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Creative User-Centered Visualization Design for Energy Analysts and Modelers

Sarah Goodwin, Jason Dykes, Sara Jones, Iain Dillingham, Graham Dove, Alison Duffy, Alexander Kachkaev, Aidan Slingsby, and Jo Wood, *Member, IEEE*

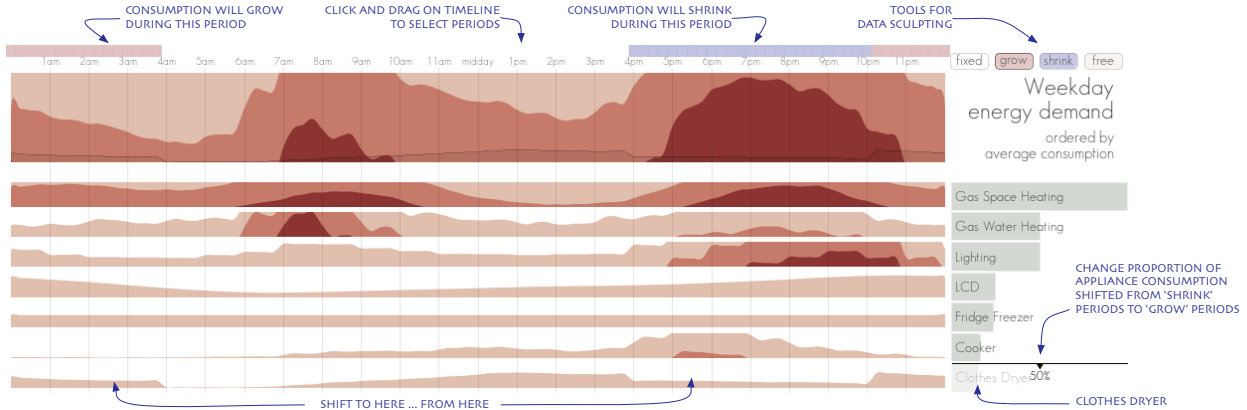


Fig. 1. *Demand Horizons* show modeled weekday energy demand over 24 hours amongst high consumption domestic appliances. *Data Sculpting* allows us to shift consumption interactively by ‘moulding’ the horizons to explore ‘what if?’ scenarios. For example, here fifty percent of ‘Clothes Dryer’ consumption is shifted from the evening peak to a period when overall demand is lower.

Abstract—We enhance a user-centered design process with techniques that deliberately promote creativity to identify opportunities for the visualization of data generated by a major energy supplier. Visualization prototypes developed in this way prove effective in a situation whereby data sets are largely unknown and requirements open – enabling successful exploration of possibilities for visualization in Smart Home data analysis. The process gives rise to novel designs and design metaphors including *data sculpting*. It suggests: that the deliberate use of creativity techniques with data stakeholders is likely to contribute to successful, novel and effective solutions; that being explicit about creativity may contribute to designers developing creative solutions; that using creativity techniques early in the design process may result in a creative approach persisting throughout the process. The work constitutes the first systematic visualization design for a data rich source that will be increasingly important to energy suppliers and consumers as Smart Meter technology is widely deployed. It is novel in explicitly employing creativity techniques at the requirements stage of visualization design and development, paving the way for further use and study of creativity methods in visualization design.

Index Terms—Creativity techniques, user-centered design, data visualization, smart home, energy consumption

1 INTRODUCTION

These are exciting times for utility companies and their energy analysts – the energy domain is data rich and globally significant. Energy analysts and modelers are now striving to effectively use the volumes of data from emerging Smart Home technologies to understand consumer behavior, conserve energy and manage supply and demand. Data visualization can offer great potential in this domain, but developing appropriate solutions presents considerable challenges, since the nature of the data are relatively unknown and the needs of energy data analysts and modelers are not yet well understood. The design brief is therefore essentially open-ended.

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Participatory approaches to user-centered design, in which users and other stakeholders are involved in co-creating requirements and designs for interactive systems can lead to solutions that are more useful and usable [35]. We have successfully used human-centered approaches in the design of visualization solutions before and have documented these in detail [27]. However, the role of *creativity* in these approaches has as yet been only implicit. Over the last decade some fields of interactive systems development have increasingly focussed on introducing elements of deliberate creativity into participatory user-centered design processes. The aim here is to enable all participants (users, designers and other stakeholders) to contribute to the exploration of new fields and the generation of requirements and design ideas for novel and useful systems [1, 6, 53]. Establishing requirements can be considered a fundamentally creative process whereby requirements analysts and stakeholders work collaboratively to generate ideas for software systems [29, 30, 32]. Indeed, Robertson [42] regards requirements analysts as inventors who bring about innovative change in designs to establish advantage. Techniques for deliberately introducing creativity into the process of user-centered design can be used effectively in this context. For example, Schmid [46] used creativity triggers [42] to help workshop participants invent requirements, whilst co-creation [45] and creativity workshops [24, 31] have been shown to be effective in generating novel requirements.

Here, we report on work in which we augment a user-centered approach to design with techniques for deliberately stimulating creative thinking when establishing context of use and developing requirements. We do so in the context of an investigation into ways in which a major energy supplier could use visualization to derive value from data that will become available following the wider adoption of Smart Home technology, by producing a series of prototypes to establish visualization possibilities. We evaluate the prototypes in terms of appropriateness, novelty and surprise and conclude that the creative impetus to our design activity had a long-term effect, contributing to designs that were found to be effective, informative and novel and a process in which creativity flourished. We offer a series of contributions that may be useful in energy visualization and beyond, namely:

- i. a *creative design case study* where a user-centered process is augmented with means of deliberately stimulating creative thinking;
- ii. *techniques* for the visualization of a new data source, including methods that contain some novelty, that may be transferable as data of this type becomes more common and voluminous;
- iii. evaluation of *creativity methods* in an applied context to support the contention that deliberately stimulating creative thinking can result in designs that are novel and useful – especially in the context of open requirements in problem-driven visualization.

