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Big Data and Occupants' Behavior in Built Environments: Introducing a Game-Based Data Collection Method

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Abstract

Energy-related human behavior in buildings is difficult to define and quantify, yet critical to our understanding of total building energy consumption. It substantially influences energy consumption and saving. In the United States, residential and commercial buildings account for more than 70% of the total electrical energy consumed in the country. This paper explores the implementation of an online energy game, e-footprints, which aims to collect data about occupants' energy consumption and saving, and summarizes the feedback related to the first prototype of the user interface. The game was tested with 110 international students. The study focuses on the visualization, aesthetics, and usability of the e-footprints energy game, as these are the important elements of a player's interaction with the user interface and greatly influence the game play and the data that can be gathered. The paper concludes with a summary of the findings and general recommendations that may be useful for others using serious online urban planning games.

Comments

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8 **Big Data and Occupants Behavior in a Built Environment:** 9 **Introducing a Game-Based Data Collection Method**

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11

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13 quantify, yet critical to our understanding of total building energy consumption.
14 It substantially influences energy consumption and saving. In the United States,
15 residential and commercial buildings consume more than 70% of the overall
16 consumed electrical energy in the country. This paper explores the
17 implementation of an online energy game *e-footprints* that aims to collect data
18 about occupants' energy consumption and saving, and summarizes the feedback
19 related to the first prototype of the user interface. The game was tested with 110
20 international students. The study focuses on the visualization, aesthetics and
21 usability of the *e-footprints* energy game as they are the important elements of
22 the player's interaction with the user interface. They greatly influence the game
23 play and the data that can be gathered with the help of the online game. We
24 conclude the paper with a summary of the findings and general recommendations
25 that may be useful for other engaging online serious urban planning games.

26 **Keywords:** online geogame, game-based simulation, data about human behavior,
27 energy consumption and saving

28 **Introduction**

29 Online serious games have been increasingly developed and implemented for urban
30 planning, public participation in planning, civic engagement and architecture. They can
31 be played either online through the Internet or any other computer network or offline by
32 downloading the application on a device. As the browsers became more sophisticated,
33 the developers started creating online browser games that use a web browser as a client.
34 Such games can be used in a variety of urban planning phases and can enable one to

35 visualize the current situation in the city and/or future developments (Gordon and
36 Manosevitch 2010, Billger et al. 2017); facilitate discussions on the issues relevant to
37 the citizens, public officials or other stakeholders (Gordon and Schirra 2012, Thiel and
38 Fröhlich 2017); visualize and discuss the proposed plans and changes in the city
39 (Gordon et al. 2011, Poplin 2012, Poplin et al. 2017); learn and reflect upon urban
40 issues (Gordon and Baldwin-Philippi 2014, Devisch et al. 2016, Poplin and Vemuri
41 2018); foster negotiation and consensus building in urban planning (Saad-Sulonen and
42 Botero 2010, Vemuri et al. 2014, Poplin and Vemuri 2018); and/or connect citizens in
43 an online virtual platform enabling them to submit their suggestions for urban designs
44 or comment on the proposed urban planning alternatives (Foth et al. 2009, Hacker and
45 Matchin 2009, Poplin 2014). Some of the efforts concentrate on creating immersive
46 playful environments for learning about cities and urban planning aimed to attract youth
47 and children (Tóth and Poplin 2013a, Tóth and Poplin 2013b, Tóth and Poplin 2014, de
48 Andrade et al. 2016). Most of these efforts in developing serious games for urban
49 planning, in their initial stages, focus on the question “How to design an online game
50 that can support one or several of the urban planning processes?” In the process of
51 designing an online game, the design of the user interface, its visualization, aesthetics
52 and usability play an important role. The designed user interface represents the first
53 contact with the user/player and may substantially influence the decision about her
54 further use of the game. It also has an impact on the frequency with which players
55 visit/play the same game and whether they keep on returning to the game or not.

56 The central research focus is built on the question how to design an online game,
57 game that can be played in a web-browser, and will enable one to collect big volumes of
58 data about occupants’ behavior in built environments. Big data in this context refers in
59 particularly to the volume of data (Keller et al. 2012, Fuller 2015, De Mauro et al. 2016,

60 Thakuriah et al. 2017) as it can be potentially collected online, 24/7 from every location
61 with Internet connection and a device with a web browser. Its focus is on human
62 behavior related to energy consumption. Annual Energy Outlook 2014 with Projections
63 to 2040 reported that in 2013 residential and commercial buildings consumed more
64 than 40% of the United States total energy and more than 70% of the electrical energy,
65 resulting in a national energy bill of \$410 billion (US Department of energy (EIA)
66 2014). Additionally, low energy buildings often fail to meet expected performance, with
67 occupant behavior (OB) contributing decidedly towards building energy consumption
68 and indoor environmental quality (Sun et al. 2014, Hong et al. 2015a, Schakib-Ekbatan
69 et al. 2015). Occupants interact with building systems in a variety of ways by opening
70 and closing windows, changing blinds, adjusting thermostats, controlling lights, all of
71 which influence the energy consumption of the building (Hong and Lin 2013, O'Brien
72 et al. 2013, Sun et al. 2014).

