Data Experiences: novel interfaces for data engagement using environmental health data

Laura J. Perovich

Submitted to the Program in Media Arts and Sciences, School of Architecture and Planning, in partial fulfillment of the requirements for the degree of Master of Science in Media Arts & Sciences at the Massachusetts Institute of Technology

September 2014

© 2014 Massachusetts Institute of Technology. All Rights Reserved

Author:

________________________
Program in Media Arts & Sciences
August 22, 2014

Certified by:

_________________________
Dr. V. Michael Bove, Jr.
Principal Research Scientist
Media Lab
Massachusetts Institute of Technology

Accepted by:

_________________________
Prof. Patricia Maes
Interim Academic Head
Program in Media Arts & Sciences
Massachusetts Institute of Technology
Data Experiences: novel interfaces for data engagement using environmental health data

Laura J. Perovich

Submitted to the Program in Media Arts and Sciences, School of Architecture and Planning on August 22, 2014, in partial fulfillment of the requirements for the degree of Master of Science in Media Arts & Sciences at the Massachusetts Institute of Technology

Abstract

For the past twenty years, the data visualization movement has reworked the way we engage with information. It has brought fresh excitement to researchers and reached broad audiences. But what comes next for data? I seek to create example “Data Experiences” that will contribute to developing new spaces of information engagement. Using data from Silent Spring Institute’s environmental health studies as a test case, I explore Data Experiences that are immersive, interactive, and aesthetic. Environmental health datasets are ideal for this application as they are highly relevant to the general population and have appropriate complexity. Dressed in Data will focus on the experience of an individual with her/his own environmental health data while BigBarChart focuses on the experience of the community with the overall dataset. Both projects seek to present opportunities for non-traditional learning, community relevance, and social impact.

Thesis Supervisor: V. Michael Bove, Jr.
Title: Principal Research Scientist, MIT Media Lab
Data Experiences: novel interfaces for data engagement using environmental health data

Laura J. Perovich
# Table of Contents

Abstract ......................................................................................................................... 3
Table of Contents ............................................................................................................. 7
Acknowledgements .......................................................................................................... 9
Prologue .......................................................................................................................... 10
Introduction ...................................................................................................................... 11
Background ...................................................................................................................... 12
  Concept ......................................................................................................................... 12
  Artifacts ......................................................................................................................... 12
  Prior work ...................................................................................................................... 12
Application & expansion of prior work ............................................................................ 18
Project outcomes ........................................................................................................... 18
Project findings ............................................................................................................... 19
Data overview ................................................................................................................. 20
Emerging contaminants ................................................................................................. 20
  Phthalates ..................................................................................................................... 21
  Alkylphenols & phenols............................................................................................... 21
  Parabens ....................................................................................................................... 21
  Flame retardants .......................................................................................................... 21
  Polychlorinated biphenyls (PCBs) ............................................................................... 22
  Polycyclic aromatic hydrocarbons (PAHs) ................................................................. 22
  Metals & particulates ................................................................................................. 22
Pesticides ......................................................................................................................... 23
  Green Housing Study data ......................................................................................... 23
  Diethyl phthalate (DEP) ............................................................................................ 23
  Benzyl butyl phthalate (BBP) .................................................................................... 23
  Diethyl hexyl phthalate (DEHP) ................................................................................ 24
Data content & meaning ................................................................................................. 24
Dressed in Data ................................................................................................................ 26
Overview ......................................................................................................................... 26


Acknowledgements

Michael, thanks for supporting and refining my ideas over the past two years. And thanks for convincing me that 3D printing is still overrated. Lathing and metal work is high on the agenda for next year!

Thanks to my readers for your advice and knowledge that helped bring this project together as well as all your research that influenced it.

Thanks to Silent Spring Institute for inspiring this research and providing me with the background needed to pursue it. And thank you for all your support that made it possible for me to be at the Media Lab.

Thanks to my UROPs Kristin Zimmerman and Dayanna Espinoza for all your hard work, patience, and new ideas.

Thanks to the John S. and James L. Knight Foundation, especially Prototype Fund manager Chris Barr, for supporting this work.

Thanks to OBMG for all of your help, from filming videos to moving MDF to yoga breaks.

Thanks to my classmates and the Media Lab community for your inspiration, diversity, and easy going attitudes. Thanks to the Director’s Fellows program for bringing in so many interesting collaborators.

Thanks to my manufacturing partners abroad, especially the Yiwu Knong Daily Commodity Factory, for trying something new. Thanks to Bunnie Huang and AQS for facilitating the process.

Thanks to Tom Lutz and John Difrancesco for the shop training and ideas about how to make things work.

Thanks to Linda Peterson, Keira Horowitz, and Kristin Hall for your help, advice, and patience.

Thanks to my friends for the Toscanini’s breaks, songs, walks to nowhere, gchat conversations, and dancing that kept me grounded through the tough parts.

Thanks to my family for helping me get here, especially mom for tolerating the fact that I don’t sew like an engineer.
“What do we really desire from our future technologies? We claim that just as in life, they should assist us in solving problems and improving our everyday efficiency. However, we further argue that technology also must prompt us to think, be curious, and wonder. If we fail or, worse yet, ignore this vital design space of wonderment for technology, we are almost certainly doomed to live amongst emotionless, servant-like, lifeless, problem solving, scientific systems.”

Paulos, E., Objects of Wonderment, 2007
Introduction

Recent advancements in statistics, computer science, and web-based interactions have led to powerful data analysis tools for researchers and a public enthusiasm for interactive data visualizations. These visualizations appear in online news media and have introduced new people to data exploration.

Yet pixels and screens limit the range of our experience. They offer great speed of analysis and significant versatility, but remain remote from the self, hard and electronic, and often somewhat emotionally distant and easy to forget. They rarely consume us or stay in our memories in the same way phenomena in the world do through their physicality and full sensory engagement. Thus to obtain a sense of wonderment in our technology, we venture off the screen back into the physical world.

Data Experiences will attempt to provide physical world form for datasets. They build from and encompass emerging work in a number of areas including data sculptures, data therapy, and wearable data (Miebach, 2014; Bhargava, 2014; Kim, 2010). I propose that Data Experiences be described by three primary characteristics. They:

- engage multiple senses
- exist at a human or embodied scale; they are immersive and present, not separate from the self like a document or a screen
- interact with the emotional and aesthetic self

Together these characteristics aim to create interactions that increase wonderment, prompt curiosity, create strong memories, and draw diverse people to engage with data.

The Data Experiences developed will be tested using environmental health datasets. Environmental health datasets collected by research organizations contain concentrations of consumer product chemicals (e.g. flame retardants, phthalates, metals) measured in homes (e.g. air, dust) and people (e.g. blood, urine). Participants almost always want to receive the results of testing because they are curious and the data is highly personal and may be relevant to health conditions (Brody, 2007). Yet researchers are often reluctant to share this information as it is a large investment of resources and expertise and institutional review boards may not support reporting data on personal chemical exposures to study participants (Brody, 2007). I created immersive physical realizations of this data that build from prior research in sharing results through paper (Altman, 2009) and digital methods (Silent Spring Institute, 2014); future work will compare outcomes of paper, digital, and experiential reporting methods.
**Background**

**Concept**

Data Experiences pull together findings from diverse fields to create a new space. They seek to take data off the screen and put it into the physical world. They may emphasize the aesthetic angle to create emotional experiences that complement analytic approaches traditionally taken with data and prompt new perspectives or engage new groups of people. A number of existing and ongoing projects begin to map out this space and inform current work, as discussed in future sections.

**Artifacts**

I created two Data Experiences using environmental health data: Dressed in Data and BigBarChart. Each Data Experience was implemented using data from Silent Spring Institute exposure studies.

Dressed in Data is a series of garments with lace patterns or prints that represent the concentrations of chemicals detected in an individual’s home. Squares in the pattern represent chemicals detected and are sized based on the amount of the chemical found. This is intended to give a “one glance” intuitive understanding of an individual’s exposures. Further information on chemical sources and health effects can be conveyed through color choices, fabric, and pattern design.

BigBarChart is a room-sized 3D bar chart that users can walk inside and interact with. Bars are constructed from custom pop-up laundry hampers extending close to 6 feet in height to create immersion and a sense of magic. Bars are networked and can change their height (motor actuation) and color (LEDs). They are responsive to interaction through tangible interfaces. BigBarChart explores immersion, human-scale, and tangible interfaces for data investigation.

**Prior work**

Prior work in tangible interfaces, art installations, data visualization, environmental health, and data ethics inform this project, as detailed below. Many of these fields are well established with ongoing research occurring at major universities while others continue to
gain momentum. Findings and best practices from these areas will be extended in developing Data Experiences.

Basic findings in data visualization were taken into account in developing these Data Experiences. For example, BigBarChart was initially chosen over BigPieChart as bar graphs have been shown to outperform pie charts in most situations (Kosslyn, 2006). The fundamental form of these Data Experiences was influenced by artistic work. In particular, large scale art installations, such as the pieces by Ana Soler and Soo Sunny Park shown below, can create a sense of wonderment that was desired in order to increase engagement and memorability. Note that these installations are static, thus do not provide a full embodiment of the final vision. This design approach for technology is formalized by Paulos (2008) who points at it as a way to engage non-experts and “incorporate the full range of life experiences” in technology.

![Image of an art installation by Ana Soler](image.png)

*Ana Soler, Mustang Art Gallery, 2012*
Additionally, a number of prior projects begin to map out the Data Experiences space. This includes data sculptures created from weather data by Natalie Miebach (2013); data sounds based on air quality data developed at Carnegie Mellon (Kim, 2009); data toys by Parsons The New School for Design (2013); 3D printed data (Jansen, 2013); data therapy and murals to increase data literacy and empowerment (Bhargava, 2012; Bhargava, 2014); breezes connecting Second Life and the real world (Segrera, 2012); wearable environmental data (Kim, 2010); wearable data advocacy (Urbano, 2013); and work on personal projection of information including the Facebook data projects by Leung (2011). Images from select projects in this space are shown below. Note that dynamism can be difficult to incorporate in these efforts. Many projects produce static displays though some—for example wearable environmental data (Kim, 2010)—provide dynamic information as well as real-time context for the information; the pattern changes as the user moves through the city, providing spatial, temporal, and contextual information.
Constraint City Vest: Gordan Savicic, 2011

Mitchell Whitelaw: Weather Bracelet, 2009
WearAir: Sunyoung Kim, et al 2010

New models of research ethics such as Community Based Participatory Research (Minkler, 2008) and data ownership (Terry, 2011) will inform and support the creation of mechanisms for sharing data through Data Experiences. Information sharing between scientists and study participants has not always been considered best practice and still may be discouraged in some fields and circumstances. For example, some institutional review boards (IRBs) do not support reporting data on personal chemical exposures to study participants if the clinical implications are not well known (Brody, 2007). IRBs believe that reporting may cause fear in study participants and lead participants to take extreme actions based on uncertain results; studies on reporting results have shown this is not the case (Brody, 2007). Ethical methods and standards for reporting results like this have been called for by the National Academy of Sciences and are still under development (Brody, 2007; Brody, 2014). Paper based reporting of household exposure data has been developed and implemented by investigators at Silent Spring Institute, Brown University, and University of California at Berkeley, among others (Brody, 2007). Digital based methods of reporting similar data are ongoing by the same collaboration in partnership with Harvard University (Silent Spring Institute, 2014).