2 APPLIED CONTEXT

Smart Meter technology enables energy consumption to be recorded for multiple appliances within the home at frequent intervals. Data are reported back to both energy supplier and consumer enabling near real-time feedback on energy use. The European Commission recommends all member states adopt intelligent meter technology with the majority to be fully equipped by 2020 [13]. The installation of Smart Meters forms a major component of the shift from passive electricity supply to ‘Smart Grids’, which use digital technologies to manage the regulation of energy demand and production, allow for flexible tariffs and provide the potential to communicate directly with Smart Homes or appliances [13]. Advances in Smart Meter technologies are consequently becoming increasingly important to both energy suppliers and consumers, whilst data yielded from these new technologies is increasing the volume and value of data available to the industry exponentially [44]. Energy data analysts and modelers are beginning to investigate opportunities to utilize the emerging data to understand consumption trends and consumer behavior [14] and to manage supply and demand effectively through optimization and flexible tariffs [4].

Data visualization and visual analytics offer real opportunities for the analysis of Smart Home data both for the energy supplier and the consumer. On the consumer side, energy use information is reported through a Smart Energy monitor. While this is seen as beneficial in comparison to the traditional energy bill [18] less intrusive forms of consumption awareness are now being investigated [43]. Visualization solutions to enable the energy industry to gain valuable insight into customer habits, identify areas where consumption can be reduced and effectively manage supply and demand levels are, however, scarcely investigated in the literature. The benefits of using visualization to study aggregated household energy use to discover patterns and trends have been highlighted [12], however the data are based on diary entries rather than volumes of frequent automated recordings.

Our research with data analysts from a major UK energy supplier begins to investigate the benefits that data visualization can bring to derive value from the data emerging from Smart Home technologies and opens up opportunities for further research. It uses the two sources of Smart Home data currently available: *live data* from a Smart Home trial and *modeled data* simulating future scenarios. The live data contains electricity and gas consumption for all appliances (e.g. refrigeration unit or television set) as named by owners of a test-bed of 130 properties participating in a Smart Home trial. The data set consists of more than 18 million recordings taken over a 14 month period. It has challenging characteristics: timings are irregular; frequency of recordings varies significantly – from minutes to days; the sample of households is small (the UK contained 24.6 million households in 2012),

self-selecting and biased in terms of geography and demographics. Householders are also inconsistent in the appliances they monitor. The model [17] uses a separate source of detailed consumption data [56] to generate appliance-based energy usage scenarios for any number of households at 15-minute intervals over a given period of time.

Both sources contain numeric information for individual households (modeled or trial participant), such as total electricity consumption, consumption by individual appliance or outside temperature, along with the time of the recording. Derived values (average, max, min, count, standard deviation) are calculated in both cases by period of time (hour, day, week etc.) and by grouping categories (such as appliance type). The model can generate large volumes of data in this form with optimized outputs simulating the shifting and reduction of demand over time. Different outputs reflecting weekday and weekend activity are also available. Daily and seasonal variations in consumption and standby options are modeled with some sophistication for certain appliances. Outputs are somewhat limited however, in that appliance use and distribution of appliances to households are determined probabilistically [17] and so may not reflect real ownership or typical household usage patterns. Appliance co-ownership relationships are therefore not realistic and neither household demographics nor geographical location are accounted for in the simulation.

3 CREATIVE DESIGN PROCESS

Our design process for exploring the possibilities for data visualization within Smart Home data analysis followed an established user-centered approach [25, 27]. However, we augmented this by applying a number of creativity techniques [24, 29, 31, 37] early on in the process. Our aim here was to see whether we could tap into the latent creativity of our target users – the energy analysts – as well as that of the design team. While designers, of visualizations and other artefacts, may be used to developing creative responses to problems or design briefs, their customers, users, and other stakeholders may not be. We have previously employed such deliberate creativity techniques with air traffic controllers [31] and the police [38], who have not been accustomed to making creative contributions to design. Through the use of techniques such as those described below, they have, in each case, been able to generate requirements and design ideas for new interactive systems that were considered both novel and useful. Here we apply these methods alongside our established means of encouraging data owners to engage actively in visualization design and development [25, 26, 27, 41, 49]. The process is summarized in Fig. 2 with the creativity techniques being inserted in the early stages with the intention of introducing a creative climate that we hoped would persist.

3.1 Creative Requirements Workshop

Creativity techniques for use in our *Requirements Workshop* were developed through two internal pilot sessions. Techniques from methodologies such as creative problem solving (CPS) [37] and Syntectics [16] were considered and additional literature reporting similar techniques was consulted [22, 34]. These included: aspirational thinking, analogical reasoning, metaphor, constraint removal, storyboarding and random combination. We tried methods out internally and adopted the techniques that were thought to be most practicable and potentially useful whilst rejecting some that might constrain – such as building a priority list or listing ideas based on their complexity. We augmented others, such as an established “*I wish*” exercise for wishful thinking [33] with prompts specific to the visualization context – “*I would like to see*”. The methods were refined in collaboration with a professional creative facilitator, who coordinated the *Requirements Workshop*.

As well as tailoring the creativity techniques, we also paid careful attention to our choice of venue, as the physical environment in which activities are carried out can have a significant impact on the creative climate [11, 23]. We therefore chose to carry out the workshop in a quiet, light, neutral venue, away from the participants’ normal places of work, with plenty of space and ample refreshment. The day long event was attended by five Smart Home energy analysts, who work together on a regular basis. They are often involved in thinking of new ideas and possibilities for Smart Home technologies, however, their

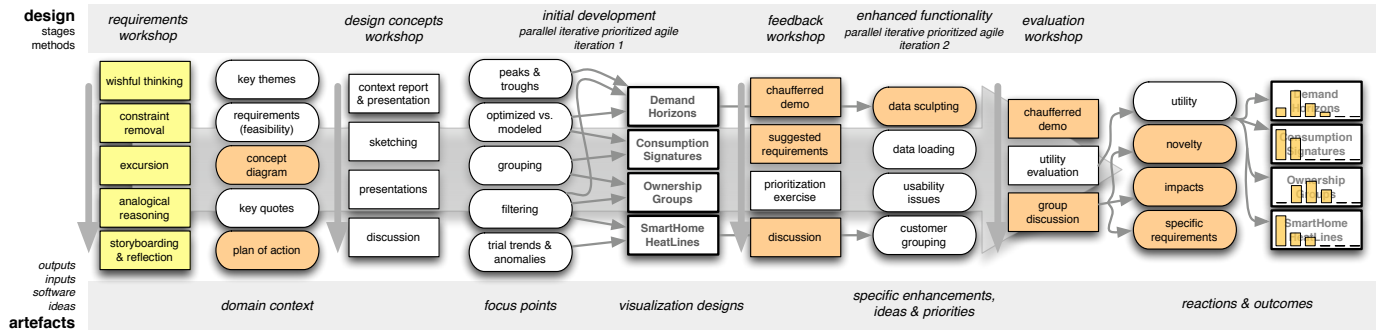


Fig. 2. The design process. Rectangles are techniques, those with thick edges represent software prototypes. Concepts are round edged. Arrows show direct links between concepts and prototypes. Other links are implicit and less direct. Yellow indicates deliberate creativity mechanisms. Orange highlights processes and concepts in which creativity amongst analysts was strong. Prototype utility is reported in detail in section 5.

knowledge of the new data sets available to them and the opportunities offered by data visualization were limited.