73 This paper has three main goals: a. It explores an implementation of an online
74 game *e-footprints* that aims to collect data about occupants' behavior related to energy
75 consumption and saving, and b. It summarizes the feedback related to the first prototype
76 of the user interface, and c. Discusses take-aways for the development for other online
77 serious games for collecting data about human behavior in a built environment. The
78 developed *e-footprints* energy game is an example of a game-based environment that
79 may help to collect big data about occupants' behavior in a built environment. The
80 paper introduces at first experimentally developed analogue version of *e-footprints*
81 energy game and the results of the initial test of its first prototype. The game prototype
82 was tested with the help of an online survey that involved 110 international university
83 students. The main goal of the test was to gain a better understanding of the elements of
84 the user interface, in particular the visualization and aesthetics and how appealing are

85 they to the young generation of players, which is often the most demanding and critical
86 generation of players. The results of this test were the foundations for the next step of
87 game development that lead into the development of an operational, web-browser
88 version of the second game prototype. This paper summarizes the results of the
89 empirical study, reflects upon the take-aways for the development for other online
90 serious games for collecting data about human behavior in a built environment, and
91 concludes with a discussion on further research directions.

92 **Online serious games for data collection about occupants' behavior in a built**
93 **environment: research focus**

94 *One of the most difficult tasks people can perform,*
95 *however much others may despise it, is the invention*
96 *of good games. (- C.G. Jung)*

97 Energy-related human behavior in buildings is difficult to define and quantify, yet
98 critical to our understanding of total building energy consumption (Hong et al. 2015a).
99 Humans interact with their built environment in order to satisfy their thermal, visual and
100 acoustic needs and preferences. These interactions are typically grounded in the
101 Humphreys' principle of adaptation (Hong et al. 2015a, Hong et al. 2015b), which states
102 that if a change occurs such as to provide discomfort, people react in ways which tend
103 to restore their comfort (Humphreys 1997). Parson (1993) distinguishes among the
104 following three main adaptive responses: behavioral, physiological, psychological. A
105 behavioral response is any type of action performed to maintain or restore a state of
106 comfort when the indoor environmental conditions cause discomfort. Such activities
107 may include turning on or off Heating, Ventilation, and Air-Conditioning (HVAC)
108 equipment, opening or closing windows, using shades and/or blinds, adjusting
109 thermostat settings, using fans, caulking, using plastic on the windows and/or doors, etc.

110 This paper concentrates on the behavioral response. A physiological response is any
111 type of unconscious reaction which allows the human body to adapt thermally to the
112 indoor environment. Example of that would be shivering when cold or sweating when
113 very warm or hot. A psychological response is any type of individual reaction to the
114 indoor environment due to discomfort, strain, pressure, motivation or adaptation to the
115 environment. It may involve the cognitive and cultural variables of each individual with
116 respect to their perception of the indoor environment (Parson 1993).

117 In the past, this type of behavioral comfort-driven responses of occupants has
118 been modeled and simulated based on some limited evidence/data gathered from field
119 studies. Such studies often use questionnaires as a research method. The problem with
120 using questionnaires is that the occupants often enter the preferred and not the actual
121 response. Some people quickly respond without too much premeditated thought.
122 Existing models of the human-building interaction include assumptions based on some
123 generic input data assuming that the occupants behave in a set way according to
124 standard deterministic design conditions such as preferred temperature, ventilation rates,
125 thermostat set points and other threshold values (Hong and Lin 2013, Sun et al. 2014,
126 Hong et al. 2015a, Hong et al. 2015b). Traditionally, in building energy modeling
127 (BEM) programs, occupant behavior inputs are deterministic and less indicative of real
128 world scenarios, contributing to discrepancies between simulated and actual energy use
129 in buildings (Hong et al. 2015a). The inclusion of the adaptive comfort model (de Dear
130 and Brager 1998) into European EN 15251 (2007) and U.S. standards ASHRAE 55
131 (2010) has promoted interest in: (1) the prediction of occupants' actions performed by
132 individuals to restore their personal comfort, and (2) the quantification of the energy
133 impact of their behavior to understand the factors driving the difference between
134 predicted and actual building energy use (Hong and Lin 2013, Sun et al. 2014, Hong et

135 al. 2015a, Hong et al. 2015b). Of particular importance are the actions of turning on/off
136 HVAC equipment, adjusting thermostats, lights, windows and blinds, and moving
137 into/out of spaces.

138 Research described in this paper explores the first prototype implementation of a
139 game-based environment that aims to capture data about behavioral comfort-driven
140 responses of occupants within the built environment. The main research questions that
141 drive this research is how to design and implement a digital online serious game for
142 collecting big data about occupants' in built environments? Serious games are games
143 developed for more than just fun, they are more often games developed for non-
144 entertainment purposes (Abt 1970, Michael and Chen 2005, Ritterfeld et al. 2009). They
145 have been used in many application areas including urban planning, civic engagement
146 in urban planning, education, energy, military and/or management. Serious games may
147 include learning elements; they may be designed as “puzzles to solve, just like
148 everything else we encounter in life” (Koster 2004, p. 34). They may include powerful
149 learning tools and are increasingly used to transform the way people learn and what
150 learning means and how it is available to them (Malone 1981, Lepper and Malone 1987,
151 Malone and Lepper 1987, Mitchell and Savill-Smith 2001, Gee 2003, Squire and
152 Jenkins 2003, Squire et al. 2005).

