

Establishment reality vs. maintenance reality: how real is real enough?

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Remote and virtual laboratories are increasingly prevalent alternatives to the face-to-face laboratory experience; however, the question of their learning outcomes is yet to be fully investigated. There are many presumptions regarding the effectiveness of these approaches; foremost amongst these assumptions is that the experience must be 'real' to be effective. Embedding reality into a remote or virtual laboratory can be an expensive and time-consuming task. Significant efforts have been expended to create 3D VRML models of laboratory equipment, allowing students to pan, zoom and tilt their perspective as they see fit. Multiple camera angles have been embedded into remote interfaces to provide an increased sense of 'real-ness'. This paper draws upon the literature in the field to show that the necessary threshold for reality varies depending upon how the students are interacting with the equipment. There is one threshold for when they first interact – the establishment reality – which allows the students to familiarise themselves with the laboratory equipment, and to build their mental model of the experience. There is, however, a second, lower, threshold – the maintenance reality – that is necessary for the students' ongoing operation of the equipment. Students' usage patterns rely upon a limited subset of the available functionality, focusing upon only some aspects of the reality that has been originally established. The two threshold model presented in this paper provides a new insight for the development of virtual laboratories in the future.

Keywords: virtual laboratory; simulation; student perceptions; verisimilitude

1. Introduction

Remote and virtual laboratories are increasingly prevalent alternatives to the face-to-face laboratory experience; however, the question of their learning outcomes is yet to be fully investigated. There are many presumptions regarding the effectiveness of these approaches; foremost among these assumptions is that the experience must be 'real' to be effective.

The creation of a realistic simulation is not a simple process. It requires experts who understand the physical phenomena being simulated. It requires experts who can build an accurate model of these phenomena. It requires experts who can build an interface to this model. Each of these tasks requires different skills, often from different people, and as the required verisimilitude of the simulation increases, the complexity of each of these skills increases.

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51 It is possible to achieve high degrees of realism in a simulation, but it is an expensive process.
52 Each of the experts involved needs to be paid for, as do the computer resources used in the
53 development of the software. For complex simulations there is further expense when they are
54 implemented – high-end computing resources become necessary to make the software function.
55 Some highly sophisticated simulations – such as flight simulators for pilot training – require
56 dedicated physical infrastructure for their use, further increasing the cost.

57 Embedding reality into a remote or virtual laboratory can be an expensive and time-consuming
58 task. Significant efforts have been expended to create 3D VRML models of laboratory equipment,
59 allowing students to pan, zoom and tilt their perspective as they see fit. Multiple camera angles
60 have been embedded into remote interfaces to provide an increased sense of ‘real-ness’. There have
61 been a wide range of impressive simulations, ranging from wind tunnels (Jia *et al.* 2006) to virtual
62 refineries (Schofield *et al.* 2004). The investment of time, energy and money into these simulations
63 shows that there is a clear perception that the increased realism of sophisticated simulations is
64 worthwhile. Whether this is in fact the case bears further inspection.

65 66 67 68 **2. The effectiveness of reality**

69
70 There is a widespread presumption that a realistic simulation is a more effective learning expe-
71 rience. There are certainly good grounds for this presumption; however, the desire for the
72 simulation to be realistic is in fact a simplification of a number of other objects that correlate
73 well with realism.

74 Students must be able to anchor what they are learning into their prior knowledge (Ausubel
75 2000), much of which will have been learned in a ‘real’ context. Thus, a realistic experience will
76 reduce the cognitive dissonance, and facilitate the assimilation of new knowledge. If the students
77 have real experiences in which to anchor new knowledge, it is logical that it will be easier if these
78 new experiences are themselves real.

79 Fidelity is an important aspect of establishing the presence of the simulation. Fidelity impacts
80 on the learner’s ability to transfer the knowledge they have learned. As practicing engineers, the
81 contexts in which they will need the knowledge will be real, rather than educational simulations.
82 Learners transfer better in high-fidelity situations, but if they do not learn in the first place, then
83 there can be no transfer (Alessi and Trollip 1991). This is the risk of high-fidelity simulations –
84 ‘increasing fidelity, which theoretically should increase transfer, may inhibit initial learning which
85 in turn would inhibit transfer’ (Alessi and Trollip 1991).

