

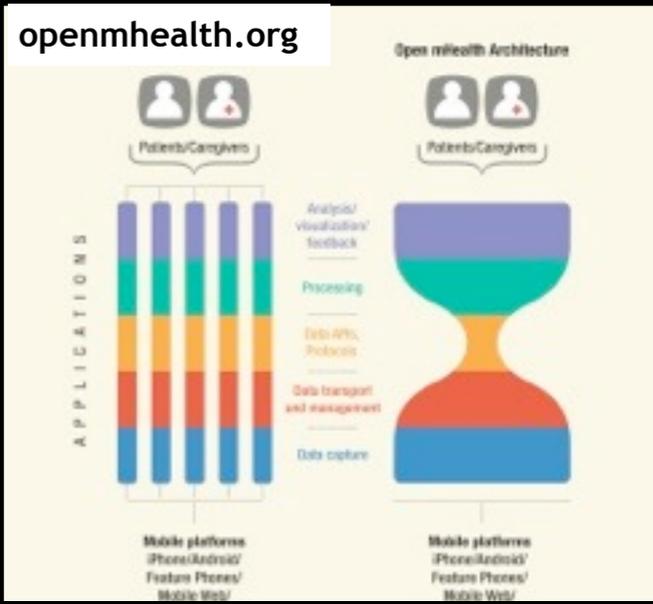
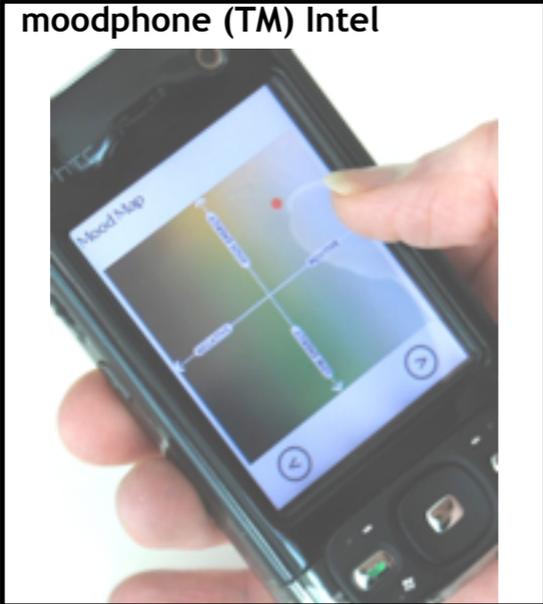
Embedding mobile computing and research in everyday life

Deborah Estrin
Professor, Cornell Tech, New York City
destrin@cs.cornell.edu

work done in collaboration with collaborators from
UCLA, openmhealth.org, iSTC, ...

Enabled by $>6 \times 10^9$ mobile phone users, increasingly with: GPS, imagers, touch screens, Internet, app stores

Motivated by 6×10^9 people on planet earth, their health needs, and economic realities



from embedded to mobile to participatory sensing

mhealth: 'personal evidence', n=me

whats next?

personal data APIs, mobile personal informatics, NEW YORK CITY!

**an end-to-end argument for driving systems research with
authentic applications**

Lessons from the field of embedded sensing 2002-2010



**Early themes: many simple
measurements
small platforms
autonomous operation**

Lessons from the field of embedded sensing 2002-2010



Early themes: many simple measurements
small platforms
autonomous operation

Midterm themes: multi-modal measurements
varied platforms
human-assisted operation

Lessons from the field of embedded sensing 2002-2010



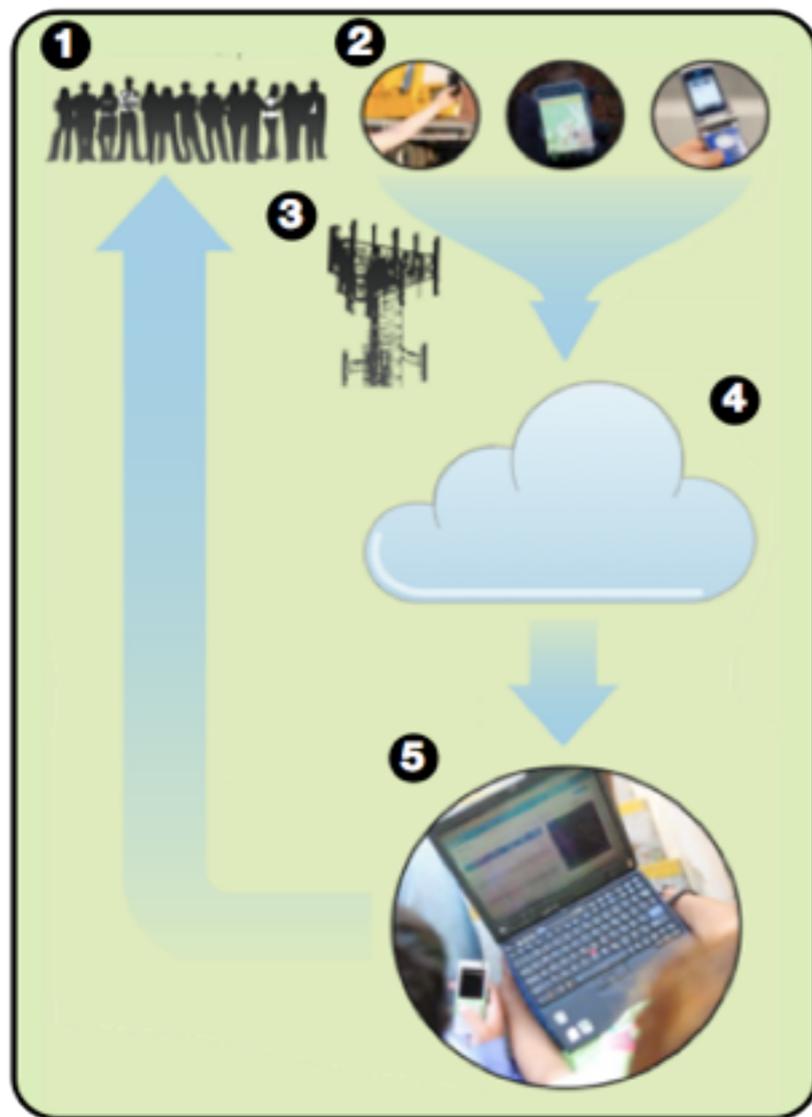
Early themes: many simple measurements
small platforms
autonomous operation

Midterm themes: multi-modal measurements
varied platforms
human-assisted operation

Eventual themes: model-based measurement
mobile platforms
assistive systems, infovis

Participatory Sensing (starting ~2006)

individuals and communities using personal mobile devices and web services to systematically explore and document their lives
(builds on methodologies of experience sampling [Csik85] and photovoice [Wang95])



Real time
(always on)

Real place
(always carried)

Real context
(historical, environmental, spatial, social)

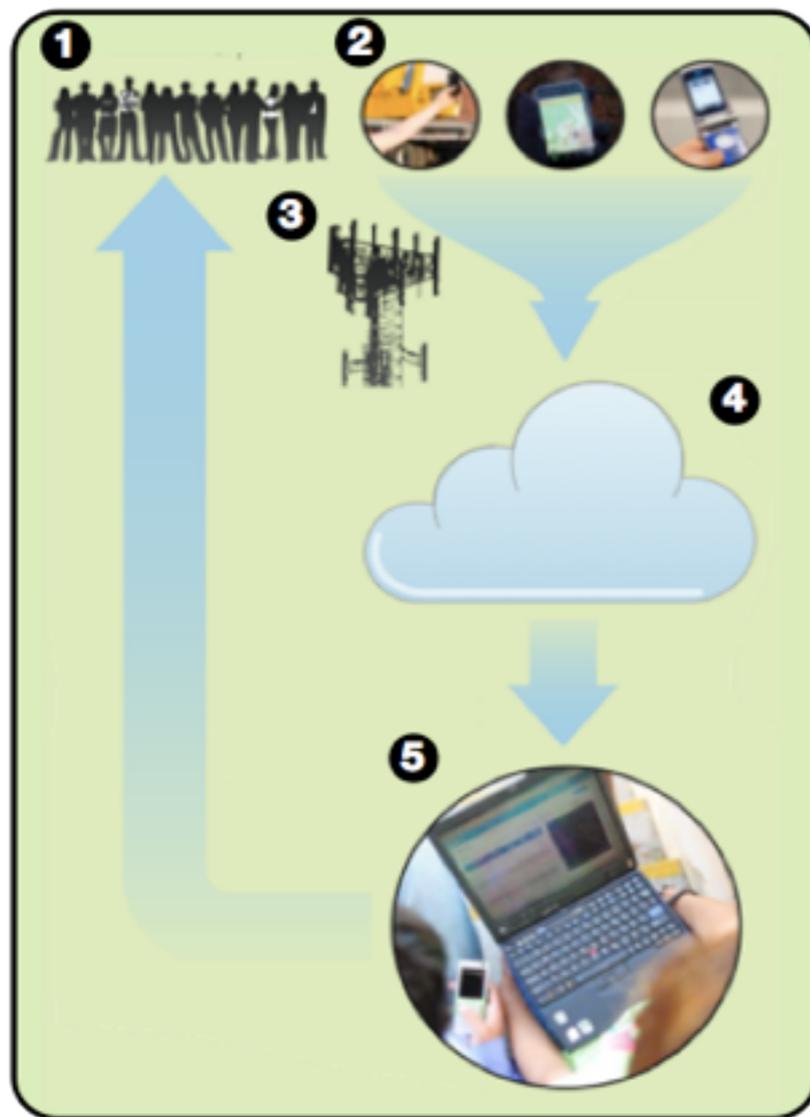
Real applications
(environment, education, community, health)



w/ Mark Hansen (Statistics/DMA), Jeff Burke (REMAP/TFT)

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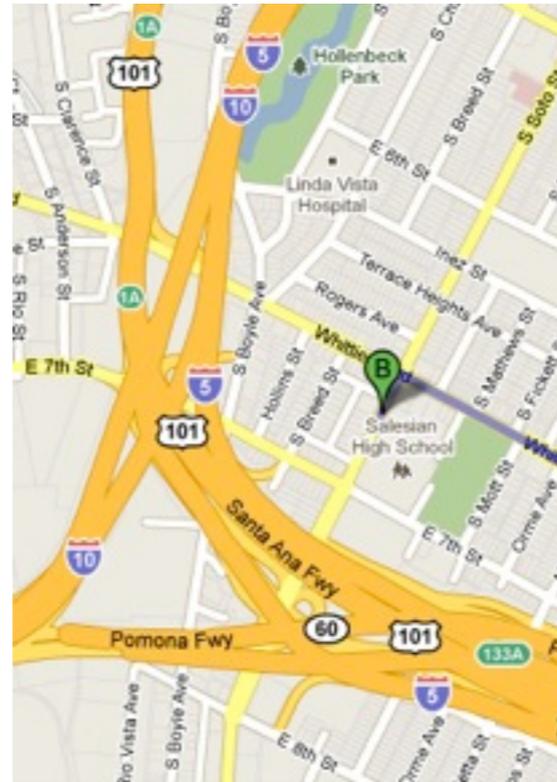
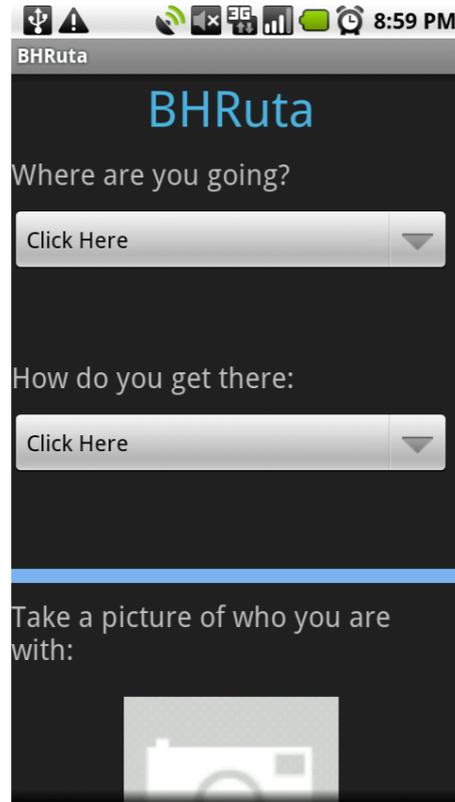
Real applications
(environment, education, community, health)