We began with some warm up activities. These included a playful introduction that encouraged participation and trust-building and introduced some analogical thinking by asking all participants “if you were to describe yourself as an animal, what would you be?” Some statements and quotations that emphasized creativity and exploration were also shared – for example, Albert Einstein’s widely reported view that: “if at first, the idea is not absurd, then there is no hope for it.”

3.1.1 Wishful Thinking

The first creativity technique employed in the main part of the workshop was wishful thinking, in which the energy analysts were asked to think about aspirations for the Smart Home programme. We captured visualization specific ‘opportunity statements’ [23] by asking: ‘What would you like to know?’, ‘What would you like to be able to do?’ and ‘What would you like to see?’ Participants worked individually on Post-it notes in a brainstorming [36] exercise, then read their answers out to the group and placed them on flip-charts. We then asked the participants to form small groups and each was tasked with selecting the Post-it in which they were most interested. To push them further in their thinking, the analysts were asked to consider ‘What next?’ and further aspirations were recorded (again on Post-its) assuming the chosen aspiration(s) had been achieved. The process continued until ideas were exhausted and some initial requirements had been teased out, revealing some of the types of innovation in which participants were interested.

3.1.2 Constraint Removal

After coffee, participants built upon this forward thinking with a constraint removal activity [24] in which barriers were transformed into a positive resource through which to create new ideas. Our energy analysts were first asked why the aspirations captured on Post-its had not yet been achieved. Once constraints were identified analysts were then asked for creative ideas about what would be possible if the barriers were removed to see whether ideas would develop further. A rapid flow of constraints resulted – from hardware technical issues, to people leading complicated lives and being difficult to understand, limited knowledge about Smart Homes, a lack of customer trust, limited time, resources and expertise as well as conflicting business priorities. ‘Removing’ some of these constraints unlocked a number of ideas about moving forward: in particular about improving and expanding the product, gaining the trust of customers and the energy industry and deriving value and knowledge from the live Smart Home data source.

3.1.3 Lunchtime Excursion

Lunch was held in an adjoining building during a lengthy break. Participants were asked to use this time to find something that had a connection (however abstract) with the Smart Home programme. This was based on the idea of an ‘Imagery Trek’ in CPS [36] or ‘Excursion’ in Syntectics [16]. Both are techniques that can help develop highly novel

or unexpected ideas and assist participants in refining or elaborating their ideas through ‘mental stretching’ [23]. The idea is that participants remove themselves from a task, take a mental or physical journey to seek images or stimuli and then bring these back to make connections with the task. Participants returned from their excursion with all sorts of artefacts including photos of a painting and the view from the lunch room and a copy of Dickens’ ‘Great Expectations’. This activity set the scene for the subsequent analogical reasoning task.

3.1.4 Visualization Awareness using Analogical Reasoning

The analogical reasoning task was an extension of the ‘Visualization Awareness’ activity that is central to our existing human-centered visualization design process [10, 25]. Here, however, we began by specifically explaining analogical reasoning and giving examples. We then asked the analysts to find analogies applicable to Smart Home visualization as they engaged in an otherwise relatively passive visual experience that introduced visualization examples by theme. Participants were given time to consider any aspects of the examples (data, layout, interactions, colors, aesthetic) that sparked a connection with the thinking that had occurred during the morning sessions. Reactions were again written on Post-its, and some of the participants created mind-maps to link the different visualizations to their ideas. In total ten analogical ideas arose while watching the visualization demos, including an idea to show wasted energy flows that was sparked by an animated visualization of millions of bike journeys [55] and an idea for using bubbles of energy consumption increasing and decreasing as used in the home, inspired by *Empires Decline – Revisited* [7]. Design requirements identified during the exercise included the need to filter, group and compare data such as by appliance type, temperature, user demographics, time and geography to understand consumption variability. Design elements identified as important included: ‘everything in 3 clicks’, ‘beautiful’, ‘engaging’ and ‘simplicity’.

This activity took longer than planned, largely due to the large number of wide-ranging and increasingly ambitious ideas that surfaced. The session ended with a highly creative *Plan of Action* envisaged for the focus of Smart Home data analysis involving a three stage process to which we could make an important contribution, namely:

1. discover – find out where energy is used;
2. displace consumption – change behavior and control devices;
3. reduce energy production – specifically by the amount needed to close a power station (power plant).

3.1.5 Storyboarding

We have used storyboarding [3] previously in creative requirements workshops in other domains [29, 30, 31, 32] to draw together and prioritize the ideas generated. Here, pairs of participants used a comic strip template, writing materials and hard copies of the various visualization awareness examples to generate artefacts (sketches and collages) showing how the ideas generated during the day might be used in practice by imagining ‘a day in the life of an energy analyst’.

Table 1. *Wishful Thinking* revealed in ‘Know/Do/See’ and ‘What next?’ Numbers show total aspirations established at the *Requirements Workshop* (Est.) and those deemed feasible by the design team (Feasible).

Activity	Aspiration Topic	Est.	Feasible
Know	Customers Habits	10	5
Know	Appliance Consumption	6	6
Know	The Value of the Data	2	2
Know	Visualization Design	2	2
Do	Improve Customer Experience	5	2
Do	Manage Energy Demand	3	3
Do	Advance the Technology	3	0
See	Data Analysis & Visualization	8	6
See	New Products and Services	1	1
—	—	—	—
Next?	Change Customer Behavior & Improve Life	5	0
Next?	Improve & Expand the Product	6	0
Next?	Understand Customer Habits	3	2
Next?	Gain Trust & Increase Customers	5	0
Next?	Educate Energy Industry & Manage Demand	5	1

Key themes that emerged from the storyboards included the need for greater understanding of consumers’ habits and the desire to understand customer behavior by grouping and comparing relevant data.

