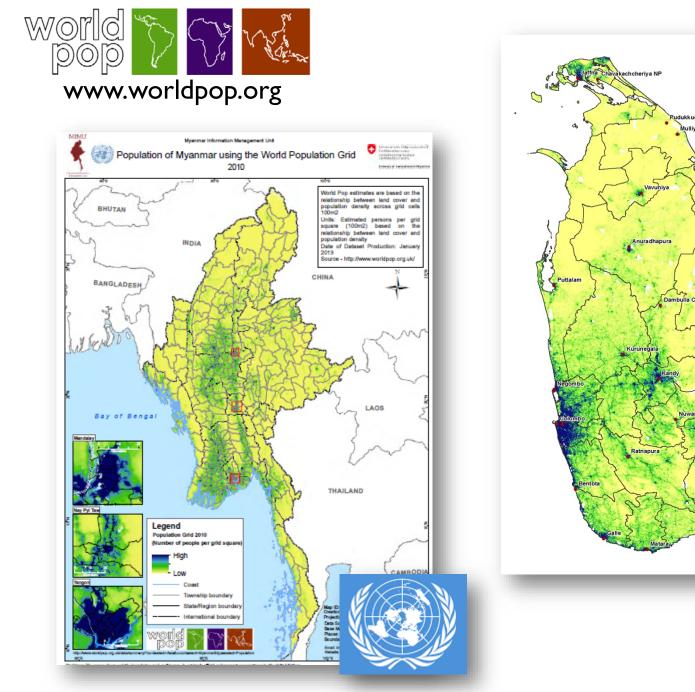


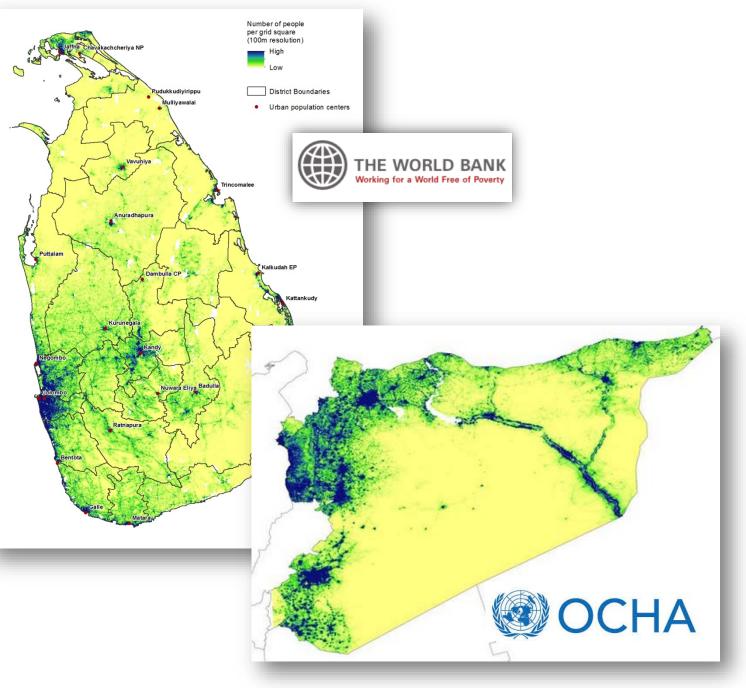
Vietnam RF Distributed Population Counts Using Non-Default Ancillary Data (OSM)



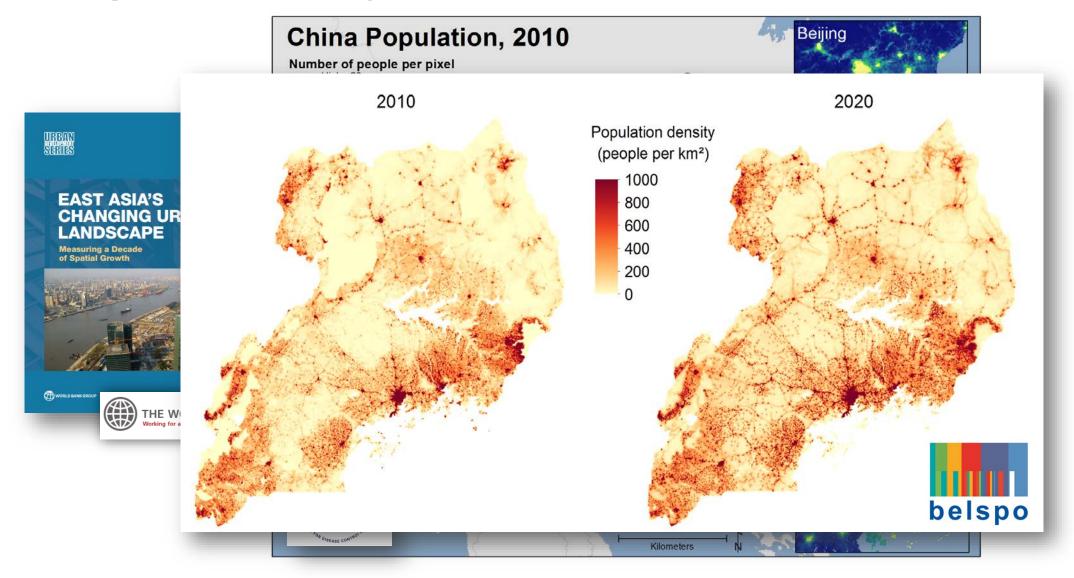
> 30 People Per Pixel

Low : 0

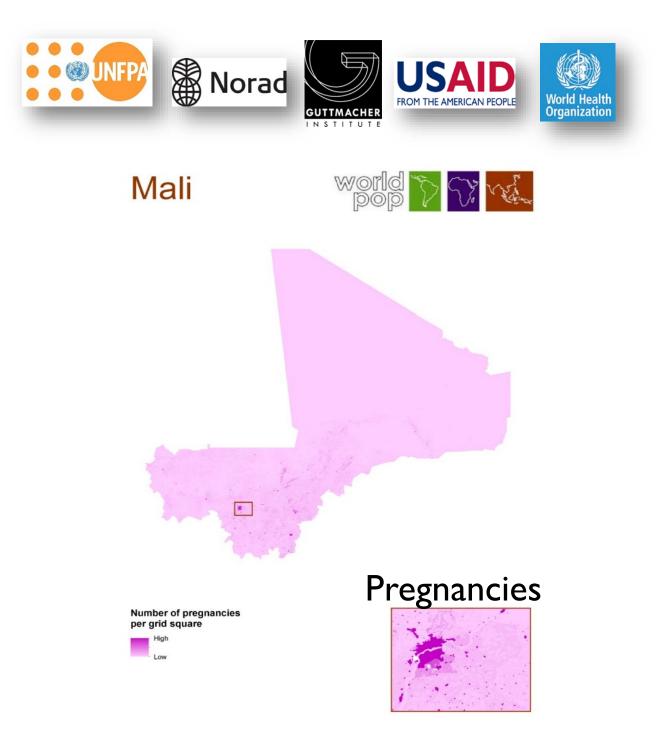




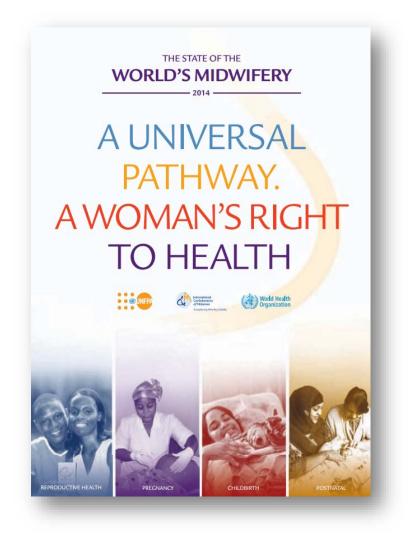
Temporal change



Gaughan et al (2015) under review; Linard et al (2013) Applied Geography; Schneider et al (2015) Env Res Letters