**Chose applications that scaled *down*...
as well as *up*: i.e., utility at small *n* so
*real use can guide iterative cycles of
innovation***

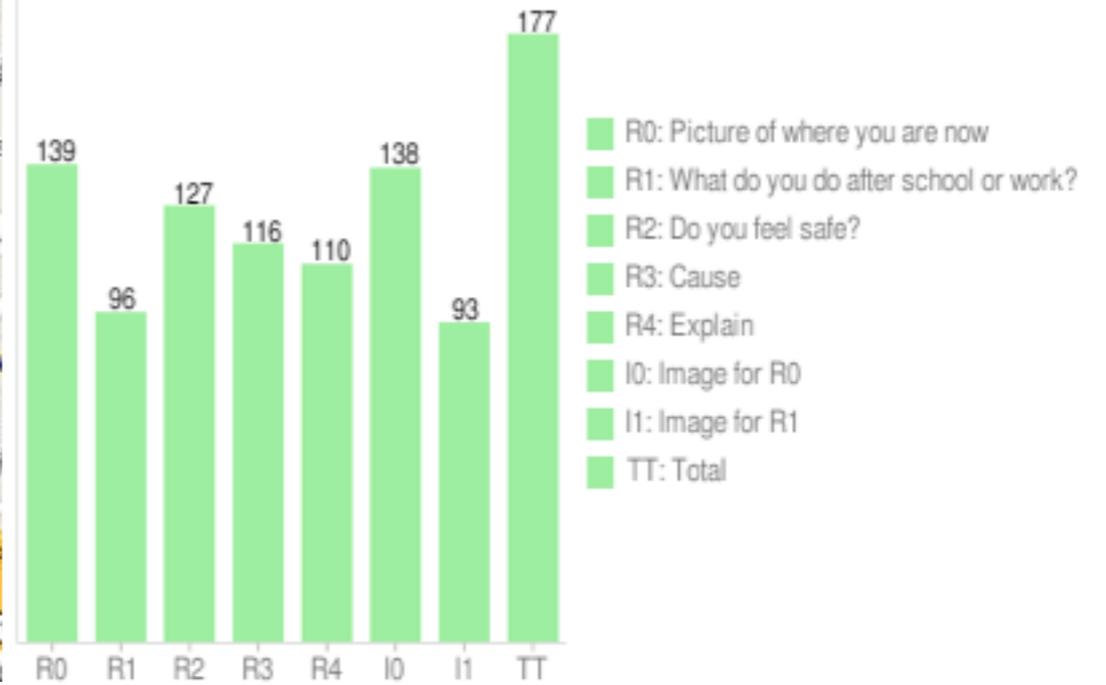


w/ Mark Hansen (Statistics/DMA), Jeff Burke (REMAP/TFT)

Assessing environmental factors: Community data gathering (Boyle Heights, 2010)



BHVecindario_Totals



20100415	raiders	2010-04-22	updat: 2010-04-22	34.06251	-118.19812
A0000022036677		22:49:17.310000-07:00	22:49:29.580879-07:00		

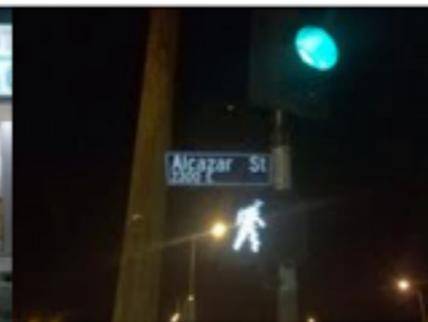
Where are you going? Errands

Place you stopped. What are you doing:
Other | groceries

What represents your neighborhood:
street name

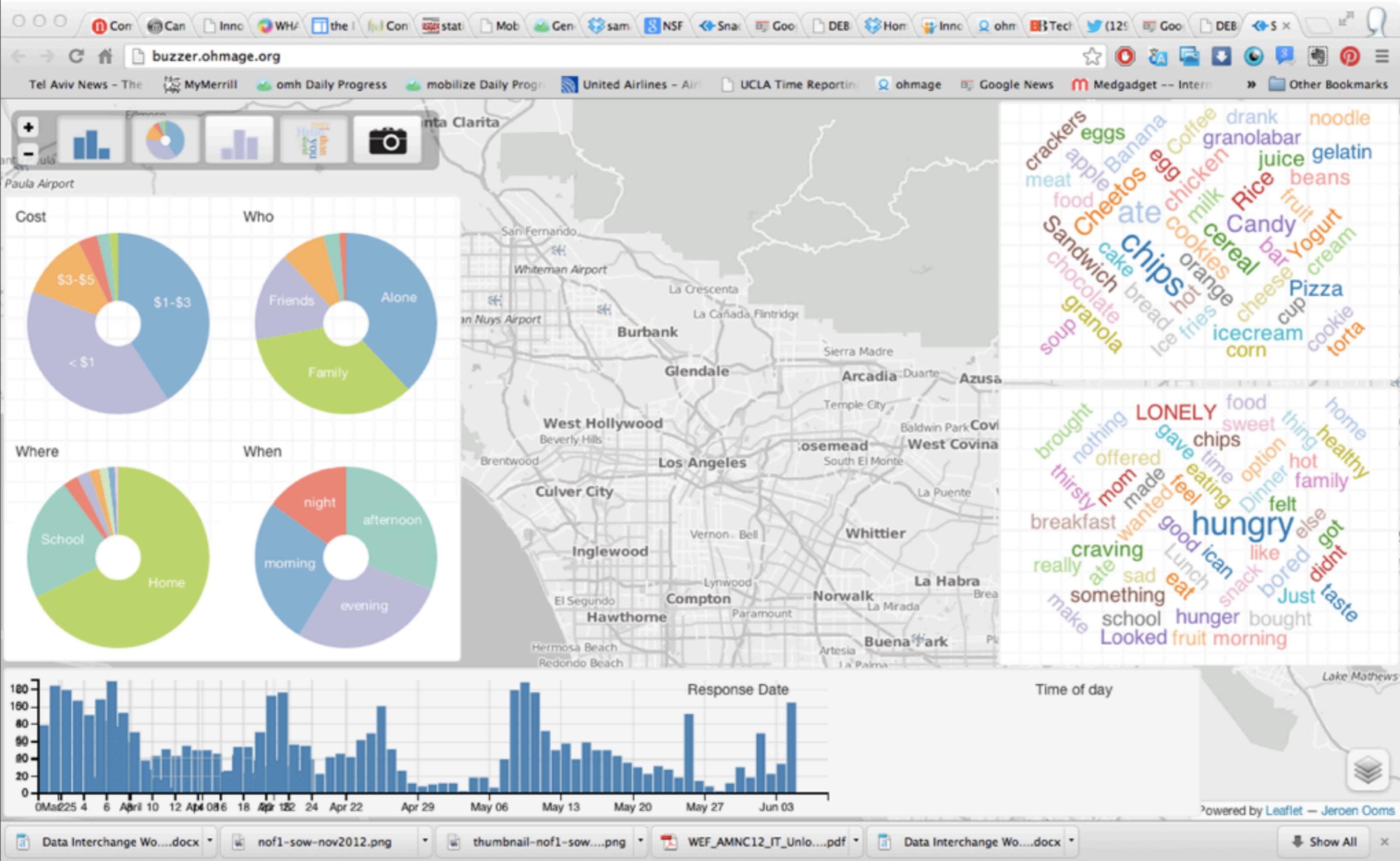
How do you get there? Walk
What where you ate along the way:
Street vendor

What did you eat along the way:
Other | tacos



Acker, Samanta, Belany et al

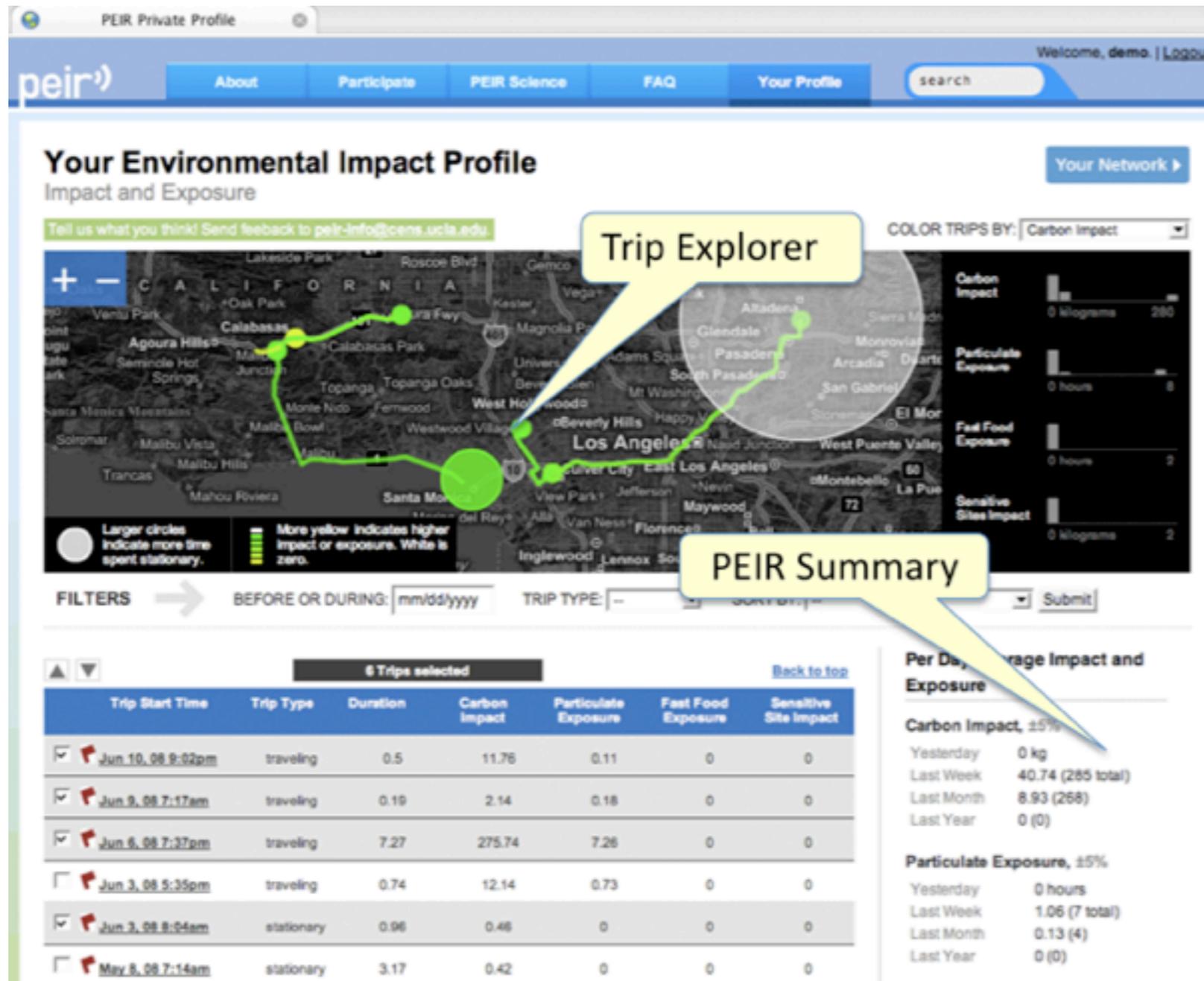
Snackboard dashboard to >3000 entries about snacking collected by high school students in 2012 (J. Ooms)