3.1.6 Reflection

To round off the workshop, participants were asked what they knew at the end of the workshop that they hadn’t known at the outset. Their responses at this point were very positive, both in regard to the possibility of developing appropriate visualizations “*It’s amazing how many techniques are applicable to energy*” and in regard to the workshop itself “*I understand more about the large scope of possibilities.*”

Overall the outcomes from the day’s activities allowed us to identify five key themes that can be seen as important to the continuation of the Smart Home programme: *Analyze the Data*: to understand more about customers’ energy habits and appliance consumption; *Develop Knowledge*: to start to prove / disprove myths and theories of energy saving and behaviors; *Communicate and Engage*: within the business, and with industry and the general public to manage demand and change behaviors; *Build Trust*: in the company and the products; *Improve and Expand Smart Products*: beyond energy to improving comfort and security. The first of these themes links directly with the first stage of the *Plan of Action*: *discover – find out where energy is used* (see end of 3.1.4), a key objective in which visualization can play an important role. Improving the understanding of customer and appliance consumption will also help pave the way to targeting some of these other themes and reaching the second and third stages in the *Plan of Action*.

The wishful thinking exercise generated 64 aspirations and opportunities of broad scope as shown through their grouping into topics (Table 1). We identified 30 of these as feasible for data visualization solutions in terms of the expertise, data and other resources available.

These key themes and feasible aspirations were reported to designers and developers in the team at a *Design Concepts Workshop*. We also presented other artefacts from the *Requirements Workshop* to establish the problem domain, describe the analysts’ needs and identify where and how effective data visualization design might be beneficial.

3.2 Design Concepts Workshop: Development Iteration 1

Development took place over a one month period with two iterations using a rapid agile approach. Within each iteration features were prioritized using the MoSCoW technique [2] with frequent meetings between designers and developers in the team to re-prioritize and discuss design decisions in light of requirements.

The first iteration began at a half-day *Design Concepts Workshop* that brought together seven visualization designers and developers (all are co-authors) many of whom had limited background knowledge of the energy industry. We began the session by presenting and sharing the domain knowledge as well as the key themes and ideas from the *Requirements Workshop*. Contextual information including the 3 stage

Plan of Action, key themes, feasible aspirations, design requirements, mind-maps and a concept diagram generated in part from these, storyboards and some direct quotes were introduced and then pinned to the walls of the room in order to prompt movement, discussion and idea generation amongst designers. The two energy data sets were also introduced and their structure, provenance and limitations discussed.

Working in pairs we generated ideas, developed sketches and reported back to the group with reference to the requirements that the idea was targeting. This enabled us to derive visualization focus points – abstract combinations of task, data and design that form a basis for ongoing development: show *peaks and troughs* in daily demand to understand when different appliances are used; *compare modeled to optimized solutions* to see whether shifting consumption could help demand management; *group and filter* consumption by appliance and types of appliance across time to identify patterns in user behavior; and, identify *trends and anomalies* in the Smart Home trial data.

These focus points were further developed during the workshop and through subsequent activity into four prototype visualization designs. These addressed generic aspirations from the wishful thinking exercise, such as: “*to know how to show the business stakeholders the data in an engaging way,*” “*to find typical patterns and make predictions,*” “*to know where energy is going*” and “*to ‘slice and dice’ the data,*” as well as specific aspirations and questions as follows:

Demand Horizons: highlight peaks and troughs in the modeled hourly energy demand during typical weekend and week days to show how each appliance contributes – “*to know what an ‘average home’ does with their energy*” and “*to better understand how different appliances contribute to the peaks in energy demand throughout the day.*”

Consumption Signatures: show how each appliance has a different signature over time-of-day and day-of-week in the modeled data by visualizing large amounts of energy consumption data in comparable form on one screen – “*how can we visualize large amounts of energy consumption data on one screen?*” and “*can we compare the energy consumption signature of appliances or groups of appliances?*”

Ownership Groups: group appliances in the modeled data by ownership, time of use and average consumption – “*to know how lifestyle links to energy demand*” and “*to better understand how the data relates to the users.*”

Smart Home HeatLines: use a per-household representation of the live Smart Home trial data to identify patterns and anomalies – “*how to visualize all the data from the Smart Home trial to understand the usefulness of the data?*”

Developing designs in parallel enabled us to address multiple focus points concurrently, present alternative techniques of potential value to the domain experts and use an established means of generating high quality and diverse outputs [9]. It also offered plenty of ‘breadth’ in terms of enabling us to explore opportunities for ongoing creativity.

3.3 Feedback Workshop: Development Iteration 2

Following the first development iteration a number of enhancement possibilities were suggested by the design team and associated effort estimated for each. These suggested enhancements were the focus of a *Feedback Workshop*, involving the four analysts who had taken part in the *Requirements Workshop*, and four others from related departments in the same organization. We presented the aspirations gathered from the *Requirements Workshop*, reflected on how we had formulated these into focus points and demonstrated our initial designs by chauffeuring the visualization prototypes in an engaging and increasingly interactive visualization session held at the company’s Smart Home test house. We then suggested enhancements (Table 2, ‘Design Team’).

Initial reactions, new ideas and other feedback were recorded for each design prototype. Our proposed enhancements and any suggestions identified by our users during the session (Table 2, ‘Analysts’) were then prioritized by the group. After the session, enhancements for each prototype were considered through a systematic re-prioritization process in terms of development complexity, time available, novelty of idea and priority through an agile procedure for planning estimation [5]. A number of key new features for each prototype were implemented (Table 2, ‘Implemented’) as described below.

