Understanding Self-Reflection: How People Reflect on Personal Data through Visual Data Exploration

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Self-monitoring

An activity of recording one's own behaviors, thoughts, or feelings

[Kopp, J. (1988) Self-monitoring: A literature review of research and practice]



Self-monitoring from the 19th century





Public scales from the late 1880s in contemporary Paris (from Crawford 2015)

HIII	M	+	W	th	F	S	-0:-	
SLeep	7	7,5	6.5	9	9	8	7.5	
exercise	35min	45min	30min	30min	60min	55min	(yoga)	
feeling-s	happy 4 relaxed	bored, unsanched	TIRED	refrished happy	excited, chipper	Laid Back	calminized organized	
symptom	neadache- 3pm ish		headache 2pm ish					
Notes		Lots of mindless work- today	2nd head- oche HIIS week, both after apple + PBsndck	went to bed early + am. workout	naif day at work!	first saturday w/ no plans in weeks	got a lot of life admin done	
						Mental Heal ttp://asweat	th Tracker. tlife.com/2016/08	3/ideas-fitne

http://asweatlife.com/2016/08/ideas-fitness-bullet-journal/











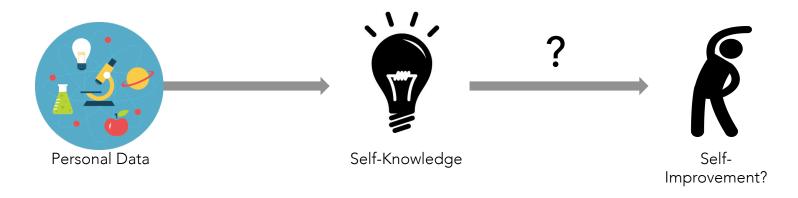




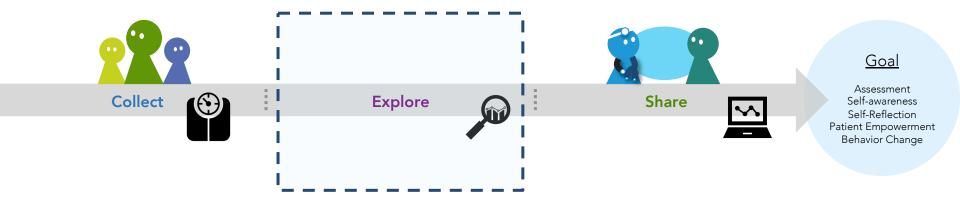


Promises

External measurement to self-knowledge Self-knowledge to self-improvement?



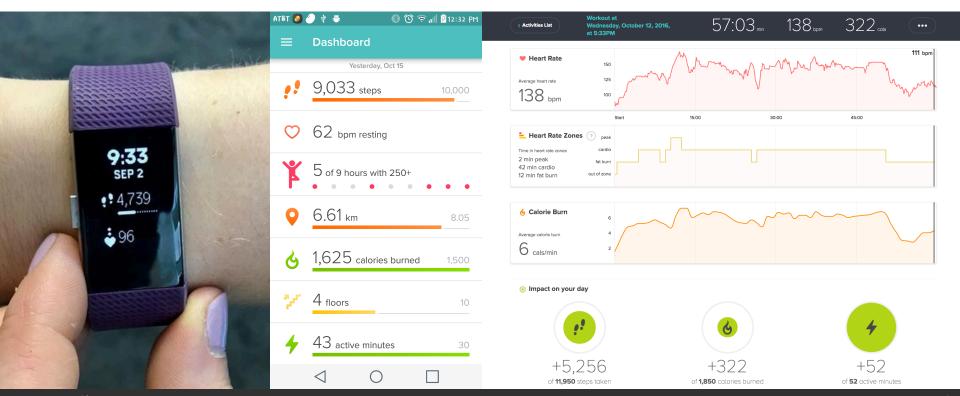
Human-Data Interaction for Self-Monitoring



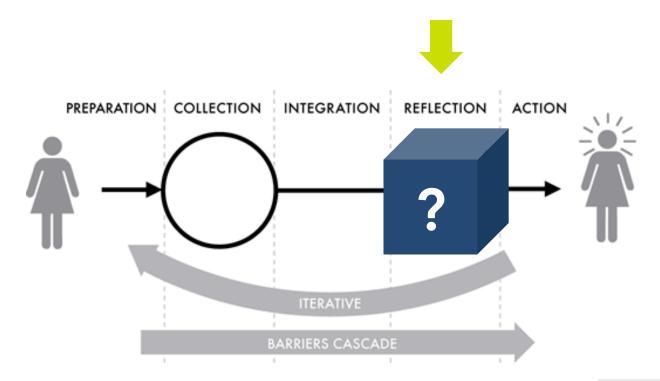
Fawcett (2015)

data exploration and analytics capabilities for personal data analysis "remain surprisingly primitive, leaving the analytical heavy lifting to the end user"...