Additionally, a number of organizations such as 23andMe (23andMe, 2014) and the Personal Genome Project (Personal Genome Project, 2014; Personal Genome Project: Harvard, 2014) have developed ways for digitally reporting genetic data related to health or ancestry, yet this has not been without controversy. In 2013, the Food and Drug Administration (FDA) forbade 23andMe from marketing its services as providing “health reports on 254 diseases and conditions,” (Food and Drug Administration, 2013) as the results were not analytically or clinically validated. The company has suspended its health-related services indefinitely while these issues are resolved (23andMe Health, 2014). The Personal Genome Project employs an open consent model that has been criticized in light of possible data re-identification and also requires participants to have a high level of prior knowledge to pass the screening quiz that is part of the informed consent process (Angrist, 2009). These cases serve as interesting models for the sharing of personal or health related data between scientists and the public, both in the controversies they draw to light around data ethics, particularly data ownership (Thorp, 2012), and the forms they use for data sharing.

While still a nascent field, work in tangible interfaces is rapidly increasing and will inform this project. Entire research groups drive forward work in this area (Tangible Media, 2014) and conferences such as Tangible, Embedded and Embodied Interaction have been established in the past decade to delineate the field (TEI, 2015; Luescher, 2013). Briefly, tangible Interfaces move digital information to the physical world thus allowing people to interact with digital information through haptic pathways (Ishii, 2008). Past work in this area includes Urp (Underkoffler, 1999), the reacTable (Jordà, 2007), PY-ROM (Chi, 2009), and I/O Brush (Ryokai, 2004), among others. Recent work from James Patten on Thumbles approaches the space explored in this thesis (Patten, 2014). These hand-sized objects allow for two way interaction between humans and computers—the robotic forms can be controlled by
both parties (Patten, 2014). Data is one of the initial intended use-cases for Thumbles and plans for 3D data representations are in the works (Patten, 2014). Thumbles draw on the idea that experiencing relations through the physical manipulation of objects is a powerful experience (Patten, 2014).

**Application & expansion of prior work**

My research will build from these models to add to the growing knowledge around Data Experiences. They will focus on immersion, development of soft and organic forms, and enduring interactions that can eventually be applied to diverse datasets. Qualitative interviews have been designed and tested for initial evaluation of forms. Future work will test revised prototypes with participants and researchers in environmental health—communities that are strongly invested, eager to receive the information, and curious about new modes of data expression.

**Project outcomes**

Two Data Experiences have been developed using environmental health data: Dressed in Data and BigBarChart. Artifacts use subsets of Silent Spring Institute’s Green Housing Study and Northern California Household Exposure Study data. Communication of individual and community data will be explored.

These Data Experiences aim to:

- serve as an approachable media for data and lower the barrier to engagement
- help participants achieve a deeper understanding of environmental health data from their community
- prompt participants and researchers to interact in novel ways and ask new scientific questions

Projects seek to achieve these aims through a number of dimensions including: immersion, interaction, artistic expression, scale, and tangibility. This thesis includes an initial prototype and preliminary qualitative user studies that will provide feedback on the efficacy of these dimensions and identify room for further growth to achieve Data Experience aims.

Note that though these Data Experiences are implemented with environmental health data they are designed to be extremely versatile. Most small to medium sized datasets that can be expressed with basic visualizations (e.g. bar plots, scatterplots) could be mapped into this space. This allows researchers from many fields to relate to the form and imagine how it might be implemented in their space. Hopefully this versatility will encourage the use of Data Experiences with other datasets and trigger new investigations.
Project findings

The fabrication process and initial user testing of these Data Experiences led to many useful findings, some in unexpected spaces. Findings are summarized here and detailed in later sections.

Environmental health partners shared important knowledge about data messaging. This included interpretations of study results and data context—e.g. sources of chemicals, health effects of chemicals, and possible particularities of the data that may cause patterns. For example, Green Housing Study (GHS) homes in Boston tend to use different flooring than GHS homes in Cincinnati causing unique exposure opportunities. This process will continue as further GHS data is available and will inform framing and future development of these Data Experiences.

Initial prototyping provided key insights on process, material selection, and parts fabrication. I thought of my prototyping process as the “Frankenstein” approach to prototyping which proved to be a useful path to developing a form that’s quite different from existing objects—like BigBarChart. The basic process was to: (1) search for objects that performed pieces of the desired functionality (2) assemble them in an ad hoc manner for initial testing (3) iterate on parts selection with help from Amazon’s “you might also like” functionality (4) create custom parts with manufacturing partners and FabLab tools and assemble them more cleanly into a next generation prototype. This process converged to a new form that embodies the initial vision.

Because BigBarChart’s first intended use is with environmental health data, it is also important to begin to consider the possible exposures it presents as a consumer product, including exposures relevant to manufacturers, prototypers, and study participants (e.g. installation visitors). An initial inventory identified a few points of concern and opportunities for improvement, including MDF use, adhesive use, textile selection, and solder selection. Fully assessing and addressing these exposures remains as future work; yet this project provides an exciting opening for a conversation on emerging contaminants in the prototyping process. Finally, initial testing demonstrated the utility of traditional methods of fabrication. In particular, though 3D printing creates precision parts that can be replicated with little effort, traditional methods of machining may be more appropriate as this approach creates stronger parts more quickly.

The fabrication of BigBarChart led to unexpected collaborations with manufacturers. Prototype fabrication was initially intended to occur in house, but testing showed that some custom items, for example large springs, could only be fabricated with specialty equipment. Thus I established partnerships with manufacturers abroad and increased capacity for communicating about the prototyping process.
Data overview

BigBarChart and Dressed in Data will be implemented using data from Silent Spring Institute studies. Dressed in Data uses data from the Household Exposure Study and BigBarChart uses data from the Green Housing Study.

The Household Exposure Study was conducted by Silent Spring Institute, University of California Berkeley, Brown University, and Communities for a Better Environment in Northern California in 2006. It builds from the Cape Cod Household Exposure Study conducted in 2000 by Silent Spring Institute. The Cape Cod study sampled indoor air and dust in 120 homes and analyzed samples for 89 endocrine disrupting compounds; over 65 chemicals were found in at least one home (Rudel, 2003). The Northern California Household Exposure Study measured concentrations of over 100 compounds in 40 homes in Richmond, California (industrial), and 10 homes in Bolinas, California (rural), in indoor air, outdoor air, and house dust. A wide range of chemicals were detected—many of them endocrine disruptors—including phthalates, alkylphenols, parabens, flame retardants, polychlorinated biphenyls (PCBs), polycyclic aromatic hydrocarbons (PAHs), pesticides, and phenols (Rudel, 2010).

The Green Housing Study (GHS) took place in 2013 in Boston, MA, and Cincinnati, OH. Families in this study lived in public housing units undergoing renovations and included a child with asthma. Data from this study includes measurements of over 50 chemicals (e.g. phthalates, glycol ethers, flame retardants, PCBs) related to health (especially asthma) in the air and dust of these homes and in the urine of study participants. Sample characterization and data analysis is ongoing, though preliminary data is available for select media and chemical groups. Early results include data for phthalates in air in multiple communities and at multiple time points; subsets of this data were used as a first implementation of BigBarChart. Additional data will be added to the prototype as it becomes available in the following year.

Emerging contaminants

Dressed in Data presents the personal data of one Northern California Household Exposure Study participant. This includes measurements of concentrations of over 100 chemicals from a variety of chemical groups. Many of these chemicals are “emerging contaminants”—their sources and health effects are not fully understood, they may have recently entered industrial use, and our exposures to them in everyday life are not sufficiently characterized. Cell data, animal data, and early exposure studies for some of these chemicals indicate adverse health impacts such as endocrine disruption, respiratory effects, and mutagenicity (Silent Spring Institute, 2007). An overview of these chemical groups, example chemicals, chemical sources, and chemical health effects is found below.
Phthalates
Phthalates are common plasticizers that increase flexibility and transparency. They are found in many products such as shower curtains, food packaging, floor tiles, toys, cosmetics, and adhesives. Phthalates are not strongly bound to plastic so they readily enter the environment as products age. Example compounds include diethyl phthalate and di-n-butyl phthalate (Rudel, 2009).

Phthalates were first found in the 1970s and have been detected in a range of environments including the Norwegian Sea. As they largely come from everyday products, phthalates are ubiquitous in the indoor environment (indoor air and dust) and indoor concentrations most often exceed outdoor concentrations. Diet is a major route of exposure, but fragrance, medical equipment, toys, and pharmaceuticals may be substantial exposure routes in some cases. Phthalates are detected in virtually all people (Rudel, 2009).

Many phthalates are endocrine disrupting compounds. Phthalates are associated with adverse effects to male reproductive outcomes such as decreased sperm count and testosterone level (Jurewicz, 2013). Some compounds have also been associated with respiratory effects and cardiac outcomes (Gillum, 2009; Rudel, 2009).

Alkylphenols & phenols
Alkylphenols & phenols are surfactants and are often used in detergents, cleaners, and disinfectants. Examples include o-phenyl phenol and nonylphenol (Rudel, 2009; Silent Spring Institute, 2007).

These compounds became common in the 1940s. They are often found in the indoor environment and in water. People may be exposed to phenols and alkylphenols through diet, product use, and air, but primary sources of exposure are not well understood (Rudel, 2009).

Many alkylphenols are endocrine disruptors, often mimicking estrogen (Rudel, 2009).

Parabens
Parabens are preservatives and anti-microbials. They may be used in personal care products, pharmaceuticals, and food. Examples include methyl paraben and ethyl paraben (Rudel, 2009).

Exposure to parabens is thought to occur primarily through the skin. They have also been measured in air and house dust (Rudel, 2009).

Parabens are endocrine disrupters that mimic estrogen (Rudel, 2009).

Flame retardants
Flame retardants are found in many consumer products including plastics, furniture, bedding, electronics, and appliances. Examples include PBDEs and FireMaster 550 (Rudel, 2009; Dodson, 2012).
Flame retardants are persistent organic pollutants that easily leach from products into the environment. Thus they are ubiquitous in the environment and in people and have been found in remote regions (Rudel, 2009).

Flame retardants are commonly found in dust and air; dust is thought to be the primary route of exposure, though diet may also play a significant role. Use of flame retardants has increased rapidly in the past few decades, but some governments have now banned or heavily restricted their use (Rudel, 2009).

Health effects of flame retardants may include neurological impacts and effects on the reproductive system (Rudel, 2009; Silent Spring Institute, 2007).

**Polychlorinated biphenyls (PCBs)**
Polychlorinated biphenyls (PCBs) were used in a variety of products including plastics, paints, electrical equipment, adhesives, floor finish, and caulking. They have been used in the US since the 1930s and were most prevalent in the 1970s until regulations began to restrict their use. Example compounds include PCB 52 and PCB 153 (Rudel, 2009).

Dietary exposures to PCBs are generally the most significant, but indoor sources may play an important secondary role as many PCBs come from the built environment or consumer products (Rudel, 2009). Exposure may also occur at contaminated industrial sites (Zhou, 2014).

PCBs are developmental neurotoxicants and endocrine disruptors and also have effects on nervous system function and cancer (Rudel, 2009).

**Polycyclic aromatic hydrocarbons (PAHs)**
Polycyclic aromatic hydrocarbons (PAHs) come from combustion. They come from burning coal, oil, gas, garbage, tobacco, or meat and are also found in roofing, plastics, and dyes (ATSDR, 1996). Example compounds include fluoranthene and pyrene (Rudel 2010).

Primary exposure to PAHs comes from air and diet, such as breathing smoke or eating burnt meat (ATSDR, 1996).

Some PAHs are carcinogenic and have adverse reproductive effects (ATSDR, 1996).