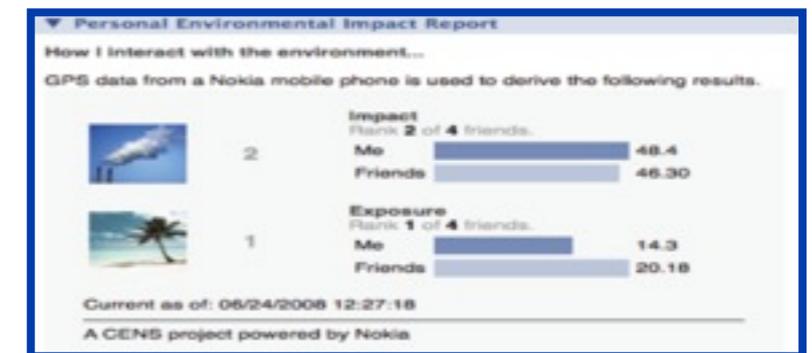
w/ Jeroen Ooms, Mark Hansen, Hongsuda T. 6

Telling traces: PEIR (2007-08)

model-based estimation using continuous location-activity-time series



Individual time-location traces used to automatically estimate daily personal carbon impact and air particulate exposure



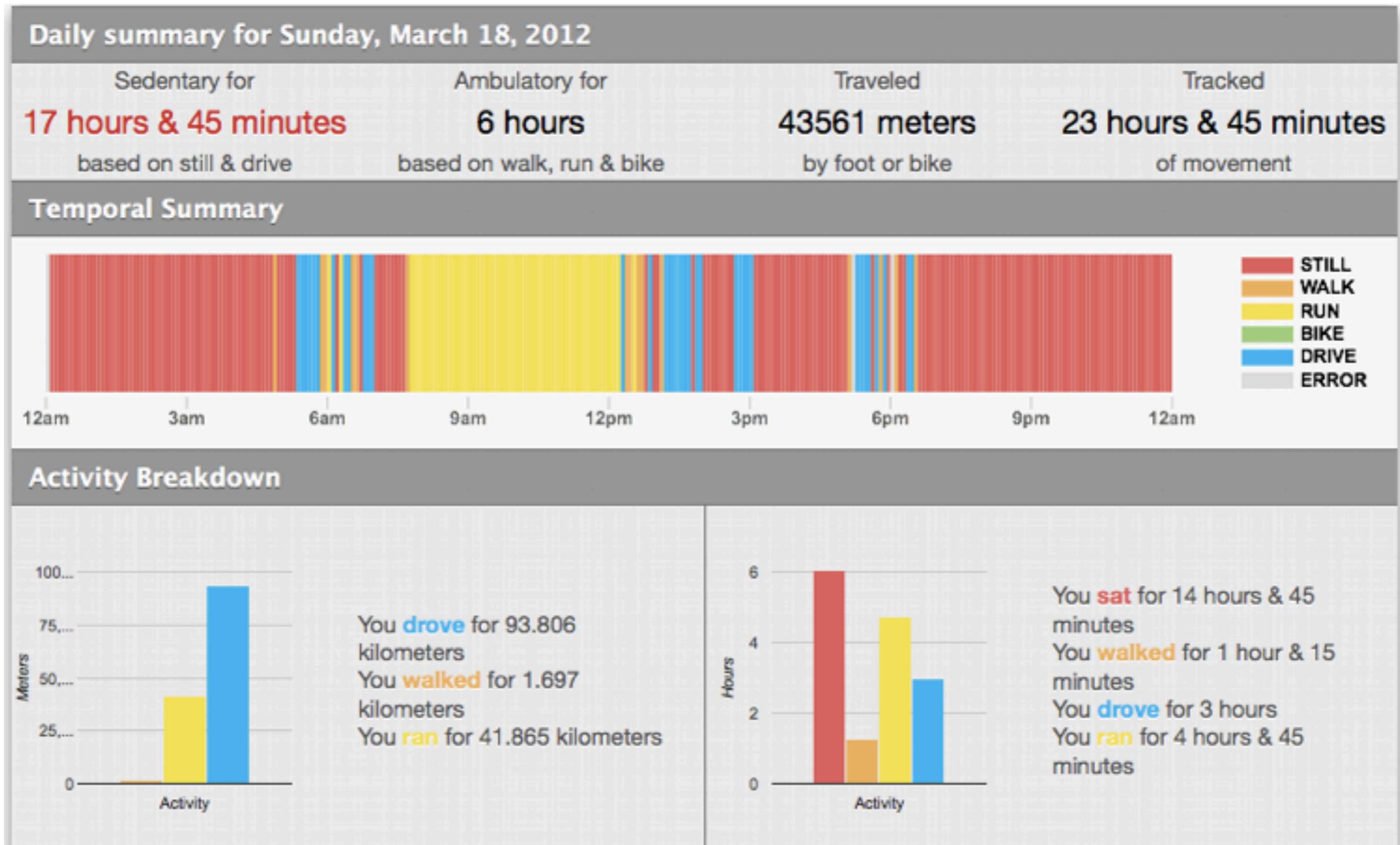
PEIR Facebook Application

<http://peir.cens.ucla.edu>

Wired NEXTFEST,
Chicago 2008

w/ Mark Hansen (Statistics/DMA), Jeff Burke (REMAP/TFT); Funded by Nokia

Pivot to focus on mobile health (mhealth)and personal n=me evidence



The promise of mobile Health (mHealth)

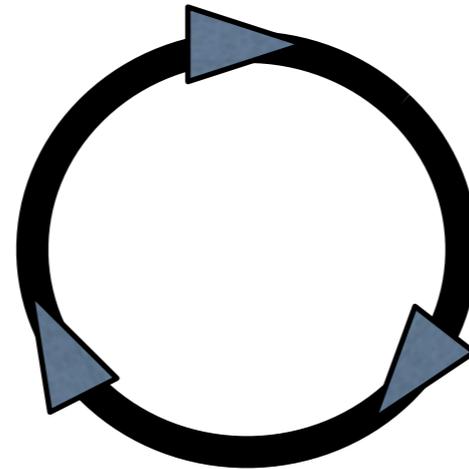
transform previously unmeasured behaviors and practices into personalized, evidence-based, and evidence-producing care



Photo: Marshall Astor, WWW

symptoms, side-effects, outcome measures, actions, activities, exposures..

capture/record activity, mobility, self-reports, tool-use, "digital exhaust"



visualize, summarize, highlight; inform, advise, persuade

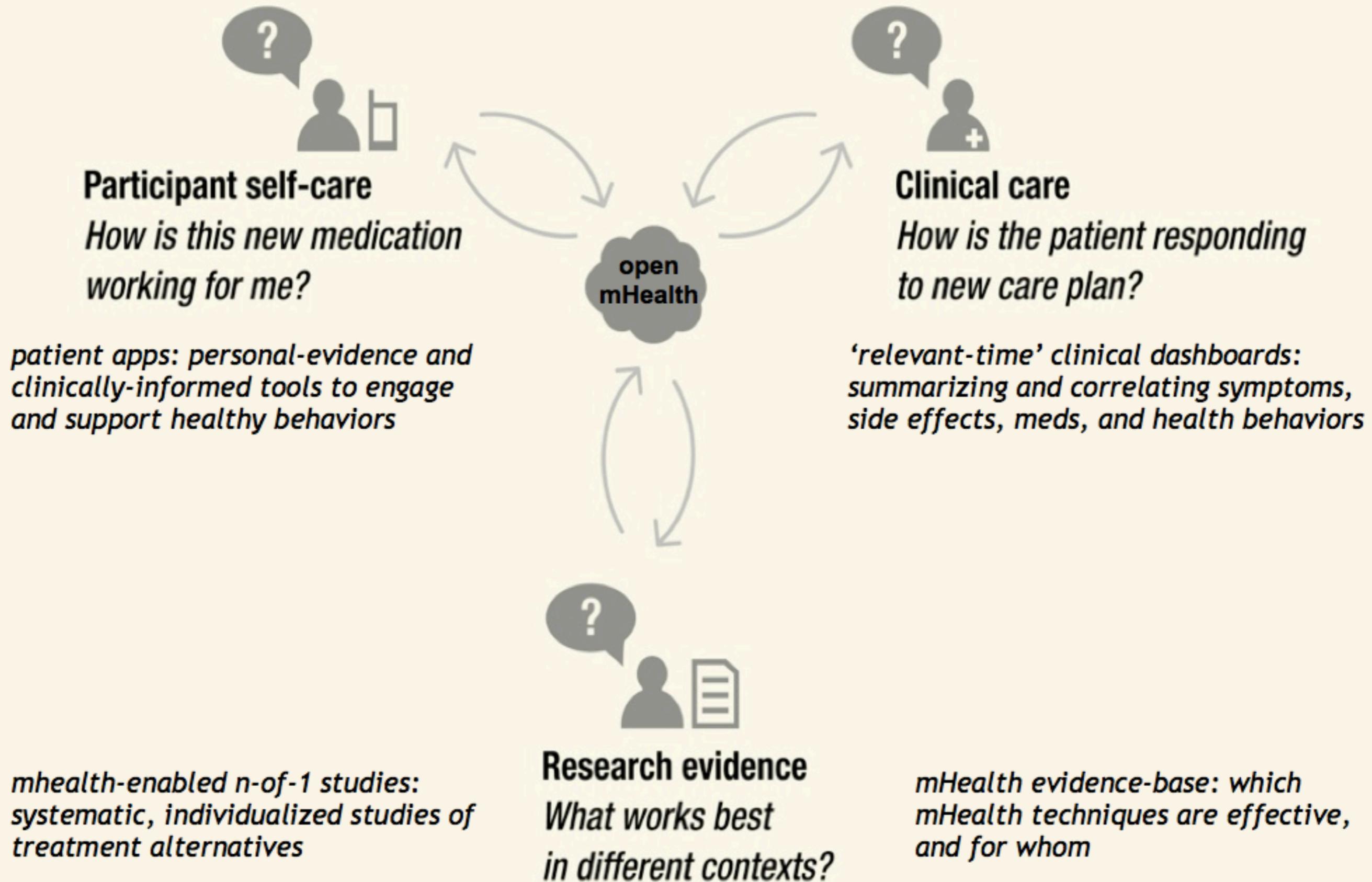
store, analyze, classify, fuse, mashup, filter, aggregate data



Why chronic disease management?

- 3 lifestyle behaviors (poor diet, lack of exercise, smoking) cause 1/3rd of US deaths; 50% Americans have 1 or more chronic diseases; age of onset getting younger.
- Over the next 20 years, Non Communicable Diseases will cost worldwide
> \$ 30 trillion; mental health > \$16.1 trillion (WEF, 2011)
- Equip individuals, families w/tools for measurement, management, understanding outside clinical setting

mHealth derived data serves 3 essential workflows



w/ Ida Sim, Open mHealth

Transformative methodological tool: recasting 'evidence'

(Complexes of)
Exposures
sertraline



strength of association?

individual



Outcome

depression



population

Transformative methodological tool: recasting 'evidence'

(Complexes of)
Exposures
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Outcome

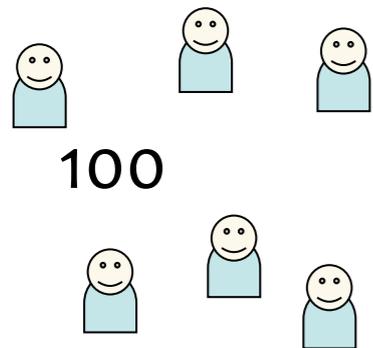
depression

individual

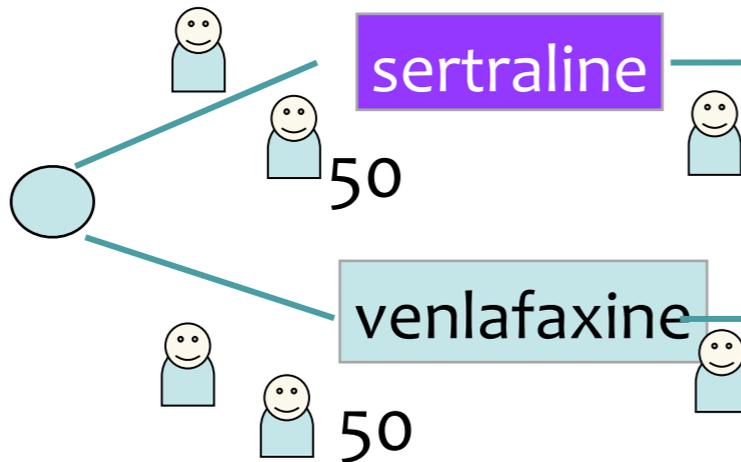


population

'does it work on average?' (RCT)



100



50

50

Depression (PHQ-9)

Depression (PHQ-9)

population

Transformative methodological tool: recasting 'evidence'

(Complexes of)
Exposures
sertraline

strength of association?