153 Creating an online game for data collection can be challenging. Successful
154 games for data collection have been developed in the domains of objects tagging as for
155 example image tagging (Hacker and Matchin 2009), music tagging (Lepper and Malone
156 1987, Turnbull et al. 2007, Law and van Ahn 2009, Mandel and Ellis 2009) knowledge
157 extraction (Siorpaes and Hepp 2008), computer vision (e.g. object detection) and
158 labeling (Carlier et al. 2012) and ranking predictions through image search (Bennett et
159 al. 2009, Law 2011). These exemplary games are often called Games with a Purpose

160 (GWAP) and are digital games in which the players' actions in the game contribute to a
161 real-world purpose outside of the game. Most of these games are based on kinds of
162 “validation” mechanisms. Their advantage is that they are generating useful data for
163 scientific purposes as a by-product of their gameplay. The issues researchers, game
164 developers and designers face with serious and GWAP games is that “the task to be
165 solved may mismatch with the game mechanics, thus decreasing the playability of the
166 game and failing to attract people and engage them in the execution of the task” (Galli
167 2014, p. vi.), which is a general issue with serious games and needs additional research
168 in better understanding of their design that can bring players into the state of flow
169 (Csikszentmihalyi 1990).

170 Developing an online serious game, that can be used in a web browser, also
171 involves the process of creating the digital appearance of the game and the user
172 interface. Game design requires a collaboration across disciplines and several iterations
173 in refining the game concept and the game story. It is challenging due to the technology
174 involved, which often enables additional functionalities (in comparison to non-digital
175 games) and the ability to access such games online. Additional challenges in creating
176 online serious games represent the design of the user interface and particularly the
177 visualization, aesthetics and usability of the designed user interface. The novel design
178 principles have been advocating for play-centric and human-centric game design
179 (Fullerton 2008, Brathwaite and Schreiber 2009, Rogers 2014, Schell 2015). One of the
180 key ideas of the play-centric approach is that “ideas [about the game and game design]
181 should be prototyped and tested early. Immediately after brainstorming ideas, we
182 encourage designers to construct a playable version of their idea.” (Fullerton 2008, p.
183 11). They should be tested in the early stages of the game development with its potential

184 users/players. This paper summarizes a response to an early prototype of the game-
185 based simulation titled the *e-footprints* energy game.

186 **Prototype of the *e-footprints* energy game for collecting data on occupants**
187 **behavior in built environments**

188 The prototype of the game *e-footprints* was developed in a research project which
189 started as an initiative of a small number of researchers from various disciplines at Iowa
190 State University. The main goal of the research was to explore the possibilities of game-
191 based environments and their use in various phases of data collection. The prototype of
192 the game *e-footprints* focuses on creating an immersive simulation of built
193 environments which invites players/citizens to experiment, explore, and make their
194 decisions about this environment in order to feel comfortable in the simulated home
195 environment. The first prototype is represented by a set of visualizations of the
196 envisioned user interface and an activity diagram demonstrating the levels of the game
197 which coincide with the levels of data collection (Figure 2).

198 **Level 1: Data collection.** In the first step, the potential players have to register and
199 create their own password. The registration page asks them about their gender and their
200 age group. In the next step the player can choose among four types of housing available
201 to her (Figure 1). It invites them to choose the house/residence similar to the one in
202 which they currently live. These steps represent the first phase of data collection related
203 to the information about the players and their house style. After finishing this first level
204 of the game and the first level of data collection, the game takes the player into a
205 simulated room environment.



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Figure 1. Select the type of a house you live in

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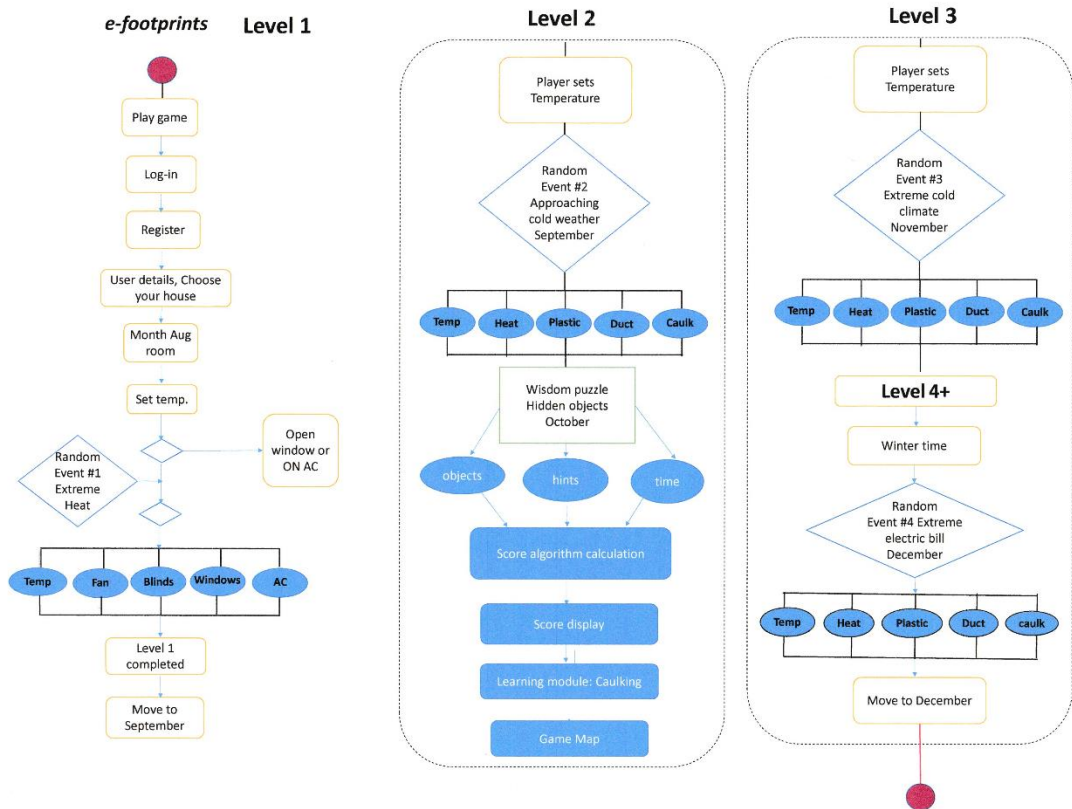
220