**Metals & particulates**
Metals and particulates come from industrial emissions, traffic, frying food, and cigarette smoke; they have both outdoor and indoor sources. Example compounds are lead, vanadium, and PM2.5 (Silent Spring Institute, 2007).

Metals and particulates have been linked to effects on the nervous system, respiratory system, and cardiac system. They have been shown to increase death rates (Silent Spring Institute, 2007).
**Pesticides**

Pesticides are used to repel or kill things, including bugs, fungus, rodents and germs. They are often used on pets or plants and can also be found in paints, furniture, and carpets. DDT, chlordane and permethrin are examples of pesticides (Rudel, 2009).

Pesticides are found indoors and outdoors. Diet and household exposures are thought to be the two major routes of exposure outside of agricultural or occupational settings. Many pesticides have been banned or restricted but some long-banned compounds can still be found today (Rudel, 2009).

Many pesticides are endocrine disrupting compounds. They also are associated with cancer and neurotoxicity (Rudel 2009).

**Green Housing Study data**

BigBarChart uses a subset of Green Housing Study data: phthalate data in air in Cincinnati and Boston at 6 months after home renovation and 12 months after home renovation. Three phthalates were selected from this data: diethyl phthalate (DEP), benzyl butyl phthalate (BBP), and diethyl hexyl phthalate (DEHP). All three compounds are plasticizers and have been linked to reproductive effects in animals and/or humans, though some may be acting as indicators for other chemical exposures (Parlett, 2013). Additional background information on each chemical can be found below.

**Diethyl phthalate (DEP)**

Diethyl phthalate is a plasticizer used in plastics as well as cosmetics and fragrances (Api, 2000).

DEP has been linked to effects on the reproductive and nervous systems (Miodovnik, 2011; Swan, 2008). It is often used as a binder for fragrances that are endocrine disrupters and may serve as an indicator for those exposures; it is unclear if related health outcomes are due to DEP or the fragrance compounds (Parlett, 2013).

Dermal contact is thought to be the main route of exposure to DEP (Gong, 2014). Consumer products are the main source of these exposures (Rudel, 2011).

**Benzyl butyl phthalate (BBP)**

Benzyl butyl phthalate is a plasticizer used in plastics and flooring (Serrano, 2014; Green Facts, 2014).

Dermal contact is thought to be the main route of exposure to BBP for most people (Gong, 2014). It is currently banned in toys, childcare objects, and cosmetics (Green Facts, 2014).

BBP is an endocrine disrupting compound (Rudel, 2011). It is also linked to respiratory effects (Scottish Environmental Protection Agency, 2014).
Diethyl hexyl phthalate (DEHP)
Diethyl hexyl phthalate is a plasticizer used in plastics (Gong, 2014) and food packaging (Serrano, 2014; Rudel, 2011).

DEHP is an endocrine disrupting compound (Rudel, 2011) linked to reproductive effects, cardiac effects, and cancer (Gillum, 2009).

Diet is thought to be the primary route for exposure to DEHP for many people (Gong, 2014). Changing individuals to a fresh food diet substantially reduced levels of DEHP found in their urine (Rudel, 2011).

Data content & meaning

Brody (2007) found that participants in environmental health studies often seek answers to the questions: (1) What did you find? (2) How much? (3) Is that high? (4) Is it safe? (5) What should I focus on? (6) Where did the chemical come from? (7) What can I do?

Often there is not sufficient research to answer all these questions for emerging chemicals, particularly in prescribing a “safe” amount. The report-back process—whether it be paper based, digital, or experiential—is meant to engage people in the scientific process by accurately conveying gaps in knowledge while also providing practical information. One method that begins to respond to participants questions is to compare the detected concentrations of chemicals in a study to amounts found in other studies (questions 1, 2, and 3). Comparison values can serve as preliminary benchmarks that provide a starting point for understanding, though they cannot indicate what levels are safe or the effects on human health. Providing information about chemical sources and health effects also positions people to consider their relative importance and presents opportunities for change. For example, an individual with asthma may note that some phthalates have respiratory health effects and try to remove sources of these chemicals from her/his home. In conveying this information it is also important to note that data on chemical sources and health effects is constantly evolving as manufacturers alter product formulations and researchers conduct additional cell, animal, and human studies on the effects of these chemicals. Both Data Experiences developed here will try to provide this type of contextual information to respond to typical participant questions and provide participants a framework for the information they are considering.

Dressed in Data first provides data context by scaling the square size in the lace patterns against the median study values. Thus if the concentration of a particular chemical for one individual is below the study median, its square size will be less than the unit square, while those above the median will have square sizes exceeding the unit square. This approach is aligned with current methods where an individual’s values are graphed on a strip plot in relation to the study median and alongside the values of others in the study. This provides
context so the information can more easily be interpreted than a mere value (e.g. “35 ug/g DDT”). The design and pattern of the clothing will begin to provide contextual information about the sources of the chemicals (e.g. data representing industrial sources will be black spots over yellow fabric, hinting at smokestack discharge obscuring the sun).

An initial design of BigBarChart will provide context in a similar manner: bar heights for a chemical are given as a percentage of the maximum value for the chemicals available; bars go from zero (minimum height = concentration of zero) to 100 (maximum height = maximum concentration). This is an initial convenience-based approach, as it ensures that all data can be displayed on the bars—bars are not required by the data to exceed their maximum possible physical height. It also provides relative information about the values as opposed to absolute values. This approach may become problematic since exposure data are often lognormal, thus maximum values can greatly overshadow the bulk of the dataset, making values around the median indistinguishable. Future work will consider other possibilities for scaling data and managing outlying points, such as showing absolute magnitudes, scaling to a median, or signifying extreme outliers with blinking lights while excluding them from the data scaling. Each bar in the room will represent one participant in the study for a common chemical. Bar height shows the amount of that chemical relative to the maximum value. The current design does not prominently display scale information and the flexible form of the bar and the inherent error in its position encourages people to understand the error associated with these measurements. This approach takes advantage of the embodied and immersive nature of BigBarChart as the heights of the bars can be read relative to the body, hopefully creating a more memorable experience.

The BigBarChart design also conveys basic health and source information for chemicals through the soft switches used to select them. Labeled icons on the switches show general health effects (e.g. reproductive) and sources (e.g. food packaging) related to each chemical. Though complete detailed information is difficult to convey in this simplified form, the switches provide a starting point for engagement.

Future BigBarChart designs will allow people to compare the bars to external studies, not just the current study. Comparison values will be introduced as additional bars, projected lines, or bar colors. Comparison values will match the scaling protocols used for the study data; they will be presented as absolute magnitudes or relative to the designated scaling factor as appropriate. Users will be able to toggle through comparison values to see how chemical levels in their community compare to other locations or time points. Bar colors may be used to classify the relative values as “higher,” “similar,” or “lower.” Heights of the bars can be compared to lines projected on walls or to additional bars. Comparison values could include national guidelines, estimated effect levels, or results from prior studies—such as median DDT concentration in the Arctic or median DDT concentration on South Carolina farms. Outlying high values would turn red to draw additional attention.

Efficacy of these approaches will be studied in detail in future work.
Dressed in Data

Overview

Dressed in Data focuses on an individual’s experience receiving chemical data from his/her own home and body.

Dressed in Data explores ways to map an individual’s exposure data onto his/her clothing. Clothing is closely tied to the body and intensely personal, just like an individual’s chemical exposures. Embedding exposure information into clothing prompts us to develop a more intuitive sense for the data and to keep it with us throughout the day. It can also serve as a point of conversation and sharing with others, achieving the Data Experience goal of increased engagement and memorability.

The project includes design of possible translations from data to patterns, code to map environmental health data to laser cut files, materials testing, and development of basic clothing patterns. Patterns focus on effectively conveying quantities and amounts of detected chemicals; initial work explores lace patterns that represent detected chemicals as squares sized based on concentration values. This provides a “one glance” intuitive understanding of an individual’s exposures. Further information on chemical sources and health effects can be conveyed through color choices, fabric, and pattern design; for example flowery prints may be used to signify that data is related to gardening or pesticide exposures.

Related work

A number of projects explore computational design for craft and clothing as a display interface, as described below. This includes works by Jacobs and Buechley (Codeable Objects, 2013) and Zoran (Hybrid Mediums, 2013) that offer paths for pairing craft and computational design. For example, FreeD (Zoran, 2013) presents a handheld milling tool that provides guidance based on a 3D model of the object. The tool adapts its feedback if the user adjusts the 3D model and the user can also autonomously control the tool to override the model to achieve the benefits of handcrafted design. Related works include Hybrid Basketry (2013) and Hybrid Reassemblage (2013). Jacobs & Buechley (2013) present a number of insights on the process of pairing craft and computational design for general use, such as the importance of simplifying code syntax to soften the learning curve for new programmers. Their users from both programming and craft backgrounds expressed engagement and continued interest in mixing craft with technical approaches (Jacobs, 2013).
Additional projects work with datasets to create fashion or art. Nathalie Miebach’s data sculptures map astronomy, ecology and meteorology data to intricate shapes. Her woven sculptures explore the role of aesthetics in expressing and understanding science (Miebach, 2014) and she has also explored the use of music for expressing data (Miebach, 2012). For example, “Recording and Translating Climate Change” compares current weather data to historical trends through pieces in both music and sculpture. Chen’s x.pose project (2014) ties fashion with data. This shirt has a 3D printed structure that frames material than can be solid or transparent. The transparency of the shirt increases as the wearer produces more shared data (Chen, 2014).

Other related projects use participatory creation of fashion through technology as a means for empowerment. For example, DressCode engages new programmers by providing an interface that displays both the code and the design output in real time (Jacobs, 2013). It reaches out to groups often underrepresented in technology, particularly young women, and provides a platform for new modes of expression and the blending of craft and computational design (Jacobs, 2013).

**Design considerations & scenarios**

Two sets of works were created: a set of garments suited for fashion and a shirt meant for daily wear.

The fashion-oriented garments are intended to be an eye catching example of data display with aesthetic consistency within themselves. They could be displayed in a public space, photographed for media, or shown in an arts setting. This serves as another avenue for bringing publicity to environmental health studies and engaging new audiences. In particular, they provide an opportunity to combine art and data and encourage communication between individuals in these spheres. Data is primarily displayed on the back of the fashion-oriented garments as the intended audience is the general public, not the user.

The second piece serves as an example of the type of garment that could be given to—or even designed and created by—participants in exposure studies as part of an effort to report findings from their homes. This garment is casual and comfortable enough for everyday use. The design is conspicuous enough that it opens the door for the wearer to discuss their results with family, friends, strangers, media, or others by using the shirt as a reference point. Yet it is sufficiently aligned with everyday styles so that such conversations are not required if the wearer would prefer to avoid them.
Software and fabrication system

These data-based garments required development in a number of areas:

- preliminary explorations of display options
- R code to transform data into a lace pattern, including multiple display options
- design files for garment patterns of various styles
- “artisanal” laser cutting approaches and garment fabrication

Code & data display

Inputs to the system are csv files with chemical names, measurements, participant ID, and auxiliary data (e.g. chemical class, measurement media). Functions sort and clean data based on user selection of chemical group and participant.

A number of data display options are supported and can be customized based on knowledge and preferences of the user. First, chemicals that were not detected in a home can be set to a variety of values before creating the lace pattern. This is particularly important with emerging contaminants since analytic methods are often still in development causing high detection limits. Thus in some cases researchers may want to assign “non-detects” to a small number below the detection limit if the chemical is thought to be ubiquitous. The code also supports selection of a variety of comparison points. Squares are scaled relative to the selected value—such as the chemical class median, chemical class maximum, or an external value—to give context to the data.

