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Explaining smart heating systems to discourage fiddling with optimized behavior

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ABSTRACT

Our work focuses on textual and graphical explanations for smart heating systems. We have started to investigate the opportunities for when to provide explanations, how we can design these explanations, and we have started to evaluate these explanations. We argue that explanations need to be carefully crafted to fit with their desired aim, in our case to encourage users' trust and reliance while *minimizing* user interactions.

Author Keywords

Explanations; smart heating systems; trust; constrained engagement.

ACM Classification Keywords

H.5.m. Information interfaces and presentation (e.g., HCI): Miscellaneous.

INTRODUCTION

Intelligent user interfaces for heating systems are becoming increasingly popular in recent years. Based on user preferences and other characteristics, modern smart heating systems like Nest, Ecobee and Lyric are able to make intelligent decisions to optimize comfort and savings on the user's behalf [9], for example to pre-heat the home or switch the heating off when it is sensed that nobody is present. While traditional heating systems are essentially directly manipulated by users, smart heating systems make some heating decisions by themselves; hence, from the user's perspective, the system has temporarily some control over what decisions are executed and how.

Not only are user interfaces for controlling heating systems getting more intelligent, the network and heat sources involved in customers' heating are also getting smarter. Two such new ways are *demand response* and *hybrid heat pumps*. Demand response means fluctuating consumption in the energy network is taken into account to reduce or shift a customer's energy consumption during peak periods. A hybrid heat pump is a heat source which can move energy from outside (either air or the ground) and "pumps" it to a unit where it can be used to provide central heating. In addition to this mode of operation, it can also switch to gas or electricity as a primary heat source. The system can, for

example, determine to run on gas if the home needs a short temperature "boost" which cannot be achieved with a heat pump that takes longer to heat up. Therefore, hybrid heat pumps are ideally suited for demand response due to their ability to shift to different energy sources without a reduction in comfort.

These modern smart heating systems are highly optimized to balance comfort and cost. Previous work involving smart heating systems has called for *constrained engagement* between the system and user [13], where the user can engage with the system to feed back important information to the system or override it if necessary but communication is constrained so it does not overwhelm the user or push itself to the front. Smart heating systems thus follow the concept of calm technology [12], in which the technology fades into the periphery. Most research has been focused on eco-feedback to manage and, ideally, reduce energy consumption [e.g. 2, 4]. Our endeavor is different; we seek to explain system behavior so that users can engage with these highly optimized systems to manage their comfort as a co-operative team. Contrary to most explainable systems, our aim is to stop users from fiddling with the system, i.e. provide explanations for these systems to encourage users' trust and reliance while *minimizing* user interactions.

Our work is set in a larger project to understand the overall value and user experience of hybrid heat pump deployment in demand response settings. This project, FREEDOM (Flexible RESidential Energy Demand Optimisation and Management), led by Passiv Systems Ltd. and funded by Western Power Distribution and Wales and West Utilities, is the first of its kind to investigate consumer engagement with these new technologies as part of a smart heating system. We describe a study to understand when and what needs to be explained, how we created designs for a smart system control app, and a study to investigate the impact of aspects of these explanations on user understanding and trust. We conclude with a brief discussion and our plans for future work.

INVESTIGATING OPPORTUNITIES FOR CONSTRAINED ENGAGEMENT AND CALM TECHNOLOGY

Drawing on previous work in intelligibility, transparency and explanations [3, 5, 6, 7, 8], we investigated how to support users to feel engaged with their smart heating system so that they feel in control and understand what the

system is doing, and the potential effects of constrained engagement on the user experience. A fuller account of the methodology and results can be found in [10]. Similar in approach to User Enactments [2], we involved 14 participants who went through 5 hypothetical scenarios to explore their reactions and ensuing interactions with a paper prototype interface which the researcher and the participant jointly constructed on the fly (Figure 1 and 2). Throughout the study, participants were encouraged to draw on the sketches and verbalize their thoughts and decisions, thus providing a rich medium to capture participant's perceptions and expectations [11].

We found that there are certain points in the system's behavior when opportunities for constrained engagement present themselves. Participants interacted with the smart heating system when their comfort was compromised or when the system actions were unexpected. Participants expected to be notified when decisions were taken, however, they can be dissuaded from intervening to override system behavior through appropriate explanations.