Personal data visualization

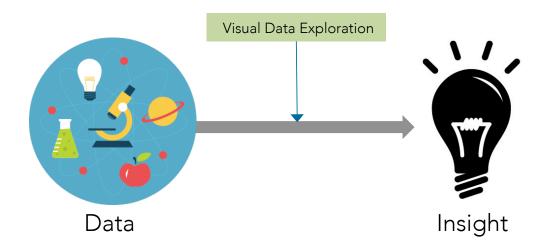


Stage-based model of PI



Visual data exploration

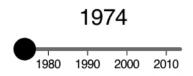
Powerful way to help people reveal meaningful insights about themselves and to facilitate self-reflection

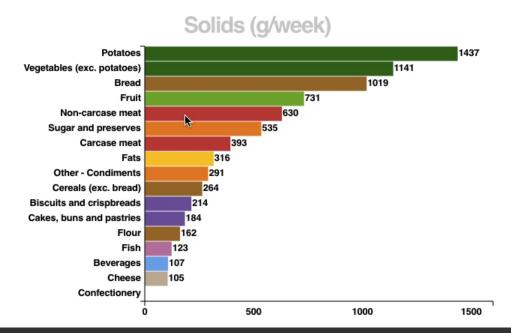


The British Diet http://britains-diet.labs.theodi.org/

✓ Trends • ¶ ¶ Typical diet

Overview · Meat · Fish · Dairy · Veg · Fruit · Carbs · Fats · Cupboard · Treats · Drinks



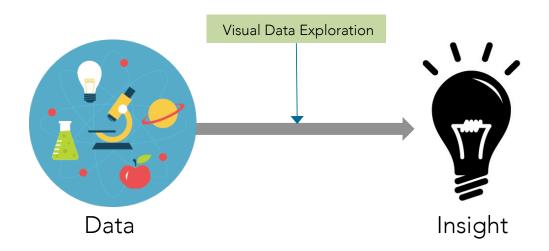


Limited support for data exploration

- Scattered data across multiple platforms (Li et al., 2011; Choe et al., 2014.)
- Don't know what to do with the data (Choe et al., 2014; Epstein et al., 2015; Lazar et al., 2015.)
- Difficult to translate questions into data attributes (Grammel et al., 2010; Huang et al., 2015.)
- Difficult to construct visualizations (Grammel et al., 2010; Huang et al., 2015.)

Research questions

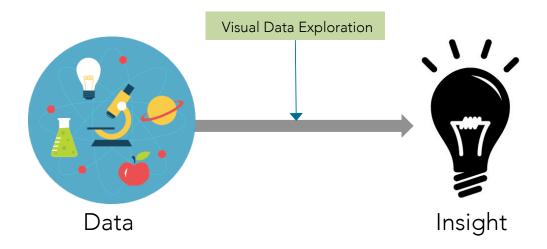
RQ1: How do people reflect on their self-tracking data? (Process)



Research questions

RQ1: How do people reflect on their self-tracking data? (Process)

RQ2: What insights do people gain from visual data exploration? (Outcome)



Insights

A key purpose of visualization Card et al., 2005

"An individual observation about the data by the participant, a unit of analysis" Saraiya et al., 2005

Characteristics of insights North, 2006

Insight gaining process Vietal., 2008

Types of personal insights

[IEEE CG&A 2015]

Personal Visualization and Personal Visual Analytics

Characterizing Visualization Insights from Quantified Selfers' Personal Data Presentations