Additionally, the code provides functions to shape the lace pattern into an iterable form. Suggestions are made based on the number of chemicals in the group. Dimensions are set to ensure that the pattern can be effectively cut in most fabrics (e.g. sufficient line thickness and square size to ensure robustness of remaining material).

The program outputs black and white JPEGs that can be used in Corel Draw to create cut files for the laser cutter.

“Artisanal” laser cutting

JPEGs of data patterns and data scaffolding were translated to cut files by using the trace functionality of Corel Draw. Files were then cut using an Epilogue laser cutter.

Copier scanners were also used to bridge the gap between handmade patterning and CAD software. The speed and precision of CAD tools and the laser cutter was critical to creating these garments. Yet in some circumstances hands-on intuition or an artistic eye was necessary for design optimization. For example, it is easy to notice poorly aligned pieces when pinning them together. One can rework these curves using scissors then scan them to JPEGs that can be traced in CAD software to refine the shapes of the pattern. These hand-cut curves can be smoothed in CAD software if needed, and pattern pieces can then
be cut in mass on the laser cutter. This hybrid of computationally constructed and handmade items is also useful in creating iterated designs or switching between materials—hand drawn sketches with a repeating pattern can be scanned and then iterated in CAD software to save time or created in an alternative material then transferred to the laser cutter for use with a different material. For example, snowflake patterns can easily be cut by hand in paper, but cutting them directly in fabric leads to distortion and fraying. Cutting a snowflake in paper, scanning it, and then laser cutting it in a fabric with synthetic content creates the desired outcome. The laser cutter is an intriguing tool for craft and warrants further exploration.

Most fabrics are well suited to laser cutting. The specific considerations, guidelines and recommended settings for cutting fabric on the Epilogue laser cutter found below come from trial and error in the Dressed in Data fabrication process.

- **all fabrics**: patterns with thin curves or unfixed edges lose their form; straight lines and fixed edges are easier to work with. Fabrics cut nicely; those with high synthetic content will seal at the edges. Be aware of this when layering fabric for cutting (layers may melt to each other) or cutting stretch fabric (sealed edges will extend but will not regain their shape after stretching).
- **silk (high fashion pieces)**: raster nicely on low settings (~10% power, 95% speed). Handle with care when placing in the laser cutter as fabric can easily twist and misalign itself though appearing flat. Pre-folding pieces and unfolding them onto the bed can be beneficial.
- **stretch blends (daily wear shirt)**: stretch fabrics with high synthetic content will melt when rastered creating non-extendable sections. Non-synthetics, such as cotton or soy blends, burn easily when rastered thus thicker fabric must be used and rastered at low settings (98% speed; 5% power) to avoiding tearing.
- **novelty fabrics (data)**: leathers are often treated and may release fumes that should be avoided. Fabrics with significant texture may lead to jagged or incomplete cuts. Faux fur burns easily and cuts very poorly creating a molting look.
- **sheer fabrics (data scaffolding)**: should be cut on very low settings (8% power, 95% speed) to avoid burning or tearing. Shapes should be sufficiently thick (~3/16”) to keep their form and avoid tears. Thicker sheers are easier to work with and stretch sheers are often less reliable.

Images of select laser cut fabric products are shown below.
rastered natural fabrics (soy & cotton stretch blend)

rastered synthetic stretch fabric (rastered regions melt and cannot be stretched)
shapes created through hybrid hand traced and CAD design methods

Garment and data pieces were assembled through sewing and gluing. Garments were sewn through traditional methods, including fasteners, trim, and sizing. Data layers were glued to one another to create the final product for the high fashion pieces. Data layers were aligned with scaffolding layers and fabric glue was applied to each data point to affix the two. The scaffolding portions of the data layer were cut away with scissors, leaving the mesh scaffolding and the data. This was an effective but time consuming process that warrants further iteration. Everyday garments included data patterns through rastering the fabric directly. Silk screening data or printing data onto fabric may also create the desired effect.

Artifacts

Two implementations of Dressed in Data were created based on data from the Northern California Household Exposure Study:

- **fashion focused designs**: four garments each representing a different chemical group and including the full set of chemicals measured in the indoor air of a single Northern California study participant
- **daily wear**: one garment representing factory related pollutants in indoor air for the same Northern California study participant
Values below the detection limit were set to zero for all garments. Sizes of squares within the lace pattern were scaled relative to the median value for the compound in the study, with non-detected medians set to the detection limit. Values above the median have larger squares while values below the median have smaller squares. For example, assume chemical A is measured at 55 ng/m$^3$ in one home and the median amount of chemical A in the study is 35 ng/m$^3$. Then the area of the data point square for chemical A is \( \frac{55 \text{ ng/m}^3}{35 \text{ ng/m}^3} = 1.57 \), giving sides of \( \sqrt{1.57} = 1.25 \). Side lengths and areas are dimensionless and measured relative to a unit square. Unit square dimensions are set at the discretion of the designer and may be scaled to the garment size or desired data print.

Square based iterated lace patterns and prints were chosen for aesthetic reasons. They were thought to more closely fit with existing clothing designs (e.g. lace, plaid prints). Data visualization research suggests that other designs, such as line-based prints that focus on length not area, may lead to better data understanding thus the merits of this design choice should be evaluated further.

Chemicals were grouped by class based on similarity in structure, source, or health effect. This is consistent with past instances of data reporting and provides structure to the data that can facilitate participant understanding. Each class was displayed on one garment. Classes include (1) polyaromatic hydrocarbons & particulates (2) phenols & alkyl phenols (3) phthalates & parabens (4) metals, ions, & ammonium. A few compounds were removed from the pattern as they were repeated measurements (o-phenyl phenol) or detection was not expected in this medium (Alachlor, Aldrin). This also facilitated dimensioning and repetition of the lace pattern (e.g. it removed prime numbers of chemicals within chemical classes).

Images of both the overall clothing design and the specific lace patterns are included below.
front of fashion dress designs

back of fashion dress designs
daily wear shirt

rastered print of factory related pollutants in daily wear
lace pattern for particulate matter, metals, ammonium

lace pattern for pesticides
lace pattern for phthalates and consumer care chemicals

lace pattern for factory related pollutants
User study protocols & initial testing

User study protocols were developed to assess the aesthetics of the garments and the information the pieces convey. Questions focus on interpreting the data results represented in the clothing, sharing overall impressions of the pieces, and assessing attitudes about data sharing. Protocols were tested on two users to acquire initial results and refine methods. Initial findings showed that users had some understanding of the data though interpreting variable attributes was confusing at times. Participants assessed the value of data sharing on a case by case basis, considering dimensions such as privacy, benefit to the community, alignment of data with personal interests, and potential harm to the individual.

Participants cited a number of factors that influence their willingness to share data including: their interest in the data, their attitudes on privacy, the social utility of the data, and the level of stigma associated with the data. Users expressed enthusiasm for showing data that was consistent with their personality, aesthetic, or vision of self, and users identified types of data that did and did not fit these criteria—examples included genomic data, social media data, and data specific to their line of work. Overall, the incorporation of data into clothing was seen as “hipster” or trendy and thus somewhat desirable, though tied up with social implications. One user expressed a desire to share environmental or other data for altruistic reasons, saying that if levels of chemicals were high then sharing that information could help cause changes that benefitted individuals or communities. Users were reluctant to share data that could be stigmatizing, providing examples such as information on diseases or genetic conditions. Attitudes were generally similar between sharing in a Data Experience setting and an online setting, though a few differences emerged. For example, one user who cited privacy as a main value appreciated that data shown in his home would only be seen by people invited into his home, providing an automatic screening process that online settings do not easily support.

Both users correctly interpreted the overall look of the outfits. The daily wear shirt was identified as more casual and something one would see on the street, while the other outfits were associated with special occasions or artistic venues. Participants had mixed past experience with textile fabrication but all expressed a general interest in the process.

Users interpreted the displayed data with some success, though the patterns were confusing at times. Users correctly identified the pattern with the largest number of detected chemicals but cited both “busyness” and number of colors in selecting that pattern, though colors did not have implications in this context. One user correctly identified the pattern with the highest relative amounts detected, but another user seemed to conflate the amounts detected with the number of chemicals detected. Users were unclear as to how to interpret the data scaffolding—they initially viewed it as an information source not merely a background grid. Colors in the data and outfits were perceived to be very important, though
these selections were based on metadata and aesthetics. Overlap between the individual squares caused confusion for one user. Future work will explore alternative data mappings, such as linear instead of area based mappings of amounts and assigning significance to color, as detailed in later sections.

**Limitations & future work**

Future user studies will test Dressed in Data designs with environmental health study participants using their own data. Studies will provide important information on clothing design, aesthetic, and perception of data that will inform design iterations to improve user experience. Future work will create garments for a set of environmental health participants based on their own data and then test these garments with them to better describe the emotional engagement.

Options for data pattern design will also be considered. Currently, concentrations for an individual are expressed as an area of a square scaled relative to a comparison value (e.g. chemical median). Expressing amount in two dimensions (e.g. square) may not be as readable as expressing it in one dimension (e.g. line or bar). Direct comparisons of chemical amounts are also not possible since concentrations are scaled relative to their respective medians. Additionally, pattern iteration and ordering of chemicals within a pattern relative to one another may influence interpretation. Currently chemicals are ordered by convenience (e.g. alphabetically) within a class; ordering chemicals by concentration, relative amount, or randomly may create a different overall impression of the data, especially as individual squares may overlap with one another when they exceed the unit square. Designers may also rescale the data pattern based on the clothing item or the user's size—this is highly valuable aesthetically but may create inconsistencies in interpretation or comparison. These possibilities will be considered further and evaluated by closely examining the data visualization literature and conducting additional user studies.

Future iterations may engage the participants in the garment design itself. Participants will have the opportunity to select colors, clothing styles, clothing sizes, and fabric type based on their personal tastes. This will act to increase engagement and memorability of the Data Experience. Advanced users could also customize data display—for example placement of pattern on the garment, iteration of pattern, shape of pattern, data subsetting, or data comparisons—and cut and sew the garments themselves. Much of the existing code and process works towards these goals and could be formalized into workshops for garment creation, as in the DressCode project (Jacobs, 2013). Involving environmental health study participants in the garment design and fabrication has the potential to increase engagement, understanding, and personalization. It is also a bridge for individuals comfortable in the artistic sphere to use to achieve better data literacy, and for those comfortable with data to gain an appreciation for the aesthetics of information. Individuals
and tools already established in both areas can help facilitate these exchanges. Individuals feel significant ownership of their clothing choices and personal fashion; leveraging this sense of ownership by including their data in the process may help them engage with the information in a space they’ve already established as their own.

Finally, future work will display data in real time, or much closer to it, to increase the real life relevance of information. Current pieces are static and based on one time measurements due to the fact that chemicals included in the data require in-depth sampling and lab work to measure, on the scale of weeks of labor. Additionally, though less of a limiting factor, the clothing fabrication process does take a number of hours. These limitations can be addressed through advances in sensors, clothing fabrication, and clothing interactivity.

