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New methods for estimating internal  
migration from Call Detail Records in  
low- and middle-income countries

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# Outline

- Detecting internal migrations from CDRs
- Bias adjustment and scaling to the total population
- Producing monthly internal migration estimates for 3 countries



Detecting an  
internal migration  
from CDRs

# What is migration?

By migration, we understand a **change of home location** by a resident for **at least one month**

The spatial resolution of a home location is the sub-regional level, usually **administrative level 2 or 3** (depending on the country)

# Challenges to detect internal migrations in low- and middle-income countries (LMICs)



**Traditionally**, internal migration or residential mobility has been studied using **surveys** or **census**



However, in LMICs, census or survey data are often **outdated** or **unavailable**

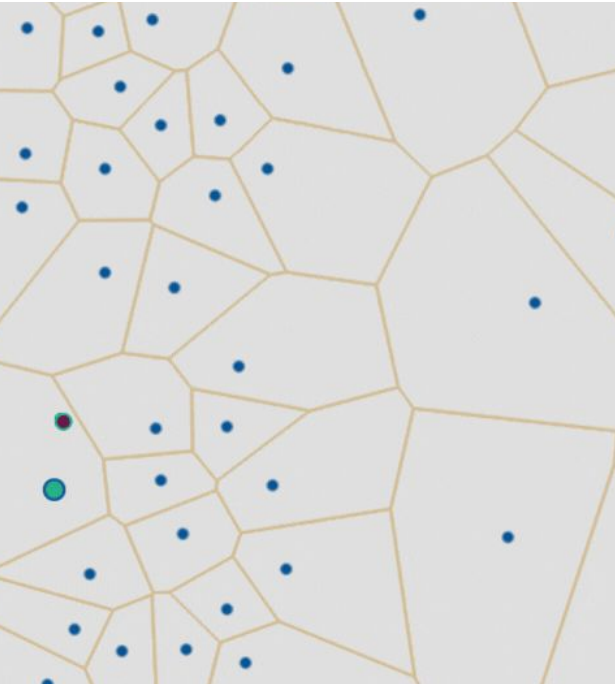


Conducting surveys in such countries can also be very challenging due to **inaccessibility** or **insecurity**