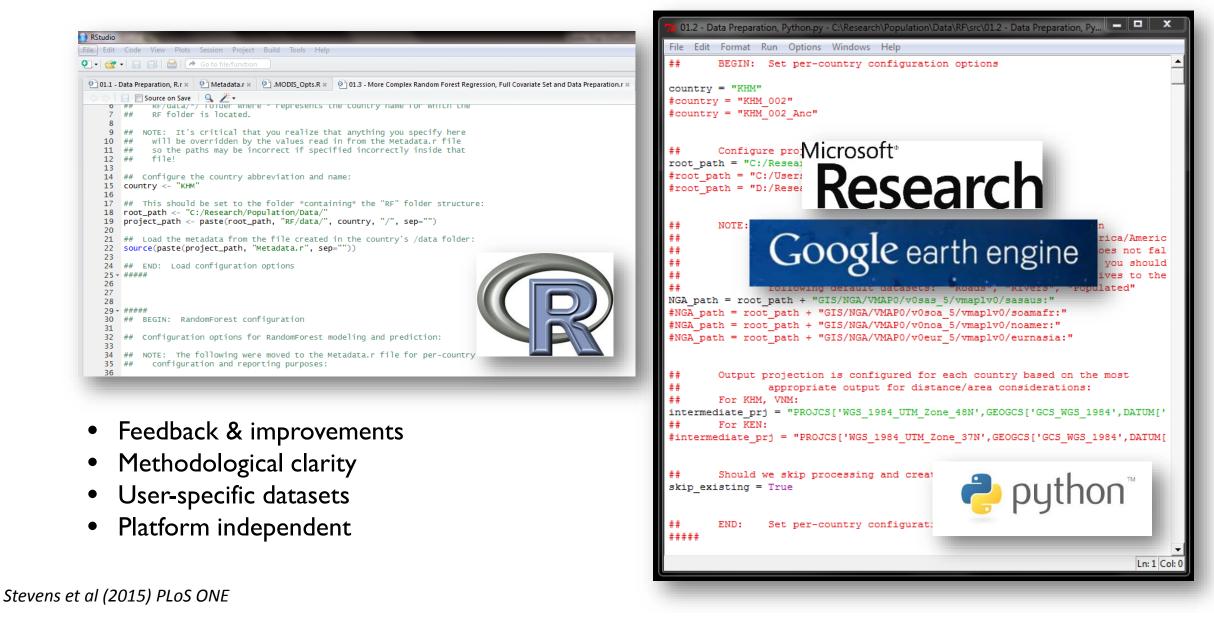


Data integration: Maternal and newborn health

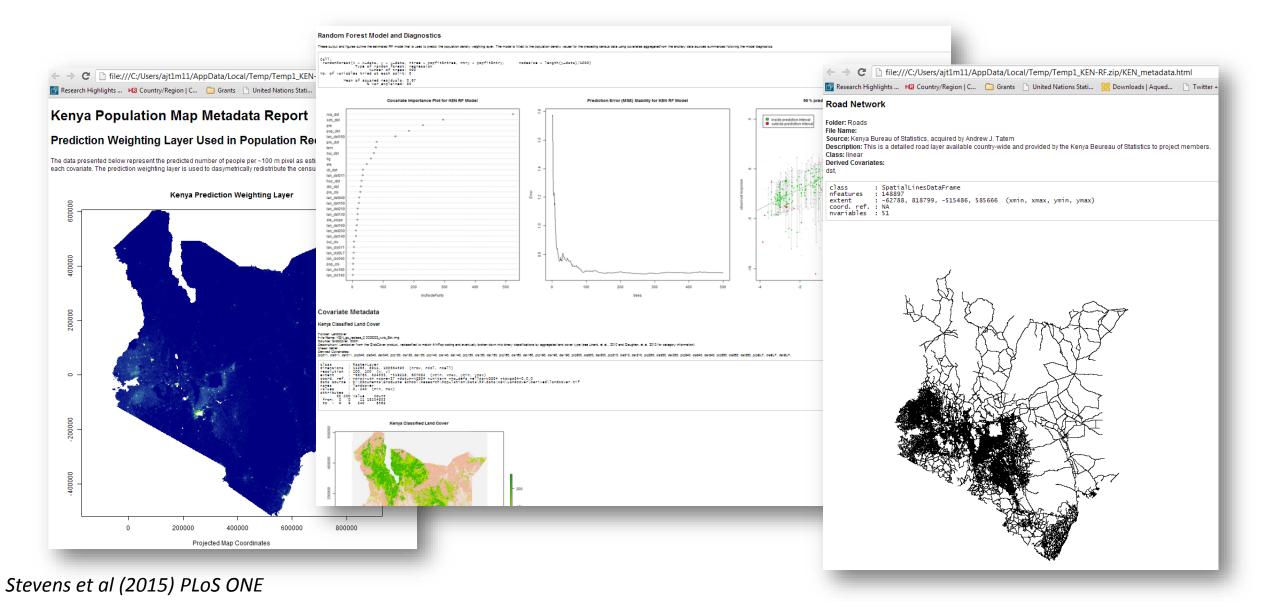


Tatem et al (2014) IJHG, Ebener et al (2015) IJHG

Open code



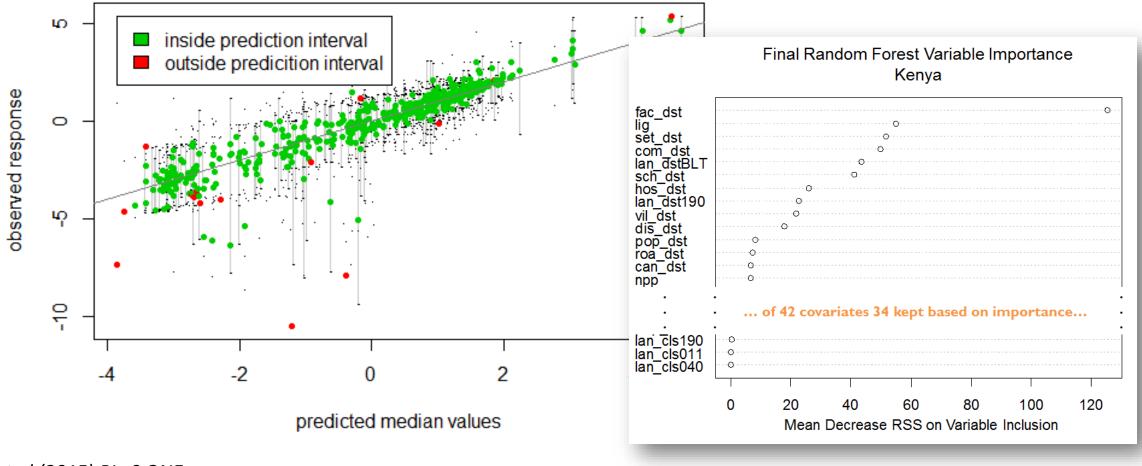
Automated metadata



Session 6. Data disaggregation and utilization challenges: Andrew Tatem (University of Southampton) - Experience of WorldPop in mapping population numbers, demographics and behaviours

Internal and external accuracy assessments

90 % prediction intervals on out-of-bag data



Stevens et al (2015) PLoS ONE

www.worldpop.org

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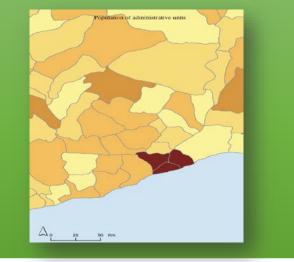
Modelling population distributions



Bottom-up: where census data are outdated/unreliable, integration of high resolution settlement mapping with survey data