A few chemicals of concern—such as particulate matter, ozone, and carbon dioxide—can already be measured in real time or almost real time with fairly reliable and affordable sensors (Taking Space, 2014; Futurlec, 2014). As more information arises about the health effects of other chemicals, researchers will find faster and cheaper ways to measure them. Multiple groups and funding sources are currently beginning to cover this space (Knight Foundation, 2014). Measurement techniques that are fast and affordable enough for daily use could be leveraged to make garments that provide a more holistic view of an individual’s exposures. Indeed, the fashion industry itself is eager to develop a “t-shirt machine” that can create simple garments without human intervention (Bevans, 2014). One could imagine being presented every morning with a shirt patterned with your exposure information from the previous day. This implementation acts as a bridge to real time data display which will be possible with better chemical sensors paired with improvements to adaptable fashion. Indeed, fashion technology is one of the next fronts in wearables. A number of fashion projects have started to explore this space of shape changing fashion and data display (Chen, 2014; Perovich, 2013). As the materials and processes advance, they will become more suited to daily fashion and can be paired with existing soft materials with specialized functions such as color changing ink, color changing thread, and smocking thread that are already established in the crafting community. One can imagine the print on your shirt changing as your chemical exposures change; your clothing changing shape and stiffness to become uncomfortable and encourage you to leave a heavily polluted area; or your clothing increasing its protectiveness by offering air filtration in polluted spaces. These types of projects would increase understanding by giving ongoing relevant information and prompting action based on these understandings.
BigBarChart

Overview

BigBarChart focuses on the community meeting aspect of data reporting and attempts to create a shared and immersive experience for the exploration of data.

BigBarChart is a room-sized 3D bar chart that users become part of and interact with. Bars are constructed from modified pop-up laundry hampers extending close to 6 feet in height to create immersion and a sense of magic. Bars are networked and can change height (motor actuation) and color (LEDs). They are responsive to interaction through tangible interfaces (soft switches) and in the future may be responsive to interactions that are direct (e.g. person entering the room), bar based (e.g. pushing down on a bar to get metadata), or digital (e.g. controlling bars and performing statistical analyses through a tablet). Bars are designed with an eye towards compactness and portability to allow for fast deployment in a variety of locations including parks, community centers, museums, and homes. A version of this project was a semi-finalist in the Knight New Challenge: Health and was funded by the Knight Prototype Fund ($35,000 / 6 months).

BigBarChart explores the use of immersion and tangible interfaces for the exploration of data. The maximum height of each bar is roughly the height of a person (~5’7”). This human size scale provides a new context for the data as well as the opportunity to consider it differently than one might on a computer screen or piece of paper. For example, one might measure values relative to themselves: the bar is up to my waist, or my knee, or taller than me. This also conveys the varying personal implications of the results; chemical concentrations have different consequences for different individuals. Children tend to be more vulnerable to the health effects of many of these chemicals and most children will be much shorter than the bars that have been scaled to adults. Additionally, the physicality and scale of the bars allow them to have real world consequences on our senses. Taller bars are visual barriers that block sight of other bars. Bars also must be navigated around. Together, they create a new memorable landscape of wonderment for individuals to explore.

A soft form was selected for both the bars and the interface for controlling the system in order to work towards the approachability design goal. The body and the objects we hold near it (e.g. clothing, bedding, some furniture) take organic soft forms that we attempt to mimic in the design as much as possible to make it welcoming to the participant. Fabric was featured to the extent possible and “machine-like” features were minimized or hidden, though some structural forms were required for support and control (e.g. telescoping poles, electronics). Person-sized bars were used to create a sense of immersion and wonderment by jarring people from their everyday scale. This presented considerable fabrication
challenges so smaller bars (standard 2 ft. pop-up laundry hamper) were used in early prototyping.

The physicality and softness of the form also begins to convey a sense of error that is part of the measurements. The form is not highly structural, rigid, or precise and it does not appear to be a machine. Instead, as the external form is fabric based, it resembles a muppet and has a soft “sketched” sense to it that invites approximation. A height is presented as data instead of a precise numeric value and the height has some variation within itself as the bars tend to be not fully level and the motor winding introduces small variations.

The project includes exploration of mechanics (e.g. base design, gearing, springs), data design (e.g. what functionality to include), electronics (e.g. networking, motor actuation, lighting), and fabrication techniques (e.g. manufacturing partnerships and communication). Results from preliminary user testing and informal conversations with participants and researchers will guide future iterations of the project.

Data considerations & scenarios

BigBarChart focuses on creating a shared experience for the exploration of data.

Part of the strength of this platform is its versatility—it can be applied to a wide range of data. It is also a dynamic interface that can react and evolve in real time, unlike Dressed in Data which is static. Even within the initial use-case of environmental health data, a variety of interactions can be imagined:

- **Community data**: Each bar represents one home in the study and bar color displays demographic information about that participant, such as city of residence. The height of the bar indicates the concentration of a particular chemical in that home. Bars react as users toggle through chemicals and demographic variables.
- **Time series data**: Each bar represents a chemical concentration. Different bars represent different time points. Bar color shows chemical properties such as chemical source, or time properties such as season. Users interact with bars to explore different chemicals and exposure events that occurred at various time points. Bars may also be arranged spatially—for example they may represent air monitors and be placed over a map of a city and change their heights in response to real time chemical levels.
- **Individual data**: The room represents all the exposures for a single participant. Each bar represents the concentration of one chemical in a particular medium (e.g. air, dust, or urine). The bar color displays chemical properties, such as health effects. Pushing down on a bar reveals information about reducing exposures.
to that chemical. Bars react to show the data of the participant currently in the room and allow users to explore different media or chemical properties.

For the purposes of this thesis, the data interactions will focus on community data for three chemicals at two time points using data from the Green Housing Study (GHS). Chemicals are diethyl phthalate (DEP), butyl benzyl phthalate (BBP), and diethyl hexyl phthalate (DEHP). Chemicals were selected based on data availability and confidence in data values. Time points are 6 months and 12 months after home renovations. Two time points were selected as an initial test of the interpretation of the movement of the bars; does seeing the process of bar heights changing from one time to the next bring meaning and might this be extended to more densely packed time series data? Each bar represents the concentration of the selected chemical in one home at the selected time point. Users employ tangible interfaces (soft switches) to toggle between variables.

Data selection and interaction designs were first simulated with computer based animations. Animations were created in R from GHS data and informed BigBarChart data use. Examples of these animations can be found in online materials.

**Hardware, software & system**

**Process**

BigBarChart is a unique form fairly distant from existing products. I thought of my prototyping process as the “Frankenstein” approach to creating new objects. I found everyday objects that approximated the desired functionality by scanning daily life with help from colleagues and friends. Initial prototypes patched these objects together casually (duct tape and hot glue) as a first test of feasibility. Promising prototypes were then iterated on by creating more precise connections and better fitted parts. Amazon’s “you might also like” feature proved to be a useful tool in the process of finding better suited parts. Finally, when a reasonably robust overall design was established manufacturers were engaged to create custom parts for pieces that could not be fabricated locally. This “Frankenstein” approach of repurposing everyday parts proved to be a productive mode of prototyping new forms.

Bars were actuated from the ground up so they could be installed quickly in a variety of places with minimal impact on the location. This is mechanically more difficult than actuation from the ceiling down, but provides benefits in mobility. Bars require only a power source and ideally a somewhat dim space for better viewing of the bar colors; no drilling or affixing structures to the ceiling is necessary.

**System overview**

The BigBarChart system contains a number of mechanical and electronic components. A brief description of each part is included below:
- **bars**: base + hampers
  - **base**: fixes parts, including hamper and electronics in place. Also positions gears and other mechanical moving parts used in actuation.
  - **hamper**: custom made pop-up laundry hamper (spring actuation).

- **electronics**: Linux machine + Arduinos + LEDs + motors
  - **Linux machine**: holds and processes data; shares data on request from master Arduino. Runs Ubuntu 14.04 with python 2.7.6 and pandas 0.14.0. Data is stored in csvs.
  - **Arduinos**: Arduino Unos control data selection and bar actions. One master Arduino controls button inputs and distribution of data to and from Linux machine to slaves. Each bar contains one slave Arduino that controls the motor actuation for that bar. Future work will add additional sensing to each bar.
  - **LEDs**: change the color of the bars. BlinkMMaxes with superbright red, green, and blue LEDs are networked with each other and Arduinos.
  - **motors**: actuate each bar (hamper) to change its height. Electronic Speed Controller (ESC) mediates communication between the motor and Arduino. Motor electronics are intended for use with hobby cars, planes, and helicopters.
  - **networking**: I2C protocol with the Wire Arduino library at a baud rate of 9600. Ethernet cables of ~2m connect bars to each other.
  - **power**: provided to the system through computer USB connector (Arduinos, 5V); power adapters (BlinkMMax; 12V); and LiPo batteries (motors; 12V). Future work will streamline this aspect of the system.

- **mechanical actuation**: strings + worm gear + motors
  - **motors**: RC vehicle based system; attached to base by custom aluminum L-bracket.
  - **aluminum adapters**: custom sized to connect the worm to the motor shaft
  - **aluminum axle**: holds the gear and rotates to wind the string to alter the height of the hampers. Fixed in bearings at either end.

- **input system**: soft switches
  - **bean bags**: stretch fabric with one side of conductive stretch fabric. Filled with lentils and color coded to match appropriate variable mat. Placing conductive side on mat completes variable selection.
  - **variable mats**: one selection mat allows user to choose a chemical, a second allows user to choose a time point. Mats are laser cut with words and icons communicating metadata such as chemical health effects and sources.
  - **future work**: objects such as doormats or bean bag chairs that are larger in scale and more bodily in form will be tested.
- **future work**: bar based inputs including users squeezing bars and compressing bars have been developed and prototyped using stretch sensors and pressure sensors; they will be integrated into the full design for increased interaction.

- **software**: python + Arduino master + Arduino slave
  - **python**: accesses csvs with environmental health data and responds to requests for data. Receives soft switch data from Arduino master, subsets data, calculates variables for each bar (e.g. motor direction, motor actuation time, color), stores existing state, and communicates necessary changes to Arduinos.
  - **Arduino master**: monitors soft switch state and sends data request to python when needed. Receives and distributes data from python to Arduino slaves for motor actuation.
  - **Arduino slave**: receives information from Arduino master and addresses motors accordingly.

Wiring diagrams for Arduino master and Arduino slave are included below.
Complete parts list, fabrication instructions, and diagrams are found in online materials. Because of the sharing goals of the project they are publicly available and are intended to encourage reproduction.

**Lessons learned in fabrication**

The fabrication process evolved throughout product development based on hands-on findings. Here are a few outcomes and changes in approach related to hardware, manufacturing, and data.

The initial set of customized hardware (e.g. axle, worm gear adapter) for BigBarChart was designed using CAD software for 3D printing. This created precise structures that could be easily aligned with other parts and replicated with little effort. Parts could also be immediately fabricated as they did not require lengthy searches for the correct source materials (e.g. rods of the correct diameter to serve as a baseline piece). Yet the limitations of these parts were also quickly apparent. First, parts required considerable print times (~6 hours) that were sometimes compounded by human or machine errors (e.g. printer failure, object scaling). Such errors wasted both material and time. Second, though parts could withstand initial testing, they were not mechanically robust enough for long term use or for more extreme cases within normal use, such a string jam in the system. Eventually, 3D
printed parts were deemed not suitable for the application and hand-lathed aluminum axles and worm adapters were fabricated (thanks Mike!). Two dimensional parts, such as the base of the bars, were designed in CAD software and cut on the ShopBot. L brackets for motor mounting presented a particularly interesting case, as they were designed as two 2D pieces and cut carefully on the water jet to preserve alignment between the two sides.

