

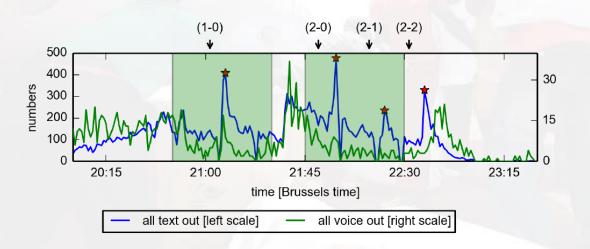




Richness of the data: dynamics

Text messages and phone calls during the international soccer match Belgium-Ivory

Coast (March 5, 2014).



Forecasting event attendance with anonymized mobile phone data C. Cloquet¹, V. D. Blondel¹,





Census



Dynamic population mapping using mobile phone data

Pierre Deville^{a,b,c,1}, Catherine Linard^{c,d,1,2}, Samuel Martin^e, Marius Gilbert^{c,d}, Forrest R. Stevens^f, Andrea E. Gaughan^f, Vincent D. Blondel^a, and Andrew J. Tatem^{g,h,i}

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Edited by Michael F. Goodchild, University of California, Santa Barbara, CA, and approved September 15, 2014 (received for review May 8, 2014)

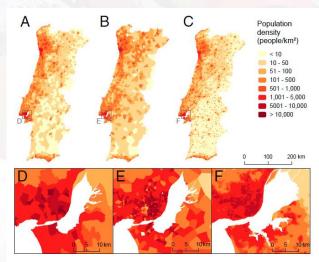


Fig. 1. Comparison of predicted population density datasets with baseline data for mainland Portugal. (A) Population density as calculated from the national census at administrative unit level 5 (ADM-5; freguesia). (B) Population density at the level of Voronoi polygons, as estimated by the MP method. (C) Population density at the level of 100×100 -m grid squares, as estimated by the RS method. (D-F) Close-ups around the capital city Lisbon.





Poverty indices

Estimating Food Consumption and Poverty indices with Mobile Phone Data

Adeline Decuyper¹, Alex Rutherford², Amit Wadhwa³ Jean Martin Bauer³ Gautier Krings^{1,4} Thoralf Gutierrez⁴ Vincent D. Blondel¹ Miguel A. Luengo-Oroz^{2,*}

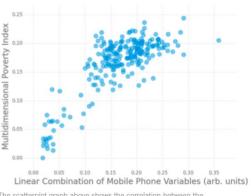


USING MOBILE PHONE DATA AND AIRTIME CREDIT PURCHASES TO ESTIMATE FOOD SECURITY

PARTNERS: UN WORLD FOOD PROGRAMME, UNIVERSITÉ CATHOLIQUE DE LOUVAIN, REAL IMPACT ANALYTICS PROGRAMME AREA: FOOD SECURITY & AGRICULTURE



FOOD ITEM (VARIABLE)	CORRELATION RANGE
Vitamin-rich vegetables (carrot, orange, sweet potato), rice, wheat, bread, sugar, meat	[0.7–0.8]
Eggs, oil, milk, butter, organ meat	[0.5–0.6]
Sorghum, ground nuts, seeds, fish, fruits, cooking banana, green leafy vegetables, beans, peas, maize, white roots, tubers, pumpkin, squash, cassava	[0.0–0.4]
White sweet potato	-0.4



The scatterplot graph above shows the correlation between the Multidimensional Poverty Index and a linear combination of mobile phone variables (including airtime credit expenditures).





Malaria

Tatem et al. Malaria Journal 2014, 13:52 http://www.malariajournal.com/content/13/1/52



RESEARCH

Open Access

Integrating rapid risk mapping and mobile phone call record data for strategic malaria elimination planning

Andrew J Tatem^{1,2*}, Zhuojie Huang^{3,4}, Clothilde Narib⁵, Udayan Kumar^{4,6}, Deepika Kandula⁷, Deepa K Pindolia^{3,4}, David L Smith^{2,8}, Justin M Cohen⁷, Bonita Graupe⁹, Petrina Uusiku⁵ and Christopher Lourenço^{5,7}

One year of aggregated movement patterns for over a million people across Namibia are analyzed, and linked with case-based risk maps built on satellite imagery.