Participants wanted detailed information about what the system is doing as well as the motivation and benefits for doing them. Textual and very simple graphical visualizations seem to be able to communicate past and future actions but if the content and form are not appropriate it might lead to users to abandon the system and might cause distrust in energy providers (instead of the user

interface providers). More complex graphical explanations were often found to be overwhelming and confusing.

DESIGNING THE EXPLANATIONS FOR A SMART HEATING USER INTERFACE

We then engaged in designing and implementing the user interfaces for a smart heating system that explains its behavior to users while trying to minimize follow-on user interactions. Our premise was that in some instances, explanations serve as *persuasive arguments* and can deter users from correcting system actions [1]. We started by identifying all the opportunities for constrained engagement in the system by listing decision points that might need to be communicated. Overall, we identified 7 different decision points that users might possibly need to have explained to them, for example, starting to preheat the home to reach a target temperature set in a daily heating schedule, maintaining a temperature set point in a daily heating schedule, etc. We then interviewed system engineers who worked on the algorithms to find out how the system makes this decision, focusing in particular what input data is used to drive the system behavior in terms of the output temperature and the heat source decisions. For example, in preheating the home to reach a target temperature setpoint at the start of the next schedule period, the system was using the current internal home temperature, the current external temperature, learned properties around the rate of heating in the home, the heating schedule set up by the user in terms of desired temperature during certain times of the day, cost versus savings modes, the weather forecast over the next 24-hour period, and energy price tariff info.



Figure 1: P3's sketch for explaining system behavior when user increased the temperature.

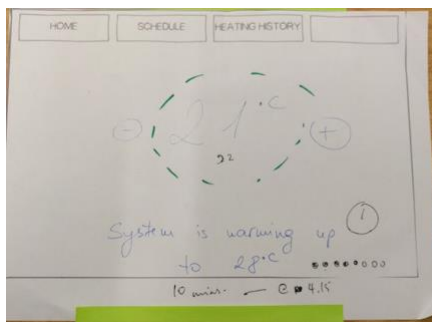


Figure 2: P7's sketch for progress indication (black dots) and an explanation how long it will take to reach a certain temperature.

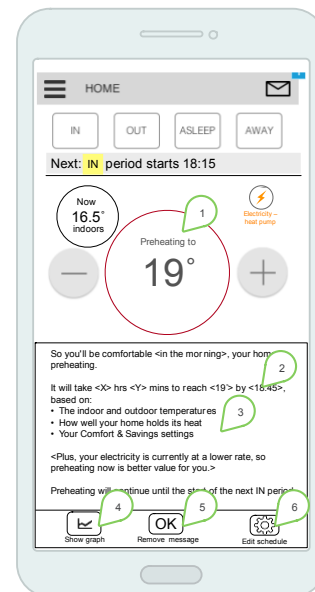


Figure 3: Textual explanations show reasons for system behavior, in this case preheating. 1) short explanation 2) system data 3) simplified overview 4) switch to graphical explanation 5) hide explanation 6) correct system.

Once we had all of this information we began to craft

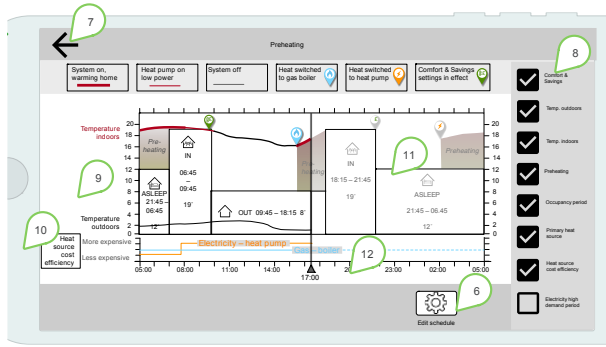


Figure 4: Graphical explanations show past and estimated future data, in this case preheating. 6) correct system 7) back to main screen 8) configure data to show in graph 9) scaled temperature 10) relative value of energy pricing 11) future is less saturated to indicate “uncertainty” 12) time with current time always in the middle.

explanations for all these decision points. Feedback during the previous study had indicated that graphical explanations were contentious and simple textual feedback also fit with the notion of calm technology. Hence, our guiding principle was that the default explanation would be a simplified textual explanation, and every time a new decision was made this was communicated (Figure 3). If the user disagrees with the system, appropriate actions, e.g. to edit the heating schedule, were supported by the explanations.

More complex graphical explanations could be shown on request which visualized the main data underlying the system behavior and its decisions. To provide flexibility, a particular combination of data can be shown in response to a system decision, for example, in pre-heating (Figure 4) a wide variety of data figures into the indoor temperature. However, the user can select (or de-select) which data is presented, i.e. the user can simplify the graphical explanation (or indeed show more data, if desired).

