Complex Networks

connections, measurements, and social systems

Sune Lehmann

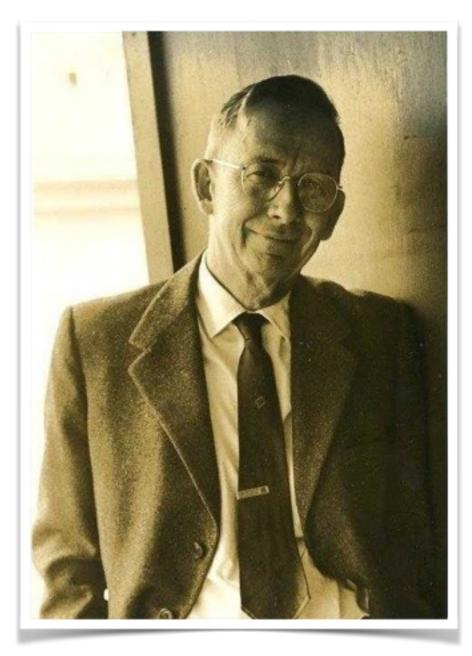
- Associate Professor, DTU Compute. Technical University of Denmark.
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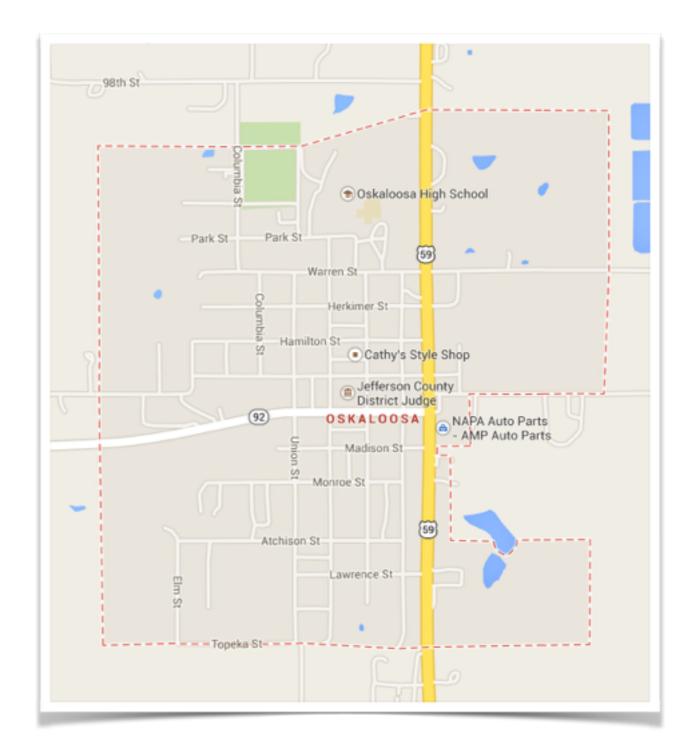




I am slightly obsessed with data quality

Roger G. Barker





Monitored "Midwest" for 25 years, using grad students, research assistants, wives - equipped with pen and paper.

Deb Roy



Human Speechome Project



PHOTO: (C) PHILIP DECAMP AND DEB ROY 2012 FROM THE HUMAN FACE OF BIG DATA

Human Speechome Project

- 11 fisheye lens cameras + motion sensors.
- 14 omnidirectional microphones
- 1000m wires connect recorders to servers in basement
- Record from 8am -10pm every day for 3 years



PHOTO: (C) PHILIP DECAMP AND DEB ROY 2012 FROM THE HUMAN FACE OF BIG DATA

Human Speechome Project

- 11 fisheye lens cameras + motion sensors.
- 14 omnidirectional microphones
- 1000m wires connect recorders to servers in basement
- Record from 8am -10pm every day for 3 years

Human Speechome in numbers

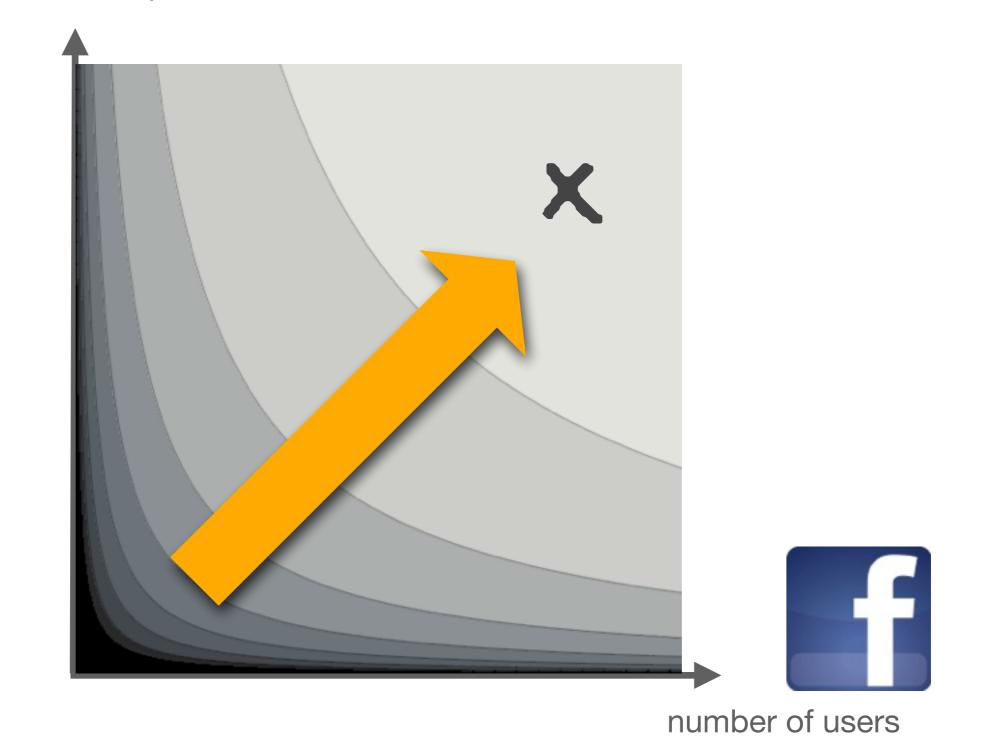
- 90,000 hours of video recorded
- 140,000 hours of audio recordings
- Approx 200GB of data collected every day
- 150 TB of raw data collected over course of project
- 70% of infants waking hours captured
- 10 to 12m words spoken
- 4m words so far transcribed

number of bits per user

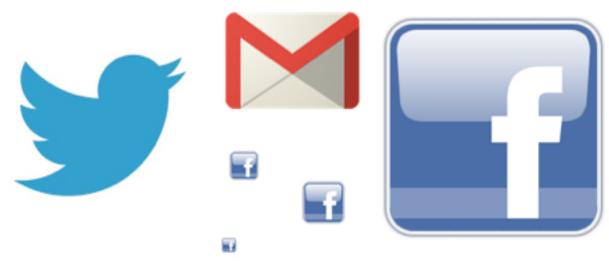




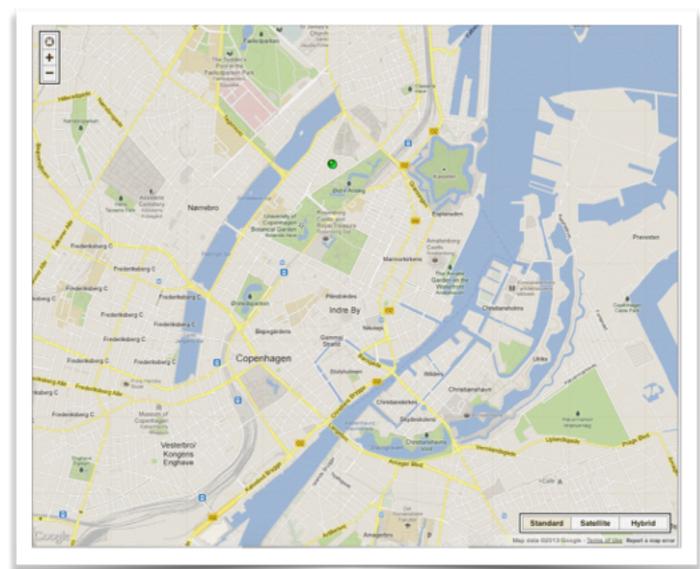
number of bits per user







Linked in







- proximity via Bluetooth
- telecommunication
- social networks
- geolocation
- demographics & personality
 live questionnaires