Eun Kyoung Choe ■ Pennsylvania State University

Bongshin Lee Microsoft Research

m.c. schraefel = University of Southampton







Bongshin Lee



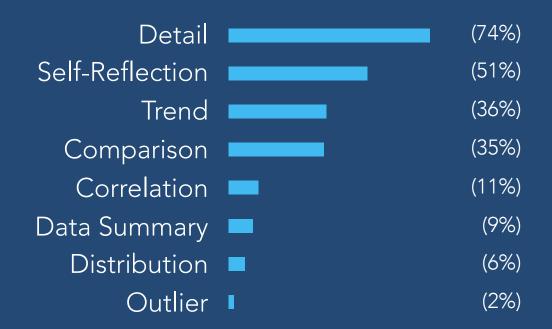
m.c. schraefel

30 video recordings of QS presentations



s a result of advances in self-monitoring as creating a line chart with two lines to convey a technology and the prevalence of low-cost correlation), and they sometimes present insights

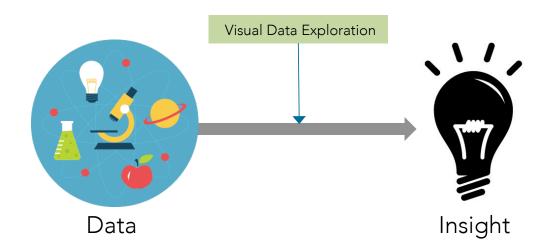
Visualization Insights



Research questions

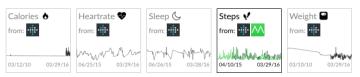
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● Visualized Self IMPORT TREND COMPARISON CONTEXT PLACES Signed in as ekdata. LOG OUT

Data Sources

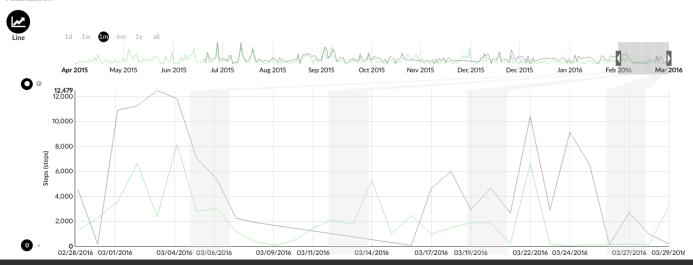


Visualized Self

Data Summary

Overall	Selected	✓ Fitbit	✓ Moves
Dates: 04/10/2015 - 03/29/2016	Dates: 02/27/2016 - 03/29/2016	Dates: 02/27/2016 - 03/29/2016	Dates: 02/27/2016 - 03/29/2016
Range: 6.0 - 27488.0	Range: 14.0 - 12479.0	Range: 73.0 - 12479.0	Range: 14.0 - 8175.0
Average: 4870.5	Average: 3184.8	Average: 5073.0	Average: 2046.3
Total: 1724140.5	Total: 98729.0	Total: 121752.0	Total: 63436.0

Visualization



● Visualized Self IMPORT TREND COMPARISON CONTEXT PLACES Signed in as ekdata. LOG OUT

Data Sources



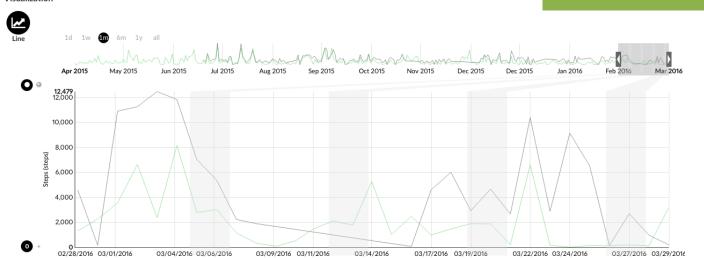
Design Rationales

Data Summary

Overall		Selected		✓ Fi	tbit	✓ M	loves	^^	į
Dates: 04/1	0/2015 - 03/29/2016	Dates: 02/27	/2016 - 03/29/2016	Dates:	02/27/2016 - 03/29/2016	Dates:	02/27/2016 - 03	/29/2016	į
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Total:	1724140.5	Total:	98729.0	Total:	121752.0	Total:		63436.0	ì

1. Support data exploration for the general public

Visualization



Wisualized Self









Signed in as ekdata. LOG OUT

Data Sources



Design Rationales

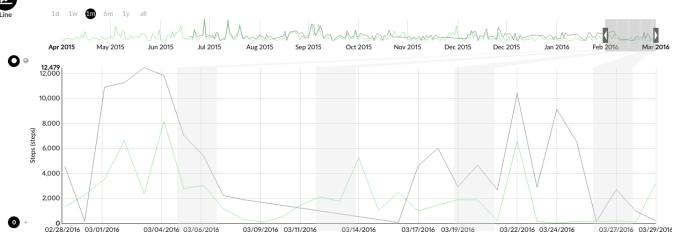
Data Summary

Overall		Select	ed	✓ F	Fitbit	✓ N	loves	^^	:
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Range:	6.0 - 27488.0	Range:	14.0 - 12479.0	Range	: 73.0 - 12479.0	Range:		14.0 - 8175.0	i
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2. Design for a personal data context