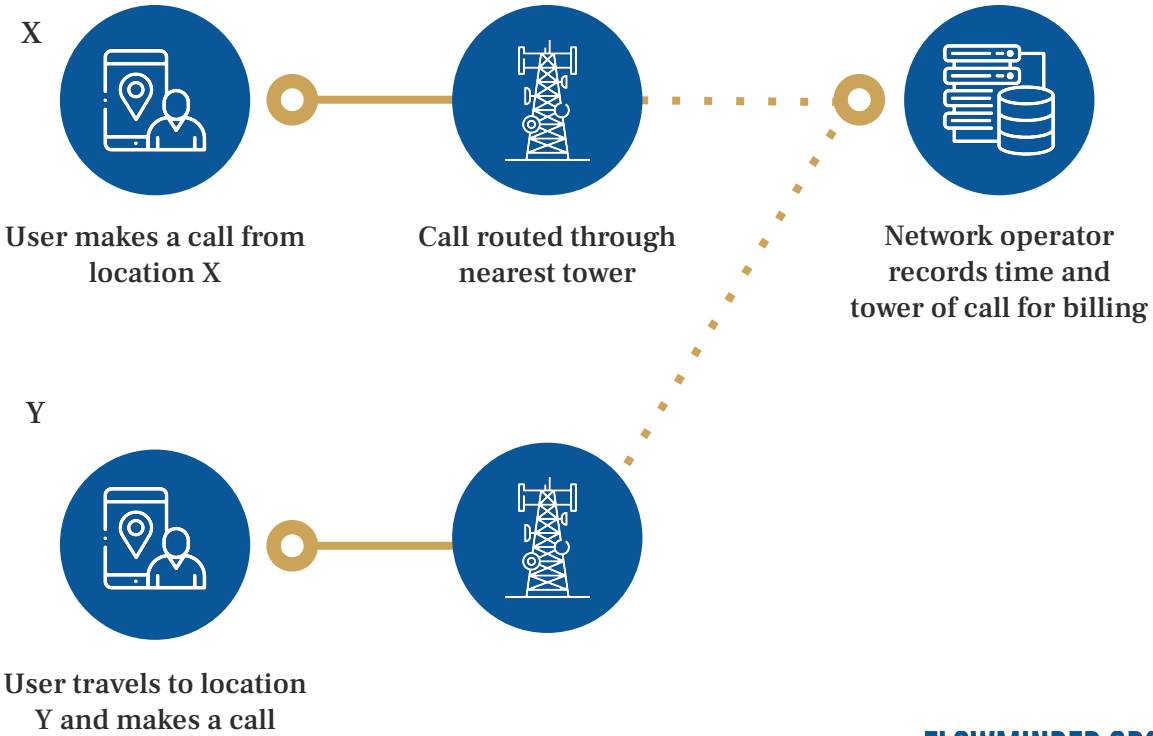


How can we estimate **residents'**  
**mobility** in such cases?

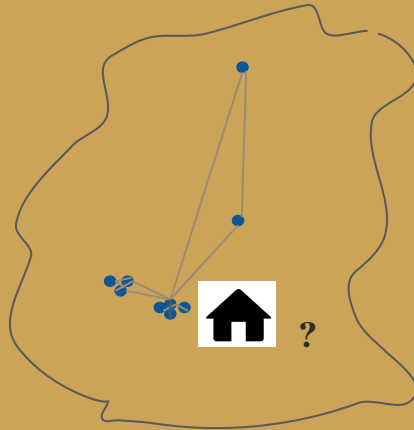
# CDRs as alternative data source for near-real time estimates of population movements & changes in population density



Subscriber movements (blue dots), Network events (red dot), Observed trajectory (green arrow), Cell tower coverage (yellow polygons), Cell towers (blue circles)

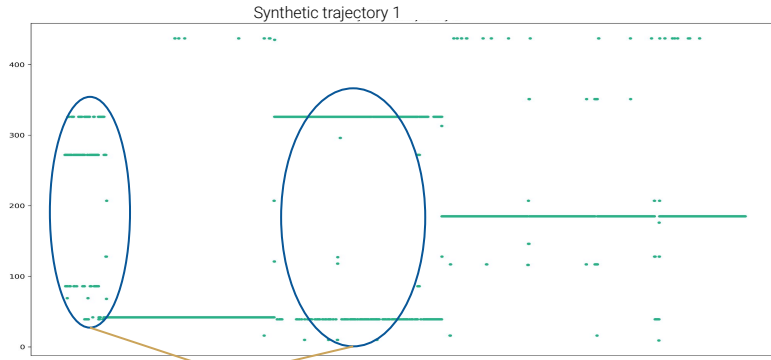


To detect **migrations** we need to  
detect **home locations**

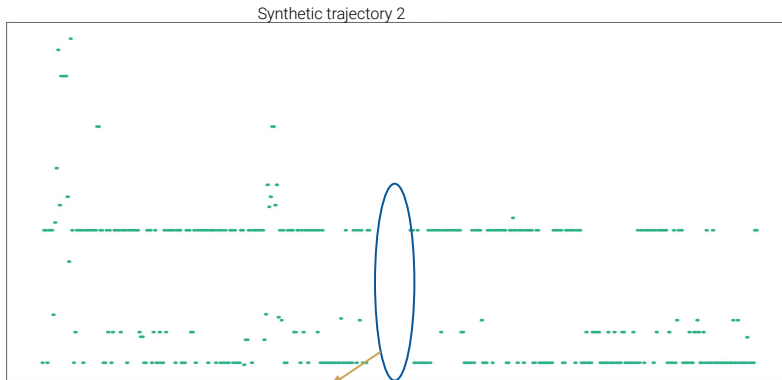




# Challenges in detecting home location using CDRs



Several potential home locations



Inactive period

Irregular calls

Infrequent calls

Short stays (1-3 days) overnight

Medium stays (4-7 days) overnight

Two or more similar frequency locations

Changing phone usage patterns

Ping-pong effect

Re-routing by a mobile operator

# How do we know where people live? (1)

## Monthly home location detection method

The **most frequent recent last-call-of-day location** in each 7 days window, moving by 1 day everyday