Credit goes to these guys (and others)

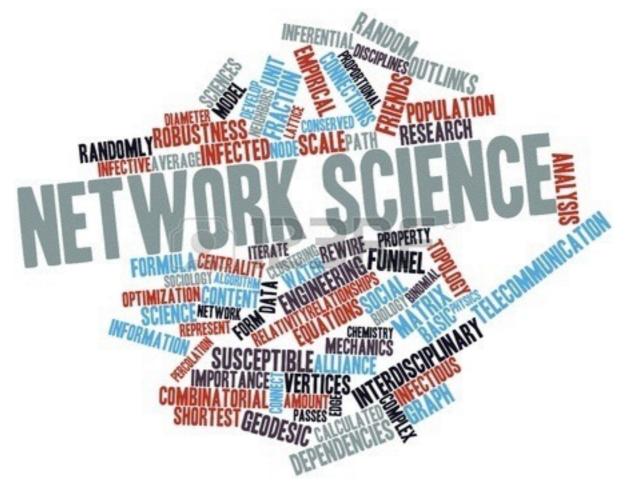


We started the project in January 2012, so what do we have to show for it?

- Ran a 200 person beta-test in September 2012.
- Rolled out the full 1000 person experiment in September 2013.
- Lots of initial work on technology: phone software, backend, visualizations
- Scientific focus so far has been
 - understanding probes,
 - privacy.
- Now we're slowly starting to do "real science".

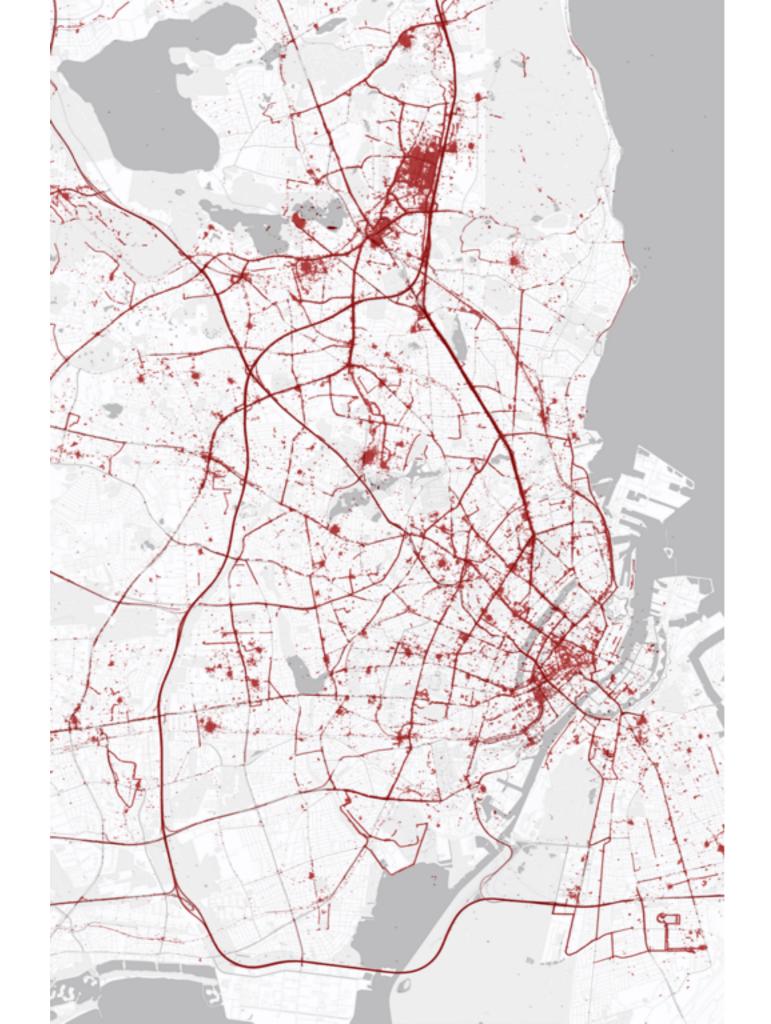
interests

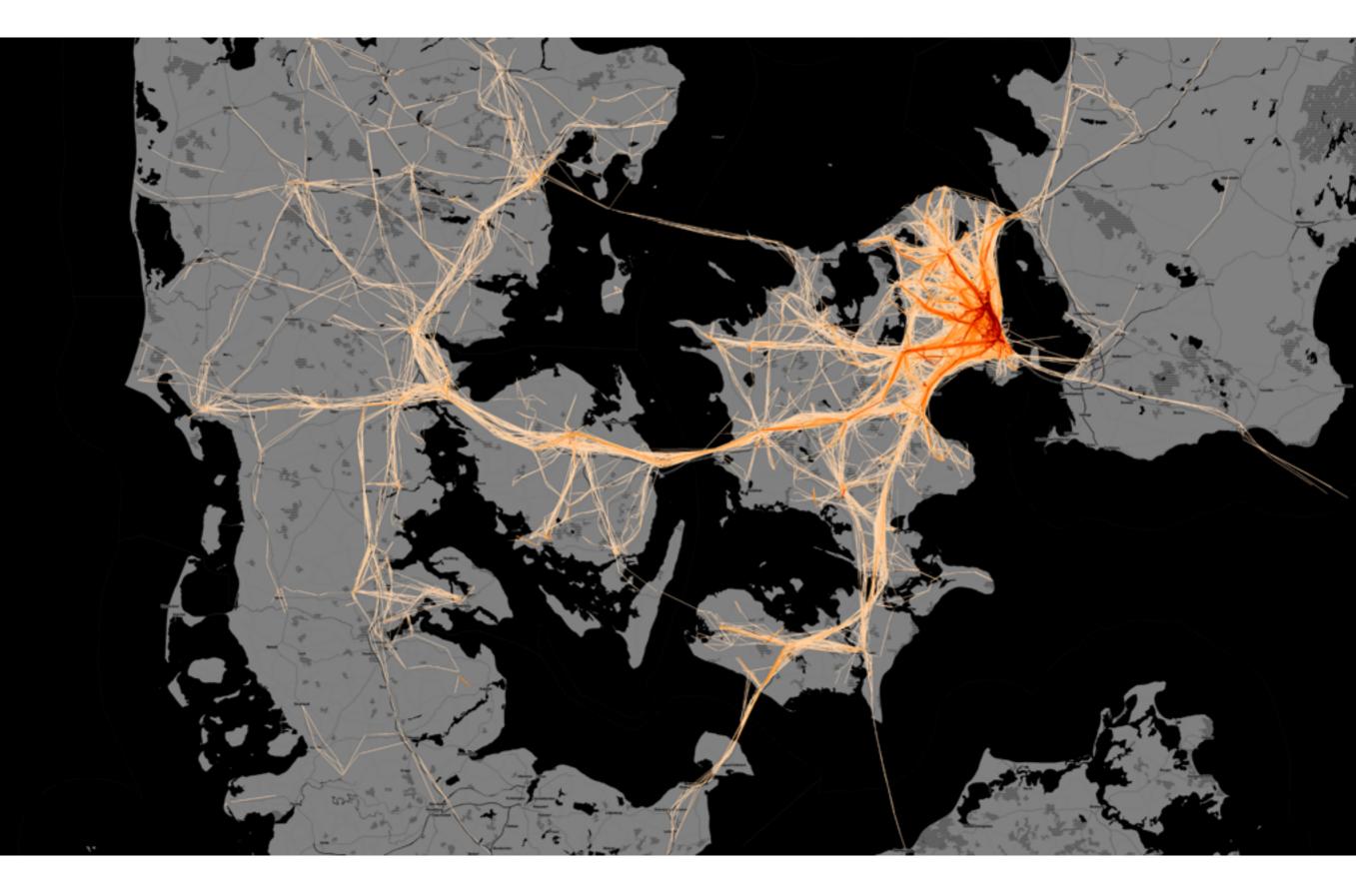
- mathematical models for human interactions
- sampling/relationship between interaction channels
- spreading processes. information/epidemics
- privacy
- the technical component



the data is flowing









Measuring Large-Scale Social Networks with High Resolution

Arkadiusz Stopczynski¹*, Vedran Sekara¹, Piotr Sapiezynski¹, Andrea Cuttone¹, Mette My Madsen³, Jakob Eg Larsen¹, Sune Lehmann^{1,2}

