

1 **Virtual strike and dip - Advancing inclusive and accessible field geology**

2

3 *Natalie Bursztyn¹, Pejman Sajjadi², Hannah Riegel³, Jiawei Huang², Jan Oliver Wallgrün², Jiayan*
4 *Zhao⁴, Bart Masters², Alexander Klippel⁵*

5

6 *¹Department of Geosciences, University of Montana, Charles H Clapp Building 126, Missoula,*
7 *MT 59812, USA*

8

9 *²The Center for Immersive Experiences, The Pennsylvania State University*
10 *University Park, PA 16802, USA*

11

12 *³Department of Geological and Environmental Sciences, Appalachian State University, 033*
13 *Rankin Science West, Boone, NC 28608, USA*

14

15 *⁴Department of Computer Science, University of Arkansas at Little Rock, 2801 South University*
16 *Avenue EIT 579, Little Rock, AR 72204, USA*

17

18 *⁵Department of Environmental Sciences, Wageningen University & Research, Wageningen,*
19 *Netherlands*

20

21 *Correspondence to: Natalie Bursztyn (natalie.bursztyn@mso.umt.edu)*

22 **ABSTRACT**

23 Accessibility and inclusivity in field geology have become increasingly important issues to address
24 in geoscience education and have long been set aside due to the tradition of field geology and the
25 laborious task of making it inclusive to all. Although a popular saying among geologists is “the
26 best geologists see the most rocks”, field trips cost money, time, and are only accessible for those
27 who are physically able to stay outside long hours. With the availability of 3D block diagrams, an
28 onslaught of virtual learning environments is becoming increasingly viable. Strike and dip is at the
29 core of any field geologist’s education and career; learning and practicing these skills is
30 fundamental to making geologic maps and understanding the regional geology of an area.

31
32 In this paper, we present the Strike and Dip virtual tool (SaD) with the objective of teaching the
33 principles of strike and dip for geologic mapping to introductory geology students. We embedded
34 the SaD tool into an introductory geology course and recruited 147 students to participate in the
35 study. Participants completed two maps using the SaD tool and reported on their experiences
36 through a questionnaire. Students overall perceived the SaD tool positively. Furthermore, some
37 individual differences among students proved to be important contributing factors to their
38 experiences and subjective assessments of learning. When controlling for participants’ past
39 experience with similar software, our results indicate that students highly familiar with navigating
40 geographical software perceived the virtual environment of the tool to be significantly more
41 realistic and easier to use compared to those with lower levels of familiarity. Our results are
42 corroborated by a qualitative assessment of participants’ feedback to two open-ended questions,
43 highlighting both the overall effectiveness of the SaD tool, and the effect of geographical software
44 familiarity on measures of experience and learning.

45
46

47 1 INTRODUCTION

48 *1.1 The “field” environment: real, virtual, and implementation for remote learning*

49 The *field* may be the single most prominent element defining geosciences. Processes relevant to
50 Earth sciences happen in the field, and their phenomenological traces are observable in that
51 physical space. Thus, fieldwork and the educational components of field trips and field camps are
52 frequently held in the highest regard (Orion and Hofstein, 1994; Elkins and Elkins, 2007; Pyle,
53 2009; Semken et al., 2018). Fieldwork remains a graduation requirement for most geoscience
54 programs despite increasing concerns over it being inaccessible to many students, predominantly
55 from underrepresented groups, as a result of financial, cultural, physical, and safety barriers.

56

57 Frequently, physical locations are hard to reach, or they may be impossible, dangerous, or too
58 expensive to access (e.g., the location is on a different continent, in a restricted area, or only existed
59 in the past) (Slater, 1999; Bowman and McMahan, 2007), or from the spring of 2020 to at least
60 into the summer of 2021, physical field experiences are hindered by the global COVID-19
61 pandemic. Furthermore, recent studies have shown that the field experience is not inclusive and
62 may be hindering retention and diversity within geoscience undergraduate programs (Hall et al.,
63 2004; Giles et al., 2020; Morales et al., 2020). In contrast, virtual field trips can allow instructors
64 to expose students to widely accessible, relevant, and authentic learning experiences independent
65 of time and space (e.g. Stumpf et al., 2008; Bursztyn et al., 2017; Mead et al. 2019; Klippel et al.
66 2020). Leveraging increasingly accessible high-resolution computing devices for education has
67 the potential to positively impact student engagement (Witmer and Singer, 1998; IJsselsteijn and
68 Riva, 2003) and efforts to integrate emerging technology into the classroom to improve
69 undergraduate success in introductory geoscience courses have further demonstrated the
70 importance of experiential learning exemplified best by field trips (Cunningham and Lansiquot,
71 2019; Dolphin et al., 2019; Lansiquot and MacDonald, 2019; Moysey and Lazar, 2019). While
72 there is some positive evidence that compares actual and virtual field trips (e.g., Klippel et al. 2019,
73 Marshall et al. under review), considering fieldwork without the field (i.e. in a virtual environment)
74 is a challenging concept for Earth science educators. Consequently, virtual and remote learning in
75 the geosciences has remained a niche product and it required the COVID-19 pandemic to explore
76 remote learning opportunities for place-based education at scale and across disciplines. We have

77 seen a dramatic influx of efforts (e.g. numerous NAGT Workshops; Earth Educators Rendezvous,
78 2020) and papers since 2020 that detail the creative ways a community, deprived of their traditional
79 educational methods, has responded to distancing constraints and travel bans (e.g., Andrews et al.,
80 2020; Bethune, 2020; Madon, 2020; Rotzein et al., 2020; Sajjadi et al. 2020; Tibaldi et al., 2020,
81 Rotzein et al., 2021; Whitmeyer and Dordevic, 2021).

82
83 Virtual environments, immersive or desktop-based, allow for creating realistic and flexible
84 experiences for (virtual) field trips and the learning activities (e.g., measuring geologic structures
85 and building mental models of spatial orientation and scale of landscape features) that are essential
86 to practice on these field trips. Examining digital twins of outcrops through magnification,
87 collecting samples, or measuring the stratigraphy are, with recent technological advances,
88 straightforward to realize virtually. Over the last four years we and others have been building this
89 capacity through combining efficient data collection in the form of 360° images, high resolution
90 images, virtual outcrop models, and simple measuring tools. What we identified as missing are
91 more complex geological tools and pedagogies for the application and practice of concepts such
92 as strike and dip. Strike and dip measurements and rock identification are the fundamental aspects
93 of any geologic map. Taking and interpreting such field measurements both require physical
94 practice and are fundamentally essential for geoscience education, but are generally not covered
95 extensively in virtual environments (see eRock; Cawood and Bond, 2018 for exceptions). Strike
96 and dip measurements allow students and professionals alike to interpret structures in the Earth's
97 crust and reconstruct deformed regional areas. It is through strike and dip that a geologist
98 understands the regional geology from deposition to deformation.