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The simulated room environment is central to the *e-footprints* energy game. The room is designed as an interactive online room in which the player/user can adjust the temperature, open up/close the window, turn on/off the air conditioner, and turn on/off the fan. The outside temperature is visible to the user at the top right corner of the screen; it gives her an idea of the temperature outside the house and a feeling for the time of the year/season and time of the day. The player can explore the room and choose the temperature conditions of the room that are the most comfortable for her, given the season and the time of the day. The data being additionally collected is the temperature the players sets in the room and her reaction to the extreme heat “random event”. She has the possibility to choose among several options as her reaction to the extreme heat: open/close the window, turn on/off air conditioning, turn on/off the fan and/or adjust the blinds. Figure 2 summarizes the activity diagram of the game play with the data collection levels clearly marked, along with the decisions available to the player on each step of the game play.



224

225 Figure 2. Levels of data collection implemented in the *e-footprints* energy game

226

227 **Level 2: Data collection.** Level two includes random event two in which the player
 228 experiences approaching cold weather and the associated result of getting cooler. The
 229 player then has some options in how she wants to react to the changes in temperature.
 230 Her options include changing the temperature in the room, install plastic on the
 231 windows and/or doors to insulate them, sealing the ducts or using the method of
 232 caulking the windows. Her decisions and preferences are recorded and saved in the
 233 back-end of the game. Additionally, a wisdom puzzle is introduced for the first time.
 234 Wisdom puzzles in the *e-footprints* energy game are designed as learning tools
 235 throughout which learning can be expanded and deepened by the players. The first
 236 wisdom puzzle aims to teach, in a fun and entertaining way, about caulking and which

237 tools are needed for caulking. The player has to find four hidden objects needed for
238 caulking and can also read more about the advantages of caulking in terms of energy
239 saving. The time needed for the player to accomplish this task is measured by a melting
240 cube and the player gets points for the completion of this task. The points are calculated
241 in an algorithm that takes into account how many hints did the player use to solve the
242 wisdom puzzle and how fast was she able to find the four hidden objects. An example
243 of a wisdom puzzle is presented in Figure 3.
244



245

246 Figure 3. Wisdom puzzle about caulking with the ghosts of Halloween

247

248 **Level 3+: Data collection.** The third, and all the next game levels (3+), introduce a new
249 set of random events and wisdom puzzles. The third level introduces the extreme cold
250 weather random event with the same options to solve the problem as the random event
251 on Level 2. The aim of the game is to capture and save the players' choices after they
252 have been exposed to the option of caulking on Level 2. Will more players choose this
253 option as a possible reaction to the extreme cold event? Level 4 introduces an extreme
254 electric bill random event. The player can respond to this random event by, again,
255 choosing among five options given by the game (Figure 2).

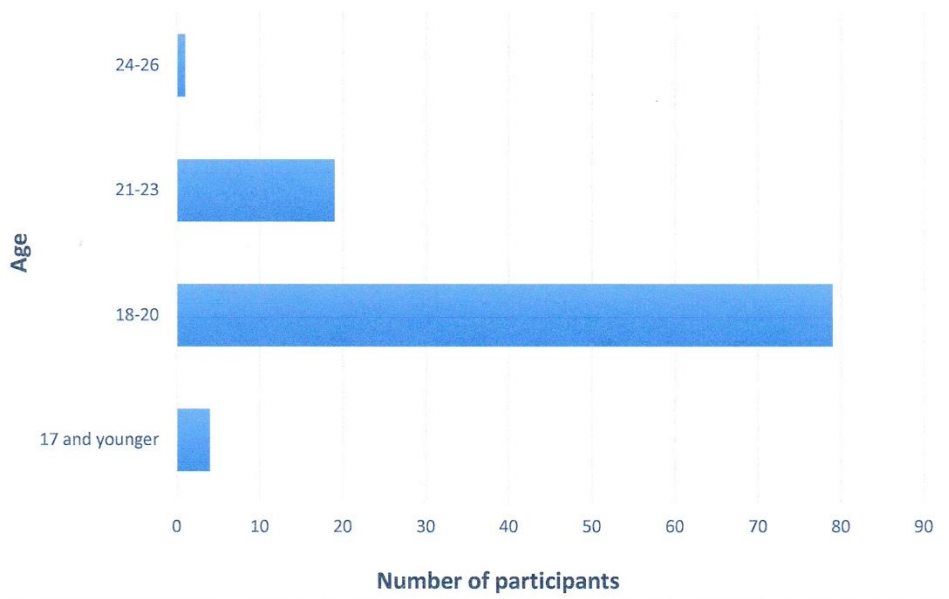
256 **Data privacy considerations.** During the game design, special attention was devoted to
257 data privacy. The first level of data collection collects data about the age, gender, and
258 the type of the house in which the player lives. The players also enter their E-Mail
259 address. This data is not personalized and will – in the further analysis – be used as an
260 aggregated dataset. The research is looking into data patterns and not an individual
261 preferences and choices. The player can skip this stage and still play *e-footprints* energy
262 game anonymously, however this is not explicitly encouraged by the game. On the
263 following levels of data collection (2+), the data collected registers the player’s choices
264 in the built environment and their chosen actions and reactions to the extreme events
265 presented to them in the game. Again, this data will be used for the analytical purpose
266 of the study that concentrates on human behavior in built environment. The main idea
267 behind is to define so called “personas” with typical behavior and use the parameters of
268 these personas for further development of an agent-based system that will enable
269 simulations of human behavior. We will explore patterns of typical behavior and search
270 for co-relations with data collected on the first level such as age, gender, type of the
271 house. A general types of human behavior will be generated based on the collected data.
272 The raw dataset will not be shared or reviled to a third party. The game can be played
273 online and there is no risk for the players associated with playing the game.