Outcome

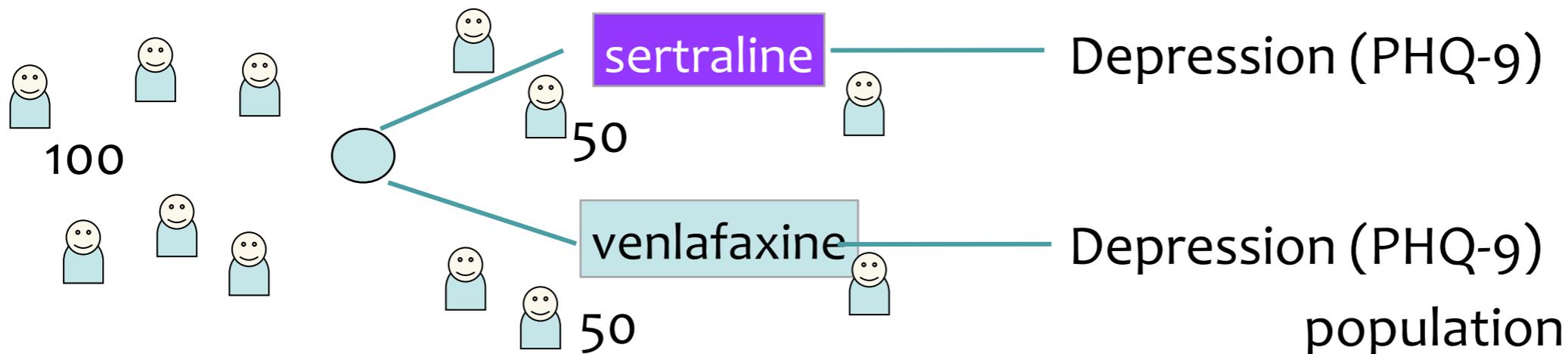
depression

individual

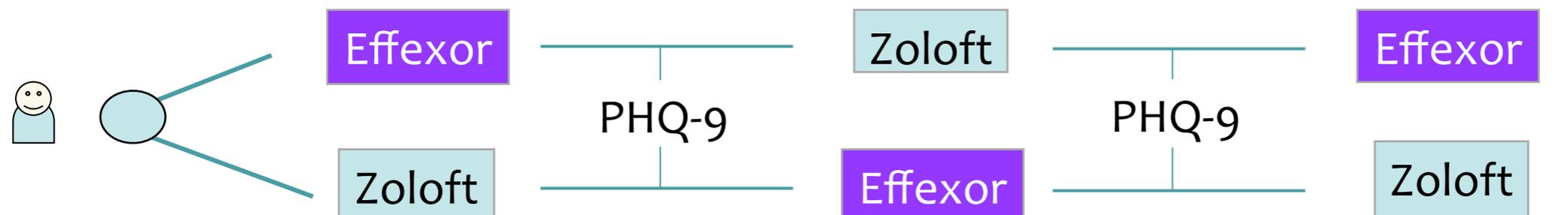


population

'does it work on average?' (RCT)



N-of-1 study design: 'does it work for Mr. Jones?'



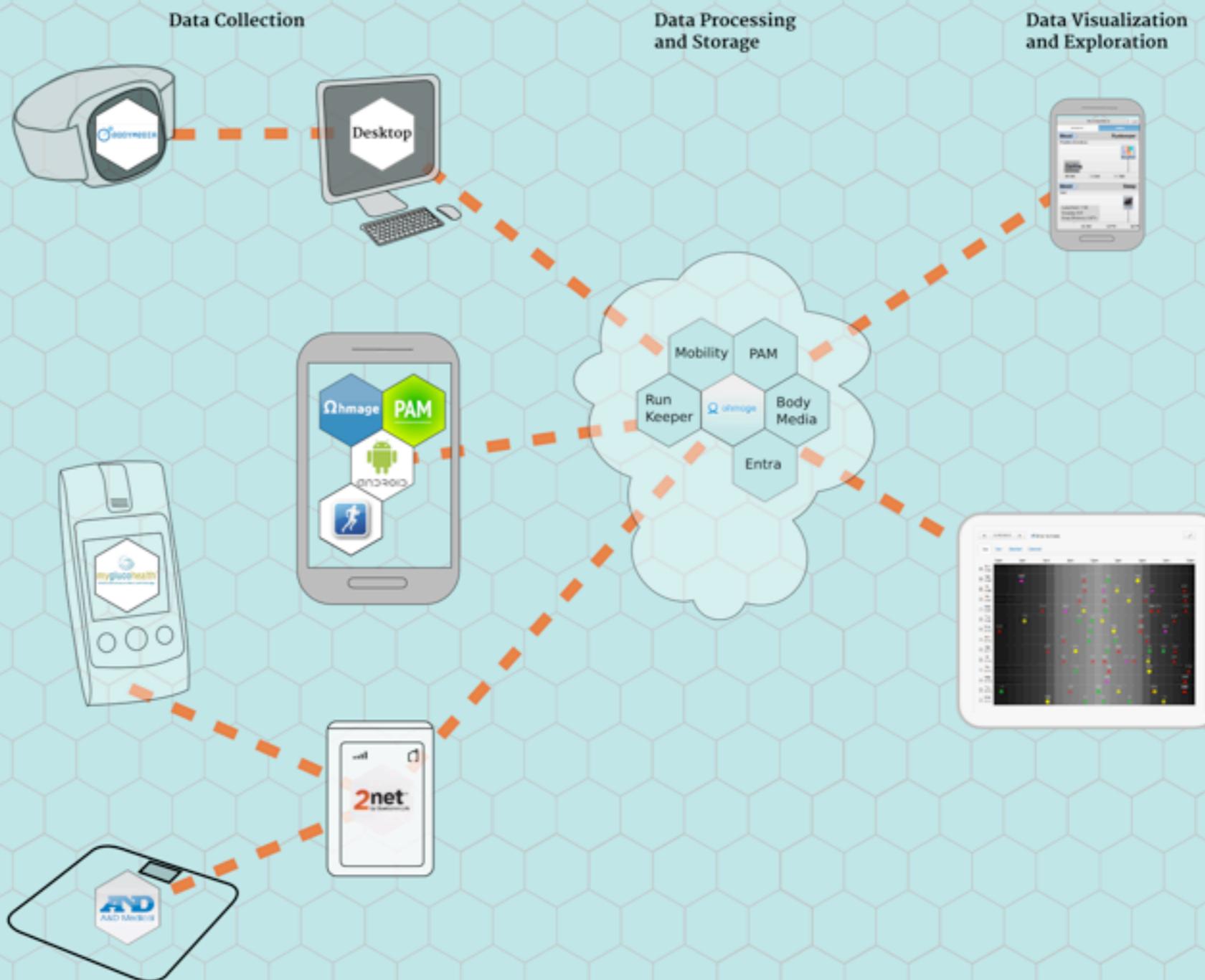
¹Kravitz, et al. Contemp Clin Trials 2009; 30:436-445

individual

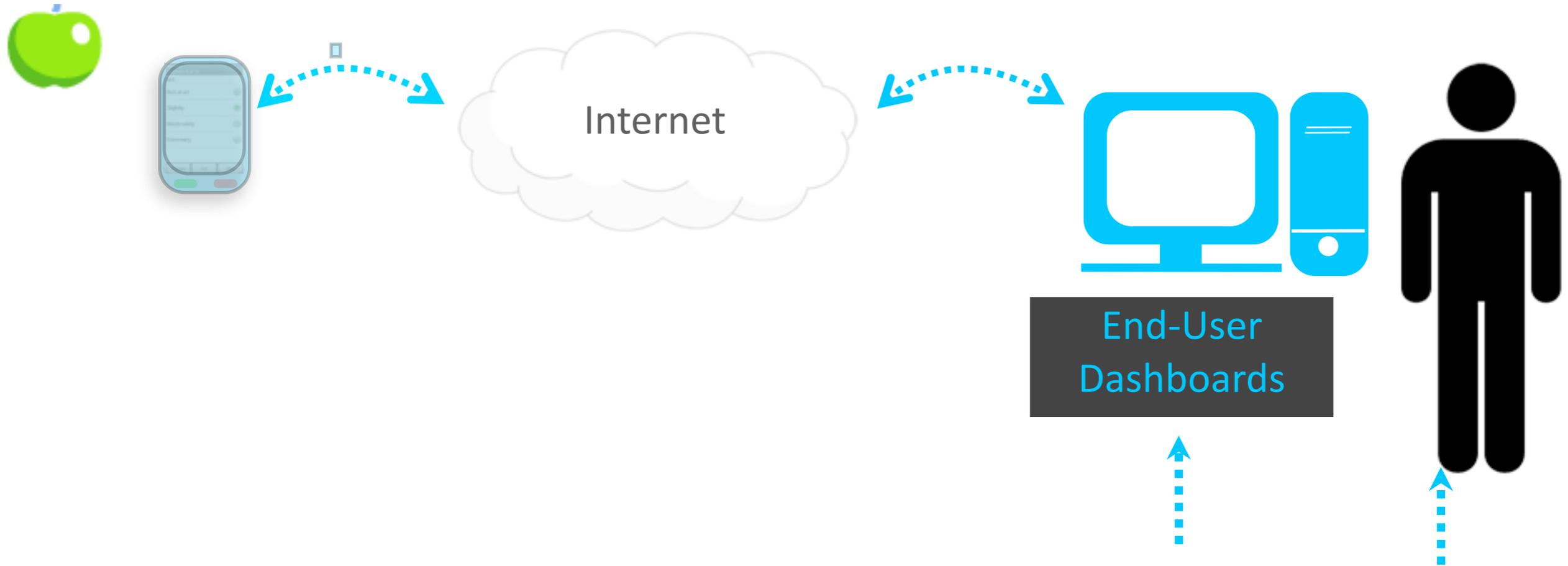
Sim, Kravitz

Open mHealth co-innovation use case: Diabetes

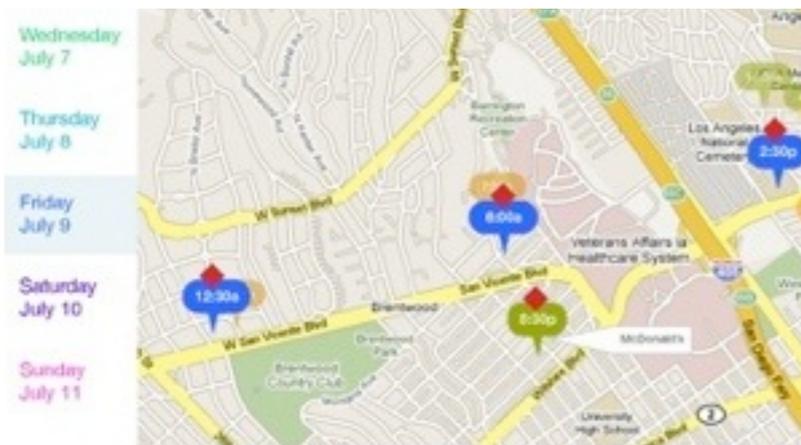
Open mHealth Case Study Diabetes Scenario