99 100 ***1.2 Why field geology? Spatial reasoning in the geosciences***

101 Students in the geosciences are frequently required to reason about objects or features that occur
102 at spatial scales too large or small to be directly observed (Gagnier et al., 2017) or hidden from
103 view (Shiple et al., 2013; Ormand et al., 2014; Almquist et al., 2018; Zhao and Klippel, 2019;
104 Atit et al., 2020). As a result, faculty frequently describe students' difficulty with spatial
105 visualization as one of the barriers to success in the geosciences (e.g. Barab and Dede, 2007; Titus
106 and Horsman, 2009; Atit et al., 2020). In particular, spatial visualization is critical to success in
107 courses such as sedimentology and stratigraphy, structural geology, and field techniques (Gagnier

108 et al., 2017). Tectonic and sedimentary processes usually form geo-spatially predictable features,
109 deducible from patterns observed in surface data when one is capable of visualizing the 3D
110 geometry (Alles and Riggs, 2011). Students who possess the spatial visualization abilities
111 necessary to succeed in these courses are also more likely to continue in the geosciences (Titus
112 and Horsman, 2009).

113
114 The development of geological reasoning skills can be scaffolded by introducing students to a
115 sequence of exercises starting with prototypical, accessible, and understandable physical locations,
116 and also by introducing more experiential practice opportunities at the lower-level prior to the
117 more challenging applications found in subsequent upper-level geoscience courses. In many post-
118 secondary institutions, the concept of strike and dip as geological measurements is introduced in
119 an introductory physical geology course. Later, students practice taking these measurements
120 extensively in a field methods course, apply these methods through different lenses of geologic
121 interpretation in subsequent focus courses, and conclude with a capstone summer field course:
122 Field Camp. Visualizing the 3D forms and structures of our planet is a critical skill for the
123 geosciences, and the foundation of this skill lies in a solid understanding of geological maps and
124 strike and dip measurements.

125
126 ***1.3 A place for virtual and immersive technologies in place-based learning***

127 Place-based learning, such as field trips, combines the practices found in problem-based learning
128 and experiential learning to foster a *sense of place* that generates an authentic learning
129 environment, something valued across disciplines from social to physical sciences. Virtual
130 environments, and especially immersive virtual environments, allow for creating learning
131 environments grounded in the same learning theories and pedagogies as place-based education.
132 Associated theories are discussed from different angles such as discovery, inquiry, and problem-
133 based learning as well as experiential learning (Kolb, 2014). The focus of this article is not on
134 learning theories and as such we are not providing an in-depth discussion of the different
135 approaches. Similarities of these approaches are grounded in a constructivist perspective on
136 learning (Winn, 1993; Dalgarno, 2002) building on the power of contextualizing learning through
137 integrating prior knowledge and experience in addition to the context in which the content is
138 embedded. Bangera and Bronwell (2014) found that benefits of these approaches include that they

139 may offer a more effective and accessible starting point for students, including minority, low-
140 income, and first-generation college students and can provide students with a greater ability to use
141 scientific thinking in other aspects of their lives. These approaches, and in particular discovery-
142 based learning, have also been found to be key to successful STEM education (PCAST, 2012).

143
144 What role can virtual and immersive technologies play in discovery-based courses and fostering
145 equity and access to STEM education such as geoscience field trips? The theoretical basis for the
146 transformative nature, especially of immersive technologies for education, is rapidly growing
147 (Dede, 2009; O’Connor and Domingo, 2017; Liu et al., 2020; Parong and Mayer, 2020; Wu et al.,
148 2020). Characteristics of virtual and immersive technologies lend themselves to realize place-
149 based learning (Semken et al., 2018), experiential and embodied learning (Johnson-Glenberg
150 2018) as well as designing environments for discovery-based learning. Placing learners into the
151 real-world with a specific problem that is relevant to a location provides a more direct connection
152 of key learning points that students can understand and use to become more engaged (Powers,
153 2004; Bursztyn et al., 2020). Designing virtual environments in which students' learning activities
154 are scaffolded by exercises and instruction is at the core of discovery-based learning (McComas,
155 2014). Geological processes can sometimes be difficult to visualize during field trips due to vast
156 spatial and time scales— this is one area in the discipline that iVR can offer a distinct advantage.
157 The blending of place-based and discovery-based learning, especially in immersive, virtual
158 environments allows for the “perceptual blending of the real and the virtual world with its place-
159 based authenticity” to enable better learning experiences (Barab and Dede, 2007, p. 2). The
160 geosciences have long been either explicitly or implicitly using experiential, place-based exercises
161 to foster discovery-based learning in their curriculum through, for example, field trips (Semken et
162 al., 2018; Atit et al., 2020). Entering the 2020 Field Camp season, a crucial component of most
163 traditional geoscience programs, instructors and students were faced with limited options: no field
164 camp, limited and socially distanced field camp, or virtual field camp. Here it is pertinent to
165 channel the virtuality momentum into constructive, critical, and empirically-grounded discussions
166 of the future and utility of VR for geoscience education. It is important to note that virtual and
167 immersive virtual experiences cannot only be designed to mimic actual field experience but that
168 they offer opportunities beyond physical reality such as reacting to the learner in real-time (Lopes
169 and Bidarra, 2011; Vandewaetere et al., 2013; Sajjadi et al., 2014; Shute et al., 2016).