86 A simulation with too much fidelity – such as an aircraft cockpit with hundreds of controls and
87 masses of feedback – can be overwhelming to a learner, particularly an inexperienced learner.
88 Another example of where higher levels of fidelity are inappropriate was raised by Aldrich (2004),
89 who observed that the best selling bird-watching guides use illustrations of birds rather than
90 photographs.

91 There is an additional danger in the use of simulations that students lose sight of the real
92 hardware being simulated, and instead get caught up in the ‘computer game’ attitude towards
93 the software (Schofield *et al.* 2004). This issue – whether the students focus on the equipment
94 being simulated, or the interface of the simulation – is known as transparency. If the simulation is
95 transparent – and thus, by implication, more ‘real’ – then the students are focusing upon the real
96 physical phenomena, rather than the artificial environment of the simulation.

97 A more realistic simulation offers many advantages in light of these desirable outcomes – the
98 mental distance between the learning experience and the students’ previous and future experiences
99 is lower. In these instances reality is good, but it must not be reality for reality’s sake – the ‘real-ness’
100 of the simulation must be focused to achieved the desired outcomes.

101 An important tool for learning through experiments is how students deal with contradictions
102 between their expectations and their observations. If the data they collect does not match their
103 expectations, then they must address their discrepancy. Students can question their data, and
104 retake measurements. Alternatively, they can question their expectations, and change their under-
105 standing of the phenomena. It is this evolution of understanding that is the objective of the
106 laboratory exercise.

107 With a simulation, there is an additional third option – to question the accuracy of the simulation.
108 Students are able to question whether the data is an accurate representation of reality, and in doing
109 so avoid having to question their own mental models of the physical phenomena. The greater
110 the sense of reality the students experience from the simulation, the less likely they are to fall
111 into this trap.

112 The challenge is to ensure that the simulation is able to provide a level of reality that best fits the
113 needs of the students. Too simplistic and there is a risk that the transparency will be compromised,
114 and the simulation becomes a computer game. Too complex and there is a risk that the students
115 will be overwhelmed with information. The challenge is further complicated by the way in which
116 students' needs change over time.

117 118 119 **3. The stages of reality**

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121 The core objective is to ensure that the level of verisimilitude matches the needs of the students
122 as they operate the simulation. The challenge is that the needs of students vary over the course
123 of their use of any given simulation. Their use can be broadly split into three phases – initial use,
124 regular use and expert use.

125 In the initial use phase, students encounter the simulation for the first time. Everything is new
126 to them, and they need to familiarise themselves not just with the physical phenomena being
127 modelled, but also with the interface, the experimental procedures and a range of other potential
128 factors. The use of standardised interfaces across a suite of simulations can help reduce the burden
129 of learning a new interface, but ultimately this orientation is unavoidable. Students are learning in
130 this phase, but this learning is mostly preparation for the objective-specific learning in the regular
131 use phase.

132 In the regular use phase, the students have familiarised themselves with the simulation interface,
133 developed a mental model of the physical phenomena, and are now able to explore how changing
134 the input parameters leads to changes in the output parameters. In this way the students can address
135 the objective-specific learning outcomes of the simulation.

136 In the expert use phase, the students seek to make their use of the simulation more efficient and
137 effective. This phase is also characterised by students finding more efficient ways to implement
138 functionality that they are already used – they seek short cuts through the simulation.

139 There is not a crisp distinction between the three phases, and students will move backwards
140 and forwards through the phases. Indeed, the expert use phase will involve exploration to find
141 new short cuts, which is a form of exploration similar to the initial use phase. The needs of the
142 students are different in each of these phases.

143 A simulation provides students with a range of interaction options – they have many branches
144 in the pathway in which they interact with the software. In the initial user phase, the students'
145 objectives are familiarisation with the simulation, which requires them to travel through many of
146 the branches, forwards and backwards. As they become increasingly familiar with the simulation,
147 they will identify which of the branches they require most often, and start to identify how some
148 of these branches link up to provide the functionality of the simulation.

149 In the regular use phase, the students tend to stick to the same regular pathways through the
150 software. They have identified how to change parameters, how to collect output data, how to

151 change perspectives of on-screen imagery, and now they use these skills to explore the physical
152 phenomena. What they are doing in this process is reducing the range of pathways they take
153 through the simulation – they are unconsciously making options in the simulation redundant.