1 DTU Compute, Technical University of Denmark, Kgs. Lyngby, Denmark, **2** The Niels Bohr Institute, University of Copenhagen, Copenhagen, Denmark, **3** Department of Anthropology, University of Copenhagen, Copenhagen, Denmark

Abstract

This paper describes the deployment of a large-scale study designed to measure human interactions across a variety of communication channels, with high temporal resolution and spanning multiple years—the Copenhagen Networks Study. Specifically, we collect data on face-to-face interactions, telecommunication, social networks, location, and background information (personality, demographics, health, politics) for a densely connected population of 1 000 individuals, using state-of-the-art smartphones as social sensors. Here we provide an overview of the related work and describe the motivation and research agenda driving the study. Additionally, the paper details the data-types measured, and the technical infrastructure in terms of both backend and phone software, as well as an outline of the deployment procedures. We document the participant privacy procedures and their underlying principles. The paper is concluded with early results from data analysis, illustrating the importance of multi-channel high-resolution approach to data collection.

Citation: Stopczynski A, Sekara V, Sapiezynski P, Cuttone A, Madsen MM, et al. (2014) Measuring Large-Scale Social Networks with High Resolution. PLoS ONE 9(4): e95978. doi:10.1371/journal.pone.0095978

Editor: Yamir Moreno, University of Zaragoza, Spain

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Funding: The SensibleDTU project was made possible by a Young Investigator Grant from the Villum Foundation (High Resolution Networks, awarded to SL). Scaling the project up to 1 000 individuals in 2013 was made possible by a interdisciplinary UCPH 2016 grant, Social Fabric (PI David Dreyer Lassen, SL is co-PI) focusing mainly on the social and basic science elements of the project. This grant has funded purchase of the smartphones, as well as technical personnel. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Competing Interests: The authors have declared that no competing interests exist.

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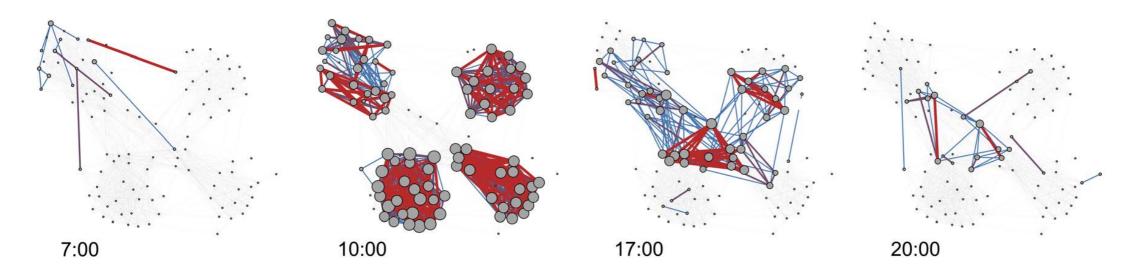


Figure 1. Dynamics of face-to-face interactions in the 2012 deployment. The participants meet in the morning, attend classes within four different study lines, and interact across majors in the evening. Edges are colored according to the frequency of observation, ranging from low (blue) to high (red). With 24 possible observations per hour, the color thresholds are respectively: blue (0 < observations ≤ 6), purple (6 < observations ≤ 12), and red (<12 observations). Node size is linearly scaled according to degree.

Face-to-face

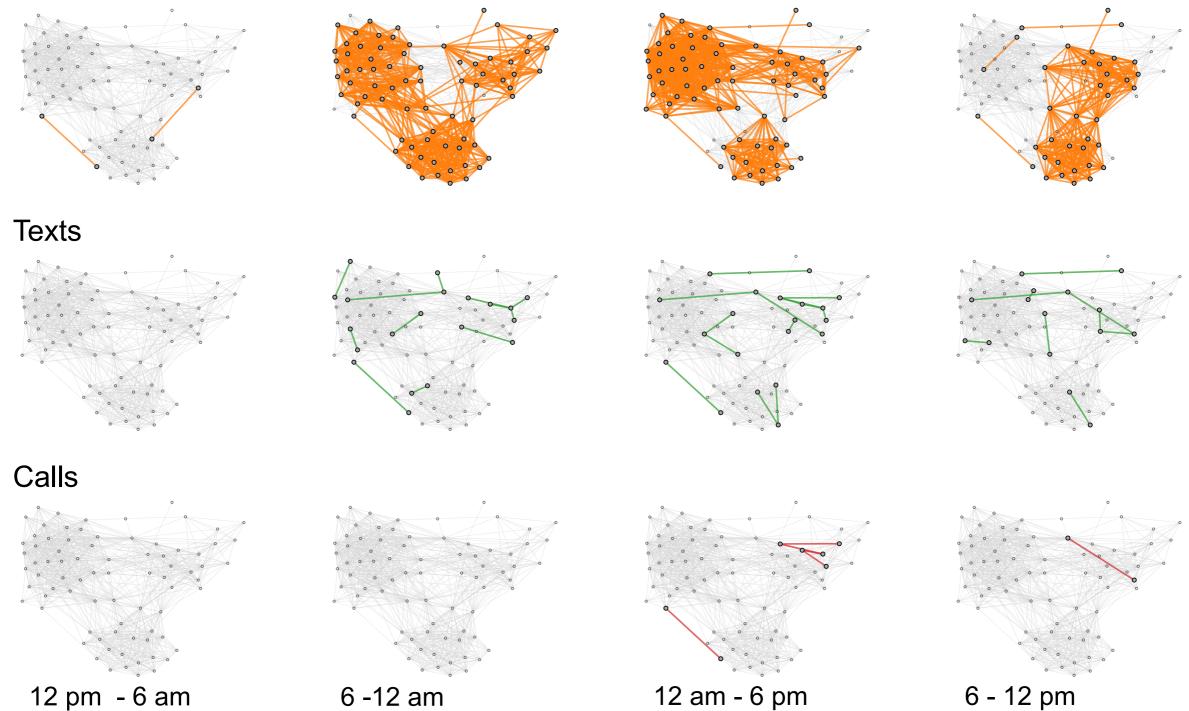


Figure 11. Daily activations in three networks. One day (Friday) in a network showing how different views are produced by observing different channels.