170

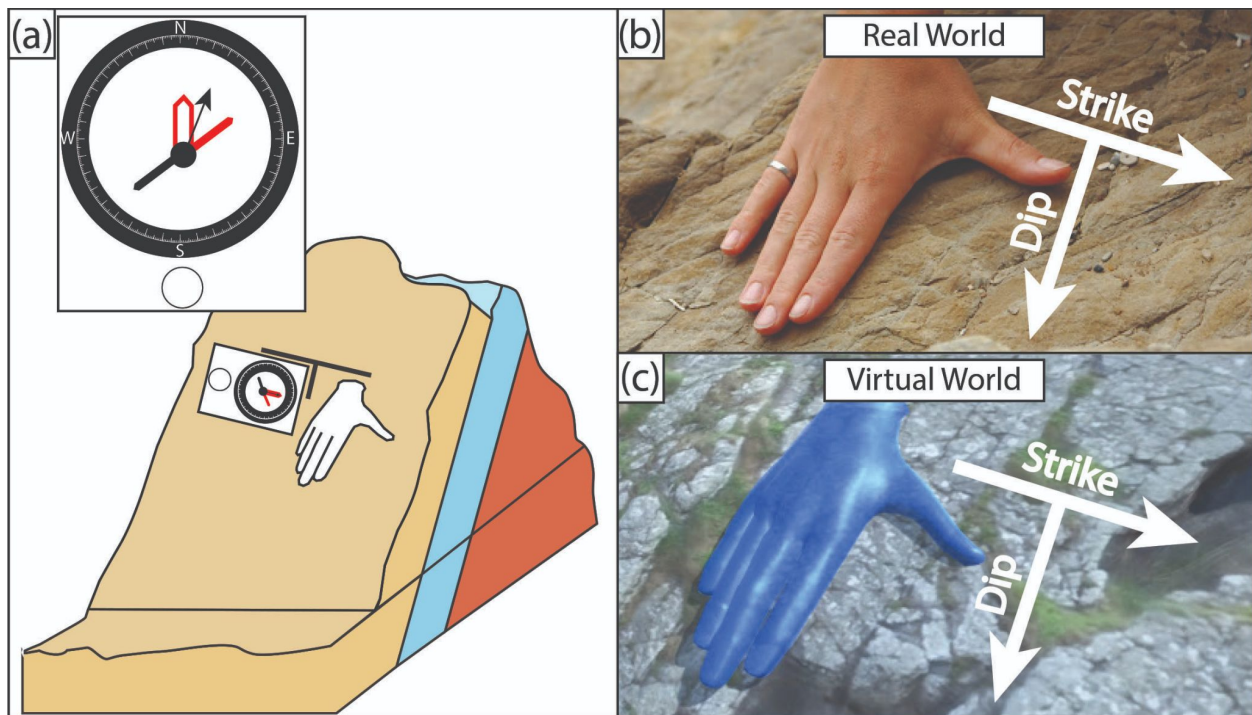
171 This paper presents a virtual Strike and Dip tool (SaD) in a web-based desktop virtual reality (dVR)
172 environment. In addition to posing many challenges, the COVID-19 pandemic induced transition
173 to primarily online teaching also presented geoscience educators with a new opportunity to
174 improve introductory field-mapping instruction to be more inclusive if we are able to recreate
175 strike and dip lab experiences through virtual environments. SaD is an interactive experience
176 created for the purpose of guiding students to think spatially for critical geological applications by
177 taking strike and dip measurements from 3D models of geological structures. The SaD tool mimics
178 an introductory geologic mapping lab where students are taught strike and dip measurements using
179 a set of angled boards with accompanying rock samples staged around a classroom (or open space)
180 to reveal an imagined geologic structure. We have replicated this experience and traditional
181 pedagogies in the virtual world with SaD and its series of digital planes and corresponding virtual
182 rock samples (high resolution 3D digital models downloaded from Sketchfab™). With this tool,
183 students can interactively learn what strike and dip measurements are, practice the basics of field
184 mapping using strike and dip, as well as practice taking measurements using a variety of geological
185 structure types. The SaD tool mimics geoscience place-based learning experiences and combines
186 them with the flexibility and scalability of dVR. A small-scale pilot assessment (eleven
187 participants) using the dVR SaD interface and an accompanying mapping assignment was
188 completed in Fall 2020 and presented at a workshop (Bursztyn et al., 2021). Building on the pilot
189 study we improved the design iteratively and rolled out SaD as a large-scale study in a 250 student
190 introductory geoscience class. We present here a more in-depth discussion of SaD, the newly
191 conducted empirical evaluation and analysis, a critical discussion of results showing important
192 considerations for the future of virtual geosciences, and our vision for future SaD and virtual
193 geoscience toolkit developments.

194 **2 METHODS**

195 ***2.1 The Strike and Dip tool***

196 The representation of 3D geologic structures in 2D form requires several standard map notations,
197 the most important of which are strike and dip measurements. New learners are typically
198 introduced to taking strike and dip measurements using the “right hand rule” (RHR) convention.

199 There are a few variations of the RHR, but a commonly used one (and the one used in this study)
200 is as follows: hold the right hand flat, with the palm down on the planar geologic feature, thumb
201 extended at 90° degrees to fingers, and fingers pointing down dip (Fig. 1). Within the SaD tool,
202 RHR is an optional feature that can be toggled off or on per user preference. Strike and dip is often
203 a challenging concept to teach to new learners of geology in the best of times, but the COVID-19
204 pandemic presented geoscience educators with a new challenge: removing the in-person field trip
205 instruction that provides guided practice in taking strike and dip measurements. Therefore, what
206 were deemed the fundamental components of in-person field instruction for learning to measure
207 geologic structures (identifying strike and dip planes and manipulating a compass to determine
208 their orientation in space), were the primary focus of the SaD tool.