These maps can aid the design of targeted interventions to maximally reduce the number of cases exported to other regions while employing appropriate interventions to manage risk in places that import them



Quantifying the Impact of Human Science. **Mobility on Malaria**



Amy Wesolowski, 1,2 Nathan Eagle, 3,4 Andrew J. Tatem, 5,6,7 David L. Smith, 6,8 Abdisalan M. Noor, 9,10 Robert W. Snow, 9,10 Caroline O. Buckee 4,11 *

Human movements contribute to the transmission of malaria on spatial scales that exceed the limits of mosquito dispersal. Identifying the sources and sinks of imported infections due to human travel and locating high-risk sites of parasite importation could greatly improve malaria control programs. Here, we use spatially explicit mobile phone data and malaria prevalence information from Kenya to identify the dynamics of human carriers that drive parasite importation between regions. Our analysis identifies importation routes that contribute to malaria epidemiology on regional spatial scales.



Fig. 2. Travel networks of people and parasites between settlements and regions. (A) Average monthly travel between regions (nodes), with edges weighted by volume of traffic. For clarity, the top 50% of routes are shown with arrows indicating the direction of movement (humans or parasites) from a

portation by returning residents, by region. (C) Average monthly parasite impor tation by visitors, where importation is only considered if the destination is receptive to onward transmission (14). Nodes are colored and labeled as described in Fig. 1.

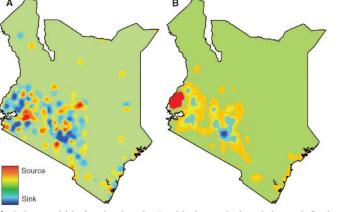


Fig. 3. Sources and sinks of people and parasites. Kernel density maps showing ranked sources (red) and sinks (blue) of human travel and total parasite movement in Kenya, where each settlement was designated as a relative source or sink based on yearly estimates. (A) Travel sources and sinks. (B) Parasite sources and sinks.

Cited By...

The ethics of big data as a public good: which public? Whose good?

Perspectives on the role of mobility, behavior, and time scales in the spread of diseases

Mass Drug Administration for Malaria: A Means to What end?

Connecting Mobility to Infectious Diseases: The Promise and Limits of Mobile Phone Data

Leveraging contact network structure in the design of cluster randomized

Assessing reliable human mobility patterns from higher order memory in mobile communications

Calling in sick: impacts of fever on intra-urban human mobility

Mobile phone data highlights the role of mass gatherings in the spreading of cholera outbreaks

Differential and enhanced response to climate forcing in diarrheal disease due to rotavirus across a megacity of the developing world

Malaria Parasitemia Among Febrile Patients Seeking Clinical Care at an Outpatient Health Facility in an Urban Informal Settlement Area in Nairobi, Kenya

Virus evolution and transmission in an ever more connected world

Impact of human mobility on the emergence of dengue epidemics in Pakistan

Quantifying seasonal population fluxes driving rubella transmission dynamics using mobile phone data

Malaria genotyping for epidemiologic surveillance

Modeling infectious disease dynamics in the complex landscape of global

Advancing digital methods in the fight against communicable diseases

Modeling internal migration flows in sub-Saharan Africa using census

Unique in the shopping mall: On the reidentifiability of credit card





Ebola

MIT Technology Review

Topics+

Top S



Biomedicine

Cell-Phone Data Might Help Predict Ebola's Spread

Mobility data from an African mobile-phone carrier could help researchers recommend where to focus health-care efforts.

by David Talbot August 22, 2014





Other infectious

The Journal of Infectious Diseases

(2016)

Connecting Mobility to Infectious Diseases: The Promise and Limits of Mobile Phone Data

Amy Wesolowski^{1,2}, Caroline O. Buckee^{1,2}, Kenth Engø-Monsen³ and C. J. E. Metcalf^{4,5}

Big Data for Infectious Disease Surveillance and Modeling

Shweta Bansal^{1,2}, Gerardo Chowell^{1,3}, Lone Simonsen^{1,4}, Alessandro Vespignani⁵ and Cécile Viboud¹

Quantifying seasonal population fluxes driving rubella transmission dynamics using mobile phone data