274 **Testing the first prototype of the *e-footprints* energy game: research** 275 **methodology and results**

276 ***Participants***

277 The participants of the survey were 110 international students. Testing of the first game
278 prototype is intentionally focused on this group of the millennials that are demanding in
279 respect of game visualizations and aesthetics. They were also quickly available for the
280 first feedback on the designed prototype needed to be executed in very early stages of

281 the game design. Among them, there were 103 undergraduate students and 1 graduate
282 student; six students did not provide information about their university program. Four
283 age groups participated in the survey: 4 students were 17 years old or younger, 79
284 students were between 18-20 years old, 19 in the age category 21-23, and one in the
285 category 24-26 years old (Figure 4). Among them were 26 female, 73 male and 5 other
286 gender. The countries of origin were Burma, Bolivia, Brazil, Canada, China, Dubai,
287 Ecuador, Egypt, India, Indonesia, Iraq, Kenya, Kuwait, Malaysia, Mongolia, Nigeria,
288 North Korea, Oman, Pakistan, Puerto Rico, Riyadh, Russia, Spain, South Korea, South
289 Sudan, Taiwan, Uganda, and United Arab Emirates.
290



291
292

Figure 4. The age of the survey participants

293 **Material**

294 The material used included the online survey and the graphics designed for the first
295 prototype of the *e-footprints* energy game. The online survey was designed as an online
296 questionnaire in Qualtrics open source software, which includes a database and enables
297 one to store the data collected, executing the online survey. The graphics of the *e-*

298 *footprints* energy game shown to the students in the online survey included four pictures
299 which represented the design of the game user interface and its graphical visualization.
300 The needed materials also included their personal devices for accessing the online
301 survey such as personal computers, tablets or smart phones.

302 ***Procedure***

303 The international students gathered in a lecture hall which has a big screen, a projector
304 and audio equipment. The chairs in the room have a small table attached to them so that
305 the students can place their laptop or a mobile device on them. The leading
306 experimenter (the author of this article) introduced the game project, explained its main
307 goals, and the goals of this survey. The goals of the project were to develop an online
308 energy game which simulates a home environment and enables one to collect data about
309 the occupants' behavior in built environments. The lead experimenter then asked the
310 students to follow the link of the online survey sent to them via email. From here on in
311 the process of the evaluation they were able to concentrate on the four graphical
312 visualizations of the game as presented in the online questionnaire. The students were
313 allowed to respond to the online survey at their own pace; there was no pressure for
314 them to finish it in a particular time frame.

315 **Empirical results of the testing *e-footprints* energy game and its further** 316 **development**

317 The survey participants were asked to evaluate the visualization and aesthetics of the
318 room, the whole image, and comment on the usability of the designed user interface.
319 The first picture shown to the students was the graphics of the room (Figure 5) with the
320 following instructions: "*Imagine you are playing an online game at your home. This*
321 *particular online game takes you into a living room. The outside temperature is 65°F*
322 */18°C, and it is in the middle of the day in October.*"



324

325

Figure 5. Introducing the room in the *e-footprints* energy game

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The more specific results for every question are summarized in Table 1. The main

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conclusion is that there is a potential for improvements of the graphics. The issue of the

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missing ceiling was a topic among many participants; over 40% of the users suggested

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changing the open roof, and additionally over 24 % seemed to somewhat agree about

331

the need to change the (lack of) ceiling (64% altogether).

# Field	Strongly agree	Somewhat agree	Neither agree nor disagree	Somewhat disagree	Strongly disagree	Total
1. I find the design of the room attractive	11 10.09%	36 33.3%	26 23.85%	24 22.02%	12 11.01%	109
2. The furniture is similar to mine at home	10 9.09%	22 20.00%	34 30.9%	23 20.91%	21 19.09%	110
3. The colors of the room are appealing	16 14.55%	28 25.45%	32 29.09%	23 20.91%	11 10.00%	110
4. There is no ceiling; this needs to be changed	45 40.91%	24 21.82%	22 20.00%	14 12.73%	5 4.55%	110
5. The emoticon makes the scene fun	15 13.89%	24 22.22%	34 31.48%	16 14.81%	19 17.59%	108

332

333 Table 1. Responses related to the graphics introducing the room in Figure 5

334

335 The students were asked to evaluate the graphics and aesthetics of the whole image and
336 not just the room (Figure 5). The image was perceived as aesthetically pleasing by 39%
337 of the students and 33% are neutral about this topic. The colors were considered
338 pleasing for 44% of the international students and 42% are neutral. The students
339 confirmed that they perceived too many objects on the image (for 55% of the students)
340 and that there is too much detail included on the image (44%). Table 2 summarizes the
341 results of the empirical study.