154 In the expert user phase, the students consciously seek to make pathways in the software
155 redundant. Rather than stick to their existing pathways, they look for newer, shorter pathways,
156 and having found them, they no longer use the old approaches. Richardson *et al.* (2006) illustrated
157 this clearly with their simulation of an electronics workbench, in which students could use hotkeys
158 to snap directly to specific views of the equipment. While in the initial use phase, students were
159 happy to pan, zoom and tilt their way across the workbench. They then quickly moved to the
160 expert phase, where they chose specific views of the equipment, and then used hotkeys to swap
161 between them – abandoning the pan, zoom and tilt functions completely.

162 Richardson's pan, zoom and tilt functions are largely redundant once students become familiar
163 with the software, but they form a critical part of the students' initial use of the simulation. These
164 functions are essential to help establish the reality of the simulated workbench, but are largely
165 redundant in the maintenance of this reality.

166 The reality will persist as long as the transparency of the interface remains, and it is only when
167 the simulation prevents an option to the user that the transparency will suffer. Many of the paths
168 through the simulation are only used in the initial familiarisation stage, and as such they are
169 only necessary in that phase. Once the students become familiar with the simulation, they narrow
170 down the range of pathways that they use – potentially allowing for these pathways to be removed
171 without the students noticing, and thus without a loss of transparency or reality.

172 173 174 **4. Different reality thresholds**

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176 Students have different learning objectives in the different phases of their use, and these manifest
177 in different usage patterns as they become more familiar with the simulation. Indeed, sometimes
178 students will deliberately compromise the reality of the simulation as part of their expert user
179 phase. This was observed by Koretsky *et al.* (2008) with their virtual silicon wafer factory, in
180 which students must set the parameters for the silicon deposit process, and then test the wafers
181 that are produced. The virtual factory contained separate interfaces for the two operations, with
182 one interface that controlled the virtual factory, and a second that controlled the virtual testing
183 rig. The first interface would generate experimental data and save it to disc, which would then be
184 available through the interface to the testing rig.

185 The students realised that although the factory interface was displaying the 'process underway'
186 message, representing the time taken for the manufacturing process to occur, the experimental data
187 was in fact already saved to disc, and available at the testing station. This allowed for students
188 to enter the production parameters, commence the process, then collect the results before the
189 simulation had in fact 'completed' the process. This is a quicker way to access the data, but it
190 undermines the 'reality' of the experience.

191 The question then becomes which is more important? Is it appropriate to make students wait
192 for data to reinforce the concept that industrial processes are not instantaneous? Or is it better
193 to disregard this aspect of the reality in order to allow them to run more combinations of the
194 parameters, and understand how each of these parameters changes the outcome of the process.
195 A simulation allows the possibility of sacrificing reality to enhance other learning outcomes.
196 Slow physical processes can be sped up, slowed down or even reversed to allow for a deeper
197 understanding to be achieved. For students in the initial use phase, these deviations undermine
198 their establishment of the reality of the simulation. For students in the regular and expert use phases,
199 however, the reality is already established, and these deviations allow them to pursue the other
200 learning objectives more effectively.

201 Effectively, there are two different thresholds for adequate realism from a simulation – one that
202 is sufficient for students to establish a sense of reality from the simulation, and a second, lower
203 threshold that is necessary to maintain the sense of reality. These two thresholds correspond to
204 the different learning objectives and contexts that students encounter during the different phases
205 of their use.

206 207 208 **5. Conclusion**

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210 Developing a simulation is a complex balancing act. There are many compromises that must be
211 made to create the most effective learning environment within the constraints of the available
212 resources. There is a general consensus that the more realistic the simulation, the better – but this
213 consensus may be flawed.

214 The learning needs of students evolve throughout their interactions with simulations, with
215 novice users having different learning objectives than experienced users. For a novice user, the
216 belief in the fidelity of the simulation is an important learning outcome, so that they are willing to
217 engage with the physical phenomena the simulation represents. While it is essential to establish
218 that the simulation is an accurate representation of reality, it is not essential that this reality be
219 maintained to the same level. Many of the features that are so impressive at first contact, and so
220 useful in establishing the reality of a simulation, are in fact unnecessary bells and whistles for the
221 experienced user.

222 The differences in the necessary levels of reality – the two distinct threshold model – offer
223 new opportunities in the design and implementation of educational simulations. By focusing the
224 resources into the areas that will provide the best return on investment, rather than into improving
225 functionality that will ultimately go unused, educational simulations can be made more efficient
226 and effective.

227 228 **References**

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