Amy Wesolowski^{a,b,c,1,2}, C. J. E. Metcalf^{d,e,f,1,2}, Nathan Eagle^{a,g}, Janeth Kombich^h, Bryan T. Grenfell^{d,e}, Ottar N. Biørnstadⁱ, Justin Lessler^j, Andrew J. Tatem^{c,f,k}, and Caroline O. Buckee^{a,b,c}

*Department of Epidemiology, Harvard School of Public Health, Boston, MA (2011s; *Center for Communicable Disease Dynamics, Harvard School of Public Health, Boston, MA (2011s; *Flowminder Foundation, 5E-113 55 Stockholm, Sweden; *Department of Ecology and Evolutionary Biology, Princeton University, Princeton, NJ 08544; *Fogarty International Center, National Institute of Health, Bethesda, MD 20892; *Department of Computer Science, Northeastern University, Boston, MA 0211s; *Department of Biological Sciences, University of Kabianga, Kericho Country, Kenya; *Center for Infectious Disease Dynamics, The Pennsylvania State University, State College, PA 18801; *Department of Epidemiology, Johns Hopkins Bloomberg School

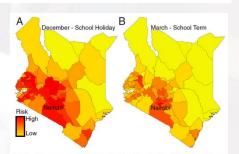


Fig. 3. The seasonal variability in the risk of importation. We analyzed the average amount of population flux per district during (A) the major holiday and a school term break (December) and (B) during a school term (March). As highlighted in Fig. 1, there are large amounts of population flux and consequently the sk of importation during school breaks (A and B). In contrast, there is a decrease in the risk of importation during the school term. During the course of the year, the districts with the largest risks vary with higher risks to western Kenya during school breaks. However, Nairobi (shown in red in both maps) consistently remains at a high risk of importation from the large oppulation flux of importation from the large oppulation flux.





Other infectious

The Journal of Infectious Diseases

SUPPLEMENT ARTICLE







Big Data for Infectious Disease Surveillance and Modeling

Shweta Bansal,^{1,2} Gerardo Chowell,^{1,3} Lone Simonsen,^{1,4} Alessandro Vespignani,⁵ and Cécile Viboud¹

¹Fogarty International Center, National Institutes of Health, Bethesda, Maryland; ²Department of Biology, Georgetown University, Washington D.C.; ³School of Public Health, Georgia State University, Atlanta; ⁴Department of Public Health, University of Copenhagen, Denmark; and ⁵Network Science Institute, Northeastern University, Boston, Massachussets

We devote a special issue of the *Journal of Infectious Diseases* to review the recent advances of big data in strengthening disease surveillance, monitoring medical adverse events, informing transmission models, and tracking patient sentiments and mobility. We consider a broad definition of *big data* for public health, one encompassing patient information gathered from high-volume electronic health records and participatory surveillance systems, as well as mining of digital traces such as social media, Internet searches, and cell-phone logs. We introduce nine independent contributions to this special issue and highlight several cross-cutting areas that require further research, including representativeness, biases, volatility, and validation, and the need for robust statistical and hypotheses-driven analyses. Overall, we are optimistic that the big-data revolution will vastly improve the granularity and timeliness of available epidemiological information, with hybrid systems augmenting rather than supplanting traditional surveillance systems, and better prospects for accurate infectious diseases models and forecasts.

Big Data for Infectious Disease Surveillance and Modeling • JID 2016:214 (Suppl 4) • S375





Natural disaters: Haïti

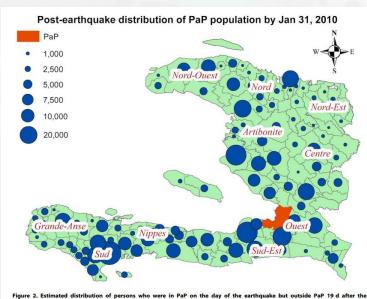
OPEN & ACCESS Freely available online

PLOS MEDICINE

Improved Response to Disasters and Outbreaks by Tracking Population Movements with Mobile Phone Network Data: A Post-Earthquake Geospatial Study in Haiti

Linus Bengtsson¹*, Xin Lu^{1,2}, Anna Thorson¹, Richard Garfield³, Johan von Schreeb¹