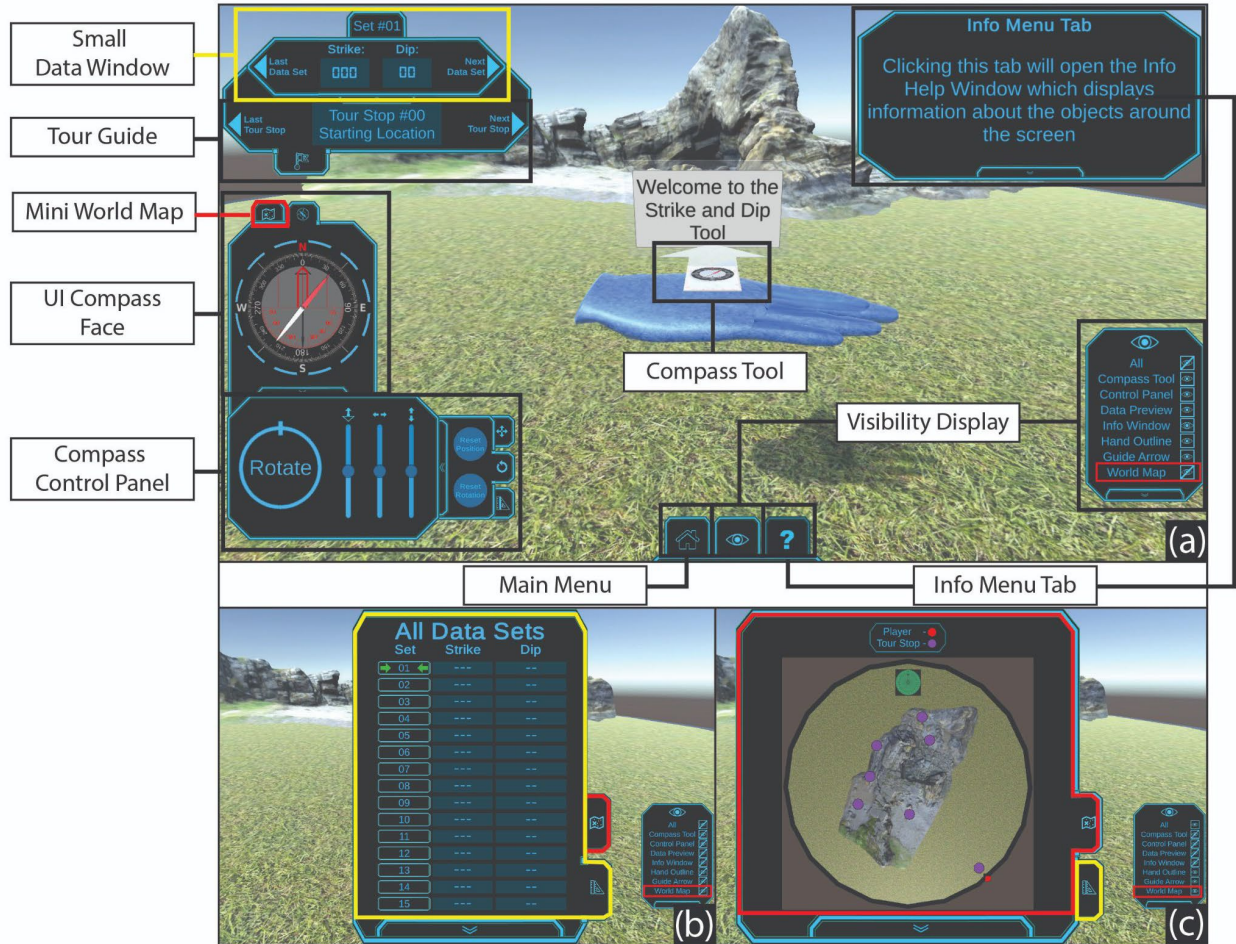


209
210 **Figure 1. A schematic of how one measures strike and dip on an outcrop.** (a) One measures
211 strike and dip on the planar surface of a rock. The strike represents the line at which the planar
212 rock surface intersects with any horizontal plane. The dip angle is the angle between that dipping
213 surface and the horizontal plane. (b) An example of one using RHR in the real world and (c) in the
214 SaD virtual field environment.

215

216 The primary components of the SaD tool are the *Compass Tool* and the *Small Data Panel/Data*
217 *Set* (Fig. 2a; 2b). The strike and dip data are recorded in the tool in the *All Data Sets* panel (Fig.

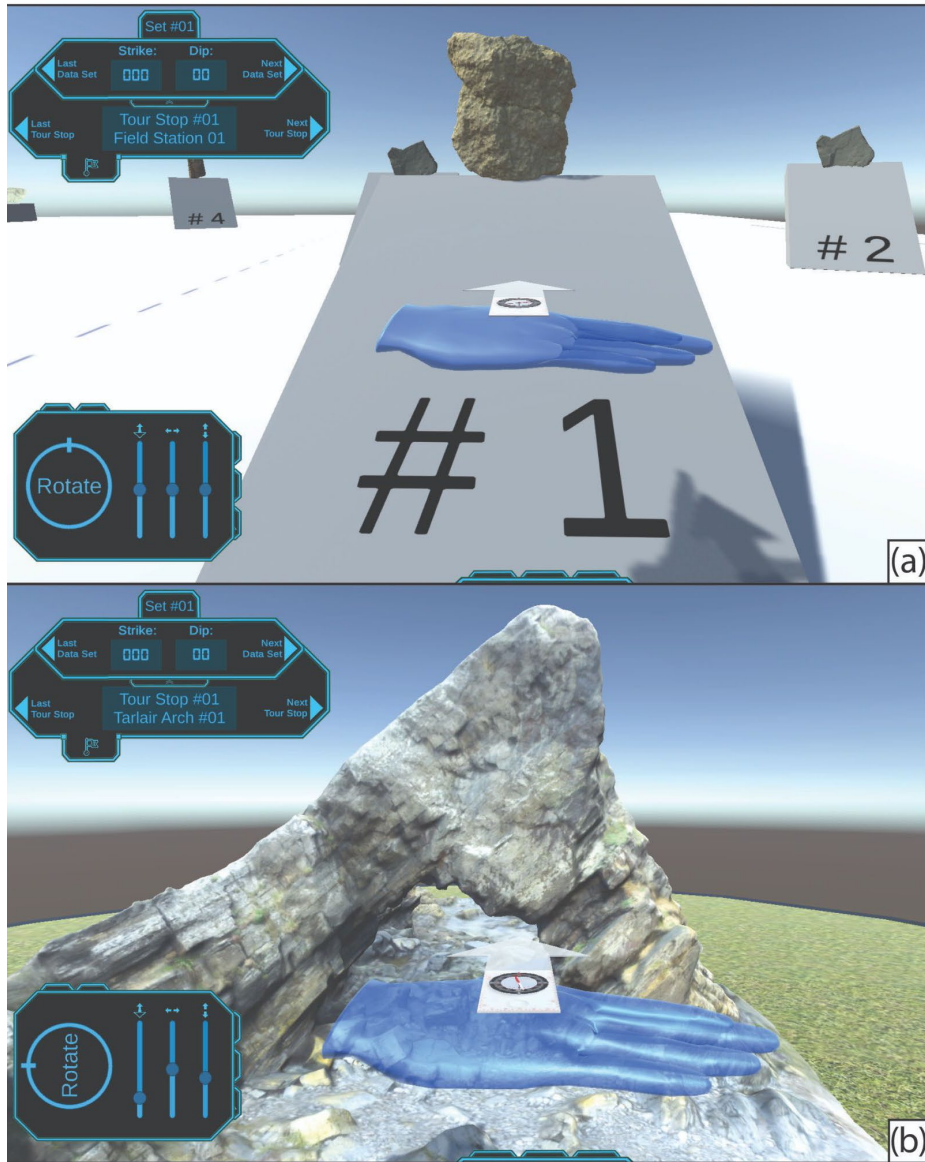
218 2b). Users can navigate around a 3D digital environment to locations where they can measure the
 219 strike and dip of various slopes (platforms or outcrops). The user can locate their position via the
 220 *Mini World Map* or full screen *World Map* (Fig. 2c). Once the user is positioned close to the slope
 221 they would like to measure, they orient the position and rotation of the compass tool (using the
 222 compass control panel) to correspond to the strike or dip measurements. In the virtual environment
 223 levels, “station locations” are specifically laid out to correspond to the assignment maps.