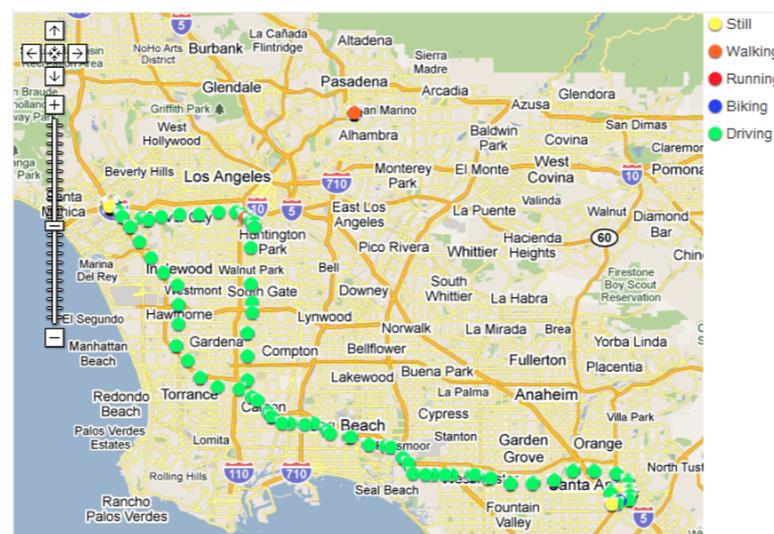
Not just a mobile app: data analysis, sensemaking, as critical and more challenging



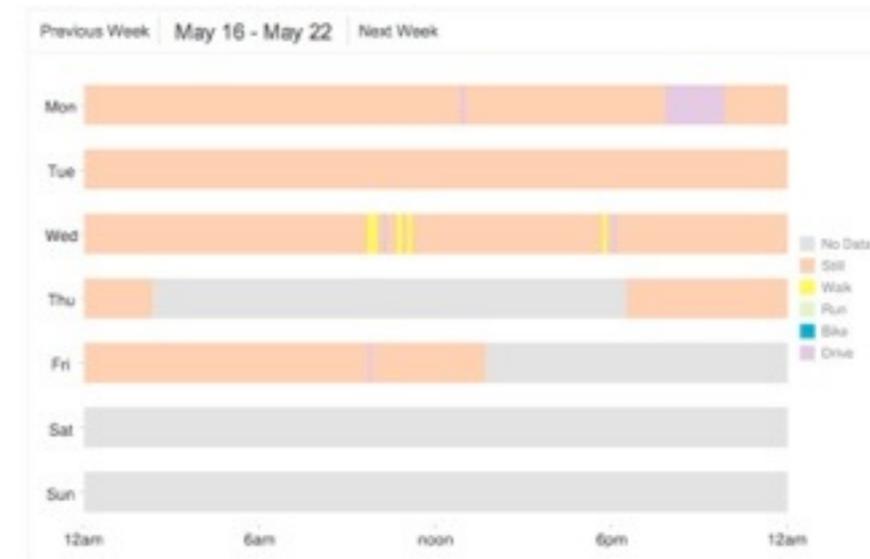
Correlations in time and space



Actigraphy over space

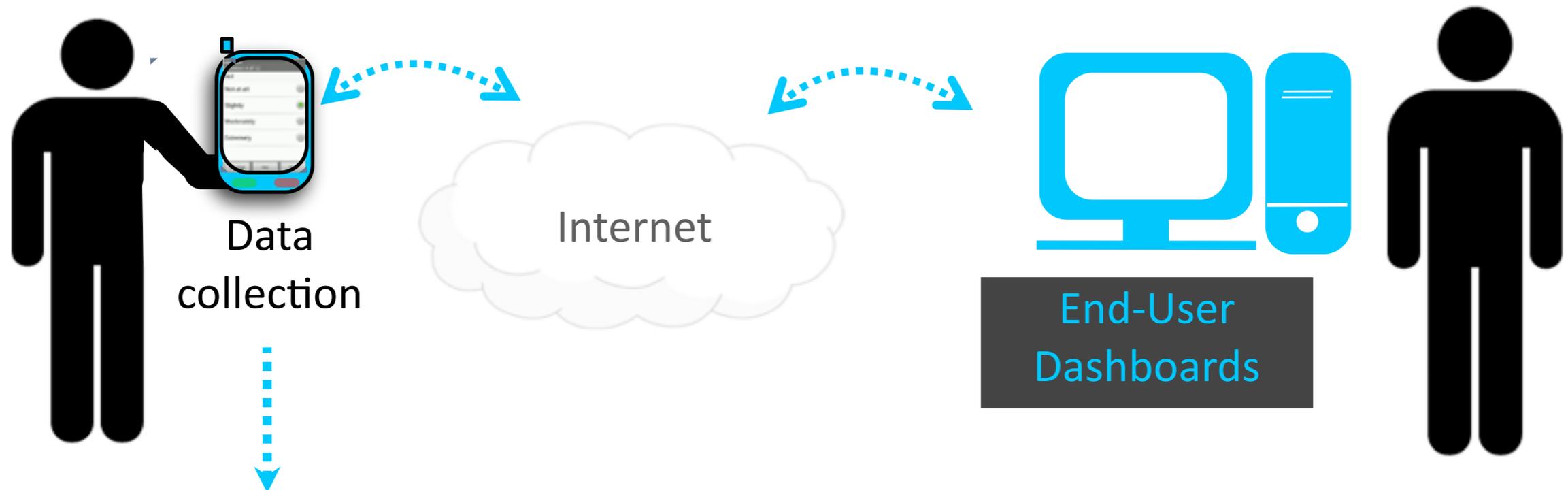


Actigraphy over time



Ramanathan, Selsky, et al

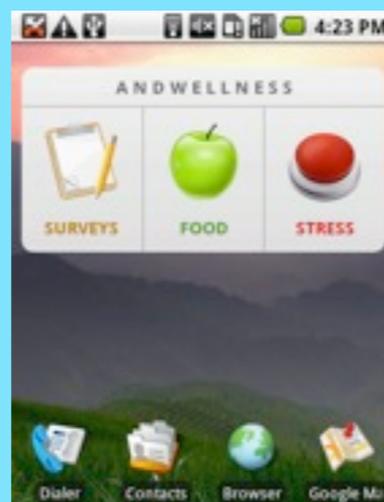
Many features apply across applications



Self Report (EMAs)

Multiple choice
Scale
Free text
Image capture
Personalization

Phonetop Buttons



Passive Monitoring

GPS, Wifi, Accel sms, calls, calendar, social media



actigraphy,
mobility, comm

Phone-based apps

Exercises/tools
Interventions
Games
Assessments

Ramanathan, Selsky, et al

“Real Sensor” streams



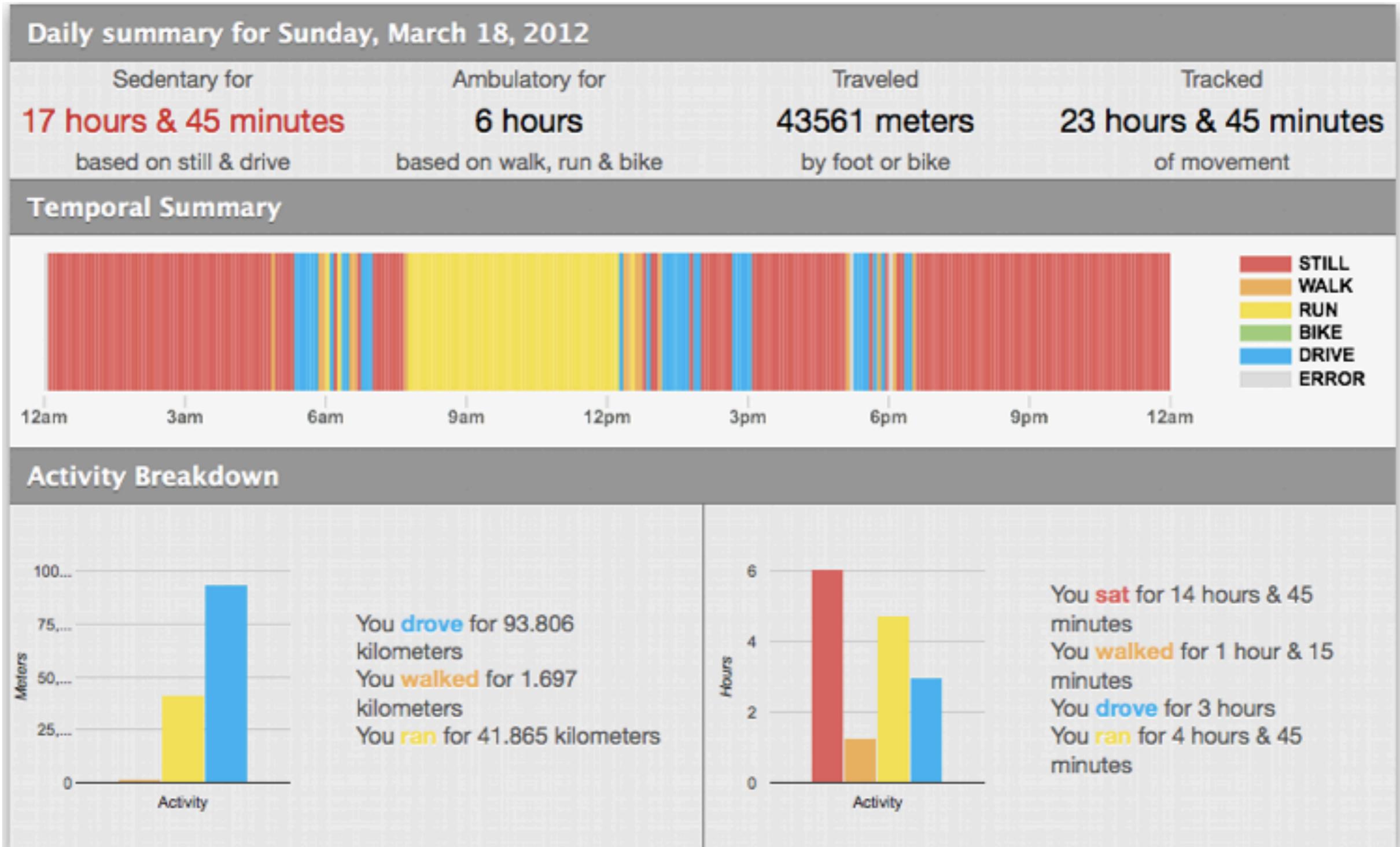
http://ginger.io/the-platform/



Last Name	Number	Status	Phone	Surveys	Contact	GBI
Battle, Ginger	W1ZP5XT	GOOD	✓	✓	📞 📄	
Blackburn, Vilhelmina	0RTBBHZ	BAD	✓	✓	📞 📄	
Dobos, Lore	ZR8K1FG	GOOD	✓	✓	📞 📄	
Evita, Et	0S6WQD	GOOD	✓	✓	📞 📄	
Garner, Angeliki	08Z1SRW	GOOD	✓	✓	📞 📄	
Gentle, Glead	38FX7TW	GOOD	✓	✓	📞 📄	
Gulyis, Hedvika	LGPFNQI	BAD	✓	✓	📞 📄	
Hoffman, Philippe	T4A7BMS	BAD	✓	✓	📞 📄	
Makar, Cenek	NB83XUJ	GOOD	✓	✓	📞 📄	
Muiryan, Dominicus	PKW7WFB	GOOD	✓	✓	📞 📄	
O'donnell, Charlotte	ZCZYULE	GOOD	✓	✓	📞 📄	
Overton, Theodora	ZYCBJQZ	GOOD	✓	✓	📞 📄	
Righi, Monique	CSSR77X	GOOD	✓	✓	📞 📄	
Roy, Weland	FORXZV	BAD	✓	✓	📞 📄	