1 Department of Public Health Sciences, Karolinska Institutet, Stockholm, Sweden, 2 Department of Sociology, Stockholm University, Stockholm, Sweden, 3 Schools of Nursing and Public Health, Columbia University, New York City, United States of America



earthquake. Circles are shown for communes that received at least 500 persons.

doi:10.1371/journal.pmed.1001083.g002

but outside PaP 19 d after the

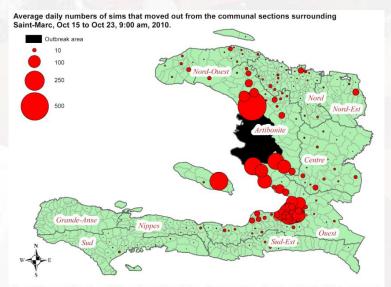


Figure 5. Average daily numbers of SIMs moving out of the cholera outbreak area. October 15 to October 23, 2010, divided per communal section of destination. The data were disseminated to relief agencies at the outset of the outbreak (October 24, 2010). doi:10.1371/iorunal.nem4.100183.005





Natural disaters



USING MOBILE PHONE ACTIVITY FOR DISASTER MANAGEMENT DURING FLOODS

PARTNERS: GOV. OF MEXICO, UN WORLD FOOD PROGRAMME, TELEFONICA RESEARCH, UNIVERSIDAD POLITECNICA DE MADRID PROGRAMME AREA: HUMANITARIAN ACTION



SUMMARY

Natural disasters affect hundreds of millions of people worldwide every year. Emergency response efforts depend on the availability of timely information, such the movement and communication behaviours of affected populations. As such, analysis of Call Detail Records (CDRs) collected by mobile phone operators reveal new, real-time insights about human behaviour during such critical events. In this study, mobile phone activity data was combined with remote sensing data to understand how people communicated during severe flooding in the Mexican state of Tabasco in 2009, in order to explore ways that mobile data can be used to improve disaster response. By comparing the mobile data with official population census data, the representativeness of the research was validated. The results of the study showed that the patterns of mobile phone activity in affected locations during and after the floods could be used as indicators of (1) flooding impact on infrastructure and population and (2) public awareness of the disaster. These early results demonstrated the value of a public-private partnership on using mobile data to accurately indicate flooding impacts in Tabasco, thus improving early warning and crisis management.

"The most calls were made from the most impacted areas"

"Flooding through the Lens of Mobile Phone Activity." Pastor-Escuredo, D., Morales-Guzmán, A. et al, IEEE Global Humanitarian Technology Conference, GHTC 2014.





Other

The emergent opportunity of Big Data for Social Good Dr. Nuria Oliver, Telefónica

Emergency Event Detection Using Mobile Phone Data Didem Gundogdu, University of Trento

oii.ox.ac.uk/events/symposium-on-big-data- and-human-development

Oxford Internet Institute: Big Data for Developement http://bigdatadevelopment.oii.ox.ac.uk





Metho: Big Data == proxies => biases, ...

On the use of human mobility proxies for modeling epidemics

Michele Tizzoni¹, Paolo Bajardi², Adeline Decuyper³, Guillaume Kon Kam King⁴, Christian M. Schneider⁵, Vincent Blondel³, Zbigniew Smoreda⁶, Marta C. González^{5,7}, Vittoria Colizza^{8,9,10,*}



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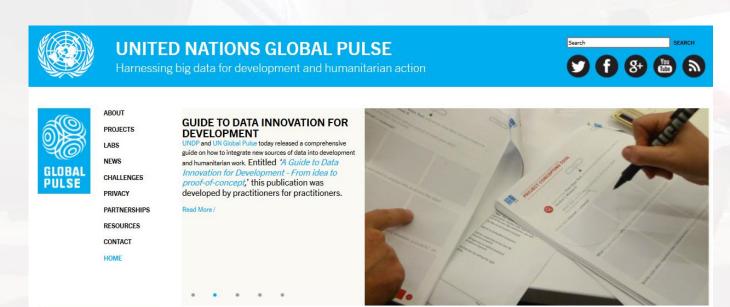
The impact of biases in mobile phone ownership on estimates of human mobility

Amy Wesolowski¹, Nathan Eagle^{2,3}, Abdisalan M. Noor^{4,5}, Robert W. Snow^{4,5} and Caroline O. Buckee^{3,6}





Global Pulse



unglobalpulse.org





Challenges D4D



First Prize and Energy Prize: Using mobile phone data for electrification planning

E.A. Martínez-Ceseña (1), P. Mancarella (1), M. Ndiaye (2), and M. Schläpfer (9)

Knowledge of local energy needs is crucial for the electricity infrastructure planning of a country. We have shown that mobile phone data are an accurate proxy of the energy needs and can be used to develop bottom-up demand models. The new methodology supports and prioritizes the electrification plans in areas with scarce information on local activities and energy consumption.