224
 225 **Figure 2. The SaD HUD (Heads Up Display).** The HUD is composed of all the tools visible on
 226 screen throughout the program. Each tool can be toggled on/off depending on user preference. (a)
 227 The main HUD displays the *Small Data Window*, where the user’s most recent strike and dip
 228 measurements are displayed. The *Tour Guide* allows the user to view which stop they are presently
 229 located. The *Mini World Map* (red outline) shows the user their location in a miniature view. The
 230 user may view the compass with more ease using the *UI Compass Face* as they are measuring the
 231 orientation of the rock with the *Compass Tool*. The *Compass Control Panel* is used to position the

232 compass on the outcrop/board to measure orientation. The *Main Menu* display allows the user to
233 adjust the speed at which they/the compass move, the level they are on, and more personalization
234 features. The *Info Menu Tab* gives brief information about each tool when the user hovers over
235 them. Finally, the *Visibility Display* allows the user to toggle on/off each tool. (b) If the user wishes
236 to view their entire strike and dip log, they can click on the triangle protractor icon (yellow outline).
237 (b) The user can also click on the *World Map* (red outline) to view their location in the environment
238 at full screen.

239
240 There are four different setting levels within the SaD tool; from least to most challenging they are:
241 *bumper cubes*, *bumper rocks*, *cubes*, and *rocks*. The two *cubes* levels have field stations set up
242 within the virtual environment as rectangular planes with a virtual hand sample rock floating above
243 (Fig. 3a). The *cubes* levels have very obvious planar surfaces for taking strike and dip
244 measurements. The two *rocks* levels have their field stations set up with rectangular planes draped
245 with rock “skins” that give an appearance closer to an outcrop (Fig. 3b). Depending on the
246 complexity of the rock texture of the “skin”, the planar surfaces within the *rocks* level
247 environments are more challenging to precisely identify. The two *bumper* levels have an algorithm
248 that flags the strike and dip measurements in red if they are greater than 10° and 5° off, respectively.
249 These flags enable self-correction by the students and facilitate only recording correct
250 measurements in the data display panel.



251
 252 **Figure 3. The two main settings: *cubes* and *rocks*.** Each may be used with the *bumper* prefix to
 253 allow the user an error-flagging buffer when measuring platform/outcrop orientation ($\pm 10^\circ$
 254 strike, 5° dip). (a) shows the level *cubes* which replicates the classroom beginner technique of
 255 using a platform to practice taking strike and dip measurements. (b) An example of the *rocks* level,
 256 which features 3D outcrops.

257
 258 **2.2.1 Participants**

259 A total of 147 undergraduate students (with an average age of 19.73) participated in this study.
 260 Out of this population, 98 students self-identified as male, 44 as female, three as other, and two

261 preferred not to answer. All students were recruited from an introductory geoscience class (Geosc
262 001 - Physical Geology) at The Pennsylvania State University in the Fall 2020 semester. This class
263 was chosen for the introductory nature of material taught including the strike and dip content
264 already in the course curriculum. The SaD experience was embedded in this course as a laboratory
265 assignment and students were awarded course credit for their participation. In essence, the
266 laboratory was conducted in a context equivalent to the traditional face-to-face environment.