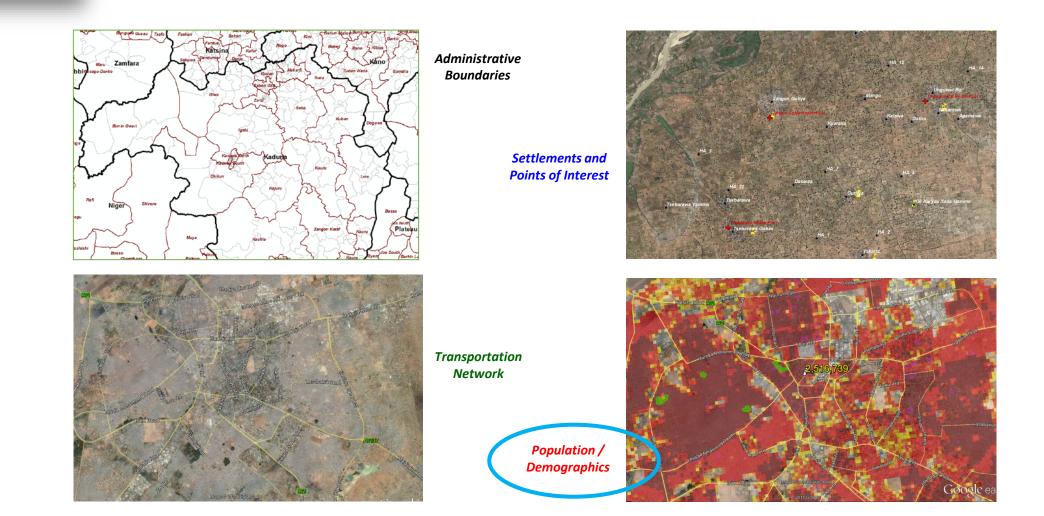


Top-down: disaggregation of admin-unit based census/official estimate counts



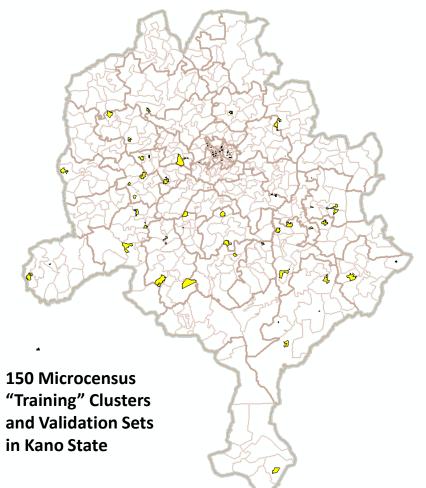


Core Data Layers

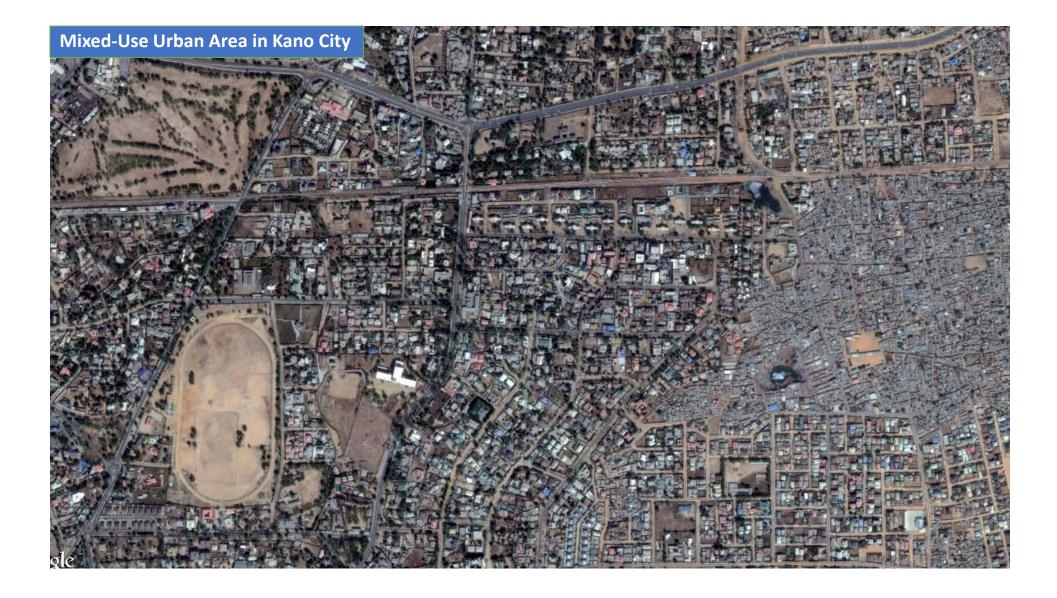


Basis for Population Modeling with Satellite Imagery

- Obtain High-Resolution Satellite Imagery
- Extract Settlement & Building Features
- Identify Neighborhood categories, covariates
- Collect Microcensus/Validation Data
- Generate & Validate Initial Output
- If Needed, Collect & Validate Additional Microcensus Data.
- Repeat until Accuracy Requirements are met.

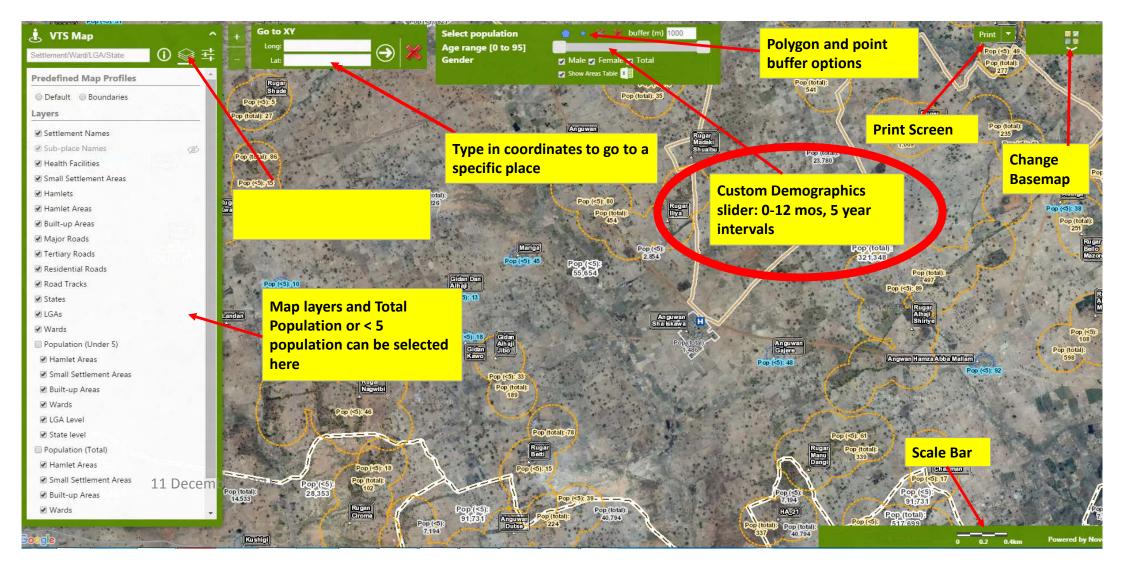






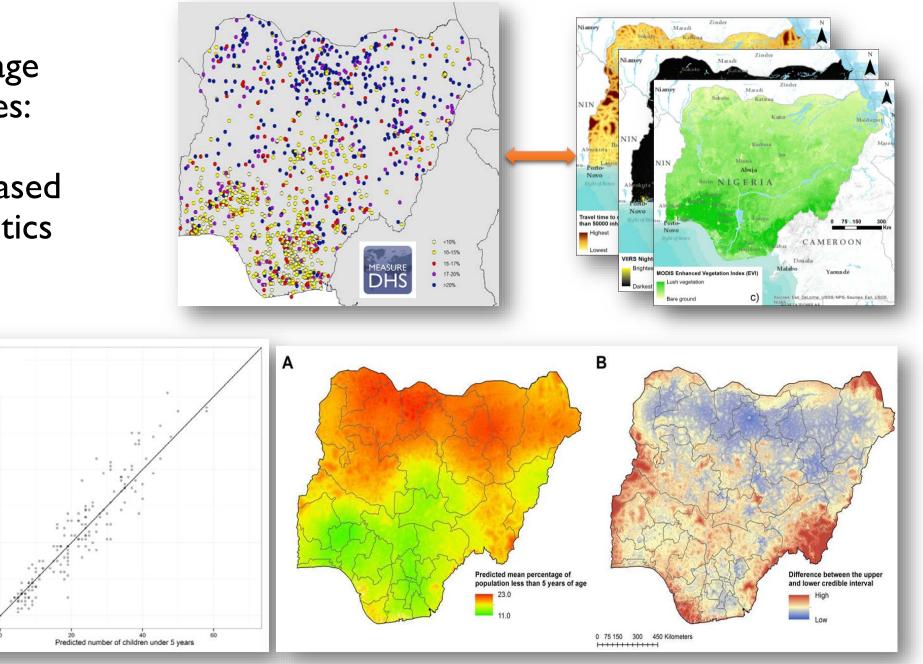
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POPULATION MODEL – USER INTERFACE OPTIONS