Visualization





Data integration from multiple sources





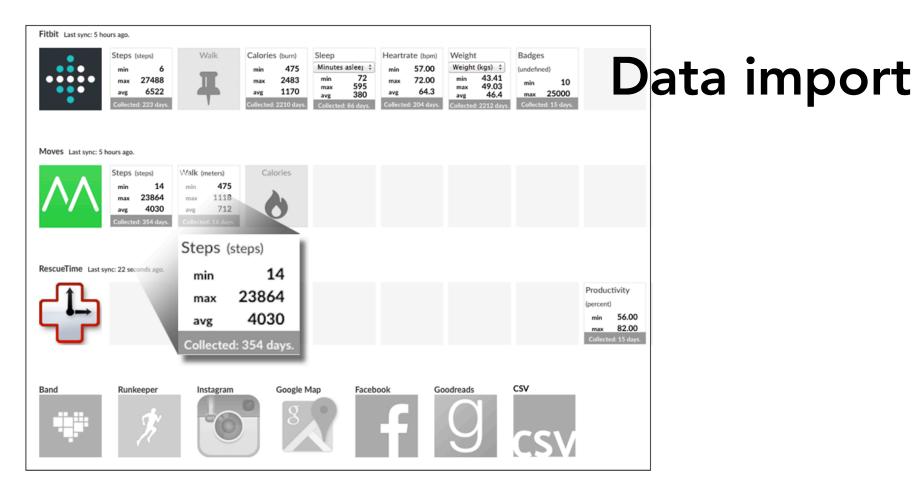


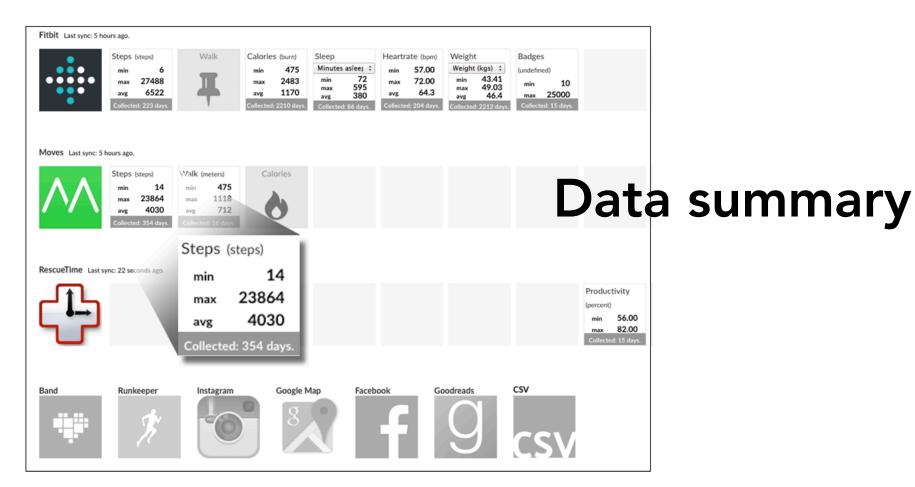


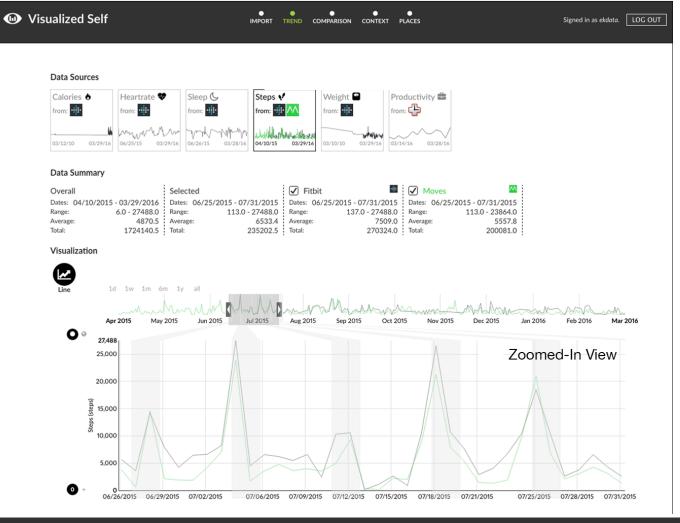


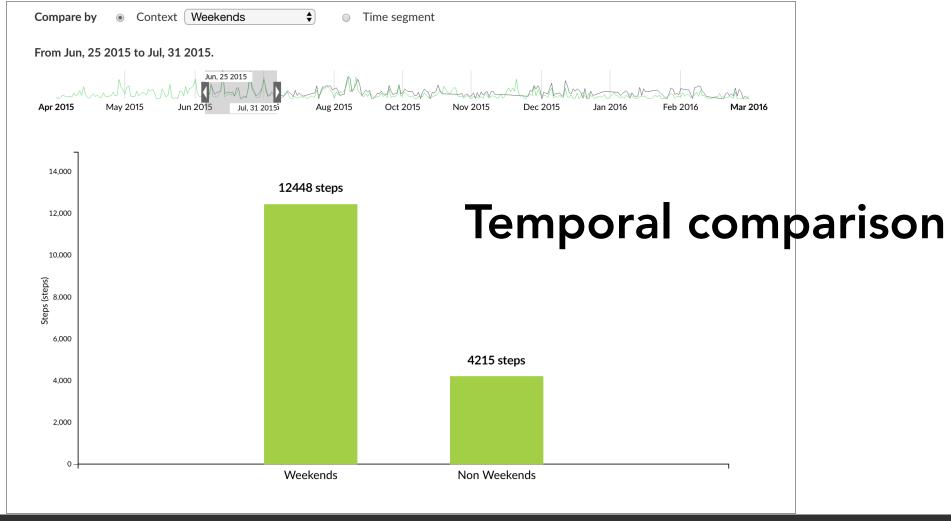




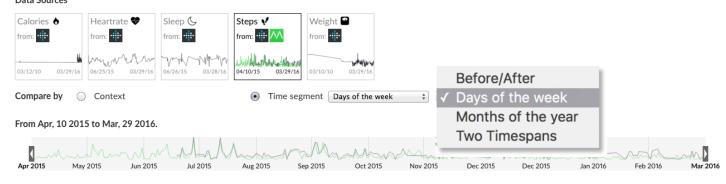


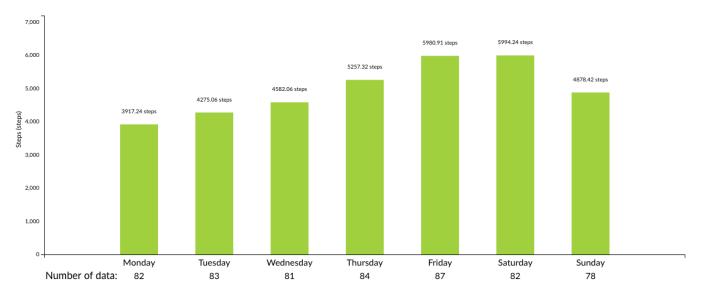