(1)University of Manchester, UK - (2) Ecole supérieure polytechnique de Dakar UCAD, Senegal - (3) Senta Fe Institute, USA



Agriculture Prize: Genesis of millet prices in Senegal: the role of production, markets and their failures

D.C. Jacques (1), R. d'Andrimont (1), J. Radoux (1), F. Waldner (1), and E. Marinho (2)

Information asymmetries are responsible for price differentials in only the few areas where the mobile phone coverage has not yet reached its full potential, which damages both poor producers and food insecure consumers. To address this issue, we have integrated it in a spatially explicit model that simulates the functioning of agricultural markets.

(1) Earth and Life Institute, Université Catholique de Louvain, Belgium - (2) Independent researcher, Rio de Janeiro, Brazil



Health Prize: Uncovering the impact of human mobility on schistosomiasis...

L. Mari (1), R. Casagrandi (1), M. Ciddio (1), S.H. Sokolow (2), G. De Leo (3), and M. Gatto (1)

Schistosomiasis is water based parasitic worm infection with debilitating symptoms affecting millions of people. We show that a relatively simple model can reliably reproduce regional patterns of schistosomiasis prevalence across the country. We use the model to study the role of human mobility on disease dynamics and to analyze intervention strategies aimed at reducing disease burden.

(1)Dipartimento di Elettronica, Informazione e Bioingegneria, Politecnico di Milano, Italy - (2) Hopkins Marine Station, Stanford University, USA



National Statistics Prize: Virtual Networks and Poverty Analysis in Senegal

N. Pokhriyal, W. Dong, and V. Govindaraju

Computer Science and Engineering, State University of New York at Buffato, USA

Poverty is a complex phenomenon, but can be approximated by observing mobile phone usages and mobility at regional is sivel and extrapolated at more granular level. Poverty maps showcasing multiple perspectives can provide policymakers with better insights for effective responses for poverty eradication.



Transport Prize: National and Regional Road Network Optimization for Senegal Using Mobile Phone Data

Y. Wang (1), G. Homem de Almeida Correia (1), and Erik de Romph (1,2)

Anonymous mobile phone traces can be fittered with an algorithm to generate a proxy for a trip origindistination matrix. This is used to develop a gravity model that predicts the future mobility in the country dependent on travel time and number of calls and messages between the departments. This information is then used to improve decision making for road network planning.

(1) Department of Transport and Planning, Delft University of Technology, The Netherlands - (2) DAT:mobility, The Netherlands



Data Crossing Prize: Using mobile phone data for Spatial Planning simulation and Optimization Technologies (SPOT)

S. Gueye (1), B.M. Ndiaye (6), D. Josselin (6), M. Poss (6), R.M. Faye (6), P. Michelon (1), C. Genre-Grandpierre (6), and F. Ciari (4)

We propose a methodology of location and relocation of amenities (home, shop, work, leisure places) for urban planning decision. Our methodology exploits mobile phone data and other veriables and point of interest on maps to propose optimal amenity locations to reduce the overall travel time or travel distance.

(1) LIA, Université d'Avignon, France - (2) LTI, ESP - Université de Cheikh Anta Diop, Senegal - (3) LMDAN, FASEG-Université de Cheikh Anta Diop, Senegal - (4) Institute for Transport Planning and Systems (IVT), Zurich, Switzerland - (5) UMR ESPACE, CNRS, Avignon, France



Data Visualization Prize: Data for Development Reloaded: Visual Matrix Techniques for the Exploration and Analysis of Massive Mobile Phone

S. van den Elzen, M. van Dortmont, J. Blaas, D. Holten, W. van Hage, J-K. Buenen, J.J. van Wijk, R. Spousta *, S. Sala *, S. Chan *, A. Kuzmickas * University of Technology SynerScope BV Sensemaking Fellowshio

Eindhoven University of Technology & SynerScope BV, The Netherlands

* Sensemaking Fellowship (MIT, Harvard University)

In our Visual analytics techniques for the exploration and analysis of massive mobile phone data, users are enabled to identify both temporal and structural patterns such as normal behavior, outliers, anomalies, periodicity, trends and counter-trends.