We carried out some limited usability testing on these explanations which showed that there were no major problems in using them. These designs were presented to the project partners to use as a basis for the design of the control app for the hybrid heat pump trial. In addition, we also designed a heating cost calculator that showed the impact of raising or lowering temperature set points in the schedule on heating bills.

SMART HEATING USER INTERFACES AND TRUST

More recently, we have undertaken a study to explore the effect of these different types of explanations on users. Recall that our premise was that increased understanding and trust might lead to *not* correcting the system. We carried out a between-group experiment to assess understanding and trust with 60 participants with one of four explanation conditions (15 participants each): Control (no explanation), Text-only (like Figure 3), Graphical-only (like Figure 4), and Both (explanation consisting of both graphical and written elements). Each participant was

presented with 5 simulated home heating scenarios: Normal maintaining, Normal preheating, Normal away mode, Preheating due to Demand Response, and Delayed heating due to Demand Response.

We captured participants’ perceived and demonstrated understanding (the latter coded from free text responses), trust ratings after each scenario, and qualitative feedback.

The results showed that explanations, in particular text-based explanations, improved demonstrated understanding, yet there was no significant effect on perceived understanding (Figure 5) based on explanation type. However, perceived understanding had a strong correlation with trust (Figure 6). Giving different types of explanations did not have an effect on trust in the system, however, participants *without* explanations were more sensitive to predictability of system behavior while participants *with* any of the explanations focused more on their perception of what the system was doing and whether they agreed with it. In order to improve trust, participants frequently asked for even more information and data to be provided especially about costs.

DISCUSSION AND FUTURE WORK

There are many options for providing explanations and controllability in a smart heating system. Here we discuss some of the findings and implications from our work so far.

Our work has explored *when* to provide explanations for a smart heating system. We believe that this work could be extended and generalized but that points for “constrained engagement” need to be investigated for each system.

We have also started to investigate *what* should be communicated in these explanations. We have focused on the “input data” and “output data” instead of the details of the algorithmic model as these aspects are more readily explained both in textual as in graphical form. However, we also noted that users expect to see reasons for making decisions and potential benefits to them. This area warrants further investigation, especially as this seems to be very dependent on the systems and its context of use. There have also been suggestions that “what-if” explanations could be very useful but so far our work has not explored this with users, and neither have we had any calls for this

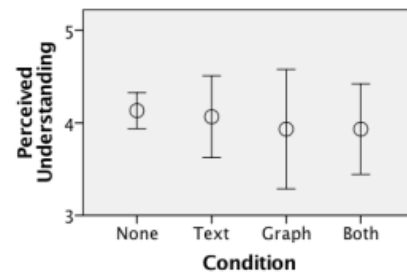


Figure 5: Mean ratings of perceived understanding with 95% confidence intervals.

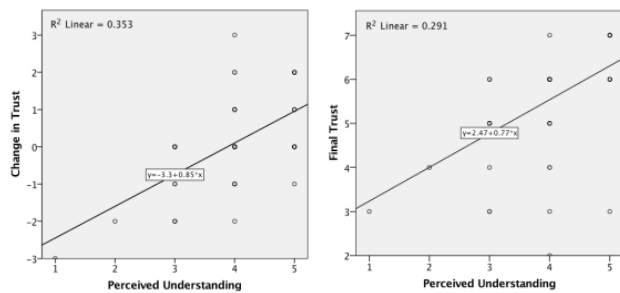


Figure 6: Correlation between ratings of perceived understanding, and change in trust (left) and final trust ratings (right).

functionality from participants in our studies.

Our designs have explored *how* to provide explanations. All evidence seems to suggest that graphical explanations need to be employed cautiously: some people love them, some people hate them. However, textual explanations, even if high-level, seem to be valuable.

We have not yet investigated the effects on perceived controllability of our UI, mainly because this is very difficult to do in a lab study since changes in temperature in response to changes in settings are very hard to simulate. However, we believe that there is a link between intelligibility, controllability and user experience but that this relationship is complex and needs to be further explored.

In future work, we will be participating in a field study over the heating season in 2017/18 during which we will have the opportunity to gather feedback from users in a real world setting after the complete smart heating system has been installed. This will provide some answers about how the explanations are perceived, and whether this might have any effect on interacting with the smart heating app.

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