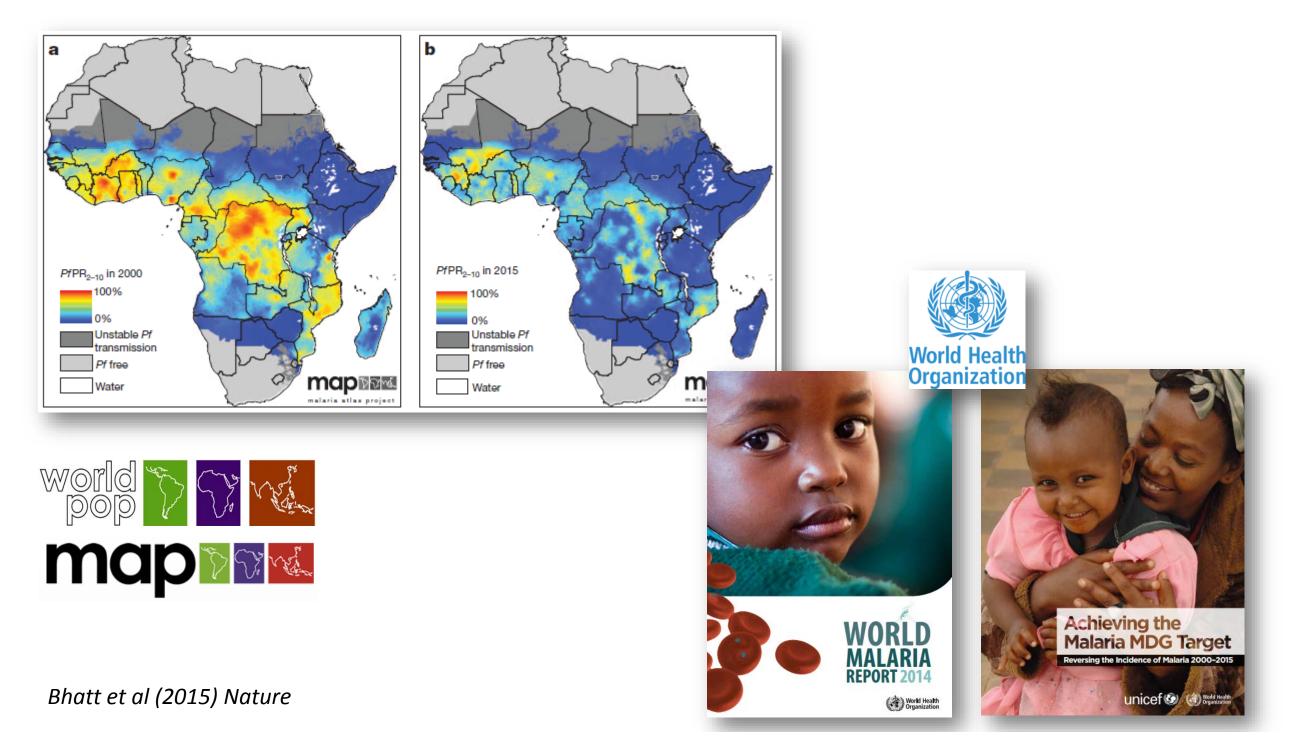


http://vts.eocng.org/

Nigeria age structures: Bayesian model-based geostatistics



Alegana et al (2015) Royal Society Interface



SPATIAL INTERPOLATION WITH DEMOGRAPHIC AND HEALTH SURVEY DATA: KEY CONSIDERATIONS

DHS SPATIAL ANALYSIS REPORTS 9



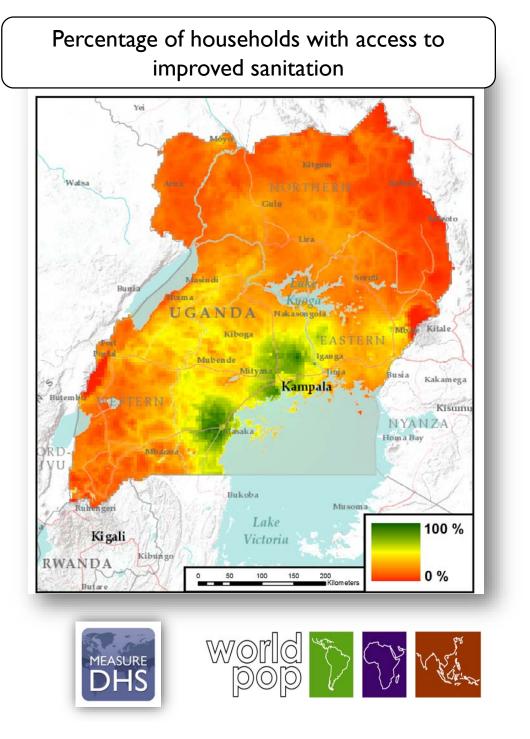
EPTEMBER 201

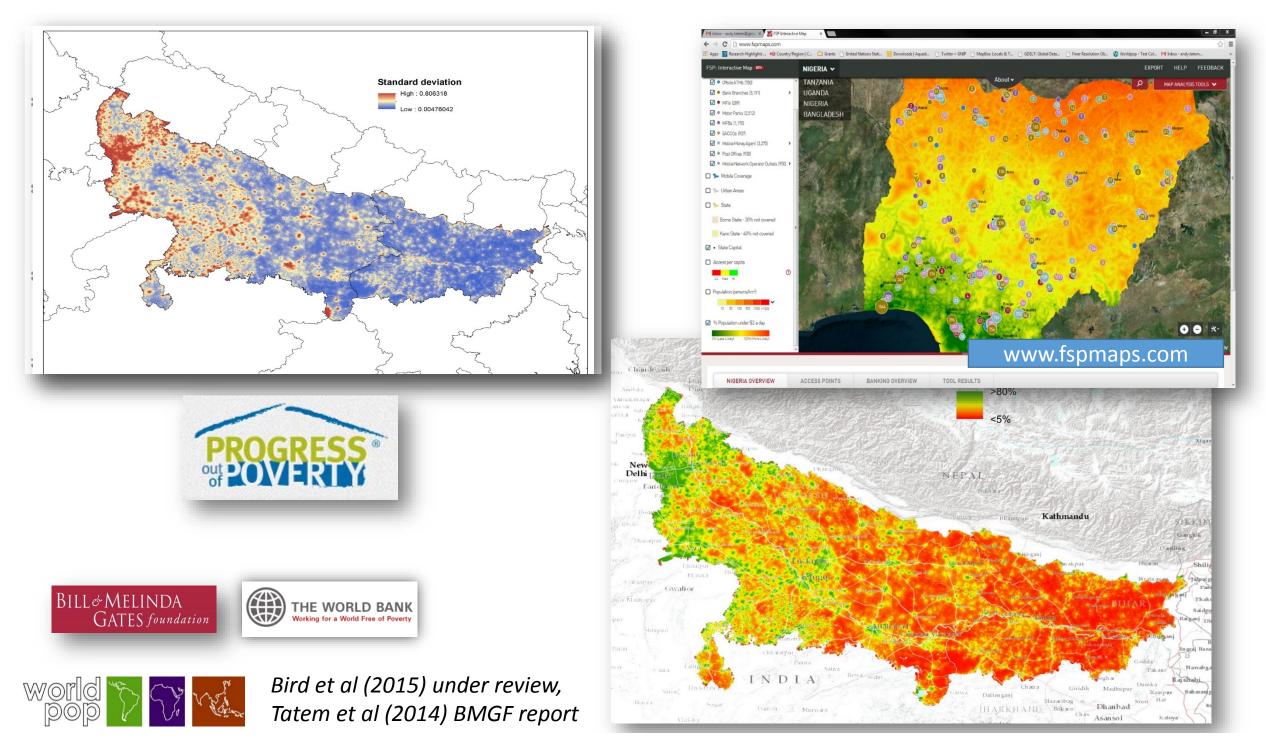
CREATING SPATIAL INTERPOLATION SURFACES WITH DHS DATA

DHS SPATIAL ANALYSIS REPORTS 11



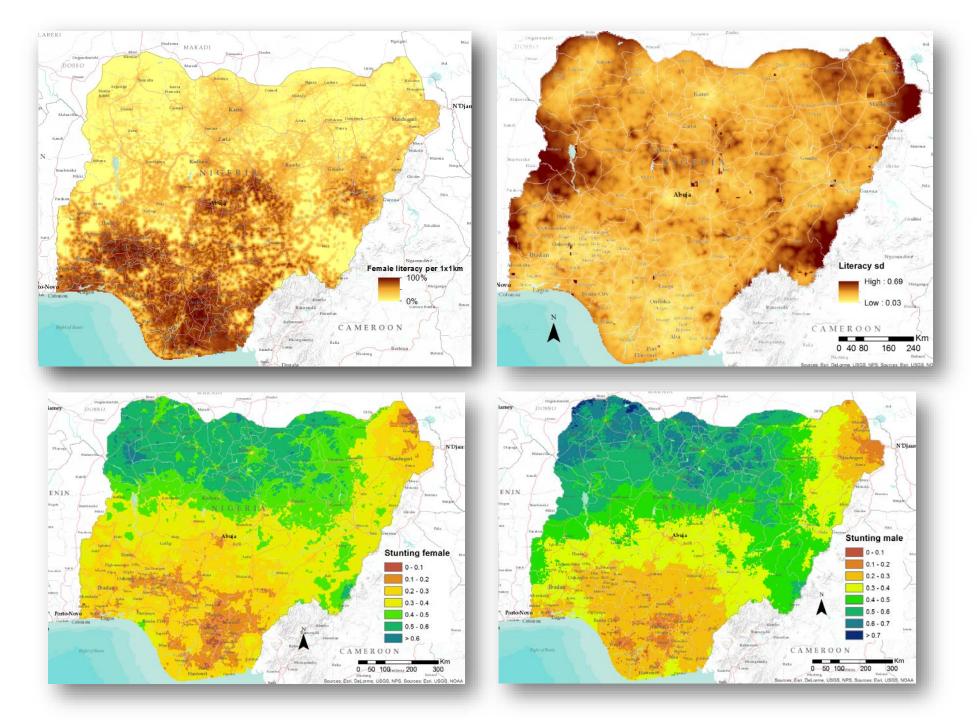
SEPTEMBER 2015 This publication was produced for review by the United Status Agency for International Development (USAID). The report was propared by Paur Gatting, Andy Tatam, Tom Bird, and Clara R. Burgert-Brucker of ICF International, Rockville, MD, USA.