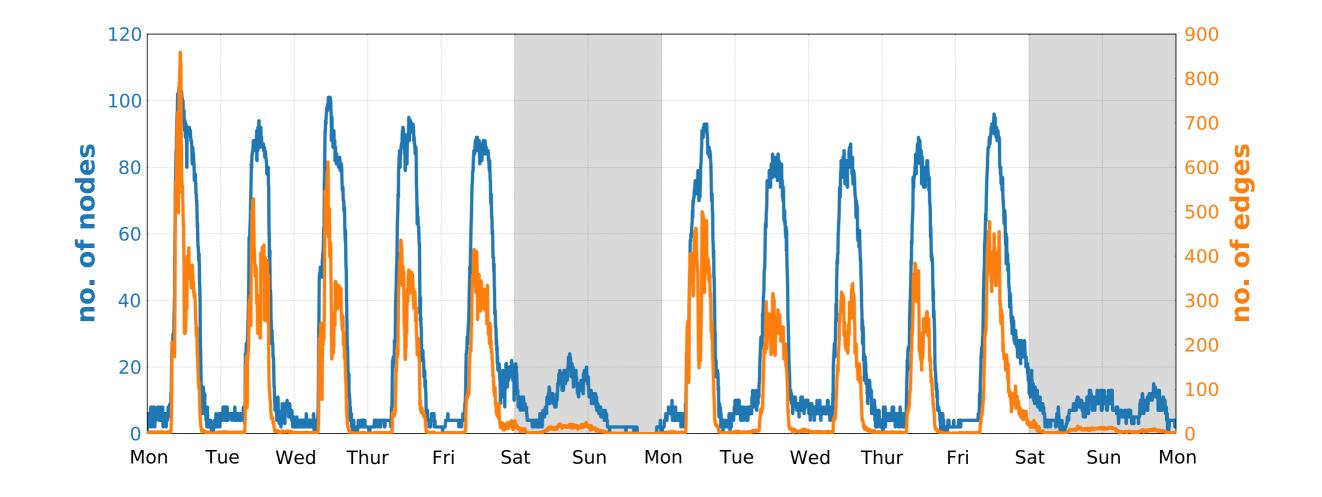


Figure 5. Weekly temporal dynamics of interactions. Face-to-face interaction patterns of participants in 5-minute time-bins over two weeks. Only active participants are included, i.e. users that have either observed another person or themselves been observed in a given time-bin. On average we observed 29 edges and 12 nodes in 5-minute time-bins and registered 10634 unique links between participants.

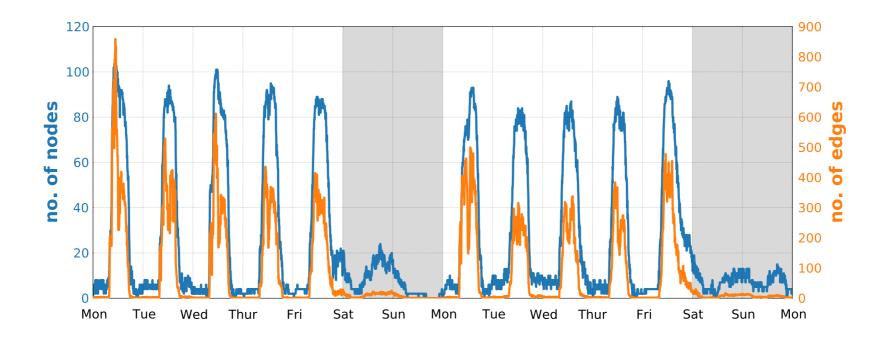


Figure 5. Weekly temporal dynamics of interactions. Face-to-face interaction patterns of participants in 5-minute time-bins over two weeks. Only active participants are included, i.e. users that have either observed another person or themselves been observed in a given time-bin. On average we observed 29 edges and 12 nodes in 5-minute time-bins and registered 10634 unique links between participants.

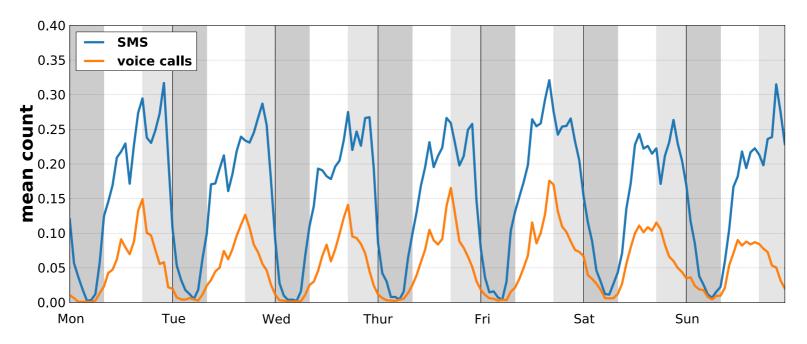


Figure 10. Weekly temporal dynamics of interactions. All calls and SMS', both incoming and outgoing calculated over the entire dataset and averaged per user and per week, showing mean number of interactions users had in given weekly bin. Light gray denotes 5pm, the end of lectures at the university, dark gray covers night between 12am and 8am. SMS is used more for communication outside regular business hours

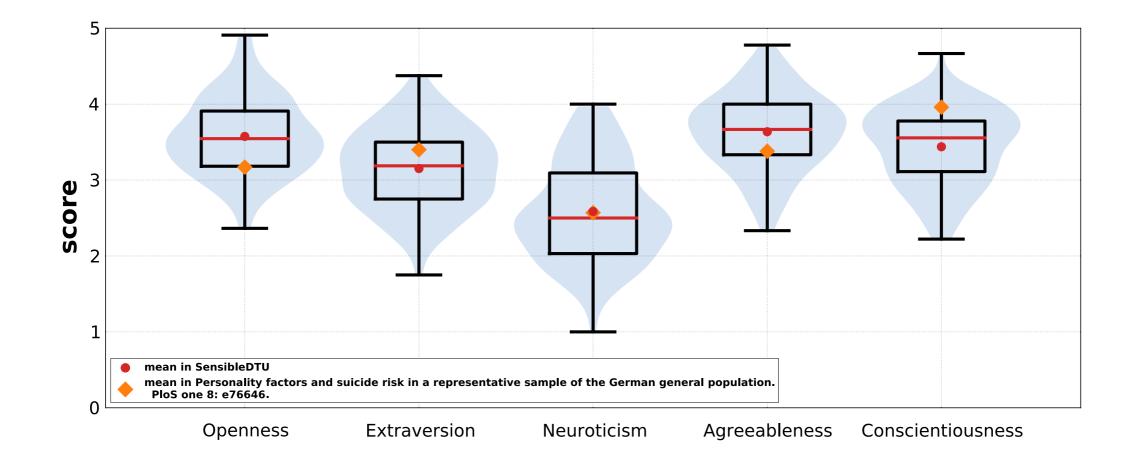


Figure 13. Personality traits Violin plot of personality traits. Summary statistics are: openness $\mu_O = 3.58$, $\sigma_O = 0.52$; extraversion $\mu_E = 3.15$, $\sigma_E = 0.53$; neuroticism $\mu_N = 2.59 \sigma_N = 0.65$; agreeablenes $\mu_A = 3.64 \sigma_A = 0.51$; conscientiousness $\mu_C = 3.44 \sigma_C = 0.51$. Mean values from our deployment (red circles) compared with mean values reported in [169] (orange diamonds).

technology & privacy



Privacy in Sensor-Driven Human Data Collection:

A Guide for Practitioners

Working Paper

Arkadiusz Stopczynski^{1,2,*}, Riccardo Pietri¹, Alex 'Sandy' Pentland², David Lazer³, Sune Lehmann^{1,4}

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- 2 MIT Media Lab
- **3** Northeastern University
- 4 The Niels Bohr Institute

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5	Privacy and Datasets	

permission from appropriate legal entities data security (attacks, physical, expiration) informed consent (control & understanding) right to choose (noise, anonymization, ownership) sharing (friends & researchers)

permission from appropriate legal entities

SØG

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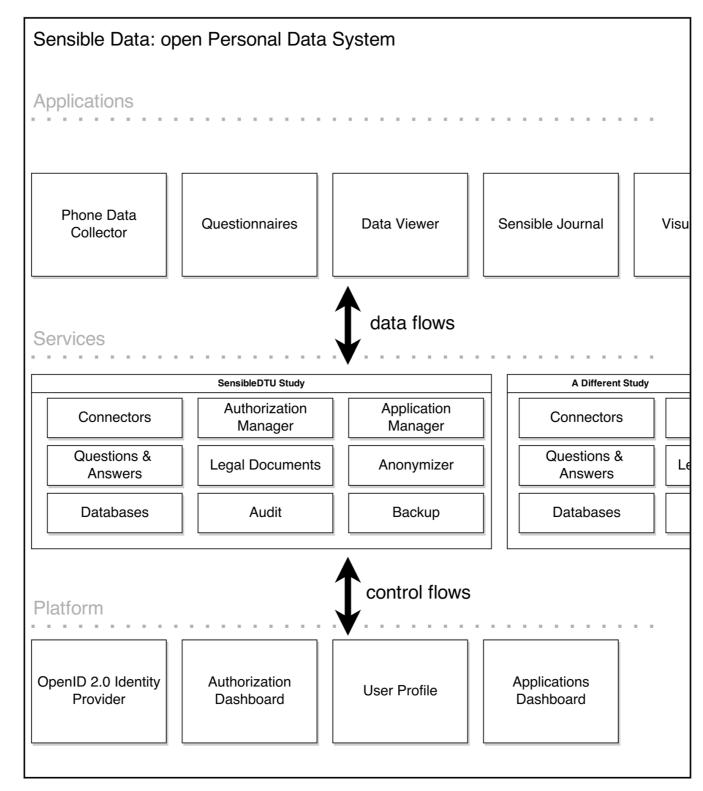
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Datatilsynets sagsbehandlingstider

permission from appropriate legal entities data security (attacks, physical, expiration)



permission from appropriate legal entities data security (attacks, physical, expiration)



permission from appropriate legal entities data security (attacks, physical, expiration) informed consent (**control** & understanding)