<http://ohmage.org>

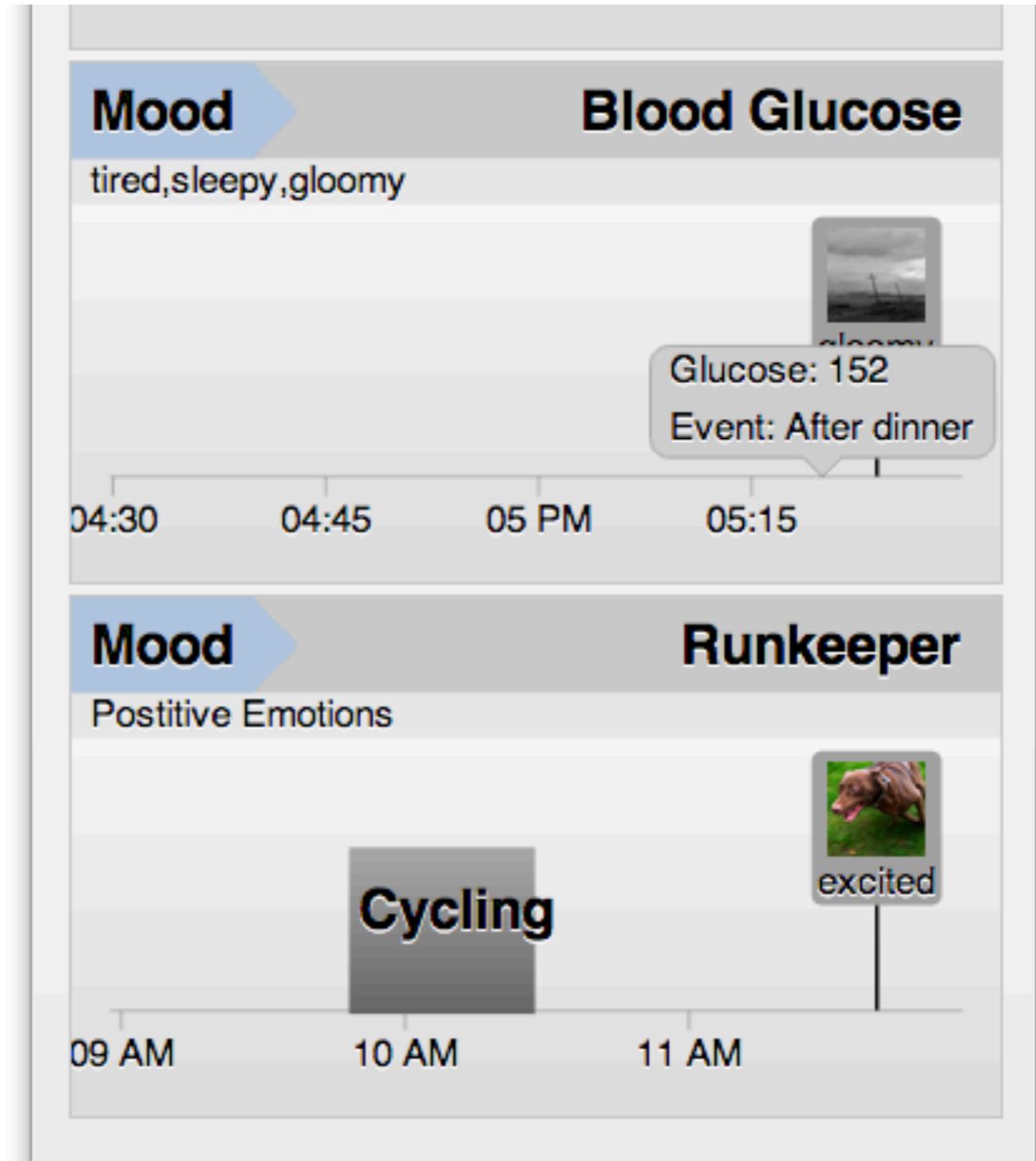
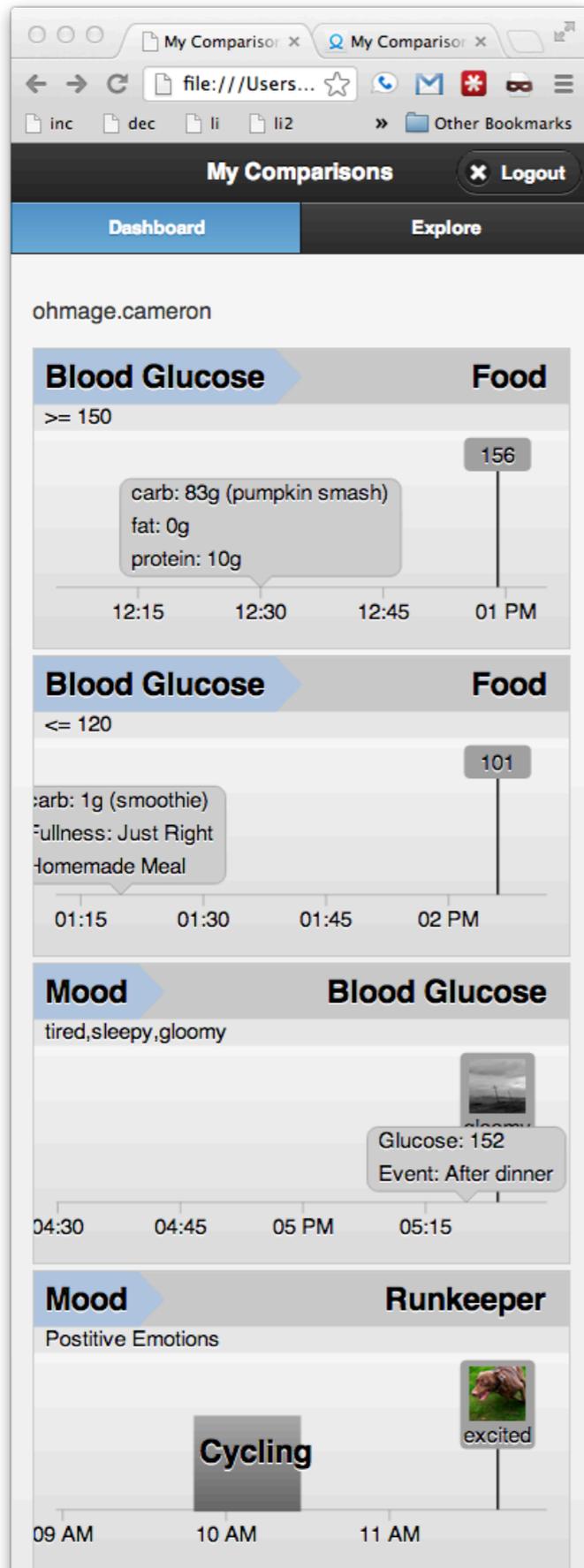
continuous activity, location traces and prompted self-report



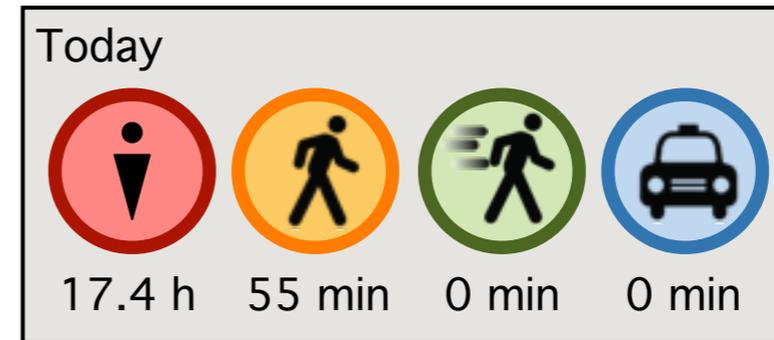
Photographic Affect Meter: PAM (Pollak et al)



My comparisons feedback screen



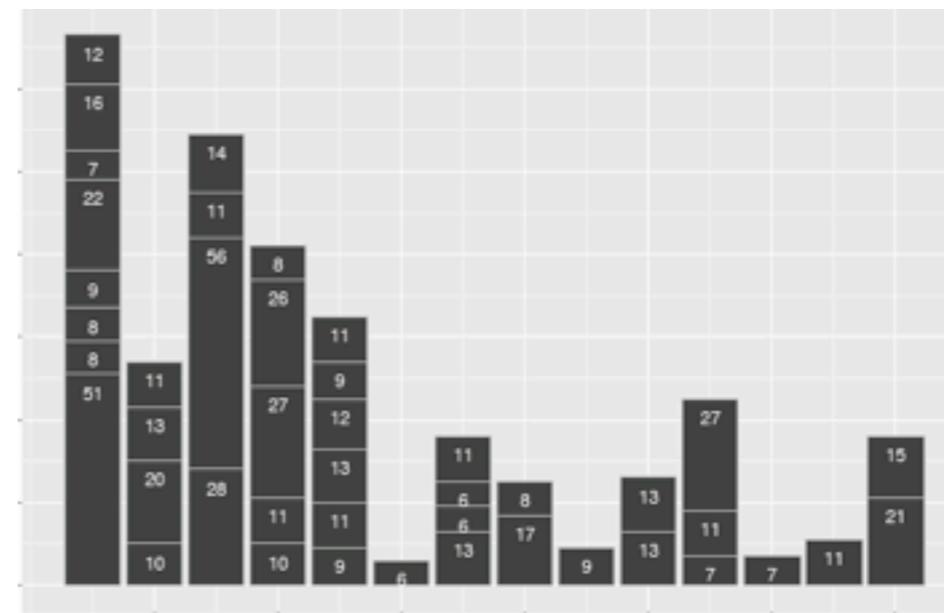
Next steps: transforming passive information into behavioral biomarkers for chronic diseases