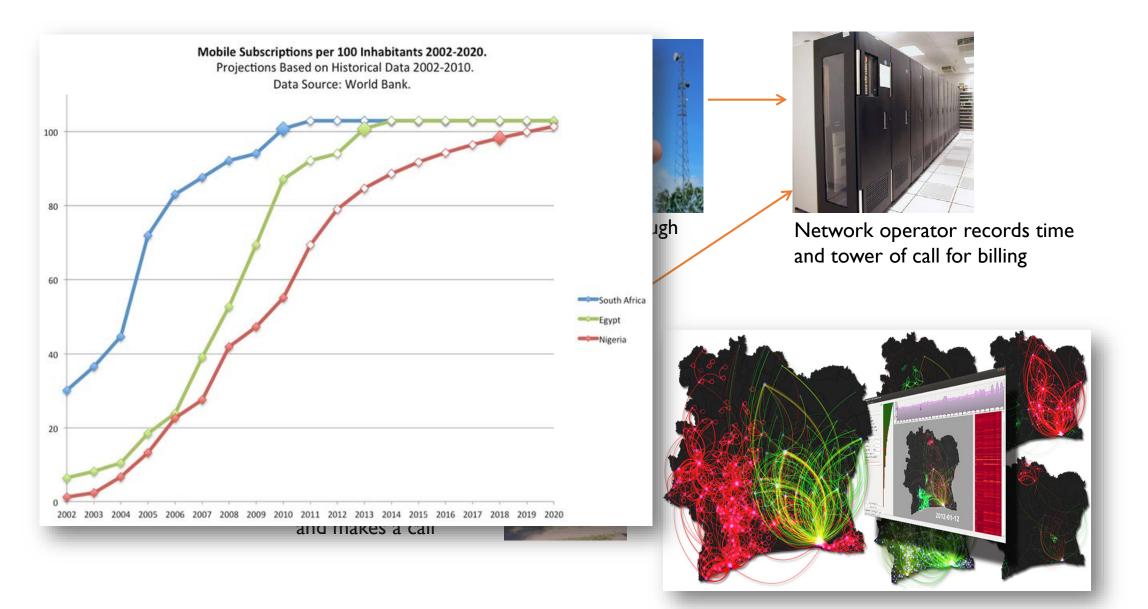


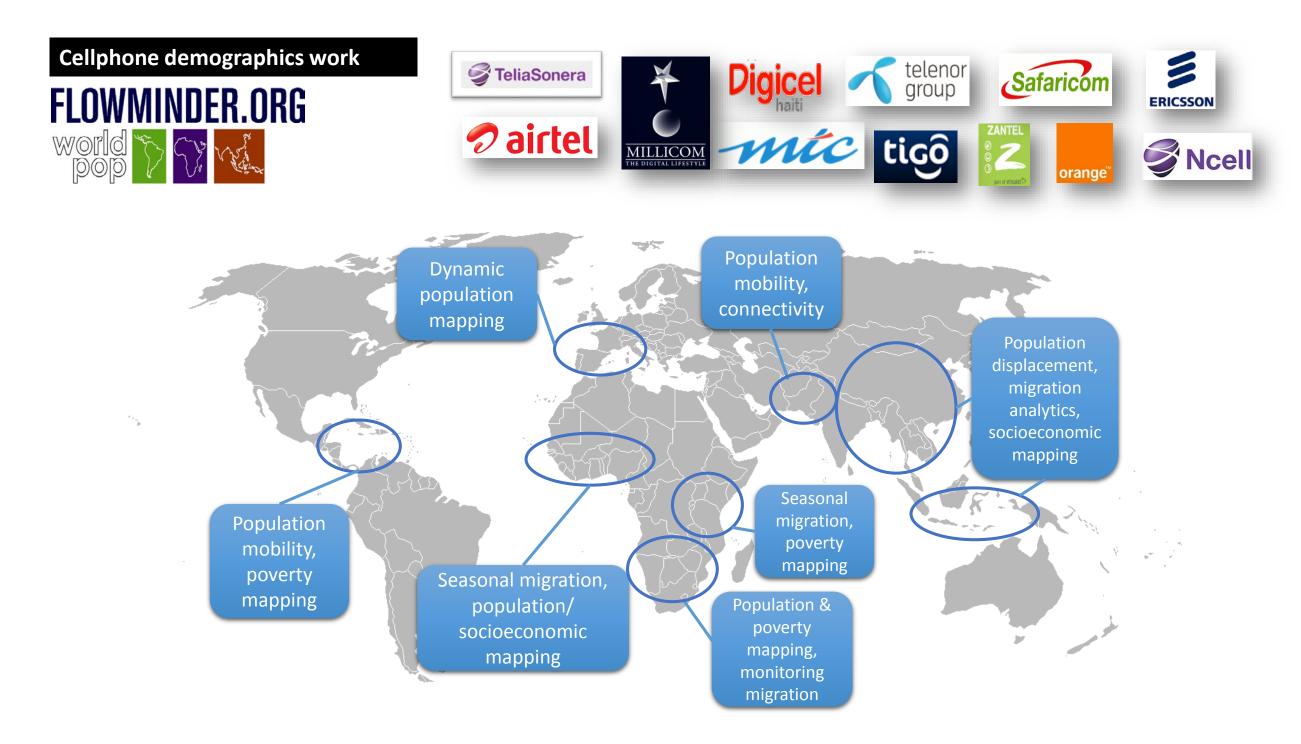


World Der.ORG

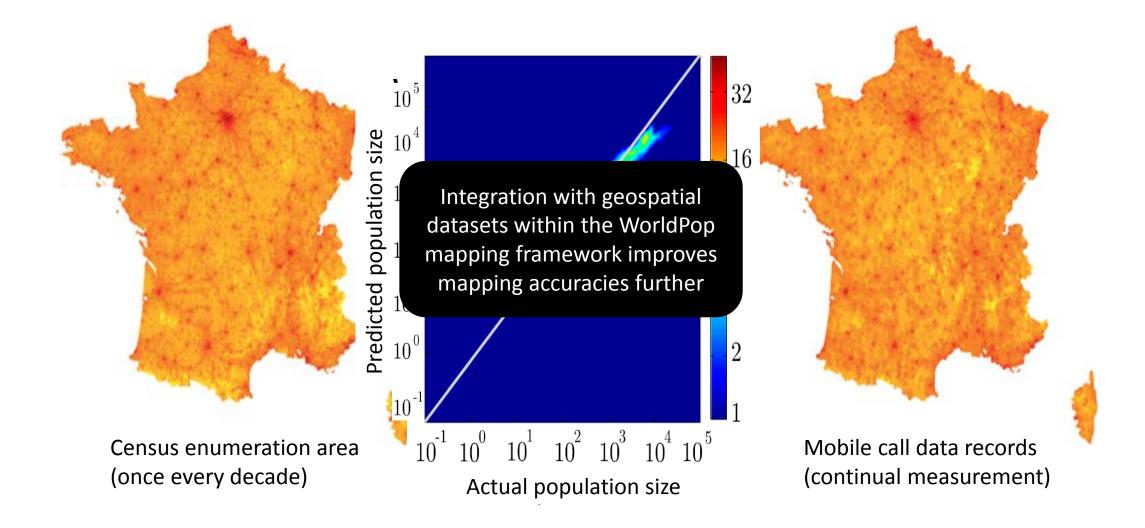
Session 6. Data disaggregation and utilization challenges: Andrew Tatem (University of Southampton) – Experience of WorldPop in mapping population numbers, demographics and behaviours

Cellphone call data records (CDRs)





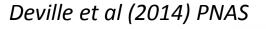
Dynamic population mapping



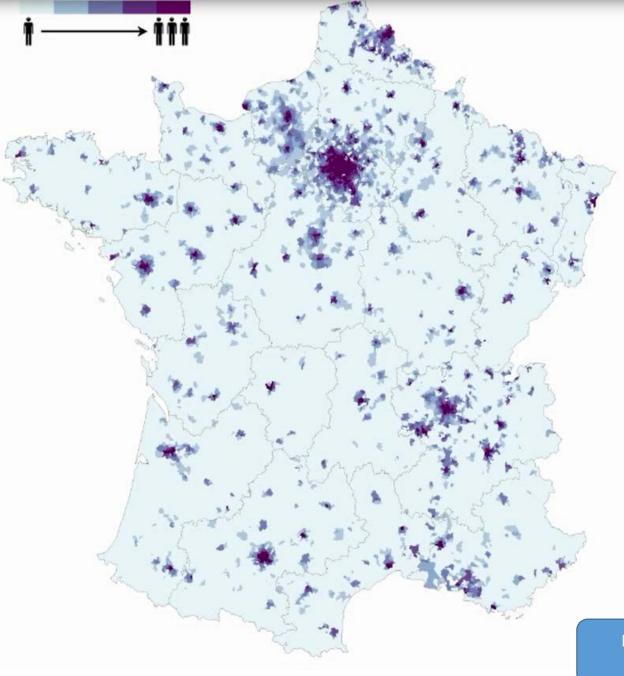


Deville et al (2014) PNAS

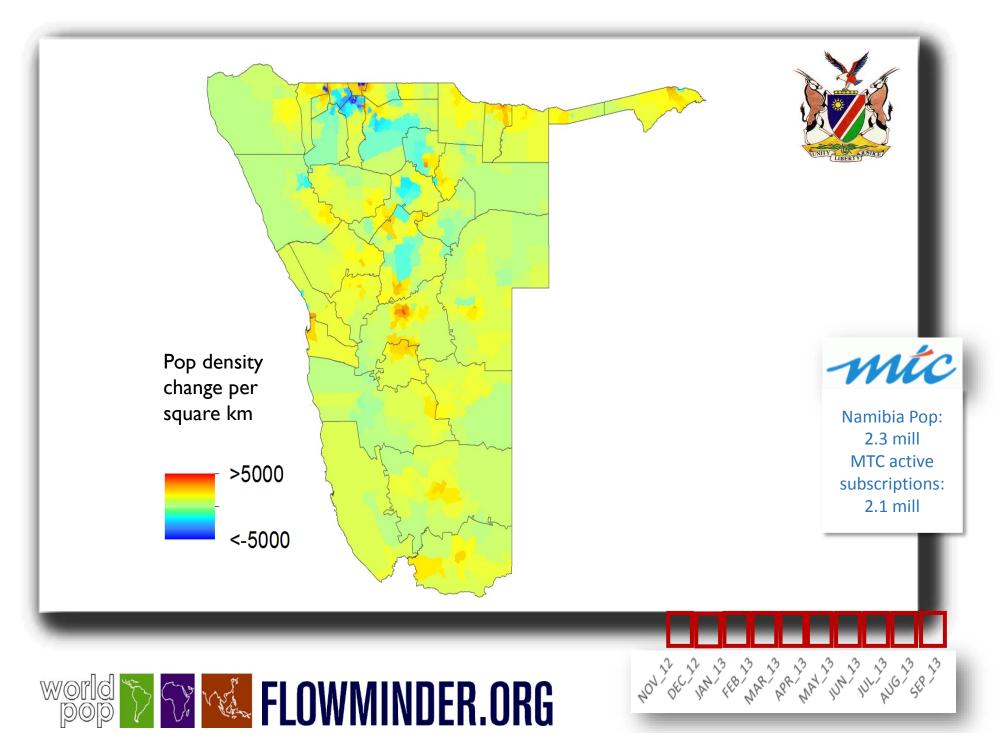
Wednesday Thursday Friday Saturday Sunday Monday