Sensible Data At Home & Profile @ About @ Help

My Projects

This page contains a list of the applications that I have given permission to collect my data (e.g. questionnaire and data collected by SensibleDTU app on my smartphone).

SensibleDTU martphones meet research in social systems!		
	IS of your applications. Applications collect data for SensibleDTU and for security reasons you must em before they are activated.	
Questionn	aires \mapsto	
he application is a que lick up the phone.	stionnaire that you must complete in order to participate in the project. After you complete the survey, you may book a time to	
not al this data does NOT in	Ilowed to submit your questionnaire data Allow >	
Phone Data Collector co	a Collector >>>	
	can submit data from your phone	
Battery, Time Offs Hardware Info Location Bluetooth scans WiFi scans	set, Screen On/Off	

My Projects

าล

This page contains a list of the applications that I have given permission to collect my data (e.g. questionnaire and data collected by SensibleDTU app on my smartphone).

SensibleDTU

Smartphones meet research in social systems!

Applications

Below there is a list of your applications. Applications collect data for SensibleDTU and for security reasons you must approve each of them before they are activated.

Questionnaires 🗪

The application is a questionnaire that you must complete in order to participate in the project. After you complete the survey, you may book a time to pick up the phone.

not allowed to submit your questionnaire data

Allow >

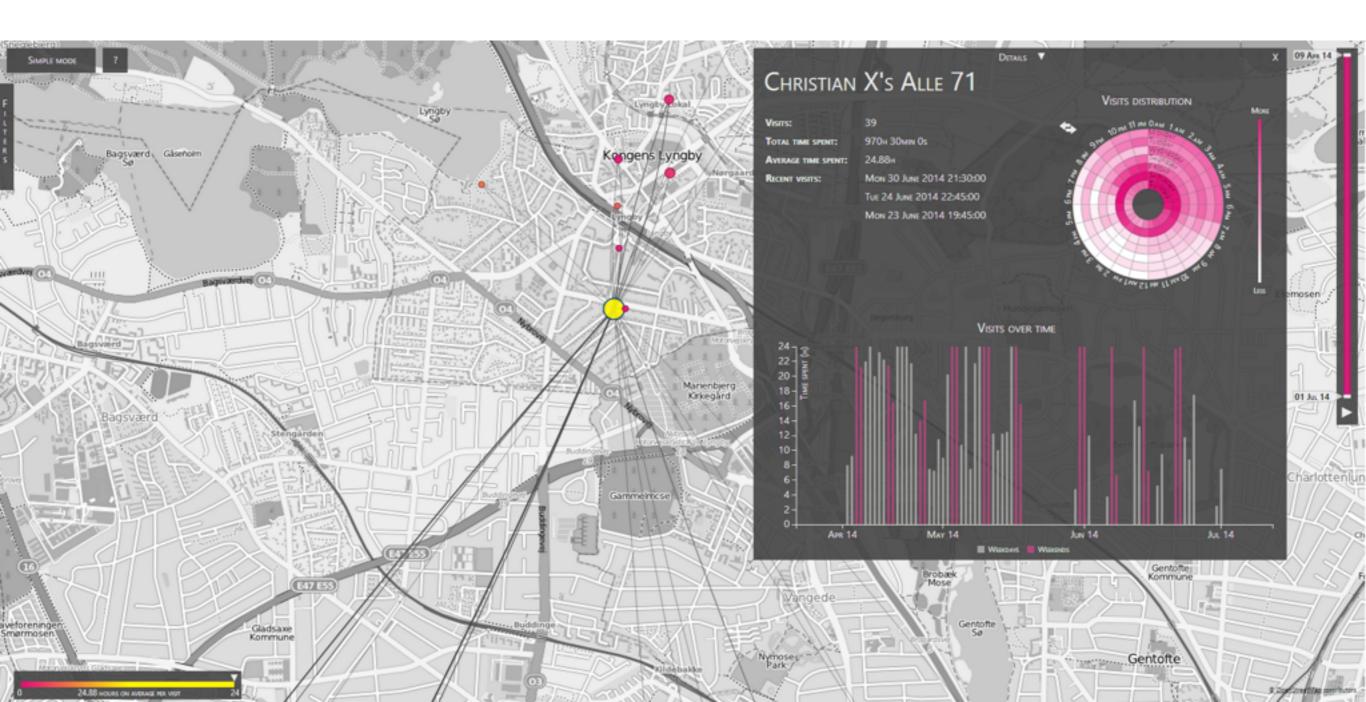
this data does NOT include your name

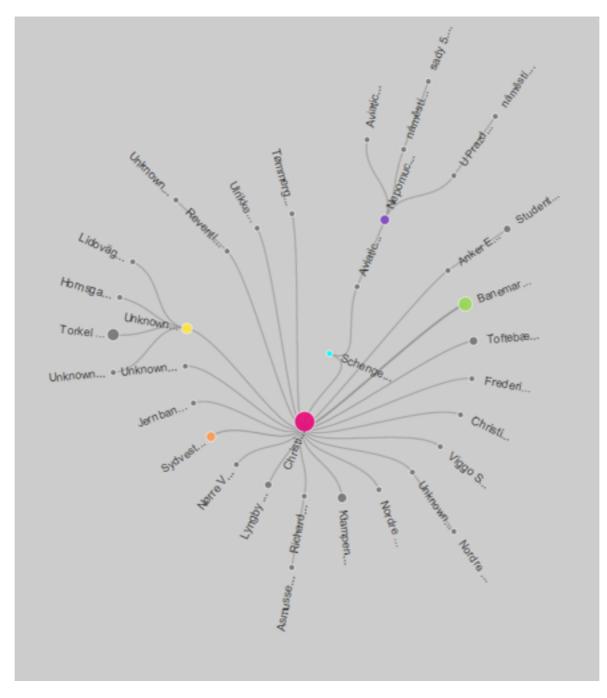
Phone Data Collector

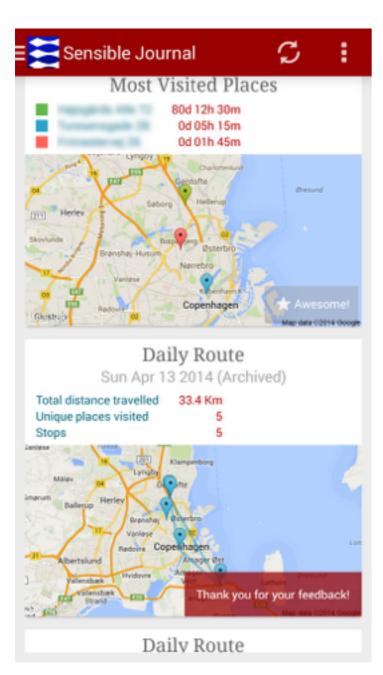
Phone Data Collector connects the SensibleDTU app on your smartphone with your Sensible Data account. You must authenticate here before the data from the phone is registered in the system. Please note that you must have your app installed on your phone before you can get started.