Data Sources





Study Session



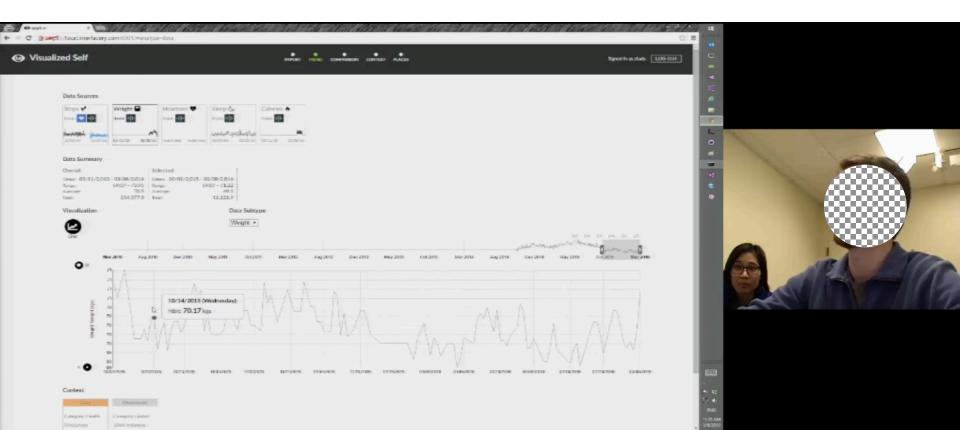
Invited 11 self-trackers to the lab

- Have been regularly tracking personal data for two months or longer
- Have been using two or more of the following devices or apps: Fitbit, Aria, MS Band, Moves, RunKeeper, RescueTime
- Age range: 24-60 (mean = 35.8)