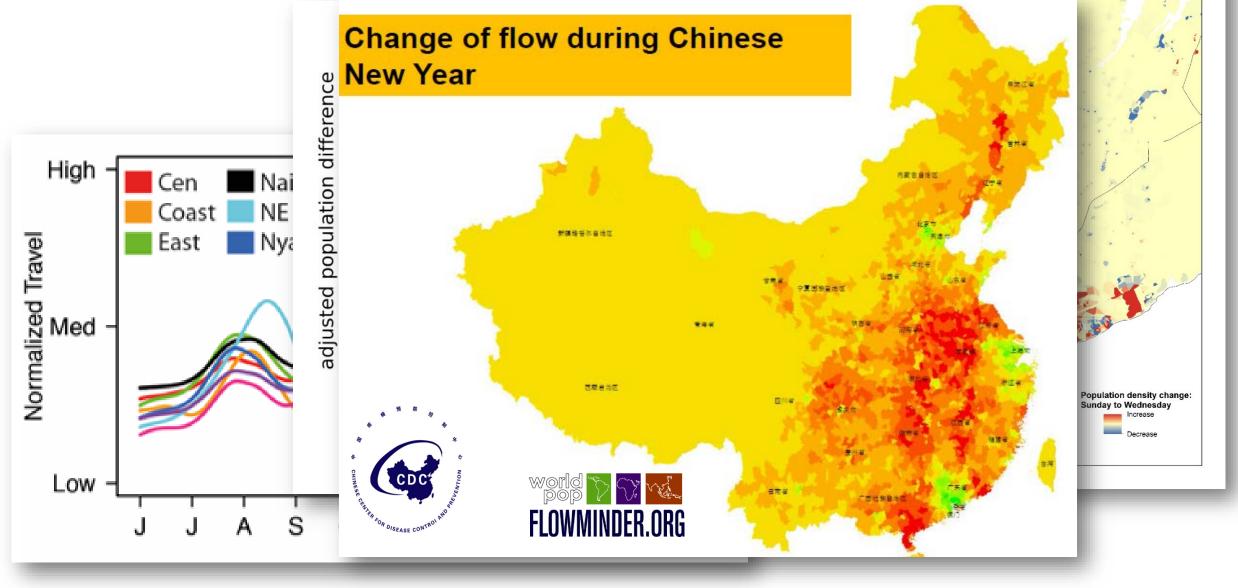




Monthly data available: www.worldpop.org



Population mobility across temporal scales



FLOWMINDER.ORG

Population displacements

Above normal inflow to each district

(negative numbers indicate less incoming people than normal)

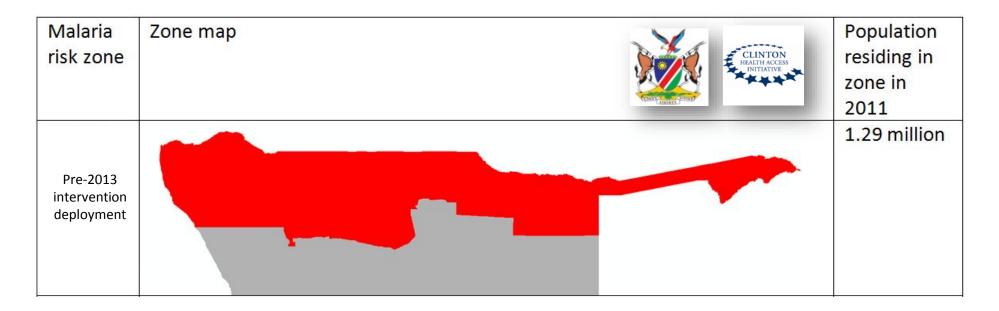


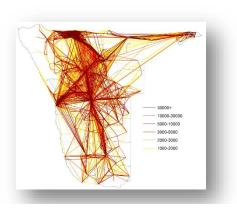
Modelling connectivity for disease elimination



Wesolowski et al (2014) PLoS Currents Outbreaks; Garcia et al (2014) Migration Studies; Sorichetta et al (2015) in review; Tatem et al (2009) Malaria Journal

Data integration: Targeting interventions

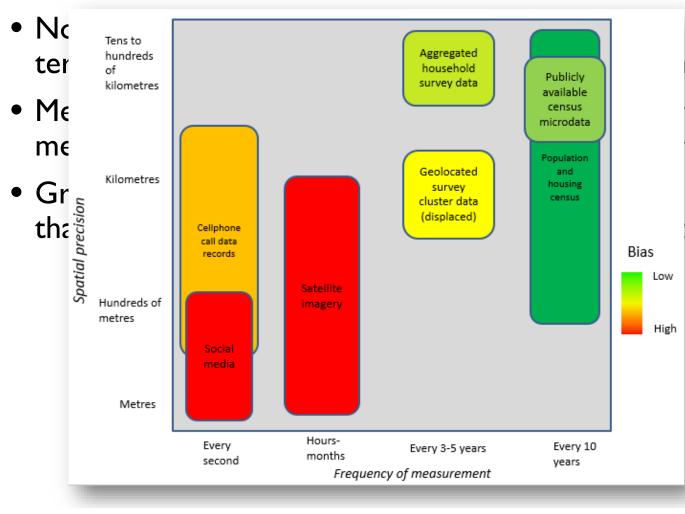




Tatem et al (2014) Malaria Journal

Summary

• In producing estimates for different geographical scales and time periods, the integration of multiple types of data to compliment traditional sources is required



each has advantages over census data in I precision

nties and providing clear nportant

es to build strong demographic databases proaches



Acknowledgements





Further information



www.worldpop.org

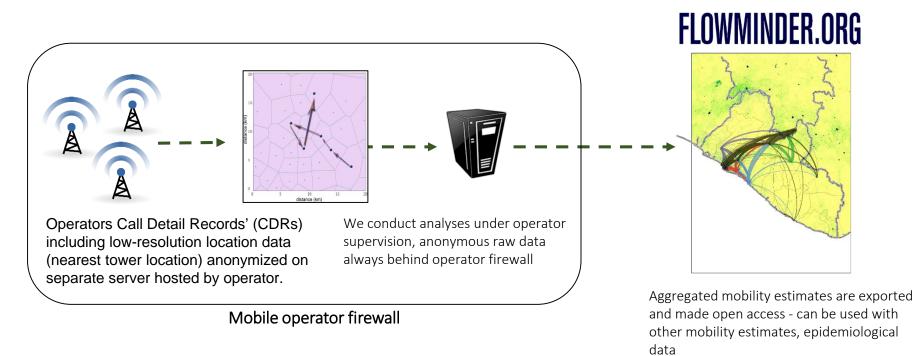


Session 6. Data disaggregation and utilization challenges: Andrew Tatem (University of Southampton) – Experience of WorldPop in mapping population numbers, demographics and behaviours

UN EGM on Strengthening the Demographic Evidence Base For The Post-2015 Development Agenda, New York, 5-6 October 2015