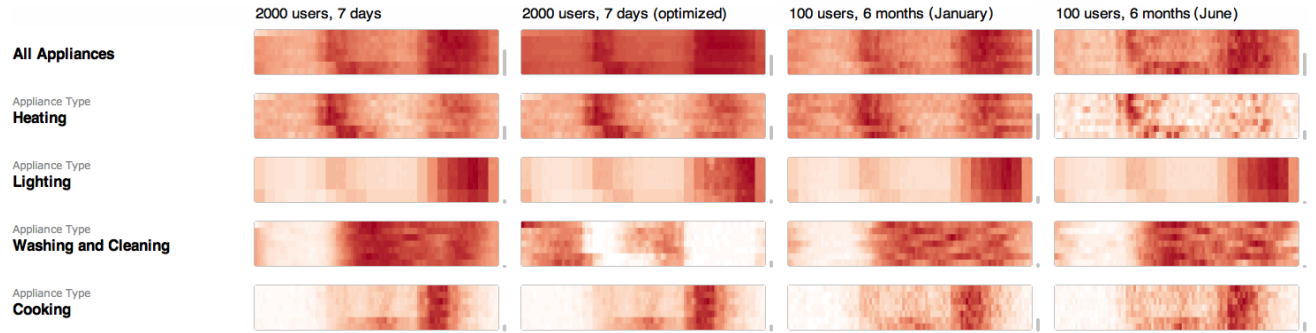


Fig. 3. *Consumption Signatures* allows modeled data to be loaded (columns) and reordered so that the weekly consumption patterns of appliances can be compared. Various coloring options scale sequential schemes by selected row, column or cell and allow diverging schemes to emphasize difference from selected items. Patterns in daily (*Lighting*), seasonal (*Heating*) and modeled (*Washing and Cleaning*) data are clear as are weekend differences (bottom two rows of each cell) such as the delay in the morning heating peak and more cooking during daytime at weekends.

Table 2. Prototype Enhancements – suggested by the design team and by analysts in the *Feedback Workshop*, and implemented in iteration 2.

Prototype Name	Design Team	Analysts	Implemented
<i>Demand Horizons</i>	11	7	6
<i>Consumption Signatures</i>	7	5	10
<i>Ownership Groups</i>	10	3	8
<i>Smart Home HeatLines</i>	10	3	6

4 RESULTS: VISUALIZATION PROTOTYPES

The four prototypes were developed with complimentary characteristics to explore different tasks, data and designs – as characterized by the focus points (section 3.2). The features are described below with detail of specific interactions explained in the supplementary video.

4.1 Modeled Data

Two of the prototypes used hourly consumption data modeled for 2000 households over a period of 30 days, with different average hourly rates calculated for households at weekdays and weekends

Demand Horizons: (Fig. 1) uses horizon charts [20] to show aggregated and appliance-based energy demand during a typical 24 hour period. Horizon charts can be switched to area graphs to aid understanding. Animated transitions [21] highlight the differences in consumption between typical days during the week and weekend. Appliances can be re-ordered by contribution to the total, morning or evening peaks and individual appliance charts can be added or removed for detailed investigation of the differences in demand between appliances and their effect on overall consumption. Several amendments were implemented in the second development iteration, including quick switching between gas and electricity appliances. In particular, a new feature was created in order to allow demand to be modified directly through the metaphor of *data sculpting*. This allows peaks to be flattened through the interface in two ways: the overall consumption of any appliance can be interactively varied to simulate improved efficiency; consumption can be time-shifted, using the *grow*, *shrink*, *fix* or *free* buttons, to simulate change in behavior (see Fig. 1 and video).

Ownership Groups: (as shown in the supplementary video) consists of a bar chart linked to a set of Tufte’s [50] redesigned Tukey box plots [51]. Bars representing each appliance are sized by the number of households that own at least one of each. Bars can be reordered to show the appliances by proportion or alphabetically. The box plots show average hourly consumption of households. Upon selection of a particular appliance these are updated to show the average consumption of the households owning this appliance. Design enhancements implemented after the *Feedback Workshop* included new selection mechanisms and three additional means of ordering – by appliance type, subtype and total power/load on the grid. Alternative views related to co-ownership of appliances were also investigated.

Consumption Signatures: (Fig. 3) visualizes the model’s highest resolution data, with records at 15 minute intervals aggregated according to time of day and day of week. Multiple outputs can be structured in to this weekly *signature* for comparison, including a six month simulation to show seasonal variation and a one week simulation with two algorithmically optimized alternatives. Multiple derived values (such as minimum, maximum and average consumption) were abstracted from the model outputs and households were sampled in the case of large data sets to ensure rapid responses. Calendar views [52, 54] visualize weekly consumption: seven rows relate to days of the week, with 96 columns – one for each 15 minute period of the day. These signatures are positioned in a matrix of small multiples in which data sets (columns) and appliances or groups of appliances (rows) are juxtaposed for comparison [15]. They are colored according to their values with two alternative schemes: a sequential scheme represents absolute values and a diverging scheme [19] shows the numerical difference between each signature and a selected item: a column (data set); row (appliance); cell (particular signature) or pixel (individual value). During the second development iteration the need to rescale the legend to the ‘best fit’ for each signature was identified and implemented.

4.2 Smart Home Trial Data

Smart Home HeatLines: (Fig. 4) represents the raw live data from the Smart Home trial. Individual households are represented as rows of values varying over time. Summaries (count, average, maximum and minimum) are calculated by household for each variable for particular time periods. Further data abstraction is available in real time as the temporal kernel can be interactively re-sized to aid pattern identification and avoid distortion due to inconsistencies in collection times. Sequential color schemes [19] are used to represent values, with a line graph to aid in the identification and interpretation of patterns and trends for any selected household. The summary statistic, source (electricity, gas or appliance) and time period (total and weekly or daily averages) can be varied interactively. Households (rows) can be re-ordered by value at a particular time period. Grouping by demographic type, sorting by similarity of profile and a map to show animated geographical variations over time were added during the second development iteration – as shown in the supplementary video.

5 RESULTS: VALIDITY AND CREATIVITY

Reflecting on both the visualization design evaluation literature [47] and methods for evaluating creativity [8, 28] we constructed a structured process to determine the extent to which both the visualization prototypes themselves and the design process through which they were generated were seen as both valid and creative (Table 3).