Daily

Assign **monthly home locations** for the current month



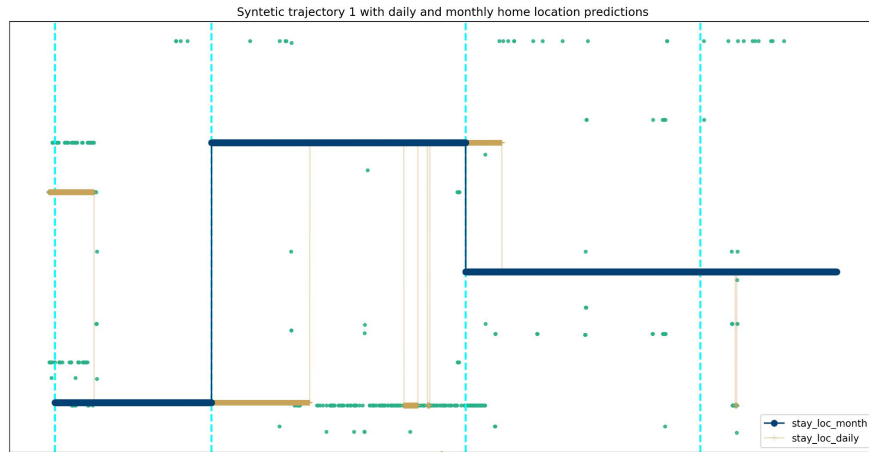
Monthly

- If **the same location** is an **absolute majority of daily home locations** (more than half) in the current month → assign it as home location
- Else if **the same location** is **more than third of daily home locations** in the current month and **majority** in the previous month → assign it as home location
- Otherwise - assign '**unlocatable**' for the current month

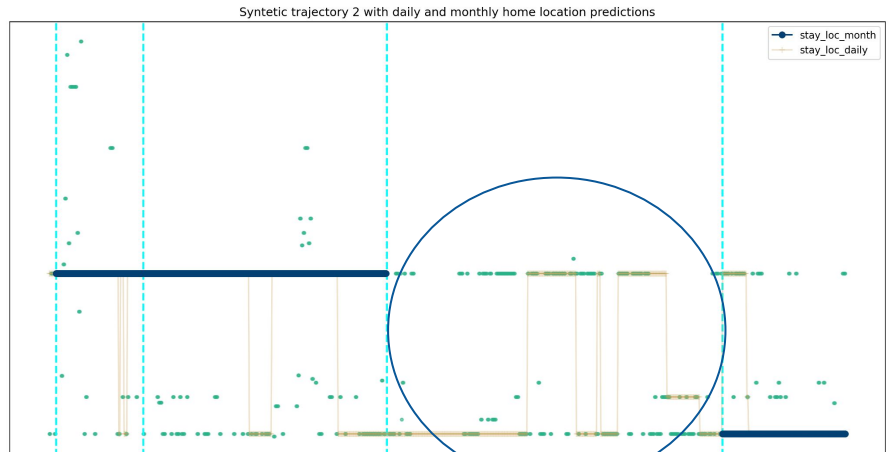
Home locations can be assigned to **all subscribers** or only to a **set of active subscribers**.

# How do we know where people live? (2)

## Monthly home location detection method: examples

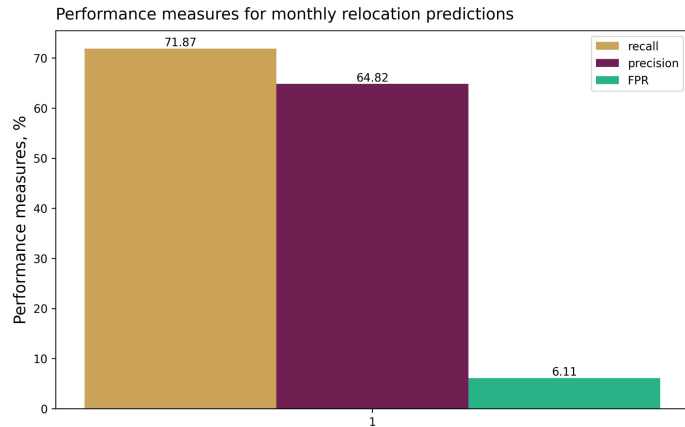
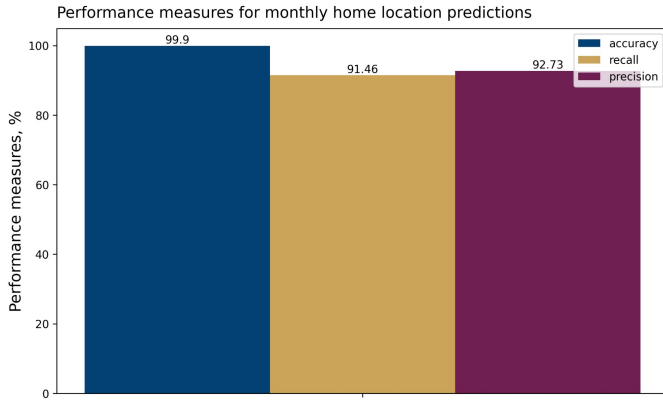