342

343

344

345

# Field	Strongly agree	Somewhat agree	Neither agree nor disagree	Somewhat disagree	Strongly disagree	Total
1. The image is aesthetically pleasing	19 17.43%	24 22.02%	36 33.03%	18 16.51%	12 11.01%	109
2. The colors are aesthetically pleasing	14 12.73%	30 27.27%	42 38.18%	13 11.82%	11 10.00%	110
3. There are too many objects on the image	19 17.59%	36 33.33%	26 24.07%	20 18.52%	7 6.48%	108
4. There is too much detail on the image	13 12.04%	31 28.70%	39 36.11%	15 13.89%	10 9.26%	108

346

347 Table 2. Responses related to the graphics and aesthetics of the user interface in Figure

348 5

349

350 The usability of the proposed user interface was evaluated very positively. The majority

351 of the students would intuitively know very well how to use it. More than 70% of the

352 students strongly agree and somewhat agree that it is clear how to adjust the

353 temperature, how to open the window, and that it is easy to read the text on the image.

354 Table 3 summarizes the empirical results.

# Field	Strongly agree	Somewhat agree	Neither agree nor disagree	Somewhat disagree	Strongly disagree	Total
1. It is clear how to adjust the temperature	51 46.36%	31 28.18%	20 18.18%	4 3.64%	4 3.64%	110
2. It is clear how to open the window	47 42.73%	40 36.36%	15 13.64%	4 3.64%	4 3.64%	110
3. The text in the middle of the image is easy to read	48 44.44%	37 34.26%	12 11.11%	9 8.33%	2 1.85%	108
4. The text in the upper corner is easy to read	27 25.71%	35 33.33%	16 15.24%	21 20.00%	6 5.71%	105

355

356 Table 3. Responses related to the usability of the user interface presented in
357 Figure 5

358

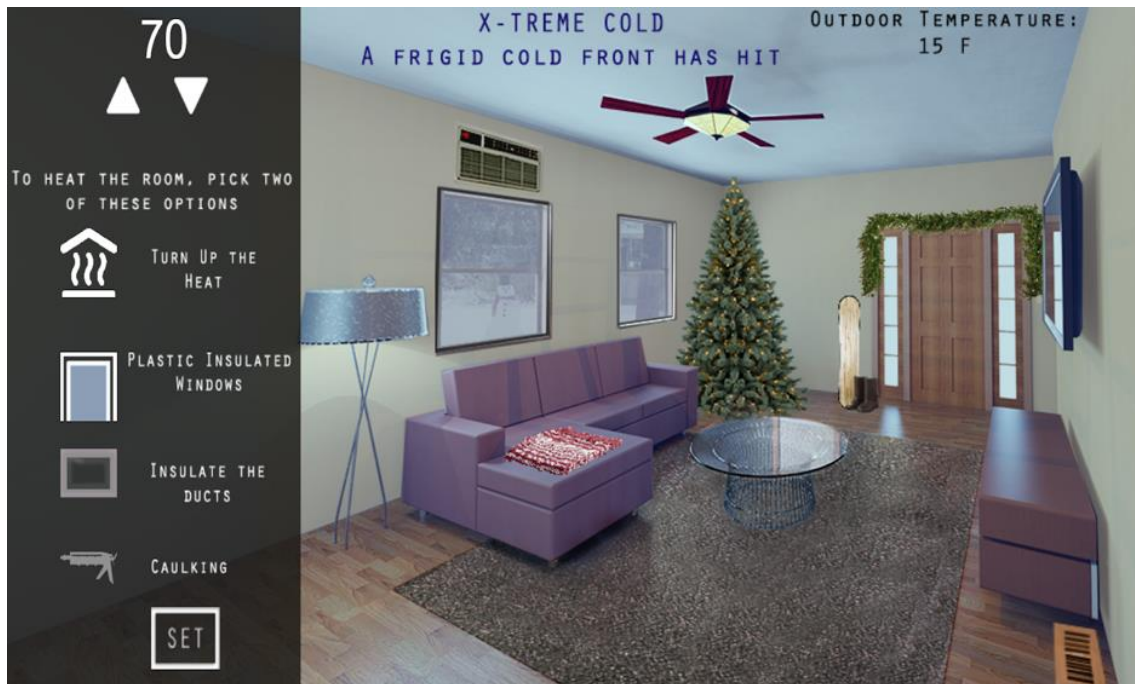
359 **Improvements of the visualizations and aesthetics.** Based on the testing results we
360 focused on the improvements of the visualization and the digital, online implementation
361 of a prototype that can be tested using web browser. Our second prototype includes a
362 Rhino 3-D model of the room and a more clean and professional design. Figure 6
363 demonstrates the improvements done in conceptualizing the main interactive room. The
364 room has the ceiling, less objects, less detail and the colors were changed. Additional
365 tests needs to be done in order to be able to understand how appealing is this room to
366 the players.