Manufacturing approaches are discussed in detail later but a few lessons learned are worth mentioning here. First, note that it may be difficult or impossible to fabricate some items in a typical machine shop as extremely specialized equipment and parts are required—this includes the large springs needed for the tall hampers. Second, US manufacturers may be accustomed to corporation sized orders or precision parts and thus be unwilling to consider smaller scale runs. Alibaba can be a useful tool for identifying manufacturers abroad for collaboration.

Before using the data in BigBarChart, some initial animations were created using R to give a general impression of the interaction. A few lessons were learned from this exercise, though the medium could not provide the full experience of the installation. First, it became evident that using buttons to adjust the data ratios (e.g. the denominator comparison value) was misleading and should be avoided. The motion of the bars suggested the values were changing, though in fact only the comparison amounts that were being adjust—e.g. PCB concentrations in GHS versus highly exposed factory workers or rural Germany. In the future, bar colors or projected axes may be used to convey similar information. Second, creating animations demonstrated the extent to which movement can attract attention. It is possible that BigBarChart would be a stronger piece if it capitalized on its motion, perhaps being used with live data or time series data. Testing of motion in the human-sized installation is needed to fully assess this.

**Manufacturing**

**Manufacturing partners**
Manufacturing partnerships were established via two channels: Alibaba and the network of Andrew "bunnie" Huang. Custom manufacturing was sought after initial searching and design testing of related mainstream products available through Amazon and an attempt at sourcing appropriate machinery locally.

The first manufacturing need was for modified pop-up laundry hampers to be used as the bars. Basic requirements of the new form were: increased height (with increased spring strength and diameter to support the height) and minor accessory modifications to the design (e.g. added ties, open bottom, partially closed top). Initially, American based spring manufacturers, metal workers, and machine shops were contacted in an attempt to fabricate large custom springs. Possible machinery included WB-4120 3D CNC wire bender (Taiwan Spring Machine, 2011), NewForm Tech: 3D wire bending machine MB-R
series (NewForm Tech, 2010), and Reivax Máquinas MAQ-3D-CNC (ReivaxMaquinas English, 2011). Organizations did not appear to have the appropriate machinery for springs of this scale (both diameter and length) and expressed little interest in pursuing this question further. Often their focus seemed to be on large industrial orders or precision work.

To find other possible manufacturers, I searched the Alibaba website for pop-up laundry hampers or similar products such as pop-up tents, crawl tunnels, and large springs. I contacted over twenty manufacturers via email or the Alibaba messaging interface with a description of the product and basic information about the desired design. Little attention was paid to minimum order requirements listed by the manufacturers when reaching out to potential partners, as almost all minimum order sizes (500 to 5,000 pieces) were substantially larger than the desired initial order quantity (~50).

Many companies contacted via Alibaba did not respond to the initial inquiry and others reported reluctance to engage with a difficult project and/or small order size. Casual conversation with individuals who have collaborated in the past with manufacturers abroad suggested that less well established manufacturers may be willing to compromise on order size and difficulty when working with MIT or other well-known American organizations in order to increase their company’s visibility and credibility in the eyes of future buyers. For those that did respond, a number of emails, often including images or drawings, were exchanged to discuss design possibilities and feasibility of the project. Additional companies opted out in this process, leaving one company: the Yiwu Kenong Daily Commodity Factory in Zhejiang Province (2014).

At this point in the process, I discovered that the Media Lab was pursuing the launch of a somewhat formalized partnership with manufacturing groups in China. Bunnie Huang was identified as the facilitator of these budding relationships and was contacted for advice on manufacturing this particular part. Using his ongoing relationship with the firm AQS Electronic Manufacturing Solutions (2014), he honed in on a couple of manufacturers that might have the appropriate expertise. I reached out to these companies, including the Yiwu Kenong Daily Commodity Factory, as found through Alibaba, with a similar project description and request as used with Alibaba contacts. Ultimately, a partnership with Yiwu Kenong Daily Commodity Factory was established.

A second manufacturing need was telescoping poles. These manufacturing relationships were built primarily through Alibaba, as well as a casual request for recommendations from existing manufacturing partners. I initially contacted over 20 manufacturers with a focus on those creating carbon fiber poles. Applications included flag poles, pointers, fishing poles, and walking sticks among others. Order requirements for telescoping poles often had lower minimums (as low as 10) than hamper manufacturing. Design approach and desired product outcomes were discussed with responding companies, including extended and closed lengths, material strength, and pole diameter. Due to the compressed timeline, product manufacturing times were also a consideration. Because the telescoping poles
were required immediately, could be ordered in lower numbers, and showed a great range in quoted prices, two factories were contracted for preliminary test work to permit prompt comparison in quality. These factories are: Jiangsu Toptek Composite Materials and Caroline Dong's unnamed company, both in Jiangsu Province, China. Ultimately, Jiangsu Toptek Composite Materials provided superior and cheaper products on a faster timeline and was selected for the full order.

**Manufacturing design process**
Successfully communicating with manufacturers throughout the design process was a challenging task. Challenges were based in logistics, language, knowledge, and philosophical approach. Strategies for optimizing outcomes were developed on the fly and met with mixed success.

Logistical challenges included time differences, shipping, and payment. Because of the large time difference between the two locations, parties were often offline during each other's work days or waking hours, thus causing a lag in email communication. This was a fairly unavoidable situation, but often caused delays in the process as questions took over 24 hours to resolve. This was minimized as much as possible by taking care in composing emails and making attempts to be available outside of normal hours during crucial periods.

Shipping could also be a challenge. One manufacturing partner could not ship directly to the US, thus AQS kindly received the package locally and arranged international shipping. This lack of direct shipping caused additional costs and delays and required the assistance of an outside organization. Occasionally, customs held packages which caused additional delays in delivery; in one case, an order of motors was returned to Hong Kong due to a faulty customs form and had to be reshipped (~2 week delay), in another instance customs officials temporarily held goods and phoned the receiver to confirm their intended purpose (~2 day delay). These challenges are not easily avoidable at this point in time, but awareness of them can facilitate planning during the fabrication process.

Arranging payments was often a time consuming process. Partnering manufacturers suggested options including bank transfer, PayPal, and Western Union for payment. Alibaba itself offers payment services, but this option was not explored, nor was it proposed by any of the factories as a desirable payment method. Other less established possibilities may be explored in the future including Stellar, bitcoin, and Stripe. When available, PayPal was the most time efficient and reliable option. Banking transfers required detailed information that neither manufacturers nor purchasers could identify and also necessitate in-person visits to the banking branch. In the end, this approach was avoided. Western Union payments were used when PayPal was not available. Unfortunately, online Western Union charges often failed for unknown reasons that customer service agents and credit card companies were unable or unwilling to divulge. Even successful online payments caused additional credit card charge fees (immediate interest accumulation and cash-back charges). In-person Western Union
payments were used on a few occasions. While they were consistently processed correctly, these payments were time consuming and required carrying significant amounts of cash to complete the payment. This would not be a feasible option when scaled. All payment methods required some level of risk and trust between purchaser and manufacturer.

Challenges based in differences in language, knowledge, and philosophical approach were quite entwined in communicating about the design and manufacturing needs. Language challenges were well known from the beginning of the collaborations, as purchaser has no proficiency in Mandarin and the English of some manufacturing contacts was limited. Limited manufacturing knowledge on the purchaser’s side included lack of correct vocabulary for parts, differences in expectations around how parts were assembled, and limited knowledge of manufacturing constraints. Manufacturers seemed to be accustomed to fabricating well established designs with concrete requirements and use-case scenarios developed; a prototyping approach for creating new objects through repeated testing, refinement of constraints, and discussion seemed outside their normal process.

A few communication approaches were useful in bridging these gaps. Helpful strategies included:

- **examine existing products**: identifying current products similar in form or function was useful in communicating the overall concept to manufacturers. Inspecting these products or product photos also served to fill purchaser’s gaps in knowledge around product components (e.g. pop-up hamper includes two material types, spring, handles), typical dimensions (e.g. inner and outer diameters of telescoping poles), product types (e.g. straight telescoping poles vs tapered telescoping poles), or functionality (e.g. use of hooks to fasten hamper closed).
- **share ideas through photos**: often labeled photos could quickly clear up confusion that was not well identified or could confirm that parties were in agreement on a course of action when language was unclear.
- **share ideas through video**: video with explanatory text throughout helped bridge language barriers and identify misunderstandings. Videos were used for demonstrations mimicking the desired outcome of the project, possible fabrication methods, and ongoing faults with the design. Note that transfer of video was sometimes difficult as file sizes were too large for email, Vimeo and youtube were not available to manufacturers, and compressed files placed on a server could not be reliably extracted by manufacturers.
- **communicate carefully**: short messages, consistent and concise language, and well defined questions or suggestions helped smooth communication. This had to be balanced against the communication lag caused by time differences; sometimes emails needed to contain multiple topics which could create confusion.
- **express prototyping expectations**: because some manufacturers seemed accustomed to receiving well outlined designs, it was useful to state and reiterate...
that some testing might be needed and multiple sample runs could be required. Further developing the manufacturer’s role in this testing would be useful.

- **request advice or knowledge**: explicit requests for input or advice early in the communication process was used to gain knowledge of the manufacturing steps and set up expectations of information exchange and dialogue around the design. This process was especially useful in selecting manufacturers and also establishing the limits and constraints of their process. It helped locate manufacturers that thought critically about problems and could provide solid feedback on the process. It also provided information about what parts were further outsourced (e.g. steel wire for hampers is contracted out; molds for telescoping poles may be contracted out).

- **confirm outcomes**: verifying all product parameters (dimension, quantity, shipping) for the final order was critical due to the high volume of requests manufactures receive and the constantly evolving designs in the prototyping process.

- **be prepared for unexpected delays or challenges**: small sample runs may be delayed by large requests from big companies or long time patrons. Parts that are contracted out (e.g. steel wire) may change in specification over short amounts of time or become difficult to attain. Try to account for these issues when developing a timeline.

**Manufacturing ethics & the environment**

Data Experiences are meant to bring data off the screen and into the physical world. It is hypothesized that the tangibility of data, its proximity to the body, and the human-sized scale of the form will bring a broader range of people to interact with the data, create a common space for conversation between researchers and the community, and induce new attitudes towards the data—“wonderment” or an artistic eye—that will be profitable to engagement and new understandings. Yet, especially when using these objects to display environmental data, it is important to note the downsides of physicality. Digital “objects” (e.g. jpegs of barplots) consume comparatively minimal resources and space. They are quickly malleable and editable, creating limited waste when there are errors in their production. Physical objects do not have these affordances; the creation of each bar requires multiple enduring objects and poorly measured material cannot be simply uncut. Clearly current Data Experiences add to the world of “consumables” that may have negative environmental impacts and must show themselves to provide benefits beyond more digital approaches such as augmented reality.

However, BigBarChart pursues an ethically important goal of data reporting to study participants (Brody, 2007). In the long term, the success of Data Experiences must be analyzed against its cost in resources. As an initial attempt to increase long term net utility, BigBarChart was created to be reusable—the barchart form is not specific to the particular environmental health data but instead can be applied to a range of datasets and types of data. The shapes and colors of the bars can be programmatically manipulated; the form takes some of its existence from the physical realm and some from the virtual realm and thus begins to better manage its use of resources. Ideally, BigBarChart and other Data
Experiences are a step on the road to Ishii’s vision of “radical atoms” (Ishii, 2012). These objects are infinitely editable and reshapable though physical in nature and thus provide the benefits of the physical and digital worlds. Follmer’s inFORM is an early example of how this approach may be implemented and applied (Follmer, 2013).