Extra slides

Preserving confidentiality

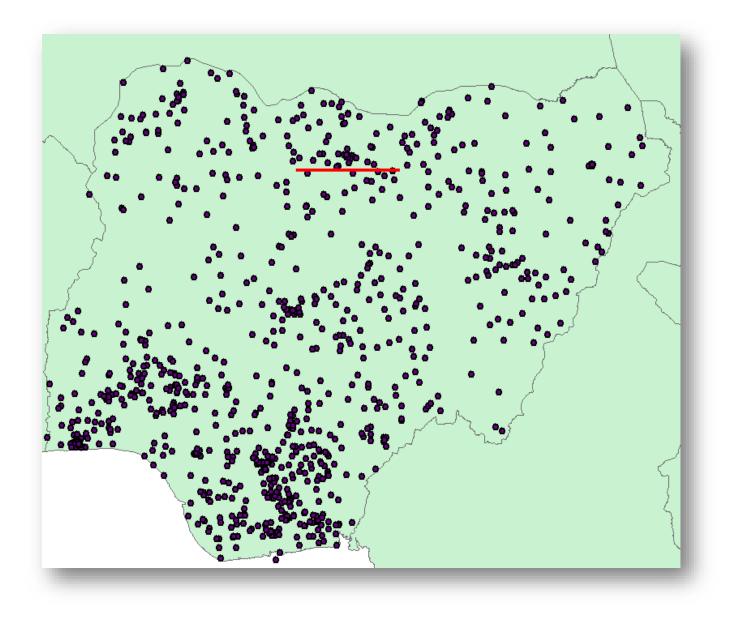


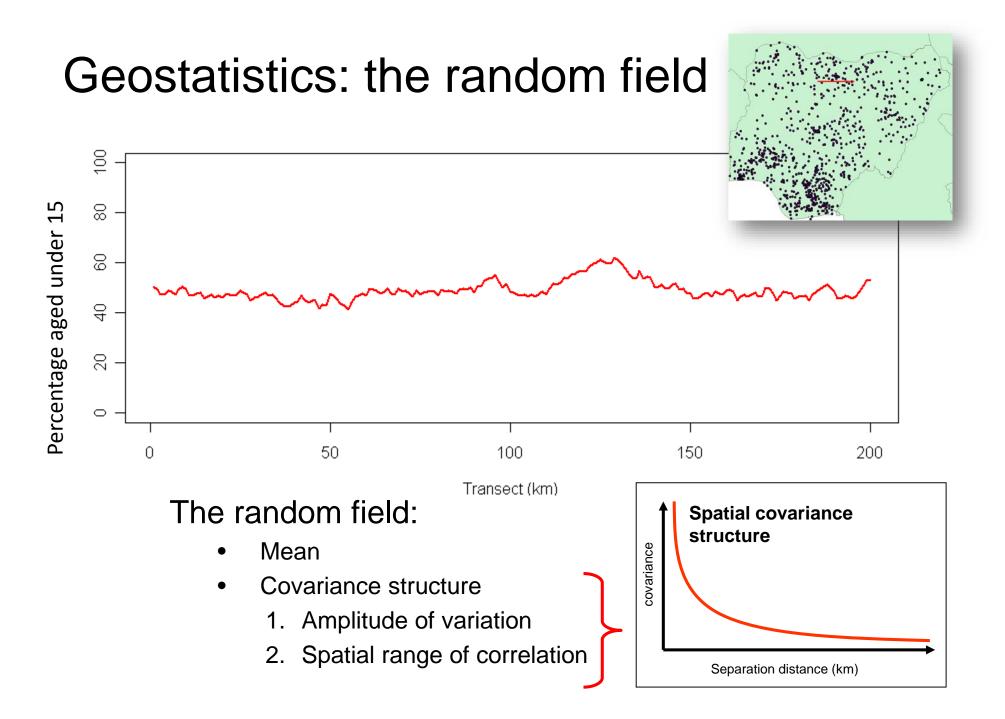
<u>Compliance with GSMA data integrity guidelines</u>: Data never leaves mobile operator's system to avoid any privacy, commercial concerns.

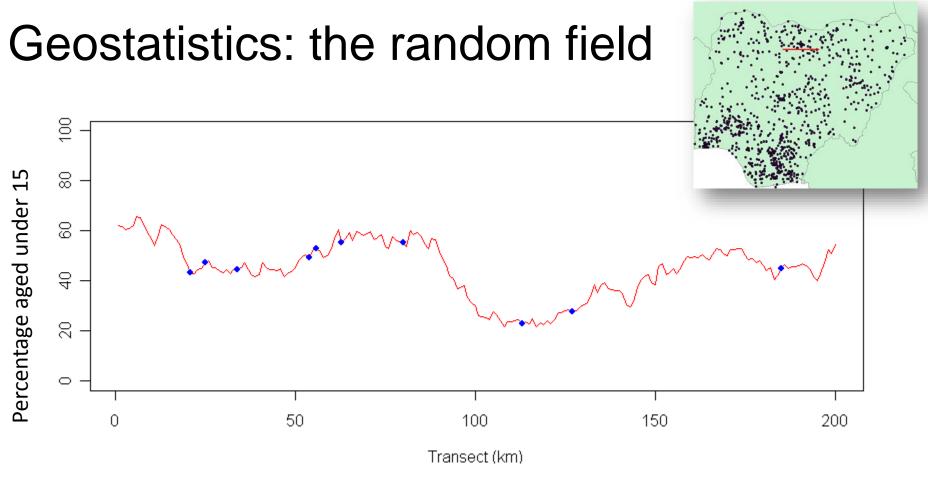
Problem statement

- We have
 - Hundreds of geolocated community surveys
 - Spread (unevenly) across areas of interest
 - Spread (unevenly) through time
- We want
 - Per-grid square estimates of age structure/household size etc
 - Must estimate these variables at locations with no data
 - Want to model uncertainty in our estimates
- Approach
 - Space-time model based geostatistics within a Bayesian framework
 Image is a set of the set of t

Geostatistics: the random field

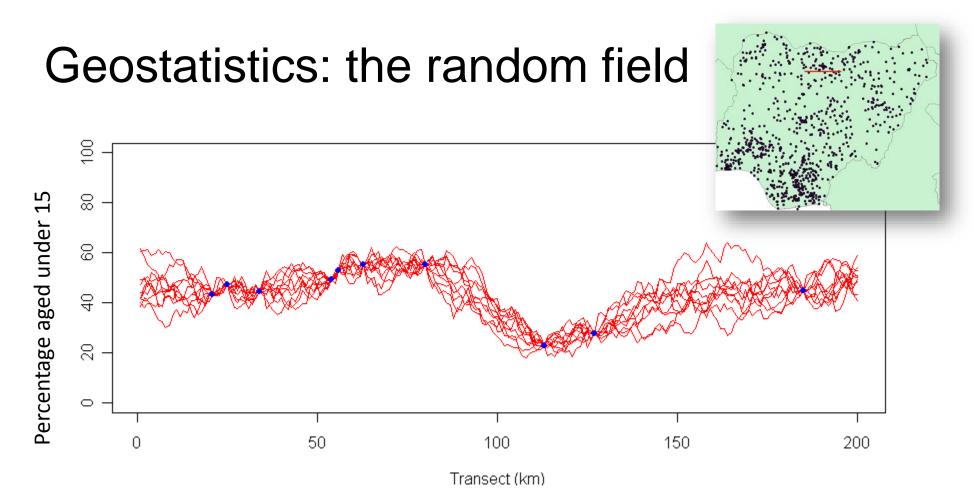






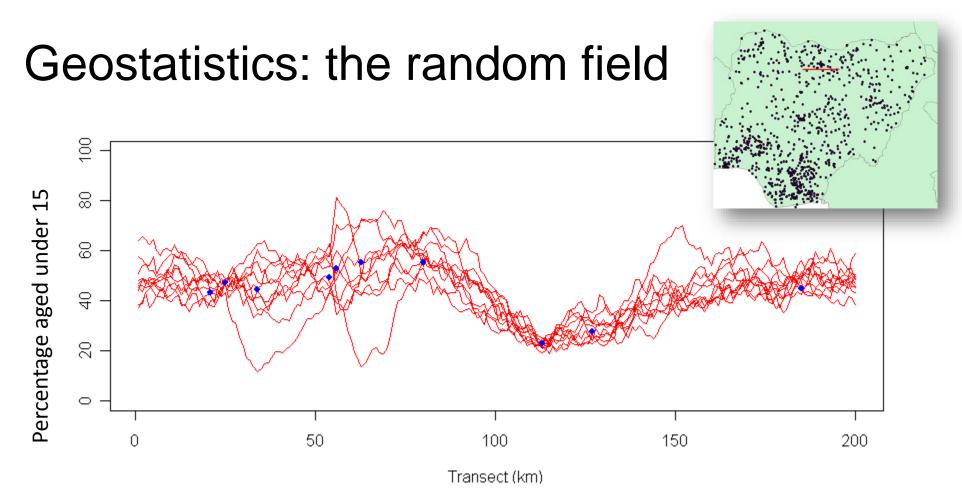
How do we use the data?

- 1. Parameterise mean and covariance function
- 2. 'Condition' on data values to predict everywhere



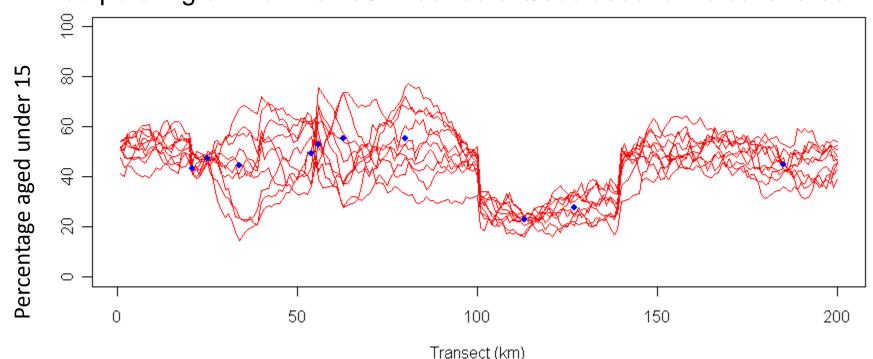
Incorporating prediction uncertainty

- Represent using multiple realisations
- Uncertainty increases away from data
- Determined by amplitude / range



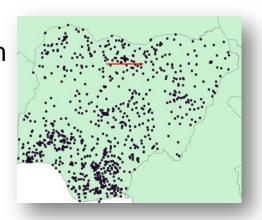
Incorporating sampling variation

- Data are sampled with error (e.g. function of sample size)
- Adopt sampling model (e.g. binomial)
- Field is better defined where data are more certain