We can detect monthly home locations and months when relocations took place



Subscriber cannot be assigned monthly home location

# Validation of the monthly home location method



- **Validation** was done on **manually labelled 781 Digicel subscribers** in Haiti
- All algorithms were run on Haiti server for privacy protection.
- Performance measures are accuracy, recall, precision and false positive rate (FPR)
- **Monthly home location detection**
  - Accuracy, precision and recall are above 90%
  - FPR is 0.06%
- **Monthly home relocation detection**
  - Both the month of relocation and the location after relocation is taken into consideration.
  - Recall is ~70%, precision is ~65%, and FPR is ~6%

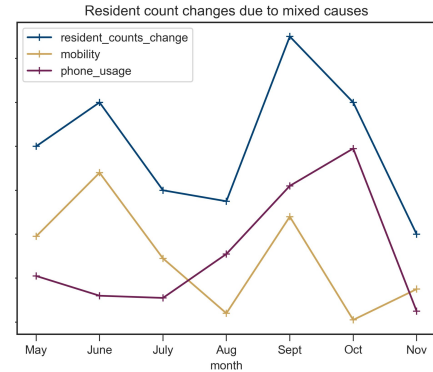
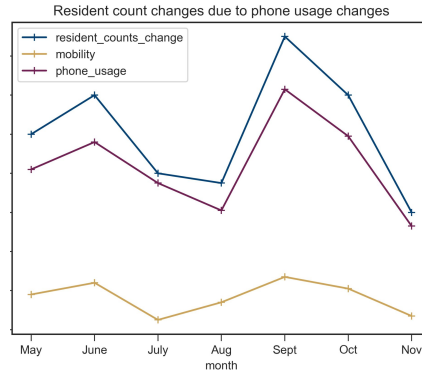
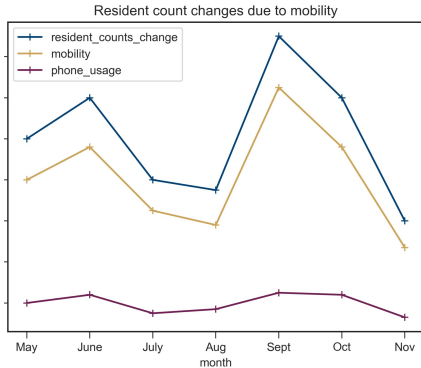
# Variations in resident counts as a measure of mobility

Reasons for variations in the resident counts between two consecutive months

Are changes in resident counts influenced mostly by **mobility** or by changes in **phone usage**?

Changes in mobility

Changes in Phone usage



# Estimating residents from net flows, subscribers

$$\text{est\_residents\_subscribers}_{an} = \text{est\_residents\_subscribers}_{a(n-1)} + \text{est\_netflow}_{an}$$

Where

- **est\_residents\_subscribers<sub>an</sub>** is the estimated number of resident subscribers in area a and month n
- **n=0** is a month corresponding to a baseline month (or a baseline period)
- **est\_netflows<sub>an</sub>** is the estimated netflow (difference between est\_inflow and est\_outflow) to area a in month n