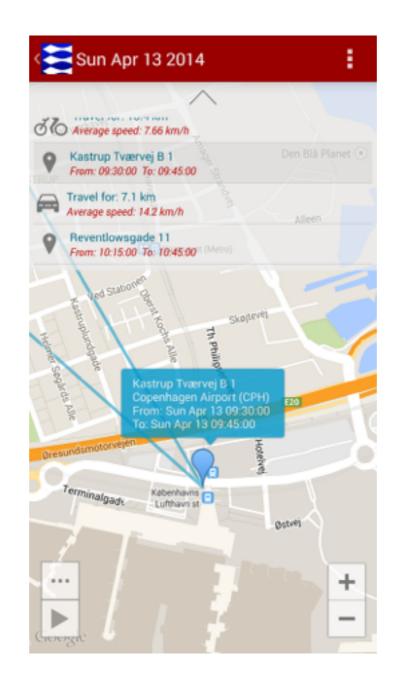
can submit data from your phone

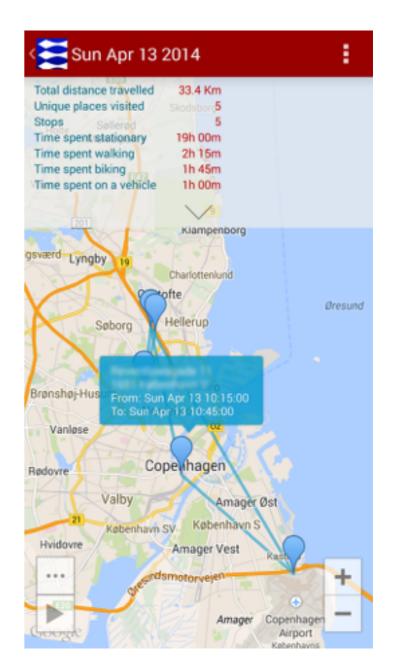
- · Battery, Time Offset, Screen On/Off
- Hardware Info
- Location
- · Bluetooth scans
- WiFi scans
- Cell ID
- Call & SMS Log (no names, no numbers, no content)
- · Contacts (no names, no numbers)











permission from appropriate legal entities data security (attacks, physical, expiration) informed consent (control & understanding) right to choose (noise, **anonymization**, ownership)



permission from appropriate legal entities data security (attacks, physical, expiration) informed consent (control & understanding) right to choose (noise, anonymization, ownership) sharing (friends & researchers)

understanding the raw data

PLOS ONE

The Strength of Friendship Ties in Proximity Sensor Data

Vedran Sekara¹*, Sune Lehmann^{1,2}

1 Cognitive Systems, Department of Applied Mathematics and Computer Science, Technical University of Denmark, Kgs. Lyngby, Denmark, 2 Niels Bohr Institute, University of Copenhagen, Østerbro, Denmark

Abstract

Understanding how people interact and socialize is important in many contexts from disease control to urban planning. Datasets that capture this specific aspect of human life have increased in size and availability over the last few years. We have yet to understand, however, to what extent such electronic datasets may serve as a valid proxy for real life social interactions. For an observational dataset, gathered using mobile phones, we analyze the problem of identifying transient and non-important links, as well as how to highlight important social interactions. Applying the Bluetooth signal strength parameter to distinguish between observations, we demonstrate that weak links, compared to strong links, have a lower probability of being observed at later times, while such links—on average—also have lower link-weights and probability of sharing an online friendship. Further, the role of link-strength is investigated in relation to social network properties.

Citation: Sekara V, Lehmann S (2014) The Strength of Friendship Ties in Proximity Sensor Data. PLoS ONE 9(7): e100915. doi:10.1371/journal.pone.0100915

Editor: Christopher M. Danforth, University of Vermont, United States of America

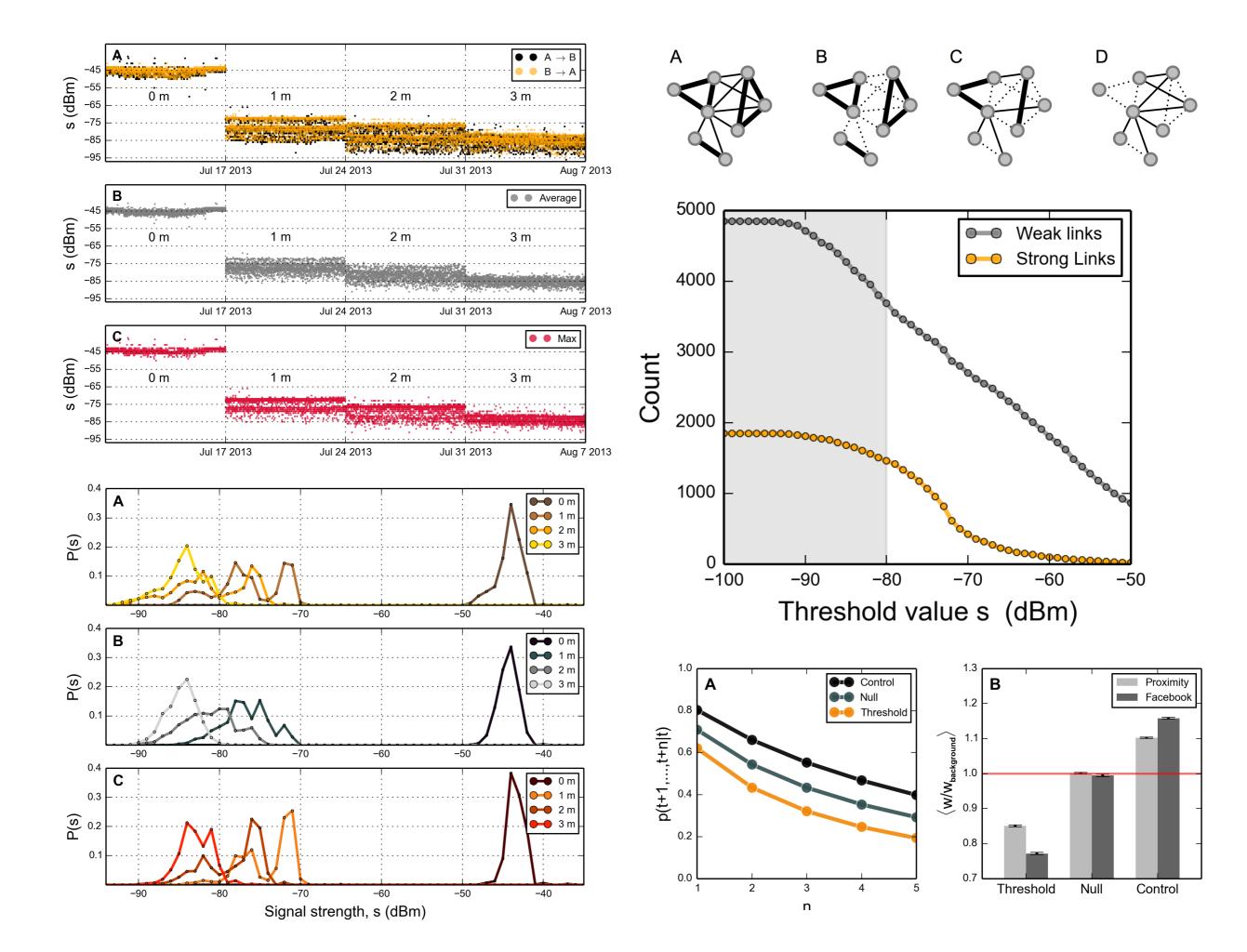
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Competing Interests: The authors confirm that one of the authors, Sune Lehmann, is a PLOS ONE Editorial Board member. This does not alter the authors' adherence to PLOS ONE Editorial policies and criteria.