Study session (1.5–2 hours total)

- Demographic / tracking experience survey
- Tutorial and demonstration of the tool
- Think-aloud session with observation
- De-briefing interview

What a session looks like



Data Analysis

- Transcribed the think-aloud session & debriefing interview
- Open coding, axial coding on the process of self-reflection (RQ1)

Data Analysis

- Transcribed the think-aloud session & debriefing interview
- Open coding, axial coding on the process of self-reflection (RQ1)
- Directed contents analysis for the types of insights (RQ2)

Levels of Reflection

R0—description

R1—description with justification

R2—exploring relationships

R3—asking of fundamental questions

R4—considering social and ethical issues

Findings

Personal insight types

Table 2. Types of visualization insights. We adopted from [8], and then revised and expanded adding new insight categories.

Type (total frequency)	Subtype (frequency)	Description	Example Quotes		
Recall (327) External context (171) Confirmation (80)		Uncaptured data provided by the self-tracker to understand and explain a phenomenon shown in the data	"I think that was soon after my surgery and that maybe would make sense cause I'd have to get up to take medicine and maybe being restless or something." [P8]		
		Collected data confirms existing knowledge	"So for the most part I mean, this graph is informative in that it does usually take me long to fall asleep. So this is confirming what I alread know about." [P9]		
	Contradiction (76)	Collected data contradicts existing knowledge	"That was the opposite of what I was expecting. I would've expected that as I ate healthier I would've been burning more calories." [P9]		
Detail (257) Identify value (105)		Explicitly specify the measured value, its range for one or more clearly identified data points, or the difference between two measured values	"And it looks like my highest since I've started using it is 7,958. I wonde what date that was." [P7]		
	Identify extreme (87)	Explicitly state the identities of the data points pos- sessing extreme values of the measure variable	"Yeah, look at this peak. 11/2014 that was a trip to San Francisco." [P:		
	Identify refer- ences (65)	Explicitly state the values of categorical variables, labels from the axes, or legends	"It says I'm taking a lot of weekend steps. That's quite surprising." [P1]		
Comparison (168)	By time segmen- tation (111)	Compare measured values segmented by time	"I could see every month this year I'm improving." [P11]		
Multiple services* (26)		Compare the same data type from two or more services	"Yeah, so the—yeah, definitely interesting to see that the two devices gave very different trends." [P6]		
	Against external data (14)	Bringing in external data for comparison	"I have data for the same period for my HDL and my LDL and my tri- glycerides. My weight is important but those are just as important as well as to how those values are changing." [P10]		
By factor (12) Instances (5)		Compare measured values by a factor (other than time)	"What was happening in February 2015? [Laughter] I bet I can tell you what those are. That's one of my son's sleep regressions." [P1]		
		Compare two specific instances	"The plan says no exercise after 7:30. And these two data points is basically validating that you've got to stick by it." [P3]		
Trend (119)		Describe changes over time	"Then there's also a time when I had surgery actually both holidays. In 2013 I had shoulder surgery in December. Everything went down." [P8]		
Value judgment* (118)		Convey positive or negative connotations about the data	"It's pretty irregular. I really wish I woke up at the same time. This is really bad." [P6]		
Distribution (41)		Explicitly state the variability of measured values	"This is around—around here is when my son was born. Second kid. So, you know, some of this stuff—you can see trending up as you're no taking care of yourself. Son gets a little older, things go back to normal. [P1]		
	By category (0)	Explicitly describe the variation of measured values across all or most of the values of a categorical variable	By category (0)		
Correlation (34)		Specify the direct relationship between two variables (but not as comparison)	"So the Calorie data matches the Steps data." [P9]		
Outlier (28)		Explicitly point out outliers or state the effect of outliers	"The min and max are interesting. But from a global perspective, they're probably outliers on asleep." [P11]		
Data summar	y (27)	Summary of collected data (such as number of data points, duration of tracking, and averages)	"So about two years, and averaging over ten [10,000 steps]. My goal is set at ten [10,000 steps], so that makes me happy." [P4]		
Prediction (14)		Predict the future based on the collected data	"Yeah, Martin Luther King. So if I go to comparison of weekend I think my weekends tend to be sluggish. Weekdays are okay." [P3]		
Total (1133)		* New insight ty	pes identified in this study in comparison to Choe, Lee, and schraefel [8]		