267

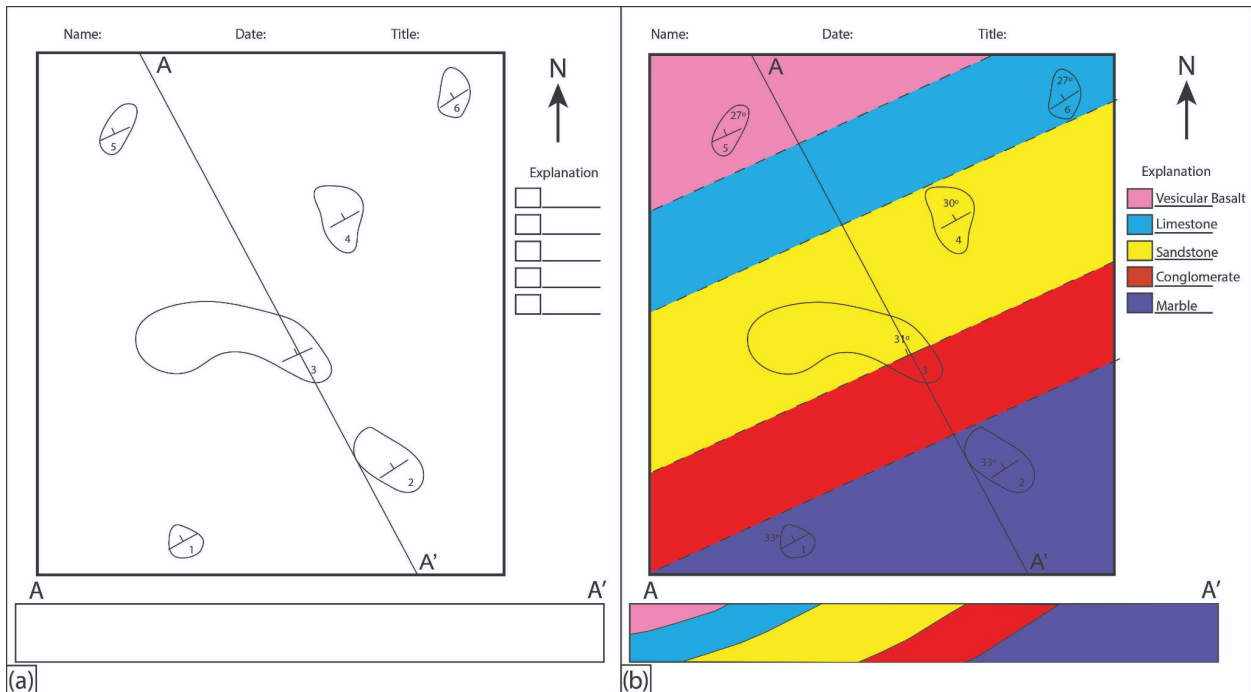
268 2.2.2 Procedure

269 The lab exercise was administered with the help of teaching assistants (TAs). Following the same
270 procedure as the traditional in-person lab that the SaD dVR experience replaced, students were
271 assigned pre-lab homework readings. During the lecture they were presented the standard
272 introductory material on geologic maps and mapping, such as how to interpret the geologic rule of
273 v's, measuring and plotting strike and dip on a map, drawing contacts, and constructing basic
274 cross-sections. Earlier in the semester students completed a geologic mapping exercise from their
275 lab workbooks for which they were provided strike and dip measurements. This lab exercise was
276 graded and returned to the students prior to their introduction to the SaD tool for their virtual field
277 mapping activity. At the beginning of the SaD lab, students were shown an introductory video
278 tutorial demonstrating how to access and utilize the SaD tool through an online dVR environment.
279 Navigation between “field stations” within the environment using arrow keys and/or mouse, proper
280 hand placement for right hand rule, measurement of strike and dip, as well as using the mini map
281 feature are all demonstrated within this tutorial video. TAs provided additional office hours after
282 the lab session and online video resources (which included a longer comprehensive tutorial video
283 and written instructions for the SaD tool as well as a video tutorial on the basics of geologic
284 mapping and drawing a cross-section). Participants in this study used the SaD tool at the beginner
285 (least challenging) *bumper cubes* level.

286

287 As for the traditional in-person lab exercise, the SaD mapping activity was completed in a single
288 3-hour lab session and consisted of two parts, both tasking the students with gathering information
289 (strike and dip, rock descriptions) with which to compile a geologic map, legend, cross section,
290 and interpretation of geologic events that formed the area. Students were given blank base maps

291 and fill-in-the-blank field notes to complete as they worked in the virtual environment. This aspect
 292 of the assignment tasked the students with transcribing the data as they would in the real world
 293 and practice active mapping. Students were also provided with the rock identifications for the map
 294 areas to reduce the number of tasks they had to complete in their single lab session. The first
 295 mapping activity (Map 1) of the assignment was an optional “practice” map with five rock types,
 296 six field stations, and relatively simple geologic relationships to interpret (Fig. 4). The second
 297 mapping activity of the assignment (Map 2) was classified as the “real” map with 15 field stations
 298 and slightly more complex geologic relationships; this is the map that was evaluated for their grade
 299 in this lab assignment. Grading of the lab exercise included evaluation of 1) the map itself, 2) the
 300 field notes, 3) the cross-section, 4) the explanation, and 5) the interpretation of geologic events
 301 that formed the area (Fig. 4).
 302
 303



304 (a)
 305 **Figure 4. The before and after examples of Map 1.** (a) Students are initially given a blank map
 306 with space to fill in the explanation and cross-section. Students are expected to fill in field notes
 307 and interpretation of geologic events on a separate piece of paper. (b) A completed map and
 308 accompanying cross-section.
 309

310 **2.3 Assessment measures and analyses**

311 The participants' experiences and learning with the SaD tool were assessed using self-reported
312 questionnaires (Appendix A). All of the questionnaire items are from established and validated
313 instruments (summarized and adapted by Lee et al., 2010 and Klingenberg, 2020). As part of the
314 demographic information, participants were asked to report on their age, gender, major and minor
315 fields of study, and year of study. Furthermore, participants were asked to report on their
316 familiarity with navigating geographical software such as ArcGIS, as well as their familiarity with
317 playing computer games. Direct student learning of geologic mapping constructs (i.e. via lab
318 grades) was not assessed for this study as the focus was on the experience of the participants using
319 the SaD tool for the purpose of learning the basics of geologic mapping.

320

321 **2.3.1 Quantitative assessment and analyses**

322 After interacting with the SaD tool, the experiences (learning and general) of the participants were
323 measured in light of *representational fidelity, immediacy of control, perceived usefulness,*
324 *perceived ease of use, motivation, control and active learning, reflective thinking, perceived*
325 *learning effectiveness, satisfaction, and self-efficacy* (Table 1, see also Appendix A for the full
326 question list). All constructs were measured on a scale of 1 to 5 and individual items were averaged
327 and collapsed into the final construct score.

328

329 In order to maintain an unbiased distribution into the low/high categories, cases where a participant
330 scored exactly the same as the median (3 for geographical software familiarity, and 4 for gaming
331 familiarity) were excluded. Using this approach 53 participants were identified belonging to the
332 low-Software Familiarity category, 41 to high-Software Familiarity, 47 to low-Gaming
333 Familiarity, and 66 to high-Gaming Familiarity. The experience and learning metrics of
334 participants were compared based on these categories using the independent samples t-test or,
335 alternatively, Mann-Whitney U test in case of non-normal distribution.

336 In addition to geographical software and gaming familiarity, we also explored the effect of gender
337 on the experiences and learning of participants. As such, the experiences and learning metrics of
338 98 male participants were compared with 44 female participants. Two-way ANOVAs were

339 conducted to explore the interaction effect between geographical software/gaming familiarity and
 340 gender on the measured experience and learning metrics reported in Table 1. All analyses were
 341 performed using IBM SPSS Statistics 22.