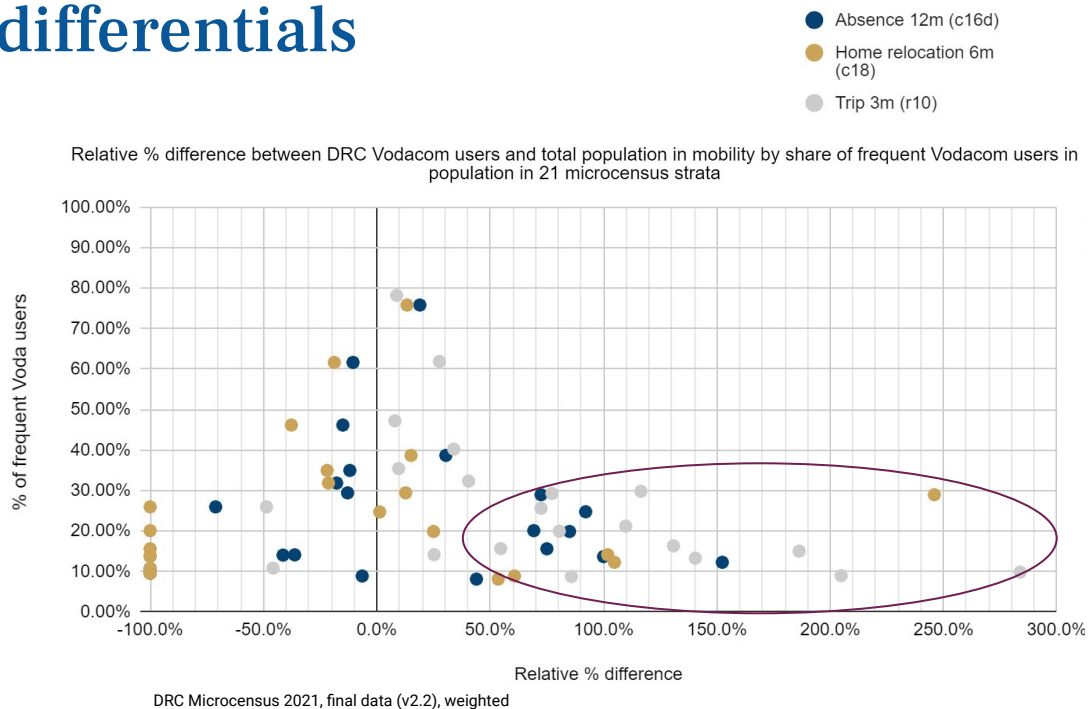
**Bias-adjustment and  
scaling of mobility  
estimates from CDR  
aggregates**

MNO subscribers are  
not a **random sample** of the  
population, nor can be assumed to be.




# Biases due to mobility differentials

- Based on survey data from the 2021 microcensus in the DRC, using three different mobility indicators, we identified some **large differences in mobility** between **Vodacom users** and the **rest of the population** (incl. non-phone-users)



- Across the three indicators and 21 microcensus strata, **15 parameters (out of 63) differed significantly** (i.e. more than the expected 5%)



Flowminder has recently developed estimation methods to arrive at **bias-adjusted & population-scaled** estimates for

- **Relocations** from sub-region to sub-region, per month
- **Residents** per sub-region, per month

## Bias-adjusted and population-scaled estimates

These estimates are based on

- CDR aggregates
- Primary & secondary survey data
- Existing population estimates
- Sub-region shapefiles

# Method for monthly residents' estimates

- The estimate of residents in area a for month n ( $est\_residents_{an}$ ) is calculated as the sum of the baseline population for that area ( $est\_base\_pop_a$ ) and by iteratively adding the cumulative sum of all net arrivals ( $est\_netflow_{amn}$ ) for all months between the baseline month and the current month, and by applying an area-specific rate of natural population growth ( $growthrate_a$ ) to each monthly sum:

$$est\_residents_{a1} = est\_base\_pop_a \quad (\text{Month 1 (baseline), } m=0, n=1)$$

$$est\_residents_{a2} = (est\_residents_{a1} + est\_netflow_{a12}) * growthrate_a \quad (\text{Month 2, } m=1, n=2)$$

$$est\_residents_{a3} = (est\_residents_{a2} + est\_netflow_{a23}) * growthrate_a \quad (\text{Month 3, } m=2, n=3)$$

$$\dots = \dots$$

$$est\_residents_{an} = (est\_residents_{am} + est\_netflow_{amn}) * growthrate_a$$