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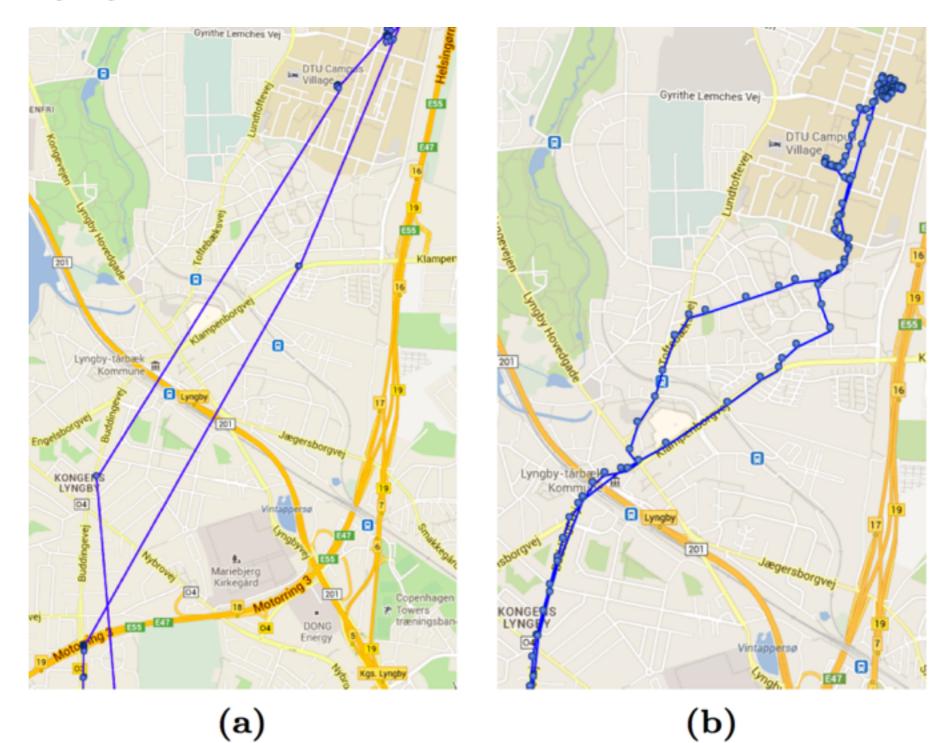
WiFi-Based Sensing of Social Systems Mobility

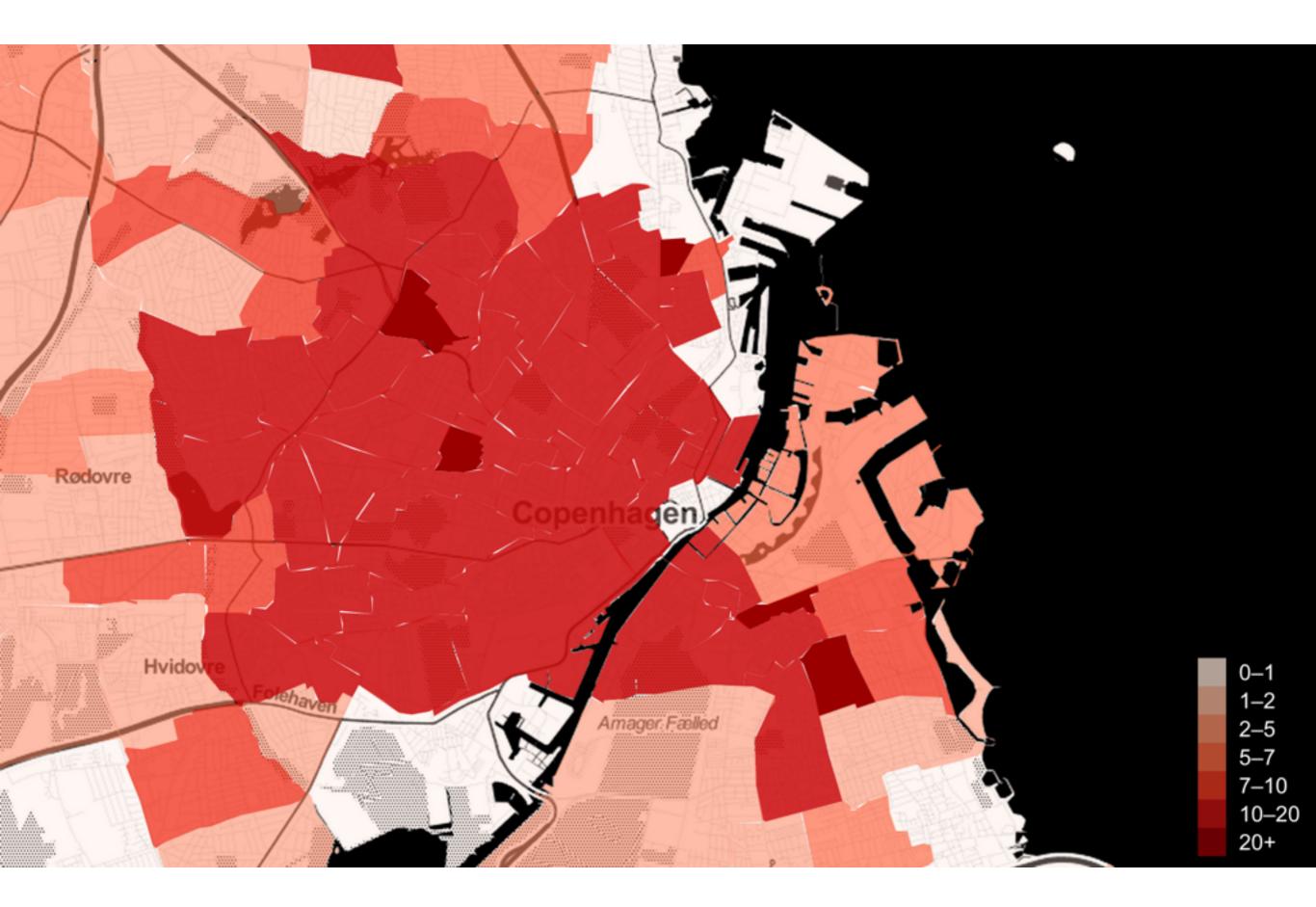
Piotr Sapiezynski¹,* Arkadiusz Stopczynski^{1,2}, Radu Gatej¹, & Sune Lehmann^{1,3}

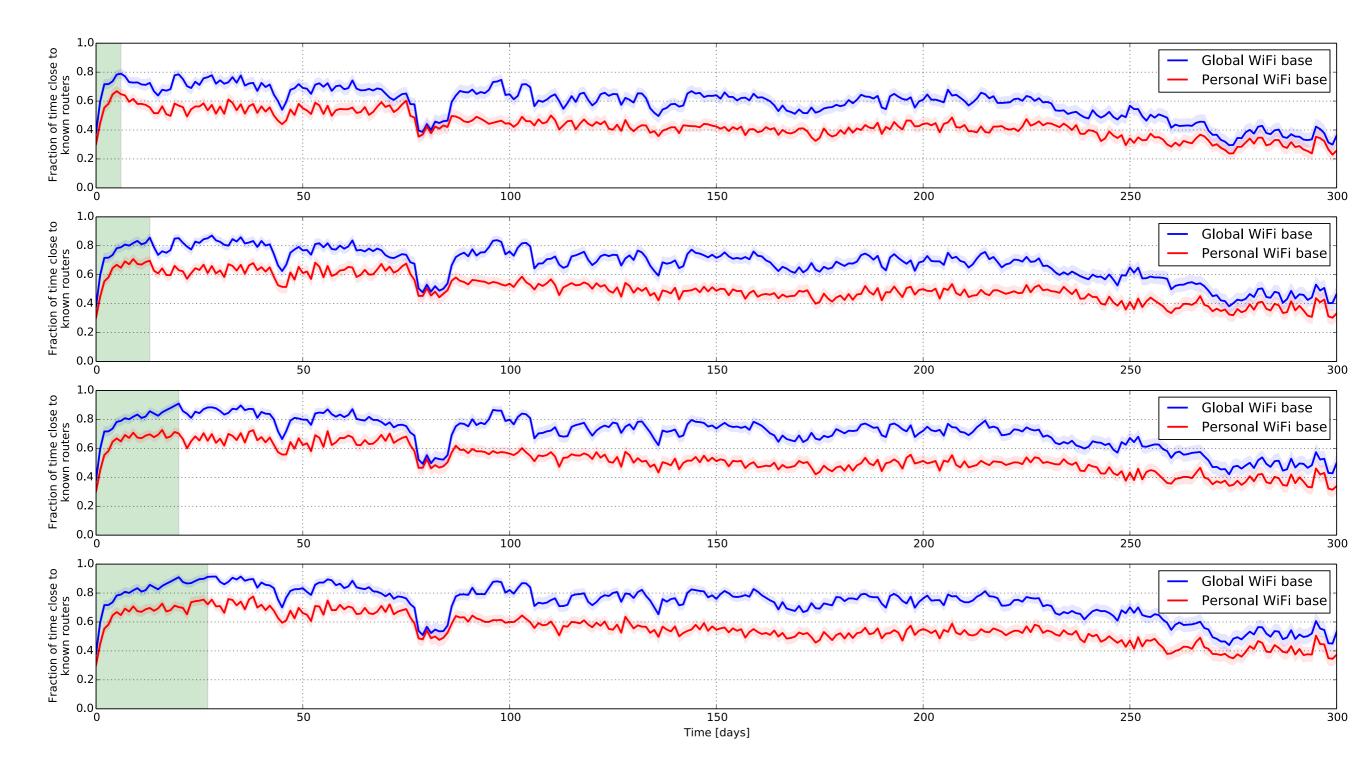
¹Department of Computer Science and Applied Mathematics, Technical University of Denmark