Hours at home per day

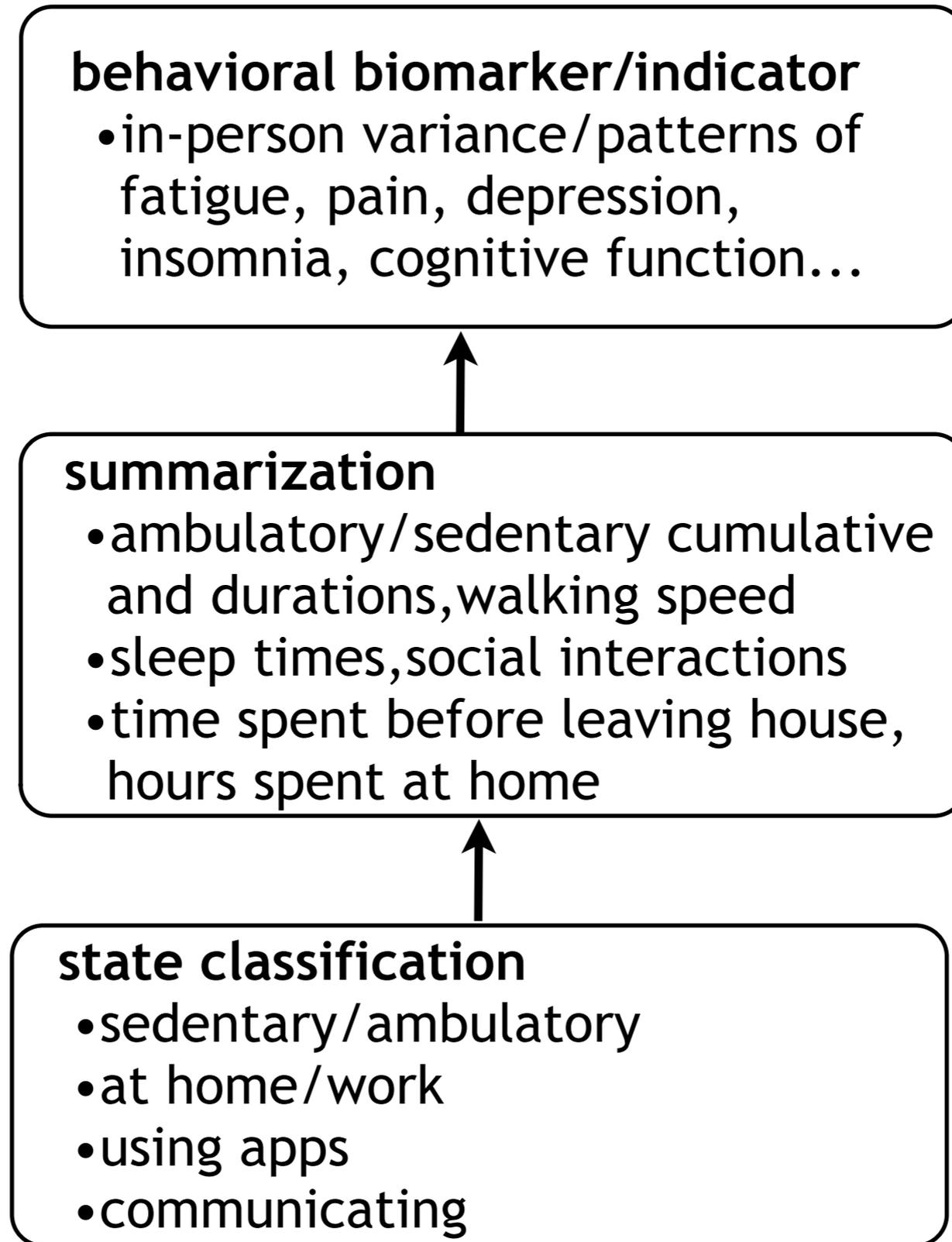


Walking periods > 6 min per day

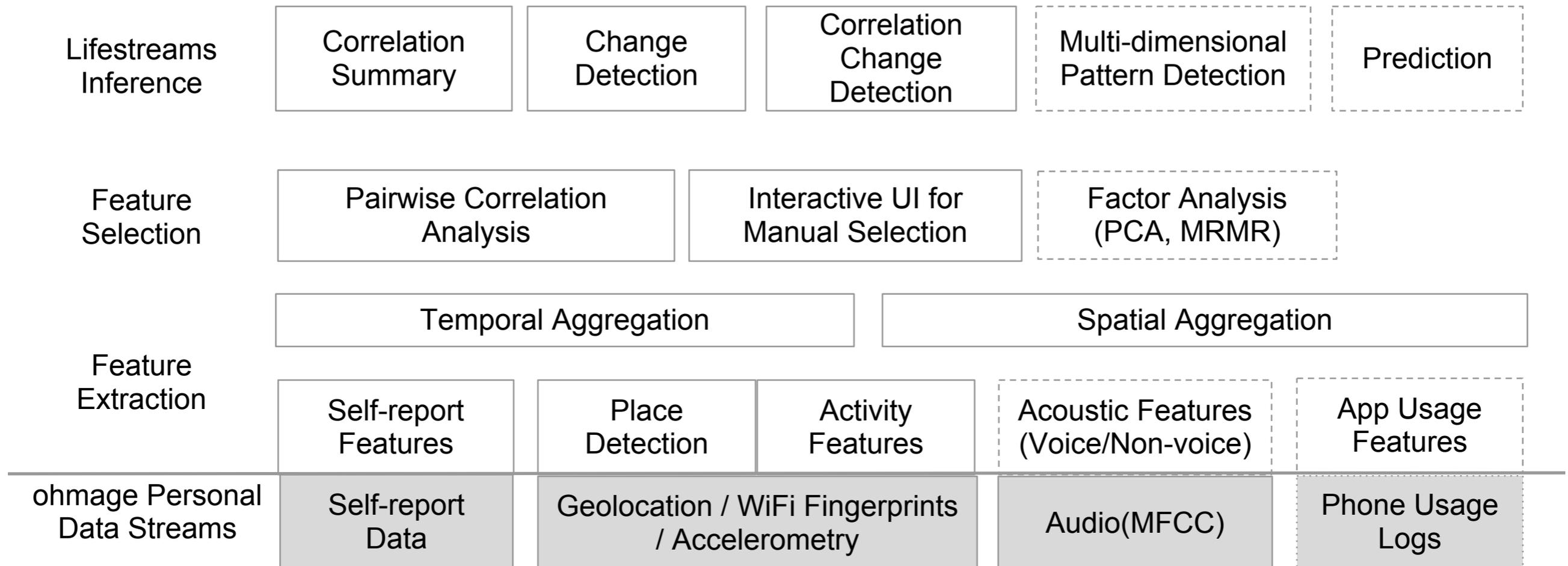


**Personalized, rapid, indicators of
health improvement, relapse, side-effects, symptoms using:
mobility logging and digital traces (vocabulary, games, spending)**

Deriving behavioral biomarkers...from app and sensor streams



Lifestreams: Modular Data Analysis Software Stack

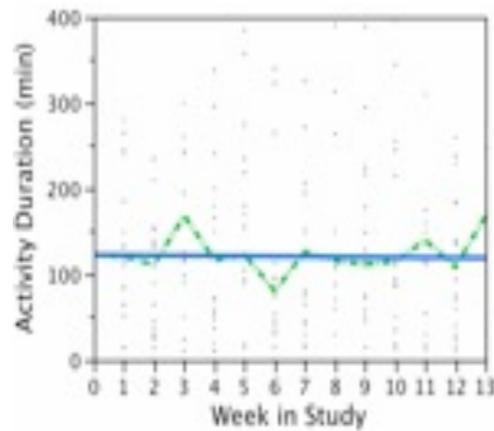


Hsieh, Tangmunarunkit, Ramanathan, et al

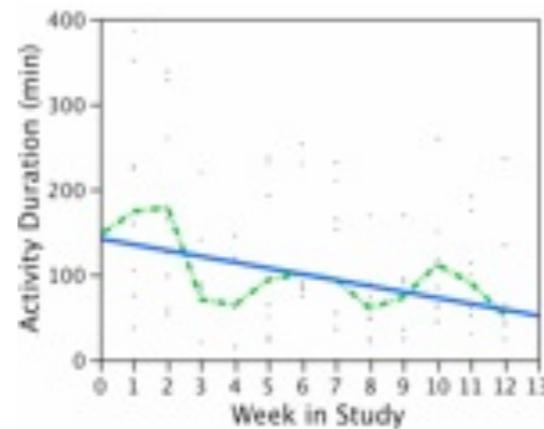
Behavioral Biomarkers can drive tailored infographics, informational incentives, feedback, game mechanics

ubifit participants who...

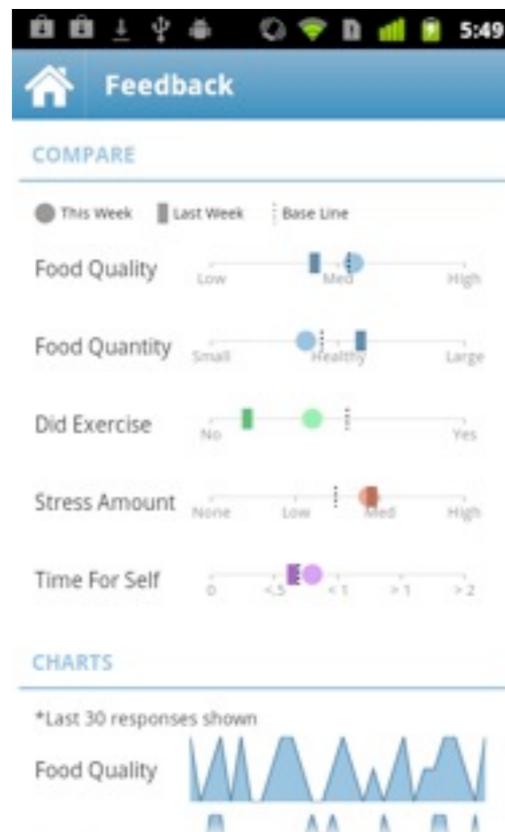
had the garden



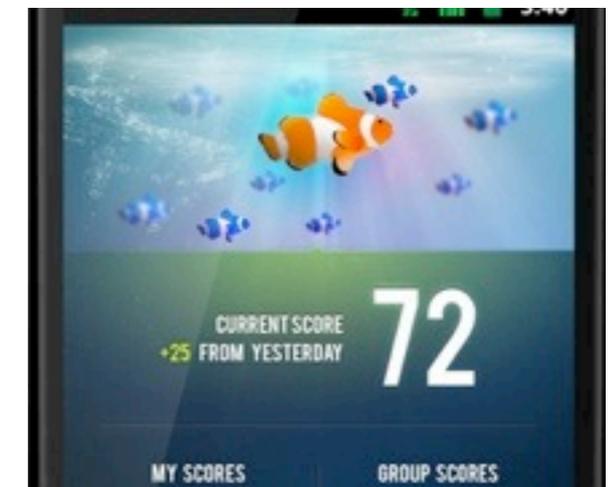
did NOT have the garden



ubifit
(S. Consolvo et al,
UW/Intel)



NIH funded new-moms study in progress
(Ramanathan, Ketcham, Estrin...)



Mobile Ambient Wellbeing Display
(T. Choudhury, Cornell)

Data reduction, selective sharing, privacy

usability and privacy
served by sharing only
extracted data features

TMI (too much information) filter

New privacy models emerging
for raw data

[http://www.weforum.org/issues/
rethinking-personal-data](http://www.weforum.org/issues/rethinking-personal-data)

<http://personal.com>

<http://wethedata.org>

behavioral biomarker/indicator

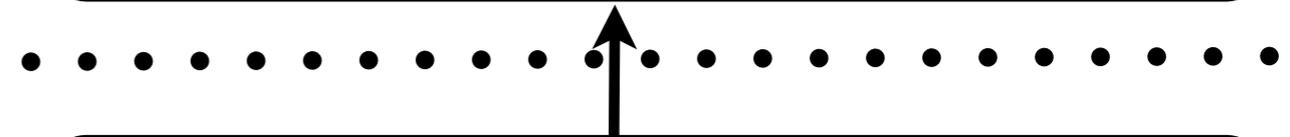
- in-person variance/patterns of fatigue, pain, depression, insomnia, cognitive function...

summarization

- ambulatory/sedentary cumulative and durations, walking speed
- sleep times, social interactions
- time spent before leaving house, hours spent at home

state classification

- sedentary/ambulatory
- at home/work
- using apps
- communicating

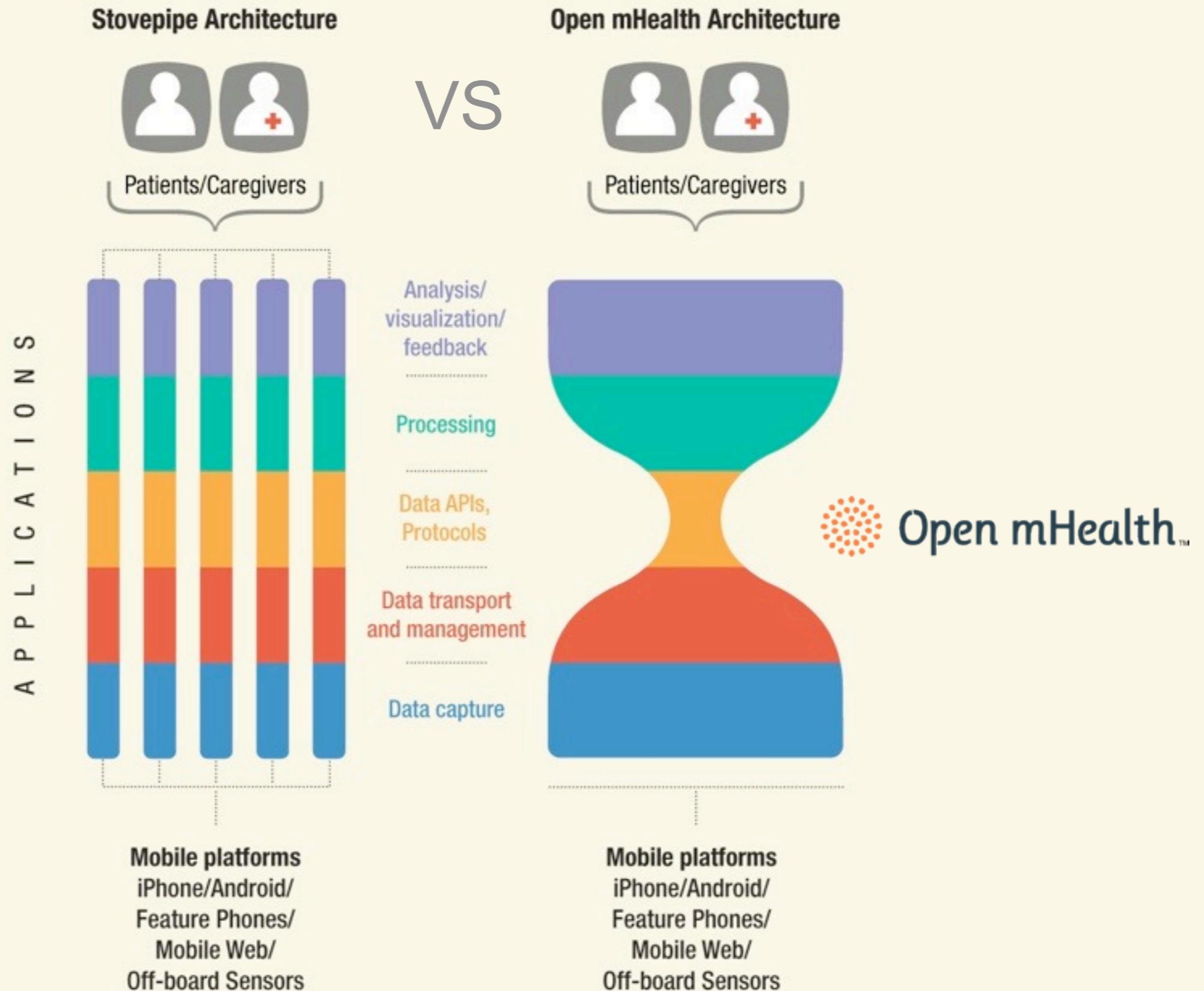


Third IT pillar of personalized, precision, medicine

“Big data”
(EHRs,
Web mining) + Omics + “small data”
(mHealth,
digital traces)