- where the net arrivals estimate for area a between months m and n is the sum of all estimated inflows to that area minus all estimated outflows from that area:

$$est\_netflow_{amn} = est\_inflow_{amn} - est\_outflow_{amn}$$

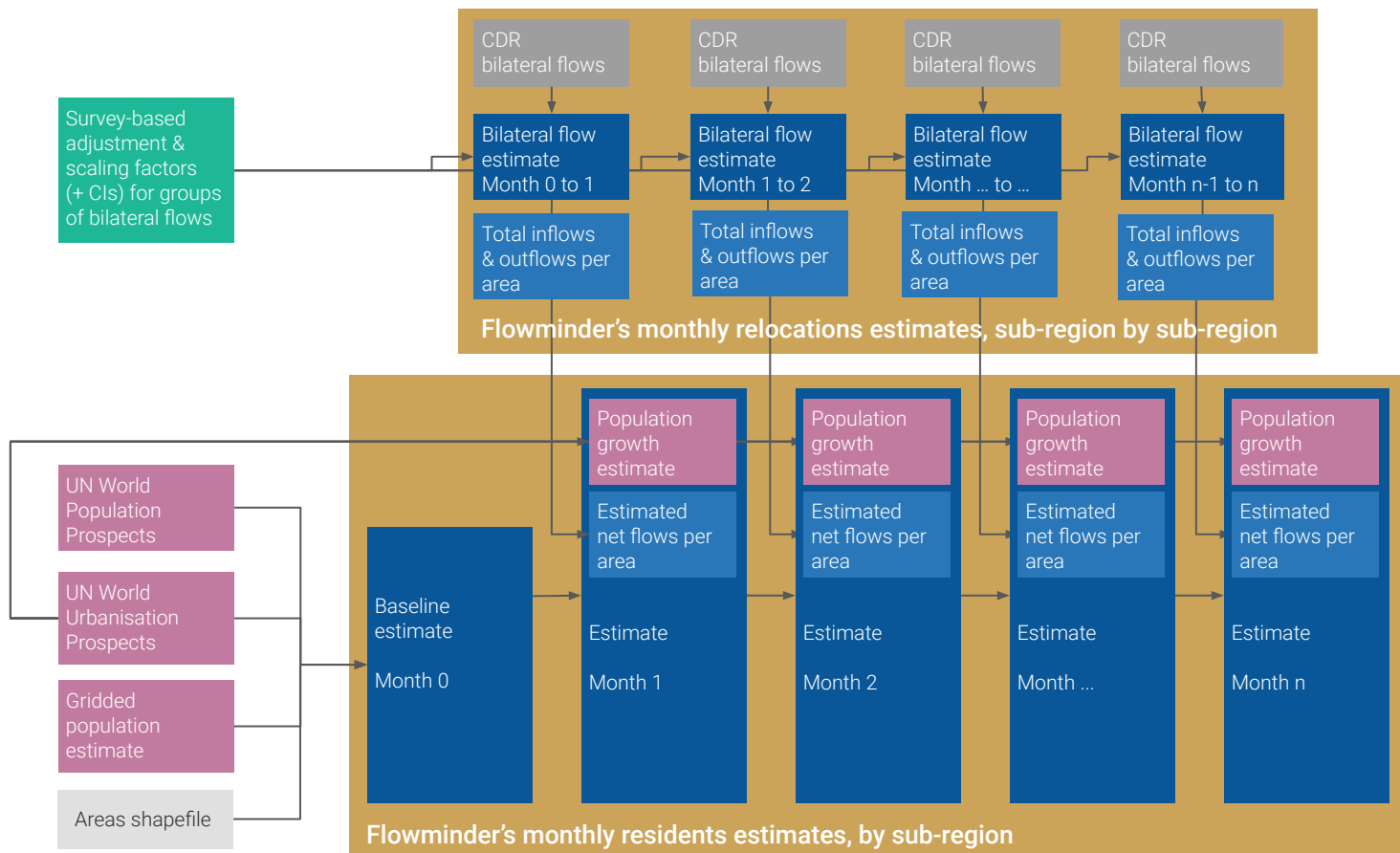
# Method for relocations' estimates

- Relocations from area a to area b between month m and month n can be estimated from CDR aggregates of relocations ( $\text{cdr\_flow}_{abmn}$ ) between those areas and months, and from a flow adjustment factor and a flow scaling factor.
- Flows are adjusted for the number of users per SIM ( $\text{users}_{ab}$ ) and the number of SIMs per user ( $\text{sims}_{ab}$ ). The flow scaling factor is the inverse of the share of MNO users ( $\text{mno\_share}_{ab}$ ) in the flows:

$$\text{est\_flow}_{abmn} = \text{cdr\_flow}_{abmn} * (\text{users}_{ab} / \text{sims}_{ab}) * (1 / \text{mno\_share}_{ab})$$