The extent to which the outputs of our process were themselves viewed as creative was a particularly important indicator of how successful we had been in our introduction of techniques for deliberately

ergy industry and in particular to the needs of a modeler: “*The way you solve a problem is by doing some visualization in your mind and these tools help you greatly to facilitate that.*”

5.2 Prototype Novelty and Surprise

The four design prototypes in general were described by one of the analysts as “*creative approaches which show us the density, variability and value of our data.*” The techniques used were “*very different*” and new to the analysts: “*the methodologies would not have come out of my head.*” Overall the designs were deemed novel and valuable: “*you have brought something that we couldn’t have thought of ... and the [Smart Home] project will be better for it.*”

Novelty and surprise were expressed in reactions to *Smart Home HeatLines* during the *Feedback Workshop*: “*I think this is brilliant*”; as well as after reflection in the *Evaluation Workshop*: “*it gives us a whole new way of analysing people,*” “*18 million data points! [It] is just impossible for us to get our head around the real value that is contained in that*” and “*I did not realize how diverse the different profiles were.*” The prototypes visualizing the less familiar modeled data also resulted in expressions of surprise and evidence of novelty. The heat mapping in *Consumption Signatures* can not be termed novel as a technique, but the sheer volume of data and the possibility to compare so much through juxtaposition and color variation was deemed by analysts to be “*really clever.*” The appliance based sorting in *Ownership Groups* was seen as both novel and useful: “*The 5 way sorting ... by category, load, subclass is not something we’ve seen before.*” Initial reactions to the animated transitions in *Demand Horizons* when shifting from weekday to weekend highlighted the novelty of this feature and the sorting of the appliances by their contribution to the peaks was seen as: “*really interesting – you just could not get that out of numbers.*” The *data sculpting* feature also received positive feedback from analysts suggesting novelty and surprise (see section 6).

Interviews conducted with the data modelers revealed that they also regarded the designs to be novel: “*they give me the opportunity to analyze the data in a different way.*” The designs also enabled the modelers to see surprising structure in their outputs: “*I didn’t expect to see these patterns*” and “*I wouldn’t be able to spot the problem before I saw this graph.*” The modelers’ view on the trial data changed completely upon seeing *Smart Home HeatLines*: “*before I thought the trial data could not be used due to errors and outliers. The visualization showed me that you can use this data and detect different patterns and user behavior.*” There were also clear opportunities identified for data visualization within the energy data modeling domain: “*it has got great potential ... to spot problems, abnormalities, see the patterns, come up with new ideas, new theories, new models.*”

5.3 Process Validity and Effect

The analysts felt engaged in the process, that they had contributed and that they had learned through doing so. They were pleased with the responses to their suggestions: “*you actually listened to our feedback, helped us shape that feedback and then delivered.*” The process of developing the prototypes was deemed to be educational and stimulating helping the analysts understand the possibilities that data visualization can offer and the value of considered visual design: “*I realize that actually this has got many potential applications and many many uses,*” “*the data is a crucial thing and the visualization of that data is almost as important to move ... from information to insight.*”

6 CASE STUDY: DATA SCULPTING

One example of novelty, as perceived by the energy analysts, relates to the ability in the *Demand Horizons* prototype to engage in *data sculpting*. Documenting the lineage of the idea through our development process draws attention to the creative processes and enables us to reflect on the impact of the creativity methods we used.

6.1 Requirements Workshop

It was evident that the potential impact of successfully implementing the *Plan of Action* (see section 3.1.4), that arose out of visualization awareness with analogical reasoning, would be significant in economic

and environmental terms: power stations are costly on both counts. The importance of the power station as a unit of production was also very clear: they are used to accommodate peaks in energy consumption, difficult to switch on and off and expensive to maintain – hence the significance of reducing peaks below the threshold at which a particular plant is needed. We thus took the *Plan of Action* to the *Design Concepts Workshop* as one of our key inputs as we had been informed that: “*the better stage 1 is, the better stage 2 and 3 will be.*” A designer explains how this inspired the development of *Demand Horizons*.

The Designer’s Story – Initial Development

I chose to design to “How can we use visualization to better understand how different appliances contribute to the peaks in energy demand throughout the day?” The objective was to design paper prototypes to meet this requirement without consideration of data or development constraints. Having some experience of developing data visualization techniques and systems, I was keen to make a contribution that fitted technique to requirement in a creative way. Knowing that many appliances might have to be shown concurrently, I was looking for a visual technique that was graphically compact, but visually distinctive. Horizon charts [20] seemed particularly appropriate as energy production jumps between discrete quanta when power stations are fired up or shut down in line with demand. This had a natural fit with the discrete ‘horizons’ of the chart. Thus the initial prototype comprised a set of horizon charts – one per appliance – and a single summed horizon chart representing total consumption. Each discrete band might represent the consumption necessary to cause a power station to be brought online (see Fig. 6).

The modeled data populating the horizon application were somewhat approximate and subject to change as the consumption model changed. This uncertainty informed the smoothed line design of the horizons as well as the smooth transitions implemented when moving between weekend and weekday consumption models (see Fig. 1).

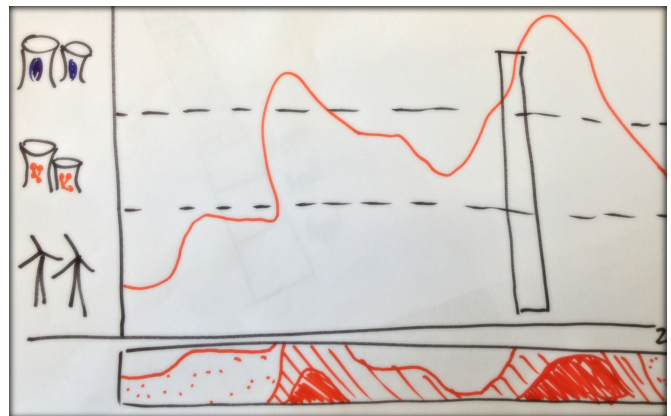


Fig. 6. Design Sketch: The *Demand Horizons* view of the *Plan of Action*.

6.2 A Creative Feedback Workshop

Initial reactions to the prototype at the *Feedback Workshop* were very positive: analysts liked the ability to play with the representation and see things change – a novelty to them in terms of their use of their data: “*there are so many touch points and ways I can move around that data – it gives you a Wow! factor*” and “*I think this is very powerful.*” Switching quickly between standard and horizon graphs helped explain the horizons and we were soon in a position where discussion about the data flowed with theories and requirements explored enthusiastically: “*if this data was live I’d like to be able to look at specific days - i.e. load shifting for tumble dryer could relate to specific days such as [when we have] rain.*” The ability to switch between weekday and weekend consumption was positively received and emphasized the

fluidity of the interface and ability to change the data seamlessly and quickly to suit particular lines of enquiry. This emphasis on fluidity and flexibility seemed to inspire some creative thinking about using the data that gave rise to interesting ideas and subsequent requirements in terms of managing energy consumption – “cooker goes off, dishwasher comes on! Can we shift the dishwasher?” – and important discussion around the timings of usage of washing machines (mainly in the morning) and driers (main usage in the evening): “[could consumers] use the washer, they leave it and then they dry it when they come home?” The significance here is that if consumers are prepared to wait to use energy consuming devices there is scope for offsetting usage to reduce the evening peak – perhaps below a power station horizon.