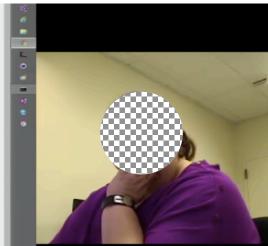
From a lower-level reflection to a higher-level reflection Data summary; Comparing

"Band has 222 days of collected data and it's saying my average is 1,464, but Fitbit has 81 days and it's saying I have 5,764 as my average. So it leads me to wonder which one is more accurate?" [P7]

multiple services; "R0" reflection







Insight gaining pattern #1

Visual data exploration



Recall previous contexts that could explain the captured behavior

"R1" reflection

Revisiting with explanation, descriptive reflection

External context

"I think that was soon after my surgery and that maybe would make sense cause I'd have to get up to take medicine and maybe being restless or something." [P8]

Insight gaining pattern #2

Recall previous contexts that could explain the captured behavior



Create an interesting question / hypothesis



Questioning; exploring relationships



Visually explore the data to look for an answer

Temporal comparison

P1: (entering Sept 15, 2015 to compare his weight before and after this date) Researcher: Why Sept 15?

P1: "That's kind of around the time I changed jobs. I was wondering if there was anything interesting there."

External context;
Comparison by time segmentation



Using External Context in Data Exploration

External Context: Uncaptured data provided by the presenter to understand and explain a phenomenon shown in the data

Calendar events, location semantics, major life events, key dates, vacation, workout types, seasons, weather...

Before/After

✓ Days of the week

Months of the year

Two Timespans

Value judgment:

"Saturday is pretty bad" [in terms of step count]



"R3" reflection

alters or transforms the reflector's original point of view

Making a resolution:

"So I need to take action to probably monitor myself to ensure that I'm at least at 2,000 [steps] or more." [P10]

Reflection on the levels of reflection

Many R0, R1, and R2 types of reflections due to Visualized Self's data summary and temporal comparison pages

Drawing higher-level reflections (i.e., R3) was less common

R3 might be an important reflection type that can potentially lead to short-term, or even long-term behavior change

Did not observe R4

Summary

Supporting self-reflection with VDE

Flexible data selection, filtering, and comparison features

Help people create interesting questions and hypotheses

Help people capture/use various contextual information

Combine system-driven and human-driven insights

Thank you!

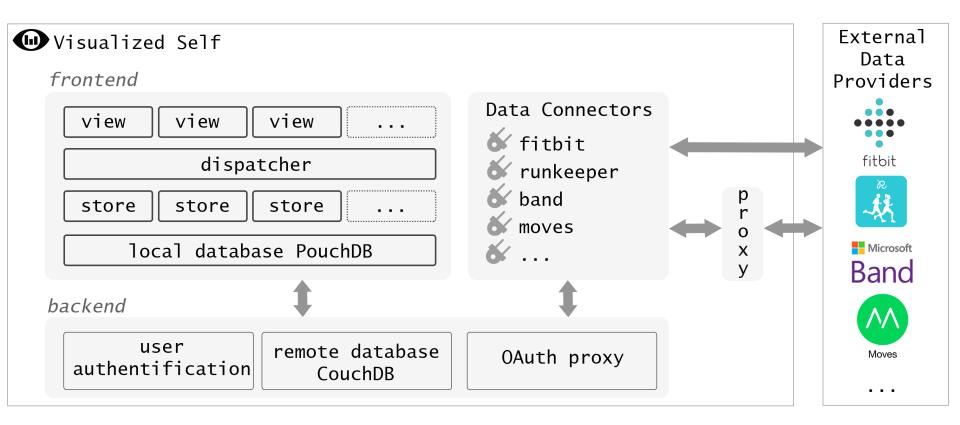
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Funding:

National Science Foundation Microsoft Research





Data ingegration from multiple sources

Fitbit Last sync: 53 minutes ago.

















Moves Last sync: 53 minutes ago.



Steps	(steps)
min	14
max	23864
avg	4030
Collected: 354 days.	