- Note: Parameters for the subset of mobile households/individuals only available at admin1 by admin1 level

# Method for monthly residents' estimates



# General caveats



Change estimates strongly depend on **baseline population number** - differs greatly between data sources



Currently only **cross-sectional survey data** used (longitudinal data needed)



Limits to **granularity of survey estimates** - admin3 by admin3 would require very large survey sample sizes or even census data



Lack of **validation data**: mobility and population estimates at admin3 are rare for LMICs

We continue to  
seek more data  
sources &  
develop new  
methodologies  
and methods.

## Next steps

- Method refinement of **home relocation detection**
  - detection of relocations on daily/weekly basis
  - detection of short and medium stays
- Use **census data** (where available) for estimation models, or for **validation**.
- Test further **estimation models** (e.g. Machine Learning, Small Area Estimation, extrapolation)

Adjustment and scaling of CDR time series aggregates are not trivial, but ultimately require **highly complex estimation models** based on multiple longitudinal & cross-sectional data sources

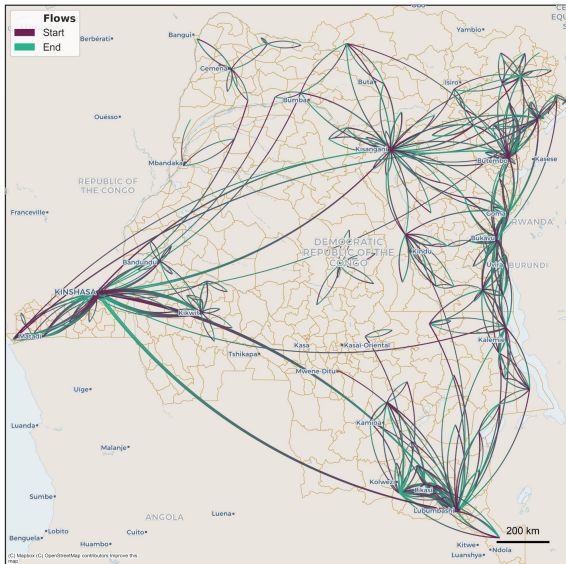


Producing monthly  
internal migration  
estimates  
for 3 countries



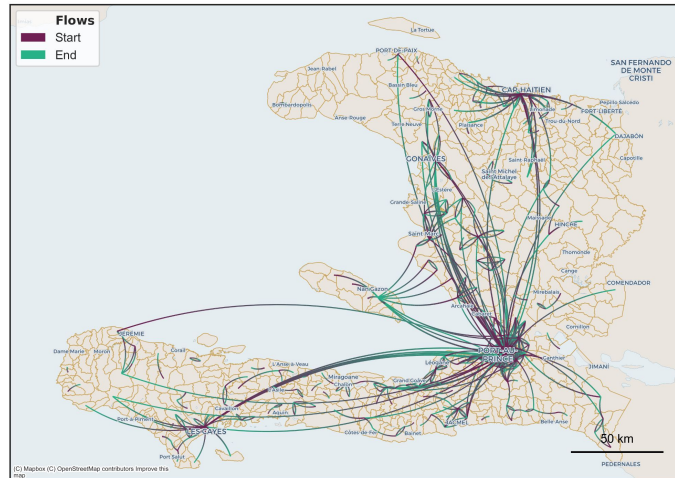
# Monthly relocations between sub-regions

## DRC



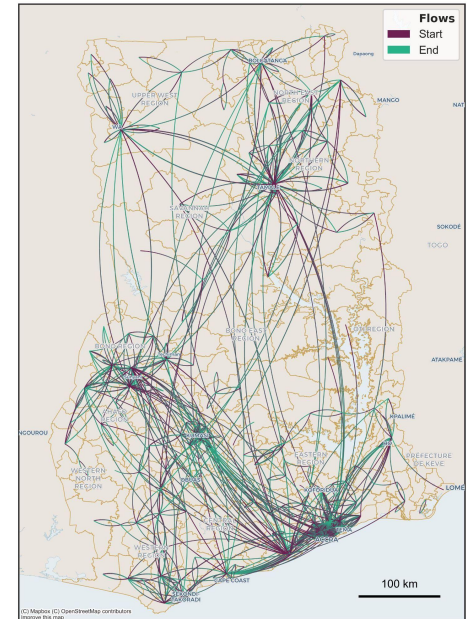
Note: estimated top 1,000 flows between health zones, median, Nov 2021 - Dec 2022