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Figure 6. New design of the room with an extreme cold random event

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Further envisioned changes include the option for the players to choose the furniture

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and other objects in the room and drag&drop them into an empty room space. This

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option is a direct response to the survey in which the participants claim that the

376

proposed furniture has not been similar to the one they have at home. The *e-footprints*

377

energy game aims to achieve immersive environment in which the players' feel

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immersed into their own home.

379

Implementation of the online prototype. After the first testing of the user interface

380

and visualizations we implemented a new version of the *e-footprints* energy game that

381

can be tested in a web browser. The front-end was implemented with Hypertext Markup

382

Language (HTML), Cascading Style Sheets (CSS), Java script and JQuery which

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enables to add complex functionalities to the web pages, and Bootstrap that made the

384

web pages platform independent from programming. The data collection, the back-end

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of the *e-footprints* energy game was implemented with the help of Structured Query

386 Language (SQL) server and Visual Studio 2015. The implemented database enables one
387 to store the players' responses in the tables that are then joined into one common table
388 responding to the game play of each individual player. ASP.net and Ajax enable the
389 communication between the front-end and back-end. Ajax was used to code the logic
390 that enables the game to send the values from the front-end to the back-end. Windows
391 10 server is used to launch the *e-footprints* energy game. The current version of the
392 online game can be tested online using the Chrome internet browser following the link
393 <http://efootprints.design.iastate.edu/>.

394 **Lessons learned and research challenges applicable for online game-based** 395 **data collection about human behavior**

396 The main goal of this experiment was to get immediate feedback on the first design of
397 the prototype of *e-footprints* energy game. Such an approach has been recommended in
398 the game design community and may lead to higher quality of game design in general,
399 which we followed in this research project. The presented results accomplished for the
400 *e-footprints* energy game demonstrated the need for the changes in the graphical
401 visualization and aesthetics of the initial prototype. The second prototype was
402 developed based on the results of the testing. However, what are the results gained from
403 developing and testing one prototype that can be generally concluded and be useful for
404 other game-based simulated environments created for collecting data about human
405 behavior? What are the main research challenges? The list of important features and
406 research challenges to be considered in the online game design for urban planning and
407 data collection about human behavior includes the following elements:

408 **Big data game-based online collection.** The game presented in this paper is
409 unique as it deals with data about human behavior that it is in its nature difficult to
410 capture. To the knowledge of the authors of this article, no comparable games exist

411 which makes this contribution unique. Researchers in human modeling and simulation
412 and artificial intelligence (Normoyle et al. 2012) that use games mostly concentrate on
413 the collection of player metrics which basically records the moves of the player and her
414 decisions in the game without intending to focus on modeling human behavior by
415 collecting this data. Examples may include the number of times a certain weapon was
416 used or the percentage the player spent performing a certain activity which is then
417 analyzed by the game designers (Ambinder 2009). Some of them focus on experimental
418 design in such a way that they can actively collect players metrics related to the
419 designed scenario in which they also concentrate on data uncertainty (Chaloner and
420 Verdinelli 1995, Settles 2012). Special concern needs to be dedicated to ethics of
421 collecting data online with or without players' knowledge and permission. Also, data
422 privacy is a valid concern for online game-based data collection applications.

423 **Modelling human behavior using Agent-Based Modelling (ABM) and**
424 **Machine Learning.** ABM and artificial intelligence (Epstein and Axtell 1996, Axelrod
425 1997) has been be increasingly implemented and used to study patterns of players'
426 behavior. In an ABM, humans are represented/simulated with the help of agents whose
427 models are designed and coded based on the knowledge of the actual behavior and
428 reactions of humans in a variety of situations. Machine learning (Mitchell 1997, Bishop
429 2006) can further on help to study patterns in the collected data. It can help identify and
430 build mathematical models based on sample data in order to be able to make predictions
431 or decisions without being specifically trained or programmed to perform certain tasks.
432 Combined with data mining it can focus on exploratory data analysis. Machine learning
433 was coined in 1959 by Arthur Samuel who was a pioneer in computer games and
434 artificial intelligence (Samuel 1959).

435 **Meaningful choices and meaningful experiences.** The ability to be able to
436 make meaningful choices is crucial for the players. Iten et al. (2017) found that
437 meaningful choices were often characterized by a combination of different options from
438 which the player had to choose one - they can be moral, strategic or emotional. Being
439 able to choose between these types of options and knowing that these choices would
440 have consequences lead to choices that were experienced as meaningful. According to
441 Rogers, Woolley et al. (2017) meaningful game experiences can be related to the depth
442 of the story and its characters; experiences which most distinguished meaningful from
443 fun experiences were those that involved moral choices and dilemmas. The fact that
444 games ask players to choose between certain options puts the responsibility in their
445 hands and makes them think about the choice at hand. In further development of *e-*
446 *footprints* energy game, the ability to choose the furniture, the location of the furniture
447 and the colors may enable the player to create the room she wants to see and feels
448 comfortable with. The ability to choose has, in general, further implications on the
449 player's attitude towards the created environment in the game. It may increase the
450 feeling of immersion and ownership (Oliver et al. 2015, Rogers et al. 2017).