This exploration of patterns in the modeled data gave rise to further creative thinking about using the interface to model changes in consumption – through changes in behavior and more energy efficient devices: “you can’t shift lighting time ... but we can remove a percentage by changing the bulbs” and “[what if we] switched everyone to a more efficient fridge freezer for example?” The aim of moving the dark peak below the upper horizon was implicit in the vigorous discussion. The design appeared to have been revealing and instructive in focusing activity on the need to reduce consumption below the levels emphasized by our horizons – much in the way anticipated by our designer, and clearly in line with the *Plan of Action*.

In turn these ideas rapidly gave rise to discussion about the interface and how we might interact with data to explore these theories: “could we drag and drop and move something from that time to another time - to imagine [model] time shifting?” We began to explore these ideas collectively: of reducing the consumption profiles of particular devices by a proportion and of moving consumption of particular devices from one time to another to remove the top horizon. Animated discussion ensued in front of the projected images with ideas being developed rapidly about how to select and represent times, percentages and shifts. This was intensive, creative design work inspired directly by data and analytic need, the latter being identified directly prior to the design ideas discussion through our prototype interface. The analysts were excited by their increased understanding and interpretation of the data, design possibilities and new ways of interacting with the models to address their objectives. This was evident in ensuing discussions about deployment and the immediate request for screen dumps to be used in an imminent internal meeting. Our focus here was very definitely on step 2 of the *Plan of Action* – *displacement* – as the data prototype had addressed much of step 1 – *discovery*. The ideas captured during this highly creative discussion at the *Feedback Workshop* were particularly useful as they were stimulated by both interface and the data analysis it enabled, in the context of an identified objective. They were communicated to our developers for the second development iteration.

The Designer’s Story – Enhanced Functionality

The requirement to allow ‘what if?’ remodeling of consumption patterns was clearly expressed, leading to the need to be able to edit the data shown in the horizon charts. Rather than separate the editing from the data exploration tasks, I combined the two processes under the metaphor of ‘data sculpting’, enabling analysts to interactively select time periods and then vary consumption levels for particular appliances with immediate graphical feedback. This idea arose in part from previous work I had seen and developed for ‘sculpting’ terrain models where interactive graphical tools are used to raise and lower parts of a gridded elevation model [39, 40]. It also follows the design pattern of ‘data as interface’ that I had found successful previously [10]. The metaphor was reinforced and partly inspired by the use of a clay colored color scheme and the smooth curves used in the charts that make the graphs look as though they are mouldable.

6.3 Evaluation Workshop

The *data sculpting* feature sparked a vibrant discussion at the *Evaluation Workshop* with plenty of ideas of possible uses. It seems to be a technique with scope for helping explain the concept of demand shifting and reduction and to explore its possibilities: “I am more confident

that internally I could use something like this to demonstrate that it [flexible demand] will work.” Known aspirations for switching cold appliances off and on were discussed, with the interface encouraging new thinking: “the fantastic thing about grow is you can grow before hand as well so you can super cool fridges or freezers.” The feature was deemed “a very useful dynamic tool” that could pave the way for a new data storage strategy to ensure that data is of sufficient resolution to allow for this kind of visualization.

The modelers also liked the idea of *data sculpting* and had not considered using visualization in this way: “this is really good. It represents what we have tried to do with the optimization tool but when I produce a model or amend it we need to re-run it. This does it instantly!” The modelers were positive when asked whether *data sculpting* would be useful to help with building and editing the optimization algorithm itself: “yes, if I had something similar to that I would definitely use that.” New ideas were also created such as relating the horizons to energy cost thresholds: “if the cost exceeds the thresholds you would have a penalty. You could visualize it and see it.”

7 REFLECTION

The evaluation and case study reported above demonstrate some success in terms of our applied designs. Approaches such as *data sculpting* in *Demand Horizons* and the comparison through color variation and alignment used in *Consumption Signatures* and the multi-scale interactive analysis through *Smart Home HeatLines* demonstrate some novelty, seem useful in this context and may be applicable in other domains and scenarios. In this section, we share the reflections of experienced designers on the extent to which the process we have undertaken can be seen as *creative*, and consider the impacts this may have had on designers and other stakeholders.

In an applied client-based project such as this, evaluating the impact of the methods used by means of a controlled study is not feasible. Our approach to gaining some initial insights on the impact of our creative methods on the process of visualization design has therefore been to reflect, as designers, on our experience in this project, in order to compare it with the numerous other projects in which we have been involved over the years. Without a control we are unable to prove that adding the creativity methods at the outset of the project had any specific impacts on the process as a whole: good visualization design projects almost always involve creativity and novelty and we actively emphasized and valued these characteristics here. However, we did feel that the creativity methods opened up particular opportunities for creative thinking. They established the true breadth of a situation in which requirements are open with familiar reference points. They took participants out of their comfort zones and enhanced the ‘away day effect’ of shared purpose. The explicitly creative activities helped visualization designers and domain experts communicate, share experiences, establish trust and work as a team. We experienced creative thinking about using data as well as about design and the creative thinking may indeed have helped us “push domain experts to discuss problems, not solutions” [47]. Based on our experience of past projects, we identify the elements where we feel the use of deliberate creativity methods had the greatest impact in Fig. 2 and discuss these further below.

Some of the simplest creativity methods seemed surprisingly effective. The animal introductions required some audacity on the part of our facilitator, but this was handled with aplomb. Developing analogies and revealing some personal information in a controlled and safe manner required openness on behalf of all participants. It seemed useful preparation for future exercises in initially putting all participants on an equal footing, establishing trust and involving surprise – suggesting that anything was possible from the outset. The excursion worked well as a preparation exercise to get participants in the frame of mind for the next activity and remind them that lunch was an opportunity to think and communicate. Everyone understood, brought something interesting back and had time to make a contribution.