Additionally, organizations that value social consciousness or serve constituents with these values may prefer to avoid Chinese manufacturing for economic, environmental, or human labor reasons. Assessing the validity of these concerns and fully treating this complicated topic is outside the scope of this thesis, but it is worth briefly addressing the environmental concerns as a prompt for future exploration since the data used with BigBarChart is environmental.

Many have documented high levels of air pollution in China, often tied to industry, and have observed connections to disease and mortality (Chen, 2013; Zhang, 2011; Lin, 2014). Outdoor air pollution in China is linked to over 400,000 deaths per year (Gong, 2012). Sources of this pollution include traffic pollution, manufacturing, and household heating among others (Gong, 2012). US air pollution related deaths are estimated at 50,000 per year (National Oceanic & Atmospheric Administration, 2014); thus these issues are not unique to China, though they may be more severe there.

Environmental regulations throughout the world are beginning to address air pollution and health. The American Environmental Protection Agency and the Chinese government have worked together through the EPA-China Environmental Law Initiative. This initiative has established air pollution standards and increased pollution monitoring (Ministry of Environmental Protection, 2013). Yet, even in the US, chemical exposure through products themselves is not closely monitored. The Toxic Substances Control Act is perhaps the primary regulatory guideline for chemicals in the US but it does not require health data for new chemicals on the market and the vast majority of chemicals in use remain untested and unidentified in products (Lohmann, 2013). Thus both countries have room for significant progress in this regard, with the European Union’s REACH legislation proposed as the best existing model for this effort (Lohmann, 2013).

**Product based chemical exposures**
BigBarChart was designed as an extension of work in data reporting in environmental health. Though its design was intentionally broad enough for use with a wide range of data, its application to environmental health prompts us to at least provisionally consider the possible exposures it presents as a consumer product. Relevant individuals and exposure pathways include: parts manufactures, prototypers, and study participants (e.g. installation visitors). Fully assessing these exposures and addressing possible points of concern is a large endeavor left as future work; this project provides an exciting opening to the conversation on emerging contaminants in prototyping and early stage fabrication.
First, note that the Restriction of Hazardous Substances Directive (RoHS) can serve as a starting point for managing chemical exposures (RoHS, 2014). This European Union regulation is intended to reduce quantities of hazardous e-waste by controlling use of six materials in electronics: lead, mercury, cadmium, hexavalent chromium, polybrominated biphenyls, and polybrominated diphenyl ethers (RoHS, 2014). Note that many of these chemicals and chemical classes—PCBs, PBDEs, metals—are included in Silent Spring Institute’s sampling protocols, though Silent Spring Institute measures a much longer list of emerging contaminants (Rudel, 2010). Under RoHS, electronics in the EU (whether fabricated there or imported) must contain less than 1000 ppm by weight for each of these chemicals or chemical classes, except cadmium and mercury which are limited to 100 ppm (RoHS, 2014). This rule applies to each component of the final product, not the product as a whole, though there are a number of exemptions, such as lead levels in high melting temperature solders (RoHS, 2014). It does not apply to non-electronic products and the United States is not subject to these standards, though California has adopted a directive that begins to move in this direction and large manufacturers may follow these regulations globally to streamline their supply chain (RoHS, 2014).

A range of goods were used in the design, prototyping, and fabrication process. Electronics were not checked for RoHS compliance; future versions will incorporate this as an initial step towards reducing electronic waste and chemical exposures. Parts used are briefly listed and categorized below with preliminary information about the exposures they could imply.

**design components:** parts used in the final design

- textiles (fabric, thread, string): may contain treatments or flame retardants; organic fabric was used in some instances (Rudel, 2003; Dodson 2012).
- conductive textiles: mix of cotton, silver, and spandex. (Sparkfun, 2014).
- mechanical parts (brackets, bolts, bearings, axles): all metal based, primarily steel and aluminum (McMaster, 2014).
- electronic accessories (cables, wire, breadboards): high plastics use, may contain phthalates, flame retardants (Wermter, 2012), or metals (RoHS, 2014).
- MDF: contains sawdust with wax and resins for binding, and often formaldehyde (Hodgson, 2002).
- electronics (Arduinos, motors, computer): many electronics contain flame retardants (Dodson, 2012). Arduinos are RoHS compliant (Sparkfun, 2014), other parts are not specified.
- hampers: had a strong smell of unknown source when received.
- carbon fiber telescoping poles: had a strong smell of unknown source when received. Also had very light dust film on surface.

**fabrication components:** goods used in the fabrication process
• bearing degreaser: MSDS identifies it as “toxic by inhalation;” company recommends it be treated as hazardous waste in disposal. Contains suspected carcinogens and cannot be shipped to California (Pedro’s, 2014; Mrak, 2013; Pedro’s, 2012).
• solder: may contain a wide variety of metals, including lead, silver, aluminum, copper, zinc, cadmium and others (Wikipedia, 2014).
• adhesives (fabric fusing material, hot glue, duct tape): may include phthalates; are applied at high heat which may present unique inhalation exposures during application (Rudel, 2003).

prototyping components: parts used for testing that were not included in the final design
• plastics (Tupperware, 3D printed parts, plastics sleds, acrylic, etc): may contain phthalates (Rudel, 2003).

These potential chemical sources have varying implications for exposures for manufacturers, prototypers, and study participants. Occupational exposures to chemicals are often orders of magnitude higher than everyday exposures (Wolff, 1985) thus factory workers may face the highest risk. Prototypers also are subject to heightened exposures during the fabrication process—for example in applying adhesive or cutting MDF—though these exposures are likely confined to a shorter stretch of time, as prototypers make only a limited number of each component. Study participants are at the lowest risk for exposure as they spend limited time with the materials and do not interact with them as closely. Yet participants may have particular concerns or chemical sensitivities that heighten the importance of these exposures; for example Green Housing Study families contain a child with asthma.

Fully addressing these exposures is outside the scope of this study but is a very important long term consideration for the success of Data Experiences in environmental health. First steps include finding replacements for items with known exposure risks, such as MDF. Other easily exchangeable items with less certain exposure risks could also be replaced; for example exclusively organic textiles could be used. Assessing and minimizing chemical exposures from items such as electronics may require action on a policy level or close partnerships with manufacturers with similar goals.

Preliminary actions have been taken in an attempt to minimize exposures during fabrication. First, hampers and telescoping poles were left open in a well-ventilated area upon receipt until their smell decreased. Telescoping poles were also washed with dish soap and warm water to eliminate the light dust-like coating. Many plastics have been removed from later designs, including Tupperware and 3D printed parts, and hot glue use has been minimized. This is beneficial from a chemical exposure standpoint, mechanical standpoint, and aesthetic standpoint as most of the parts have been replaced with metal pieces that create stronger and more elegant connections. Alternative modes of creating fast adhesive connections that are strong enough for short term testing would be a great benefit to early prototyping efforts. Additionally, basic precautions were taken in creating
prototypes to minimize fabrication-related exposures. This includes frequent hand washing, use of exhaust fans and dust collectors, and situation of products in well ventilated areas. Lead free solder was used when available and dermal contact with solder was minimized.

Prototype outcomes

Images of the major components of BigBarChart are shown below. Additional images and video can be found in online materials.

*soft switches for selecting time variable (laser cut)*
soft switches for selecting chemical variable (laser cut)

bean bags for soft switches; one side conductive fabric
custom motor bracket (water jet) and motor adapter (lathed)

motor system: motor, battery, electronic speed controller
bar electronics: Arduino, BlinkMMax, connections to motor, I2C, power

bar electronics including motors
tapered telescoping poles

bar: custom human-sized pop-up laundry hamper
bar base: worm gear, motor hardware, telescoping poles

fully assembled bar (extended)
fully assembled bar (compressed)

set of networked bars: functional demo
Evaluation

Initial evaluation and iteration of BigBarChart prototypes was done through informal conversations with Media Lab researchers, Media Lab member companies, and environmental health researchers and a small user study. A wide range of member companies and partners expressed interest in using BigBarChart, including financial corporations, technology companies, and design organizations. Many individuals mentioned particular use cases such as educating clients, breaking down barriers during official meetings, and creating a space for playfulness. As this form of data communication is quite novel it may require additional iteration before converging to a stable prototype. Protocols for qualitative user studies were developed and tested on a small sample to begin to assess user impressions of both the scientific and artistic aspects of the form. Information from these initial studies will inform future prototypes that will be tested with users closer to the data, specifically environmental health study participants and researchers engaging with BigBarChart using their own data.

Full evaluation of BigBarChart was delayed until a prototype aligned with the initial aesthetic vision could be achieved. A change between the sample run and the final run of custom hampers caused structural issues with the hampers so the design is no longer self-supporting. The cause of this issue remains unclear, though it may be due to alternations to the springs used in the hamper by the manufacturer’s metal supplier or a slight change in dimensions. Process debugging is ongoing and expected to be resolved shortly. Findings from a small scale user study are presented below.

Users’ overall perceptions of the look of the bars seem to align with the intended aesthetics even though the tested prototype did not fulfill the full vision due to delays in manufacturing. The bars seemed to capture the attention of the users who described them with words like “large,” “colorful,” and “fun.” Users also expressed interest in the fabrication, materials, and actuation mechanism for the bars and the soft switches.

Users interacted with the soft switches that control the bars with minimal guidance, though further design iteration and labeling could clarify this process. One user was unaware that both sets of switches could control the bars, instead assuming only the set closer to the user could be employed. The primary divisions between chemicals or time points within the mats and the intended placement of the bean bags was also initially unclear to one user who attempted to place the bean bag on the sources and health effects icons. After some trial and error both users successfully manipulated the bars. At this point, the overall structure of the switches seemed to become clear to the users and they correctly identified the health effects and sources of chemicals. Improved design (e.g. use of color or fabric matching to indicate where the bean bag should go), labeling (e.g. mat with chemical options has overall label), and further verbal description of the dataset would allow users to understand the interface more quickly.
Fluidity and clarity of interaction with BigBarChart could be improved by increasing the responsiveness of the bars. The current prototype includes a number of delays and bars take over a second to start responding to a change in the switch. This delay caused users to pick up and replace the bean bag or to move it again as if they doubted that it was working. Users also did not observe the color changes because of the slow response times (colors change after the motors), limited strength of the LEDs, and difficulty in seeing colors through the colored fabric; these issues are addressed in future designs. Overall, the interface must meet people’s expectations of immediate reaction time for real world objects. It is expected that many of these delays can be removed in future iterations and that during any remaining delays a blinking light or equivalent physical counterpart to Apple’s spinning pinwheel can be used to assure the user that something is happening.

Users immediately inquired about the scale of the data, while also recognizing the relative changes (e.g. the data was less than, more than, or almost the same as the previous set). One user expressed interest in knowing more about the context of the amounts; they wondered whether values were dangerous and expressed interest in including a comparison bar where the hazardous levels could be displayed. Future work will explore this approach as well as bar color and wall projection for providing context.

**Limitations & future work**

Future work on BigBarChart includes technical and interaction modifications to create a next generation prototype for larger scale testing.

Technical modifications will increase the durability and aesthetics of the system and decrease its exposure impact. The MDF base will be replaced with a healthier alternative, solder options will be explored, and textiles will be replaced with organic alternatives where available. Electronics will also be replaced with RoHS compliant electronics. This will decrease the chemical exposures of prototypers and study participants, a particularly relevant concern when displaying environmental health data to study participants who may have chemical sensitivities.

Technical modifications to the code and the electronics will increase robustness and interaction possibilities. I2C networking may be replaced with an option more suitable to distance. Wireless communication will be explored to increase the mobility of the bars. Motor, LED, and Arduino powering will be streamlined to one battery source to further reduce wiring. Mechanical components, such as telescoping pole support system, will be further refined to increase the reliability and aesthetics of the system. A second round of user testing will occur with ~10 units to gather feedback on aesthetic and experiential goals.