²Media Lab, Massachusetts Institute of Technology

³Niels Bohr Institute, University of Copenhagen







questions and answers for **big data**

Inferring Human Mobility from Sparse Low Accuracy Mobile Sensing Data

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Abstract

Understanding both collective and personal human mobility is a central topic in Computational Social Science. Smartphone sensing data is emerging as a promising source for studying human mobility. However, most literature focuses on high-precision GPS positioning and high-frequency sampling, which is not always feasible in a longitudinal study or for everyday applications because location sensing has a high battery cost. In this paper we study the feasibility of inferring human mobility from sparse, low accuracy mobile sensing data. We validate our results using participants' location diaries, and analyze the inferred geographical networks, the time spent at different places, and the number of unique places over time. Our results suggest that low resolution data allows accurate inference of human mobility patterns.

ACM Classification Keywords

H.2.8 [Database Management]: Database Applications – data mining

Author Keywords

Spatial data mining; mobile sensing; place discovery; location sensing; mobility;

"real science"

SCIENTIFIC REPORTS

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SUBJECT AREAS: APPLIED PHYSICS COMPUTATIONAL SCIENCE APPLIED MATHEMATICS

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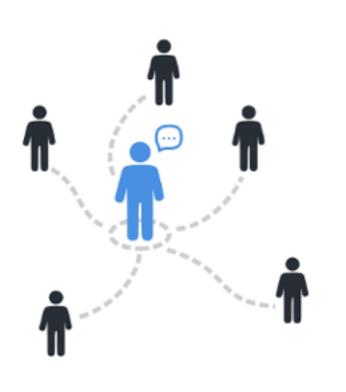
Correspondence and requests for materials should be addressed to

The Strength of the Strongest Ties in Collaborative Problem Solving

Yves-Alexandre de Montjoye¹, Arkadiusz Stopczynski^{1,2}, Erez Shmueli¹, Alex Pentland¹ & Sune Lehmann^{2,3}

¹Media Lab, Massachusetts Institute of Technology, ²Department of Applied Mathematics and Computer Science, Technical University of Denmark, ³Niels Bohr Institute, University of Copenhagen.

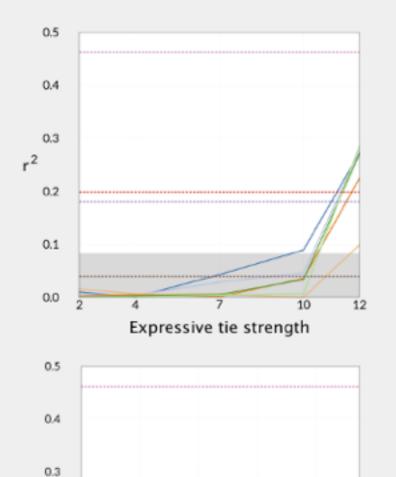
Complex problem solving in science, engineering, and business has become a highly collaborative endeavor. Teams of scientists or engineers collaborate on projects using their social networks to gather new ideas and feedback. Here we bridge the literature on team performance and information networks by studying teams' problem solving abilities as a function of both their within-team networks and their members' extended networks. We show that, while an assigned team's performance is strongly correlated with its networks of expressive and instrumental ties, only the strongest ties in both networks have an effect on performance. Both networks of strong ties explain more of the variance than other factors, such as measured or self-evaluated technical competencies, or the personalities of the team members. In fact, the inclusion of the network of strong ties renders these factors non-significant in the statistical analysis. Our results have consequences for the organization of teams of scientists, engineers, and other knowledge workers tackling today's most complex problems.



Why networking doesn't work?

Yves-Alexandre de Montjoye, Arkadiusz Stopczynski, Erez Shmueli, Alex 'Sandy" Pentland, and Sune Lehmann

"Networking," creating social ties in the workplace, generally increases performance. It is common career advice for executives, entrepreneurs, and academics. More companies are investing in common spaces and team building events to support networking.



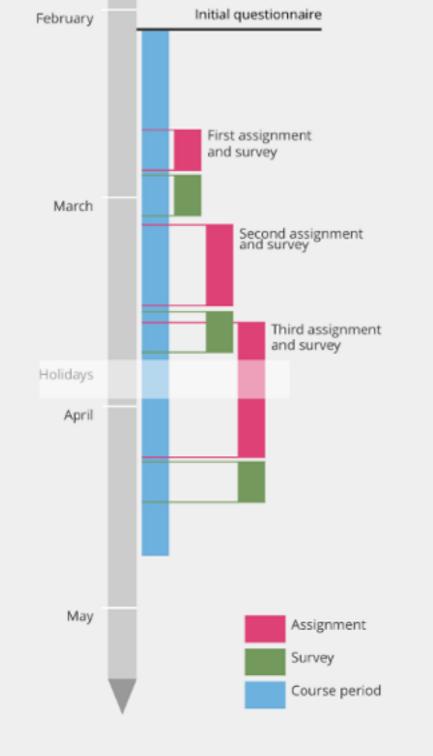
But networking doesn't seem to improve **performance in competitive environments**. Our experiment at a large western university showed that when faced with a complex problem, the **weak ties of team members did not have any effect** on their performance. Frequency of interactions with the people we'd consider to be acquaintances, and even with those we'd would consider friends, did not help your performance.

Network Measures

Mean/Max team degree Mean/Max team in-degree

A Priori Competencies

- ····· Self-evaluated technical competencies
- Measured technical competencies



Methodology of the experiment

We measured expressive ties, personality, and technical competencies through a questionnaire at the beginning of the experiment. We assigned participants to teams based on their expressive ties. Participants were assigned to three different teams over the course of the experiment. We used the logs of the university wifi system to infer the time participants were spending together. The effect is strongly non-linear: only the strongest ties in both expressive and instrumental networks have an effect on performance. Networks of both instrumental and expressive strong ties explain more of the variance than any other considered factors, such as measured or selfevaluated technical competencies, or the personalities of the team members; and the inclusion of the network of strong ties renders these factors non-significant in the statistical analysis.

The Strength of the Strongest Ties in Collaborative Problem Solving

de Montjoye, Y. A., Stopczynski, A., Shmueli, E., Pentland, A.S., Lehmann, S. (2014). The Strength of the Strongest Ties in Collaborative Problem Solving. Nature S.Rep., 4.

Contact: yva@mit.edu



Technical University of Denmark work in progress

Inferring friendships from WiFi data