342

Table 1. Metrics from participant questionnaire and their respective explanations

Metric	Explanation
Representational fidelity ¹	The degree of realism within the virtual environment.
Immediacy of control ¹	The ability to change position/direction and manipulate objects within the virtual environment.
Perceived usefulness ¹	Two metrics for "usability" where 1) usefulness relates to the terms: important, relevant, useful, valuable; and 2) ease of use relates to the terms: convenient, controllable, easy, unburdensome.
Perceived ease of use ¹	
Motivation ¹	Intrinsic interest based on autonomy and competence; within virtual environment derived from user control over what/when is viewed.
Control and active learning ¹	Active involvement in the learning process; learners make their own decisions about the pace, order, and flow of learning activities while completing the task.
Reflective thinking ¹	The generation of curiosity or confusion about what is seen being used as a catalyst for learning new concepts by making sense of observations.
Perceived learning effectiveness ¹	Two metrics for "learning" in the affective domain where 1) perceived effectiveness relates to generation of understanding, meaning, and interest in the topic; and 2) satisfaction relates to gaining knowledge through the virtual environment, including appreciation for the learning experience.
Satisfaction ¹	
Self-efficacy ²	The degree of confidence in understanding of the topics practiced through the virtual experience.

^{1,2}metrics derived from ¹Lee et al., 2002 and ²Klingenberg, 2020

343

344 2.3.2 Qualitative assessment and analyses

345 Within the survey, two open-ended questions were asked from the participants about their
 346 experiences with the SaD tool:

347 1) “How was your learning experience using this tool? Describe how you felt about practicing
348 geologic mapping in a virtual environment.”

349 2) “How did your experience using the strike and dip tool change between the first and second
350 mapping activities? Explain within the context of the technology (ease of use, functionality, etc.)”

351 Combined with the quantitative analyses, qualitative analyses provide deeper insights into how the
352 SaD tool was perceived by the participants. Based on the structured content analysis approach
353 proposed by Schreier (2012), two independent coders examined the responses of participants and
354 inductively generated codes that would capture their content. The coders reached agreement by
355 grouping and rearranging the codes into the final schemas (one for each question) based on the
356 most frequent codes. Inter-rater reliability tests based on Cohen’s Kappa were also conducted for
357 the finalized results. To further understand these results, we examined the associations between
358 geographical software familiarity and gaming familiarity groupings (high/low) and each of the
359 codes using a chi-square test of independence and a post-hoc test with Bonferroni correction
360 (resulting in an adjusted alpha of 0.0125).

361 **3 RESULTS**

362 ***3.1 Quantitative analysis***

363 We first looked at the scores for the different measured metrics (Table 1) averaged over all
364 participants to analyze the overall assessment of the SaD tool. The results summarized in Table 2
365 show slightly above-average scores for the *representational fidelity* and *motivation* metrics, and
366 well-above-average scores for *immediacy of control*, *perceived usefulness*, *perceived ease of use*,
367 *control and active learning*, *reflective thinking*, *perceived learning effectiveness*, *satisfaction*, and
368 *self-efficacy*. These scores indicate a positive overall evaluation of the SaD tool, implying that it
369 succeeded in eliciting a good experience for users, and therefore can be considered an effective
370 learning instrument.

371 As a second step, we were interested in how the experience with the SaD tool was impacted by
372 individual differences between the participants related to past exposure to geographical software
373 and video games. The sampled population reported a slightly above-average score for familiarity
374 with navigating geographical software (M=2.86, SD=1.25), and a well-above average score for

375 familiarity with gaming (M=3.91, SD=1.23). The results from the analyses comparing the survey
 376 scores of participants based on their software and gaming familiarity groupings are reported in
 377 Tables 3 and 4 that follow.

Table 2. 5-point scale survey results

Metric	Mean	S.D.
Representational fidelity	2.96	0.99
Immediacy of control	3.36	1.02
Perceived usefulness	3.25	0.99
Perceived ease of use	3.28	0.8
Motivation	2.95	0.83
Control and active learning	3.33	0.91
Reflective thinking	3.16	0.97
Perceived learning effectiveness	3.11	0.92
Satisfaction	3.12	0.92
Self-efficacy	3.37	0.84

378

379 Our results indicate statistically significant differences (by a combination of independent samples
 380 t-tests and the Mann-Whitney test in the case of non-normal distribution) for almost all the metrics
 381 in the general and learning experiences of students grouped by low and high software familiarity.
 382 For *representational fidelity*, scores of the high software familiarity group were higher than those
 383 in the low software familiarity group (M=3.46, SD=0.95 and M=2.59, SD=0.92, respectively; $t(92) =$
 384 $4.461, p < 0.001$). For *immediacy of control*, scores in the high familiarity group were higher
 385 than in the low familiarity group (M=3.7, SD=0.89 and M=3.21, SD=1.17, respectively; $t(92) =$
 386 $2.188, p = 0.026$). For *perceived usefulness*, scores in the high familiarity group were higher than
 387 in the low familiarity group (M=3.56, SD=1 and M=3.01, SD=1.07, respectively; $t(92) = 2.536, p$
 388 $= 0.013$). For *perceived ease of use*, scores in the high familiarity group were higher than in the
 389 low familiarity group (Mdn = 3.75 and Mdn = 2.75, respectively; $U(N_{low} = 53, N_{high} = 41) = 554.500,$
 390 $z = -3.979, p < 0.001$). For *perceived learning effectiveness*, scores in the high familiarity group
 391 were higher than in the low familiarity group (M=3.45, SD=0.82 and M=2.95, SD=0.95,
 392 respectively; $t(92) = 2.728, p = 0.008$). For *satisfaction*, scores in the high familiarity group were
 393 higher than in the low familiarity group (M=3.4, SD=0.92 and M=2.9, SD=0.97, respectively; $t(92)$
 394 $= 2.570, p = 0.012$). Lastly, scores for *self-efficacy* were greater in the high familiarity group than

395 in the low familiarity group (M=3.64, SD=0.83 and M=3.16, SD=0.89, respectively; $t(91) = 2.651$,
 396 $p = 0.01$). For a complete reporting of these results refer to Table 3.