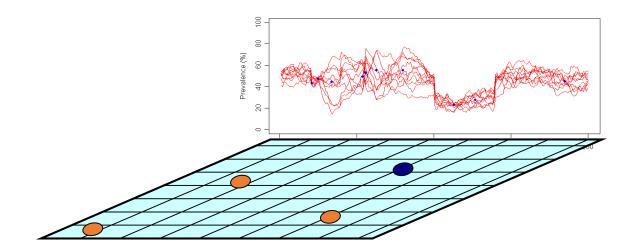


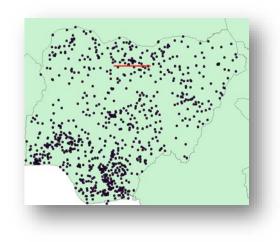
2. Incorporating environmental/infrastructure/socioeconomic covariates

- E.g. Effect of urban vs rural areas
- R/U/Other status of data and predictions assigned from satellite data analyses
- Allowed underlying mean to vary between classes
- Multiple covariates selected in multivariate model

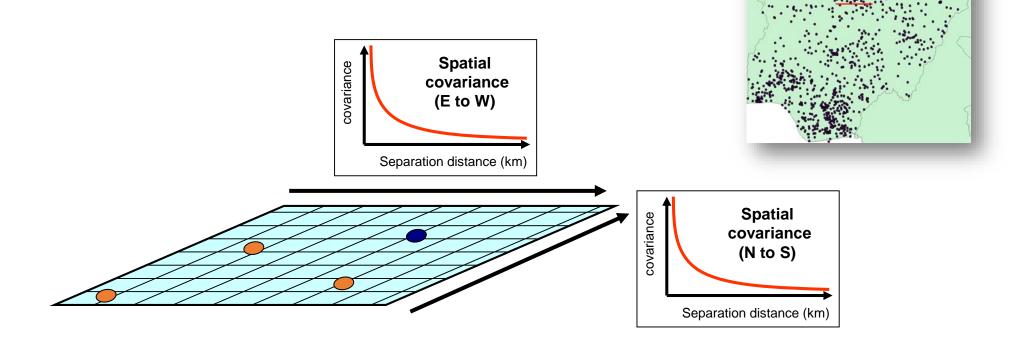


3. Incorporating spatial anisotropy

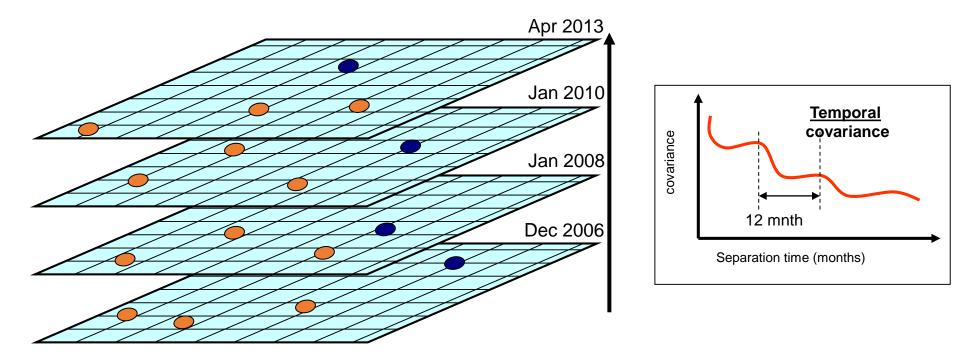






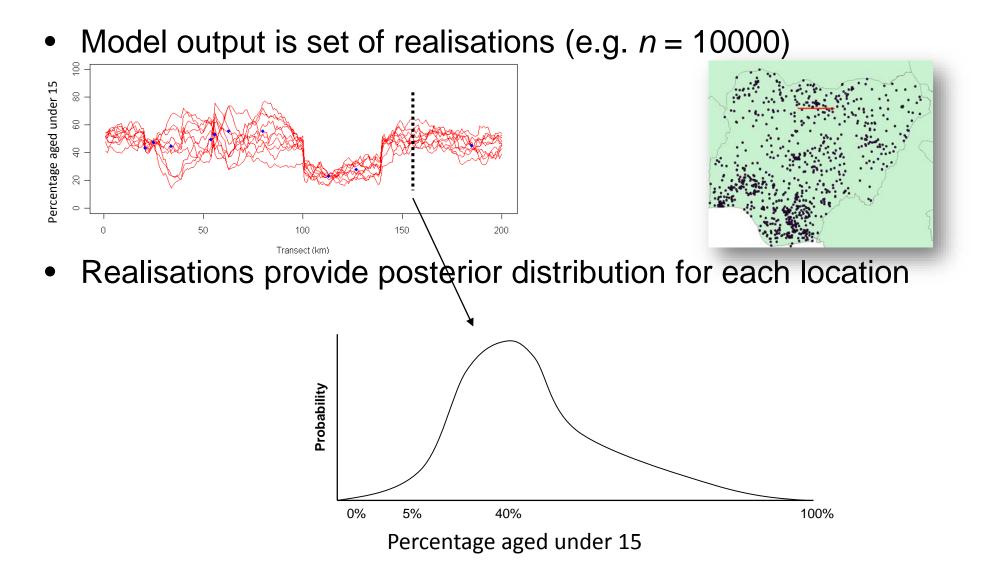


4. Full space-time dimensionality



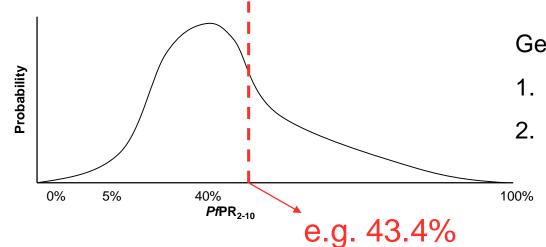
- 2d spatiotemporal covariance function: temporal axis
- Sinusoidal seasonal component...? (see mobility mapping later)

Geostatistics: making maps



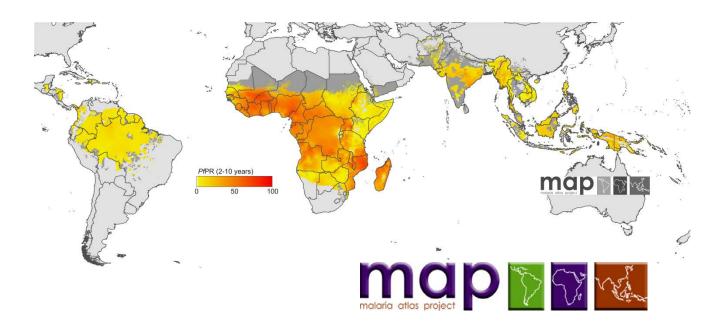
Geostatistics: making maps

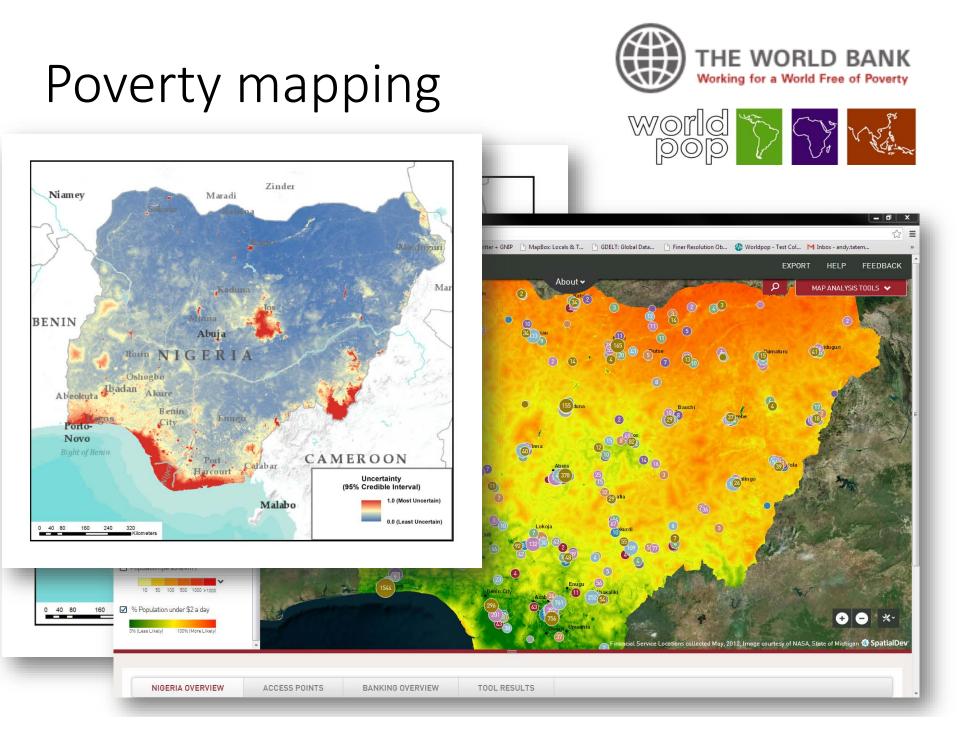
Generating maps from posterior distributions of PfPR₂₋₁₀



Generated various summaries:

- 1. A 'point estimate' (mean)
- 2. A summary of uncertainty (e.g. St.deviation)





Modeling Approach Overview