Later, a further refined system will be fabricated on a larger scale (~30-50 units) to increase wonderment and display larger datasets. This includes a more complete version of the
GHS data as well as new datasets, for example live streamed time series data, financial data, and personal health data. Bars will be enhanced with additional interaction capabilities in response to user feedback and to suit the new datasets, perhaps showing relationships between variables such as traffic and pollution. Tangible interfaces for variable selection will also be scaled up further (e.g. bean bag chairs) to increase the bodily nature of the interaction.

Future user testing will occur with environmental health study participants and researchers to assess the impact of the experience when interacting with one’s own data. This also allows comparisons with other methods of reporting, such as paper based or digital report-back systems currently in use and under development. As BigBarChart is designed to promote community interaction and engagement, the system will be tested with multiple concurrent users, including study participants and environmental health researchers, to look for effects on engagement levels and styles (e.g. length of interaction, depth of interaction). These studies will help assess whether BigBarChart achieves its goal of lowering the barrier to data engagement, helping participants achieve a deeper understanding of environmental health data from their community, and prompting participants and researchers to interact in novel ways. They will also evaluate the extent to which BigBarChart succeeds in being immersive, interactive, artistic, and tangible.
Conclusion

Data experiences in environmental health

BigBarChart and Dressed in Data are early manifestations of a growing field of Data Experiences created using environmental health data from two Silent Spring Institute studies. Dressed in Data uses data from the Household Exposure Study conducted in Northern California in 2006 and BigBarChart uses phthalate data from the Green Housing Study conducted in Cincinnati and Boston in 2013.

More broadly, BigBarChart and Dressed in Data serve as example Data Experiences that help define an emerging space. They differ from existing physical manifestations of data by being highly immersive, flexible, and human-scaled. These particular Data Experiences are also a somewhat “neutral form”—the objects displaying the data are not exclusively designed for environmental health data and so can be quickly used with a range of small to middle sized datasets, from social media data to finance data to health data.

The development of Data Experiences responds to research needs in environmental health and science. Science communication and public engagement is of increasing interest as demonstrated by the citizen scientist movement and community based research practice. Specifically, multiple research agencies have called for increased results-reporting in environmental health and study participants are eager to receive this information in an engaging and empowering manner (BEST, 2006; The National Conversation on Public Health and Chemical Exposures, 2011). Future implementations of BigBarChart and Dressed in Data with Green Housing Study participants will serve as a starting point for new methods of interaction between researchers and communities.

Future work

BigBarChart and Dressed in Data suggest many areas of future work in Data Experiences. This includes open source Data Experiences, iterations on current Data Experiences, field-specific Data Experiences, and further testing of Data Experiences. Furthermore, both Data Experiences could add features based on user feedback and hardware and software could be iterated to develop a fully deployable product.

Immediate next steps along these lines for both Dressed in Data and BigBarChart include:

*Full open source documentation and support:* Code, parts lists, and fabrication files and instructions are available for BigBarChart and Dressed in Data through the website http://dataexperiences.media.mit.edu. This provides a strong starting point for groups to
fabricate their own data experiences and explore new directions and implementations. Yet this documentation may not be sufficient for community groups or organizations without engineering or technology resource—both in personnel and machinery. Future work includes creation of accessible documentation of these processes including tutorials and physical resource documentation to empower a broad range of groups to use these approaches on their own data (e.g. high school students creating clothing expressing their Facebook data; BigBarChart used in financial meetings). Partnerships may be established with interested organizations to create example projects that can serve as a template and inspiration for others.

Design iteration based on increased robustness and design values: Fabrication and initial researcher testing brought to light functional weaknesses in the prototype in normal use scenarios. For example, communication between bars in BigBarChart can fail or become prohibitively slow when more than eight bars are networked. Also, two dimensional areas were initially used to compare concentrations in Dressed in Data to create an aesthetically pleasing pattern, but research suggests this pattern may not be as legible as one-dimensional representations (e.g. lines or bars). This information will be integrated into the next generation prototypes before full user testing.

User studies and full scale implementation as a report-back tool: Protocols (see appendix) for initial qualitative user studies of these Data Experiences have been developed and tested. Full studies will be conducted as future work, and will provide a launching point for design iteration so Data Experiences can be used to report environmental health study results to participants. Report-back testing will be conducted with study participants as well as environmental health researchers. This phase of the study will include semi-structured interviews similar to those used to measure and describe the success of paper and digital report-backs in similar household exposure studies.

Topically focused data experiences: BigBarChart and Dressed in Data are designed to be flexible platforms for use with a range of data beyond environmental health data. This has the benefit of bringing Data Experiences to broad audiences and encouraging brainstorming across fields. Yet it does not capitalize on specifically themed objects that can strengthen messaging and promote behavior change. Future work will consider forms specific to emerging contaminants in the indoor and community environment. Partnerships with advocacy organizations and designers will help develop appropriate action-based Data Experiences related to environmental health. For example, the forms of each bar in the barchart could be made to represent the source of the particular pollutant (e.g. smokestack for industrial pollutants, hairspray can for personal care products). Or, a Data Experience around exposure to personal care products could be a bathroom mirror that reflects both the user’s image and their internal “chemical image” as make-up and other products are applied, showing the user the potential harms of their beautification routine.
Supplemental

Online documentation

Progress on this and other projects was documented through a daily image or video and one line description. Hints at failed paths and the process of converging to successful approaches can be found at: www.lookimadeyouathingtoday.com.

BigBarChart and Dressed in Data are open source hardware and software projects. Documentation including videos, code, fabrication files, materials lists, and assembly instructions can be found at: http://dataexperiences.media.mit.edu.

Manufacturing partners

Factory information for manufacturing partners is included below:

- **Pop-up laundry hampers**
  Contact: Alice Yuan; alice@zjywcy.com
  Fotang Town, Beiyuan Indrustry, Iwu City, Zhejiang Province
  Yiwu Knong Daily Commodity Factory
  Tel: 86-579-85203790; 86-15825780336
  Skype: chenyusales

- **Telescoping poles**
  Contact: Lynn Wang; lynnwang1962@hotmail.com
  Jiangsu Toptek Composite Materials Co., Ltd.
  69 Dongfeng Road, Rugao City, Jiangsu Province, China
  Tel: 86-513-87307123, 86-15950820196
  Skype: lynnwang1962
Semi-structured interview: BigBarChart

Interview preparation:

- discuss and sign informed consent
- begin audio recording of the interaction with permission of the participant
  (*future versions may include video recording*)

Interview structure:

- Phase 1: ~10 minutes of participant interaction with interface
- Phase 2: ~15 minutes of questions for the participant
- N.B. Interview is semi-structured and phases may overlap.

Phase 1: introduction and interaction

This interface is called BigBarChart. Right now it’s showing data on chemicals found in homes in a study conducted in Boston, MA.

Feel free to explore BigBarChart now and see what it does. Please speak aloud your thought process as you do so. We’ll do this for about ten minutes, or you can let me know any time if you’re done exploring.

*Prompt if needed: is there a way to change what BigBarChart is showing?*

Phase 2: questions

How would you describe BigBarChart to a friend?

*Prompt if needed: how would you describe what information BigBarChart shows?*

What did it feel like to look at data like this?

Did you learn anything from seeing BigBarChart?

*If yes:* What did you learn?

What is your favorite thing about BigBarChart?

What would you like to change?

Was there anything in BigBarChart that was confusing?

Was there anything you wished BigBarChart could have done that it didn’t?

What information did BigBarChart show?

*If confusing:* What parts of the data could you look at with it?
How would you describe the bars? What three adjectives would you use to describe them?

Is there anything you would like to add about your experience with BigBarChart?

**Interview close: thank you and questions**

Thank you for participating in the study. If you have any questions about the study or the interface you just interacted with, I'm happy to answer them now.

Thank you for your participation. [$15 Amazon gift card].
Semi-structured interview: Dressed in Data

Interview preparation:
- discuss and sign informed consent
- begin audio recording of the interaction with permission of the participant
  (*future versions may include video recording*)

Interview structure
- Phase 1: ~5 minutes of garment introduction and observation
- Phase 2: ~15 minutes of questions for the participant
- N.B. Interview is semi-structured and phases may overlap.

Phase 1: introduction

These outfits display information about chemicals found in one person’s home. The data is represented in the lace pattern on the outfits. The squares represent individual chemicals found in the person’s home. The size of the squares represents the relative amount of the chemical that was found. Each garment looks at one type of chemical—for example, the red outfit shows data about chemicals from plastics. Overall, data from over 100 chemicals are included in the outfits.

This shirt shows the same data that’s in the lace pattern on the yellow factory pollutants outfit. We won’t consider it for the first questions, but we will look at it in the later ones.

Phase 2: questions

At first glance, which garment seems to have the most chemicals? What makes you pick that one?

At first glance, which garment seems to have the highest relative concentration of chemicals? What makes you pick that one?

What do you think when you look at the lace pattern?

If you touch the garments, how do they feel?

Where do you think a garment like this might be seen or worn? At what sort of occasion?

Have you ever designed, sewn, knitted, crocheted, or made other things from textiles before?

*If no:* Is this something you find interesting or you think you might learn in the future?
If yes: What sort of things have you made?

Would you consider wearing a garment like this that was about chemicals found in your home or your town? What if it showed data about something else? Why is it or isn’t it an appealing idea to you? [daily wear]

Would you consider sharing data on chemicals detected in your home online—say on social media, on a blog, or as part of a research study—if it was available to you? Why or why not?

What if the pollution was measured outside your home—say somewhere within a half mile of your home? Would you consider sharing it on social media, a blog, or as part of a research dataset? Why or why not?

Do you have any other thoughts, comments or questions you’d like to share?

Interview close: thank you and questions

Thanks so much for participating in the study. If you have any questions about the study or the garments you just saw, I’m happy to answer them now.

Thank you for your participation. [$15 Amazon gift card]
Bibliography


Bevans, Chrisopher. Personal communication, January 2014.


Chen, R., Peng, P.D., Meng, X., Zhou, Z., Chen, B., Kan, H. Seasonal variation in the acute effect of particulate air pollution on mortality in the China Air Pollution and Health Effects Study (CAPES). Sci Total Environ. April 15; 0: 2013


Chi, P., Xiao, X., Chung, K., Chiu, C., Ishii, H. Burn Your Memory Away: One-time Use Video Capture and Storage Device to Encourage Memory Appreciation. CHI 2009.


Follmer, S., Leithinger, L., Olwal, A., Hogge, A., Ishii, H.. inFORM: Dynamic Physical...
Affordances and Constraints through Shape and Object Actuation. Proceedings of the 26th annual ACM symposium on User interface software and technology, 2013.

Food and Drug Administration. “Inspections, Compliance, Enforcement, and Criminal Investigations.”


Kim, S., Paulos, E. Listening to Air Quality. EISE, 2009.

Kosslyn, Stephen; *Graph Design for the Eye and Mind*. Oxford University Press, 2006.


Luescher, S. *Beyond Visualization – Designing Interfaces to Contextualize Geospatial Data*. MIT Media Lab, 2013.


Vande Moere, A. Beyond the tyranny of the pixel: Exploring the physicality of information visualization. In IV’08, 2008.


Vande Moere, A., Patel, S. Analyzing the design approaches of physical data sculptures in a design education context. In Visual Information Communications International (VINCI’09), 2009.