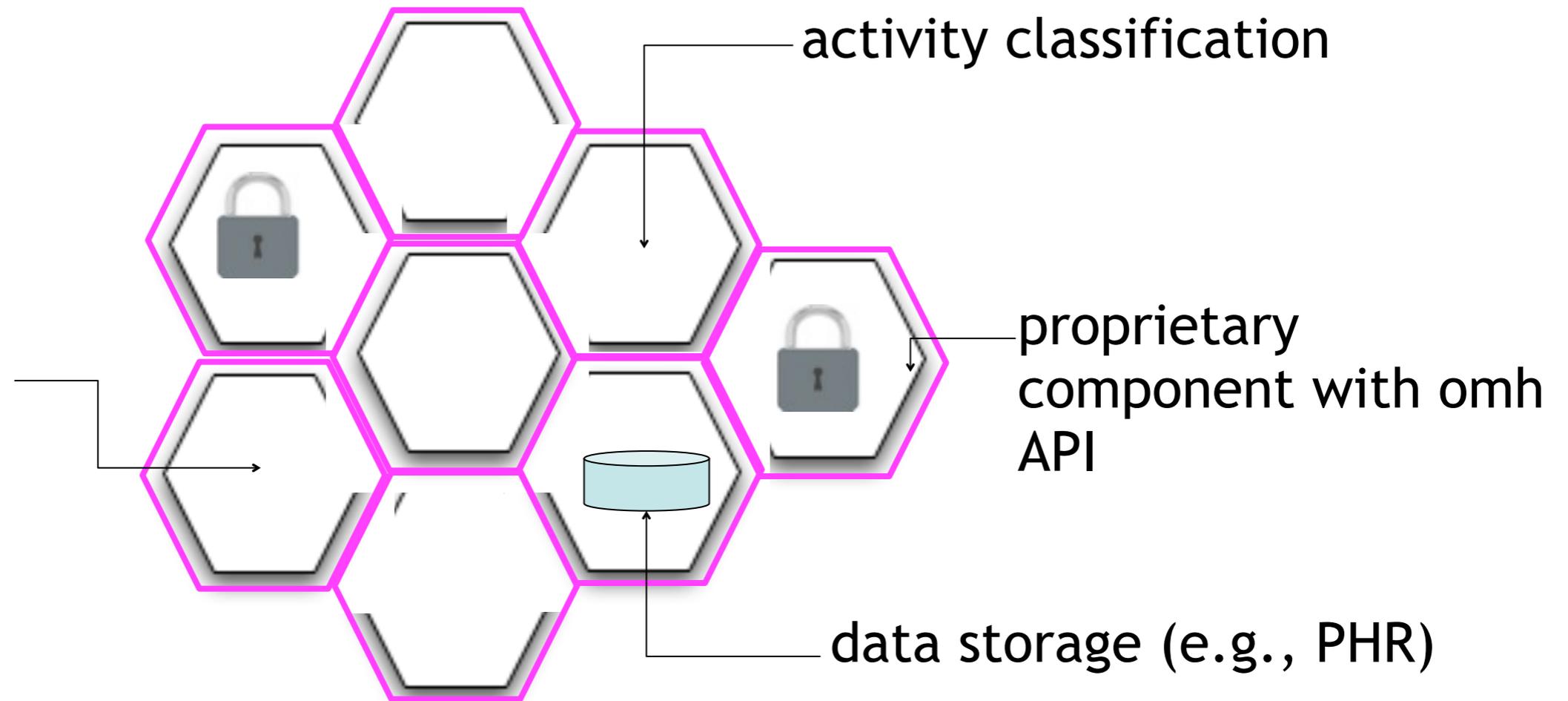
Open architecture and community

so mHealth solutions can integrate best available apps and techniques



open architecture for mobile health

graphing
significant
changes in
mobility



a small set of common principles/practices by which these modules are described and interface to one another

Estrin, Sim. *Science* 2010; 330:759-760
Chen, et al. *JMIR* 2012; 2012;14(4):e112.

Open mHealth light-handed approach to semantics

Foster broad ecosystem of software components that can process or visualize a single payload ID

- competition, different algorithms, new approaches ...
- after all ... this is all 'new stuff' ... processes and products will be iterating rapidly as we learn

Enable a data standard process that supports rapid evolution

- fits the desired and unavoidable dynamics of a *learning healthcare* ecosystem

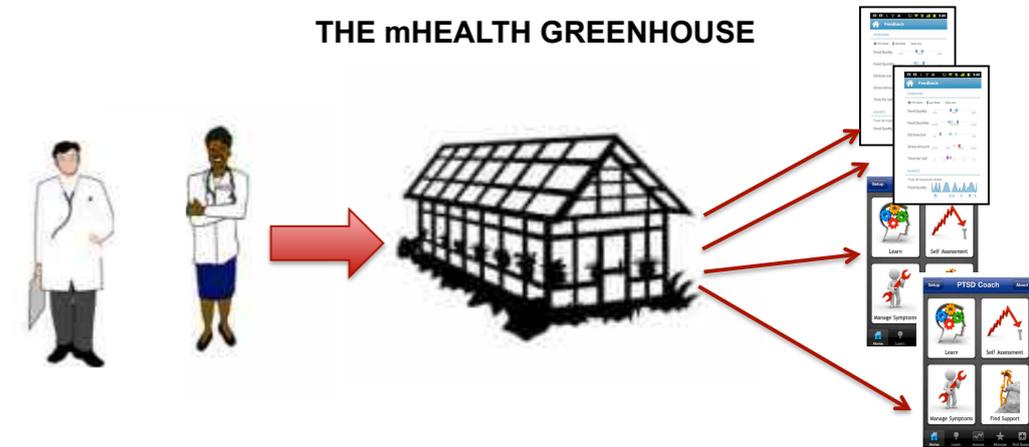
Data payload defined by Schema ID, version, lightweight schema

- utilize payload IDs to represent existing standards as well
- accommodate both existing software and existing standards

Whats next ...

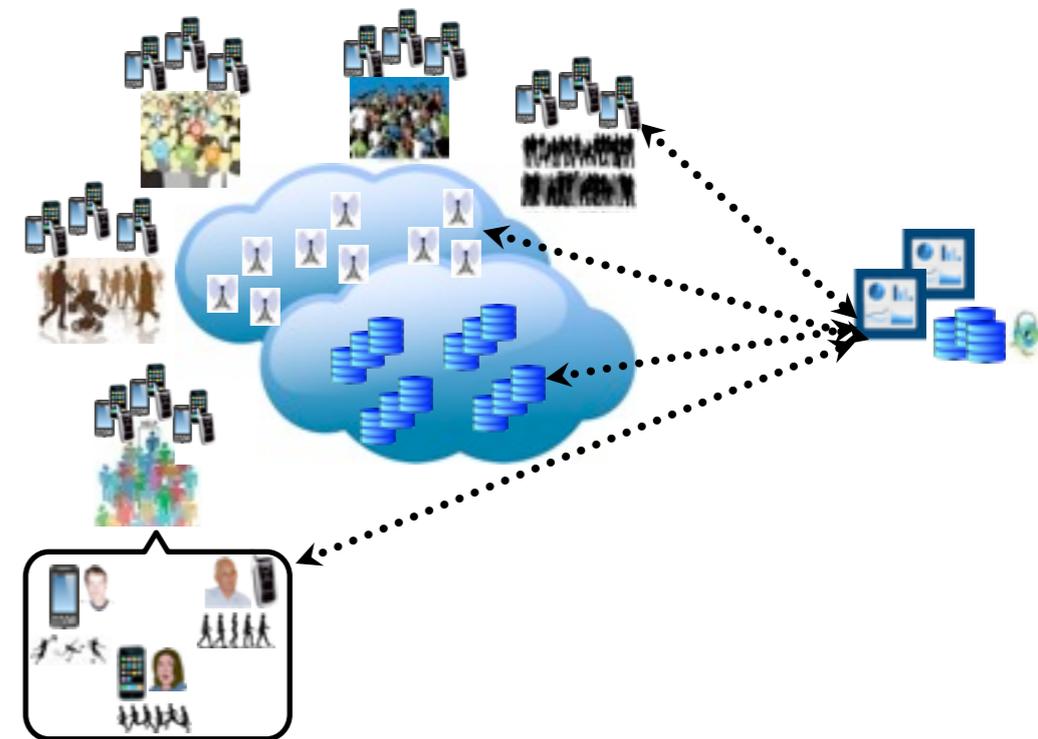
mHealth Greenhouse

- collaborate with innovative clinicians to develop new ways to address patient care by leveraging mobile data: behavioral biomarkers
- tools to support patients' disease-management
- support rapid, iterative prototyping and piloting



mpire: mobile personal informatics research and experimentation

- Open up programmatic access to individuals to obtain their personal digital traces from mobile, search, social, e-commerce, games, apps
- Personal Data APIs to foster personal services/apps
- Testbed in NYC with access to 1000's of mobile subscribers for experiments with privacy and detailed analytics



End to end arguments in systems-*research* design: *a case for including authentic applications in experimental systems research*

Original argument [Saltzer, Reed, Clark, 1981]

- “...functions placed at low levels of a system may be redundant or of little value when compared with the cost of providing them at that low level.”
- “The argument appeals to application requirements, and provides a rationale for moving function upward in a layered system, closer the application that uses the function.”

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Implications for systems-*research*/innovation

- authentic applications needed as part of systems research exploration to keep functional and performance requirements on a purposeful track
- need to build...and use...

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More general lessons from systems

- architecture, modularity, well-defined interfaces, analytics are critical
- enable rapid, iterative, automated, learning and sharing across applications, institutions, markets
- importance of shared robust open infrastructure

Acknowledgments: Collaborators and Sponsors

Collaborators

Technology Collaborators :

Deborah Estrin, Tanzeem Choudhury, Mark Hansen, JP Pollak, Nithya Ramanathan, Joshua Selsky, Hongsuda Tangmunarunkit

Application/domain expert faculty/PIs (Health science):

Jessica Anker, Jacqueline Casillas, Mary Charlson, Scott Comulada, Julia Hoffman, Patricia Ganz, Cary Reid, Mary Jane Rotheram-Borus, **Ida Sim (UCSF)**, Fred Sabb, Katie Shilton, Dallas Swendeman, Michael Swiernik

Students, Staff:

Staff: Betta Dawson, John Jenkins, Cameron Ketcham,
Graduate students: Faisal Alquaddoomi, Hossein Falaki, Brent Flagstaff, Andy Hsieh, Jinha Khang, Donnie Kim, Min Mun, Sasank Reddy, Vids Samanta

Sponsors and Partners/Collaborators

Cornell: Cornell Tech, WCMC CHiP

UCLA: CENS, GCFC, CHIPTS, Health Sciences, JCCC

Federal funding: NSF STC and CISE, NIH

Corporate funding: Google, Intel, MSR, Nokia, T-Mobile

Foundations/NGOs: RWJF, CHCF, CRA