Practical Application Prize: Mobile Data as Public-Health Decision Enabler: A Case Study of Cardiac and Neurological Emergencies E. Mutafungwa M, F. Thiessard M, M. Pathé Diallo M, R. Gore M, V. Jouhet M, C. Karray M,

N. Kheder (4), R. Saddem (4), J. Hämäläinen (1), G. Diallo (1)

The objective of the study is to show the areas in which the absence of a nearest hospital can result in death or serious aqueals. The identification of areas at high risk in case of stroks of myocardial inferction, requiring rapid intervention, could help Public Health decision makers to priorize investments.

(f) Department of Communications and Networks, Aatho University School of Electrical Engineering, Finland - (2) EFIAS INSERM U697, ISPED. Université de Bordeaux, France - (3) Vriginia Modeling Analysis and Simulation, Old Dominion University, USA - (4) Faculté des Sciences de Tunis, University of Univ. Tunisla



Scientific Prize and Ethics Mention: Construction of socio-demographic indicators with digital breadcrumbs

F. Bruckschen (1), T. Schmid (2), T. Zbiranski (1)

We show that socio-demographic indicators such as population, age, literacy, poverty, religion, ethnicity, electricity supply and others can be estimated in unprecedented detail and virtually ad-hoc using antennato antenna traffic data only. We offer a uniform approach that can be easily extended to other variables. Results are tested for spatio-temporal robustness and visualized as heat maps.

(1) Humboldt Universität Berlin, Germany - (2) Freie Universität Berlin, Germany

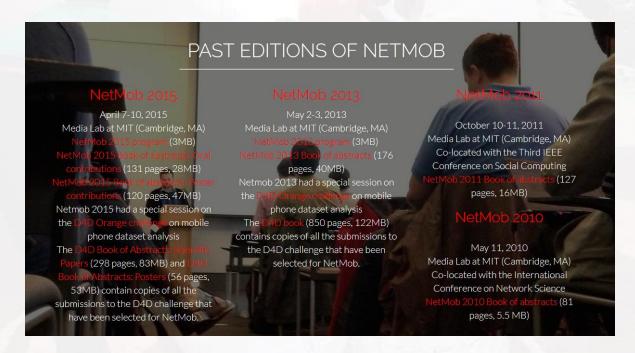




NetMob



The main conference on the scientific analysis of mobile phone datasets
5-7 April 2017 Vodafone Theatre, Milan, Italy





Non-profits / private companies







Accélérer des ONGs par l'innovation technologique et le big data



dataforgood.fr



data4sdgs.org

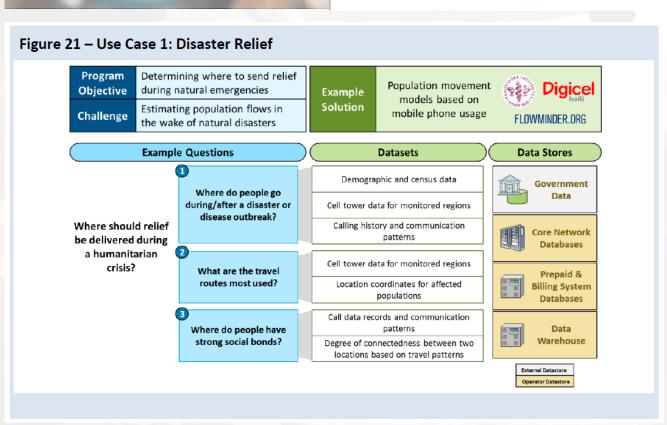




Questions & challenges

Using Mobile Data for Development









Questions & challenges

Linnet Taylor, The ethics of big data as a public good: which public? Whose good?

Ethical review of the Orange D4D Challenge Senegal