451 **Conclusions and further research**

452 This paper summarizes the empirical results gained from an online survey of the user
453 interface designed for the first prototype of the *e-footprints* energy game, a game that
454 concentrates on collecting data about players'/occupants' behavior in a built
455 environment. This game is a unique contribution to the online methods for collecting
456 data about human behavior that can be used by many internet users across continents. It
457 was designed to collect high volume of data about human behavior in built
458 environment. The focus of the study presented in this paper was on the visualization of
459 the first prototype of the *e-footprints* energy game, aesthetics of the graphical images,

460 and the usability of the presented user interface. All these elements substantially impact
461 the game play and the players' ability to repeatedly return to the game. Only with a high
462 quality user interface and its functionalities that players like and are willing to return to,
463 an online game can be successful.

464 The empirical results demonstrate that additional resources in improving the
465 graphical visualization and aesthetics of the *e-footprints* energy game need to be
466 devoted in the next stages of the game development. The game is currently in the phase
467 of redesign and further development. This phase includes a complete redesign of the
468 graphics included in the *e-footprints* energy game which will also slightly change the
469 functionalities of the game. We are working on the new graphical style of the game,
470 choices of new colors, changes of the text and its style, and the visualization of the
471 room itself. The mechanics of the game will allow for more meaningful choices for the
472 player; being able to choose the type of the room objects and their placement would
473 enhance this game and the players' immersion into the built environment. This step of
474 testing in the initial phase of development was crucial for the next steps of prototyping
475 of the *e-footprints* energy game.

476 This paper is a continuation of the research on data analytics and urban planning
477 games. In general, the community game-based data collection is a growing research
478 area and represents a unique potential for further research, development and
479 implementation. It may bring together computer scientists, architects, machine learning
480 experts, urban planners, urban informatics researchers and experts and geoinformation
481 scientists. It is a truly an interdisciplinary effort to design and develop a successful
482 serious online game.

483 There is a need for a more systematic approach to the topic. Establishing a
484 specific conference, dedicated workshops or seminars on the topic would be very

485 valuable. Collaboration with computer scientists who could advise on the recent trends
486 in the implementation of such games, and game designers who could contribute their
487 knowledge about successful methodologies and techniques of the design would bring
488 insight and improvements of participatory methods needed in urban planning. Novel
489 ways of data collection still need to be discovered; ways in which data collection
490 becomes fun, playful, engaging and truthful in the ways data has been presented and
491 analyzed. Research has a ways to go to improve our understanding of the player-game
492 interaction, the design of the user interface, the dynamics of games, and how to create
493 and stimulate the feeling of “flow” for the players. The masses of collected data can
494 enable to study the patterns in the collected data, employ analytical methods such as
495 machine learning and data mining and improve our understanding of human behavior
496 which can then consequently lead to smarter buildings in our future oriented smarter
497 cities.

498 **Data Availability**

499 c. Some data, models, or code generated or used during the study are available
500 from the corresponding author by request. Items available: the complete results of
501 the online survey performed to evaluate the first prototype of the *e-footprints*
502 energy game.

503

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777 **Tables**

778 Table 1. Responses related to the graphics introducing the room in Figure 5

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# Field	Strongly agree	Somewhat agree	Neither agree nor disagree	Somewhat disagree	Strongly disagree	Total
1. I find the design of the room attractive	11 10.09%	36 33.3%	26 23.85%	24 22.02%	12 11.01%	109
2. The furniture is similar to mine at home	10 9.09%	22 20.00%	34 30.9%	23 20.91%	21 19.09%	110
3. The colors of the room are appealing	16 14.55%	28 25.45%	32 29.09%	23 20.91%	11 10.00%	110
4. There is no ceiling; this needs to be changed	45 40.91%	24 21.82%	22 20.00%	14 12.73%	5 4.55%	110
5. The emoticon makes the scene fun	15 13.89%	24 22.22%	34 31.48%	16 14.81%	19 17.59%	108

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789 Table 2. Responses related to the graphics and aesthetics of the user
 790 interface in Figure 5

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# Field	Strongly agree	Somewhat agree	Neither agree nor disagree	Somewhat disagree	Strongly disagree	Total
1. The image is aesthetically pleasing	19 17.43%	24 22.02%	36 33.03%	18 16.51%	12 11.01%	109
2. The colors are aesthetically pleasing	14 12.73%	30 27.27%	42 38.18%	13 11.82%	11 10.00%	110
3. There are too many objects on the image	19 17.59%	36 33.33%	26 24.07%	20 18.52%	7 6.48%	108
4. There is too much detail on the image	13 12.04%	31 28.70%	39 36.11%	15 13.89%	10 9.26%	108

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805 Table 3. Responses related to the usability of the user interface presented in
 806 Figure 5

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# Field	Strongly agree	Somewhat agree	Neither agree nor disagree	Somewhat disagree	Strongly disagree	Total
1. It is clear how to adjust the temperature	51 46.36%	31 28.18%	20 18.18%	4 3.64%	4 3.64%	110
2. It is clear how to open the window	47 42.73%	40 36.36%	15 13.64%	4 3.64%	4 3.64%	110
3. The text in the middle of the image is easy to read	48 44.44%	37 34.26%	12 11.11%	9 8.33%	2 1.85%	108
4. The text in the upper corner is easy to read	27 25.71%	35 33.33%	16 15.24%	21 20.00%	6 5.71%	105

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