Our impression following the visualization awareness activity was that use of analogy was very evident. Participants applied many of the ideas shown in visualizations from other domains creatively and effectively to their own area of interest. This activity spurred on a long and

interesting conversation about what was possible with the data to hand and might be achievable given the visualization examples presented. It seemed that these ideas generated after the visualization demos were stimulated by the morning's activities. We regarded them to be more numerous and creative than is the norm in these sessions and the outputs – such as mind maps developed during the awareness activity – were sophisticated. The storyboards produced in the activity that followed were not as useful as we had hoped. This may have been due to a lack of energy or the fact that previous discussions meant that we were overrunning – partly because graphical summaries were already being produced as participants took the initiative to generate mind maps in response to the analogical reasoning activity. Sketches or stories that are more data focussed may be more useful in our domain and we are likely to encourage the mind-mapping as a visualization storyboard during analogically focussed awareness activity in the future.

The novel ideas established at the subsequent *Feedback Workshop* are not easy to attribute directly to the initial use of creativity methods, but were rare in our experience of user-centered visualization design in terms of their quality, relevance and originality. The expressions of novelty and surprise (see section 5.2) were particularly embedded in organizational context, including evidence of insights, and realizations of new capacity and scope for the group. Possible changes in the way that the organization stores and uses data were suggested. Our sense was of a strong link and our activity felt focussed with participants particularly engaged and able to make excellent and sometimes unexpected suggestions for design possibilities throughout the process. We claim above that creativity may have persisted throughout the one-day *Requirements Workshop*. We also suggest that the early use of creativity methods may have had longer lasting effect through our study. Equally, being explicit about our desire and efforts to be creative may have been beneficial – a positive example (in design terms if not in experimental terms) of the experimenter effect in an *in vivo* situation where controls are not feasible. The *Designer's Story* (see section 6.1) offers some evidence to support this suggestion.

In terms of process, the analysts felt that they had made beneficial contributions and been able to communicate effectively with the design team. They reported benefits in terms of both understanding the data and visualization possibilities (sections 3.1.6 and 5.3). We felt that levels of engagement and learning were high and would associate this with the persistent sense of creativity that we are reporting. We acknowledge that this sense of contribution and ownership may have an effect on the evaluation – a positive bias being highly likely. However, it may also have an effect on uptake, which could be evaluated through a longitudinal study post implementation [48].

Our designs were not wholly successful in terms of analyst reactions however. *Ownership Groups* quickly revealed that the modeled data did not capture the kinds of relationships between users and appliances that we had hoped to explore. The lack of a realistic pattern emerging meant that analysts were less engaged with this application than the others, reinforcing established findings [27]. Reflecting back, it seems that we may have been collectively over-optimistic in anticipating that we could either find or imagine patterns where our data did not support them (see the data description in section 2). Perhaps the creative nature of our *Design Workshop* resulted in some inefficiency and inappropriate design. Perhaps explicitly creative visualization design processes may produce more 'misses' than standard approaches and thus be particularly costly. Perhaps – but benefits may also be associated with this cost. We captured plenty of suggestions that our prototypes were relevant beyond the original use cases and target group (see section 5), with various ideas for *Smart Home Heat-Lines* and *Demand Horizons* being used in other organizational and customer facing contexts. Additionally our designs were deemed useful by the modelers, who used them to develop insights and expressed interest in building aspects of the prototypes into their workflows (see section 6.3). We are unable to establish whether this is due to the open requirements, unknown data and design to focus points rather than formal task analysis (all used in previous design studies), or the parallel design or the creative approaches used in this case. Further work is needed to explore these various possibilities and any effects.

8 CONCLUSION

Our experience of using deliberate creativity techniques in the visualization design process has been very positive. We present reactions from the domain experts – energy analysts and data modelers – and reflect on our own experiences to support this view. We describe a series of candidate designs for energy visualization that have been developed through intensive user-centered collaboration. They have been enthusiastically received in most cases in light of initial requirements and expectations and have resulted in insights about data, new knowledge about analytical and visualization possibilities and potential behavior change in individuals and within organizations. They may be more widely useful as energy visualization becomes more widespread. Our evaluation supports the conclusion that they constitute a successful exploration of possibilities for analytical Smart Home data visualization.

Energy analysts and modelers found the designs novel and useful. Designers also developed methods they deemed novel in collaboration with and response to analysts. We claim, through reflection informed by our experience of what has been a lengthy and intense process, that *the explicit use of creativity methods is likely to have contributed to the development of novel and effective solutions that are well aligned with established need*. This is particularly significant in a situation where requirements are open and data largely unknown. We cannot trace back through the hundreds of prioritized requirements and captured reactions, the hours of discussion and the piles of sketches to establish a direct causal link between the creativity sessions and our designs – we don't think this is how it works. Visualization design is much more holistic, taking ideas from all sorts of influences often in parallel – just as good visual thinking uses multiple stimuli concurrently to generate ideas and make decisions. Indeed, we suspect that *the very fact that we were explicit from the outset about creativity being a focus in the project may well have made us more creative in our approaches*. The *Designer's Story* (section 6.1) suggests that this may well be the case.

We conclude that *the deliberate use of techniques to enhance creativity early in the visualization design process can contribute to success in terms of process and outcomes*. In our experience this proved highly likely to be the case in: establishing a creative working environment; developing requirements; pushing designers and developers to novel solutions; and building a sense of trust, common purpose and ultimately achievement in a diverse team. Furthermore we suggest that *using creativity techniques early in the visualization design process may have longer term positive effects on creativity and satisfaction that persist throughout a design process and perhaps beyond*.

In applied design projects domain experts' time is limited and valuable. We find real benefit in encouraging them to be as creative as possible early in the process as our experience suggests that creative methods challenge mental and social barriers, can enthuse and energize participants and engage them in design. Carefully facilitated, visualization focussed, use of *wishful thinking*, *constraint removal*, *excursion*, *analogical reasoning* and *reflection* may be straightforward 'discount' methods that contribute to buy-in, satisfaction and the efficient use of participants' time. We see room for using these creativity techniques and others, such as creativity through random combination [37], at various stages through the design process to explore their effects. Indeed, we plan to use creativity techniques in future projects as they seem to provide a low cost means of establishing a beneficial creative climate. We call on others to do the same. Perhaps documenting and reflecting upon the creative aspects of the design and indeed analytical processes in a series of projects will be the best way to share and assess experiences. We may then begin to understand more about the specific effects of creativity on user-centered visualization design.

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