## Haiti



Note: top 500 flows between communal sections, median, Feb 2020 - Feb 2022

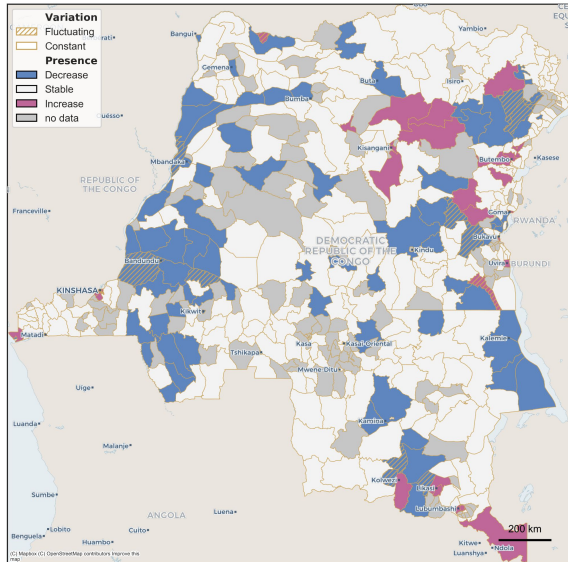
## Ghana



Note: unscaled top 1,000 flows between districts, median, Jan - July 2021

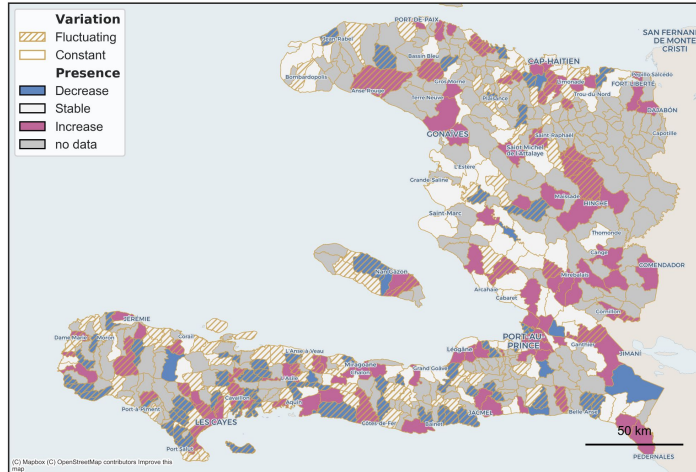
# Monthly population change per sub-region

## DRC



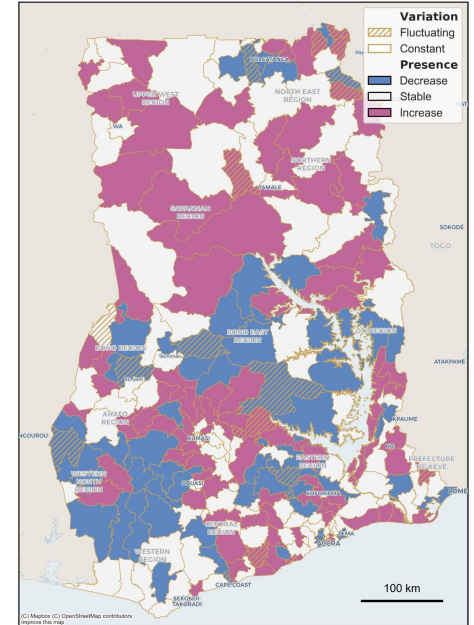
Note: average monthly variation in residents by health zone, Nov 2021 - Dec 2022, relative to baseline residents in each area

## Haiti



Note: average monthly variations in residents by communal section, Feb 2020 - Feb 2022, relative to baseline residents in each area

## Ghana



Note: average monthly variations in residents by district, Jan - July 2021, relative to baseline residents in each area



Thank you!

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