Table 3. Results of independent samples t-test comparing students grouped by software familiarity

Metric	Software Familiarity	N	Mean	Std. Dev.	P
Representational fidelity	Low	53	2.59	0.92	<0.001**
	High	41	3.46	0.95	
	Total	94	2.97	1.02	
Immediacy of control	Low	53	3.21	1.17	0.026*
	High	41	3.7	0.89	
	Total	94	3.42	1.08	
Perceived usefulness	Low	53	3.01	1.07	0.013*
	High	41	3.56	1	
	Total	94	3.25	1.07	
<i>Perceived ease of use</i>	Low	52	2.98	0.76	<0.001**
	High	41	3.68	0.77	
	Total	93	3.29	0.84	
Motivation	Low	53	2.87	0.88	0.3
	High	41	3	0.89	
	Total	94	2.93	0.88	
Control and active learning	Low	53	3.2	0.97	0.1
	High	41	3.56	0.86	
	Total	94	3.36	0.94	
Reflective thinking	Low	53	3	0.99	0.2
	High	41	3.33	0.84	
	Total	94	3.19	0.94	
Perceived learning effectiveness	Low	53	2.95	0.95	0.008**
	High	41	3.45	0.82	
	Total	94	3.17	0.93	
Satisfaction	Low	53	2.9	0.97	0.012*
	High	41	3.4	0.92	
	Total	94	3.12	0.97	
Self-efficacy	Low	53	3.16	0.89	0.010*
	High	40	3.64	0.83	
	Total	93	3.37	0.89	

* $P < 0.05$; ** $P < 0.001$; italics denote metrics with non-normal distribution for which Mann-Whitney test was also used

397 A similar trend in the results was observed for students grouped by gaming familiarity. Our results
398 indicate statistically significant differences (by a combination of independent samples t-tests and
399 the Mann-Whitney test in the case of non-normal distribution) for almost all the metrics in the
400 general and learning experiences of students grouped by low and high gaming familiarity. For
401 *representational fidelity*, scores of students belonging to the high gaming familiarity group were
402 higher than those in the low gaming familiarity group (Mdn = 3.25 and Mdn = 3, respectively;
403 $U(N_{low} = 47, N_{high} = 66) = 1167.500, z = -2.266, p = 0.023$). For *immediacy of control*, scores of
404 students belonging to the high gaming familiarity group were higher than in the low gaming
405 familiarity group (Mdn = 3.75 and Mdn = 3, respectively; $U(N_{low} = 47, N_{high} = 66) = 959.000, z = -$
406 $3.467, p = 0.001$). For *perceived usefulness*, scores of students belonging to the high gaming
407 familiarity group were higher than in the low gaming familiarity group ($M=3.42, SD=0.74$ and
408 $M=2.96, SD=0.8$, respectively; $t(111) = 2.483, p < 0.05$). For *perceived ease of use*, scores of
409 students belonging to the high gaming familiarity group were higher than the low gaming
410 familiarity group ($M=3.42, SD=0.74$ and $M=2.95, SD=0.8$, respectively; $t(110) = 3.459, p < 0.01$).
411 For *control and active learning*, scores of students belonging to the high gaming familiarity group
412 were higher than the low gaming familiarity group ($M=3.5, SD=0.85$ and $M=3.12, SD=0.9$,
413 respectively; $t(111) = 2.253, p < 0.05$). For *perceived learning effectiveness*, scores of students
414 belonging to the high gaming familiarity group were higher than the low gaming familiarity group
415 (Mdn = 3.43 and Mdn = 3, respectively; $U(N_{low} = 47, N_{high} = 66) = 1147.000, z = -2.357, p = 0.018$).
416 For *satisfaction*, scores of students belonging to the high gaming familiarity group were higher
417 than the low gaming familiarity group (Mdn = 3.42 and Mdn = 3, respectively; $U(N_{low} = 47, N_{high} =$
418 $66) = 1122.000, z = -2.504, p = 0.012$). Lastly, for *self-efficacy*, scores of students belonging to the
419 high gaming familiarity group were higher than the low gaming familiarity group ($M=3.55,$
420 $SD=0.78$ and $M=2.86, SD=0.92$, respectively; $t(110) = 3.296, p < 0.01$). For a complete reporting
421 of these results refer to Table 4.

422 With respect to gender, our results indicate that male students ($M = 3.48, SD = 0.83$) reported
423 significantly higher scores for self-efficacy than female students ($M = 3.12, SD = 0.85$), $t(139) =$
424 $2.329, p < 0.05$). No other significant differences for gender were shown to exist.

425

Table 4. Results of independent samples t-test comparing students grouped by gaming familiarity

Metric	Gaming Familiarity	N	Mean	Std. Dev.	<i>P</i>
<i>Representational fidelity</i>	Low	47	2.69	1.06	0.023*
	High	66	3.13	0.92	
	Total	113	2.95	1	
<i>Immediacy of control</i>	Low	47	3	1.08	0.001**
	High	66	3.7	0.86	
	Total	113	3.4	1	
Perceived usefulness	Low	47	2.96	0.8	0.015*
	High	66	3.42	0.74	
	Total	113	3.23	0.98	
Perceived ease of use	Low	47	2.95	0.8	0.001**
	High	65	3.42	0.74	
	Total	112	3.25	0.8	
Motivation	Low	47	2.77	0.92	0.131
	High	66	3.03	0.81	
	Total	113	2.92	0.86	
Control and active learning	Low	47	3.12	0.9	0.027*
	High	66	3.5	0.85	
	Total	113	3.34	0.89	
Reflective thinking	Low	47	2.93	1.1	0.05
	High	66	3.32	0.9	
	Total	113	3.15	1.01	
<i>Perceived learning effectiveness</i>	Low	47	2.82	0.98	0.018*
	High	66	3.27	0.88	
	Total	113	3.08	0.95	
<i>Satisfaction</i>	Low	47	2.86	0.92	0.012*
	High	66	3.28	0.89	
	Total	113	3.1	0.92	
Self-efficacy	Low	47	3.01	0.91	0.001**
	High	65	3.55	0.78	
	Total	112	3.32	0.88	

* $P < 0.05$; ** $P < 0.001$; italics denote metrics with non-normal distribution for which Mann-Whitney test was also used

427 Finally, we were interested in investigating the possible interactions between geographical
428 software/gaming familiarity and gender on the experience and learning metrics of participants.
429 Two-way analyses of variance (ANOVAs) were conducted for this inquiry (see Table 5 for
430 complete results) and revealed no statistically significant results.

431

432 Our results indicate that the individual differences among students in light of their prior familiarity
433 with navigating geographical software as well as their familiarity with gaming has a pronounced
434 effect on their experiences. The unveiled trend indicates that higher familiarity with either
435 geographical software or gaming leads to a significantly better experience with the SaD
436 tool. Importantly, no effects of gender or significant interactions between software/game
437 familiarity and gender on the experience and learning metrics of participants were observed.

Table 5. Results of ANOVA examining interaction effects of geographical software familiarity and gender and gaming familiarity and gender

Metric	Software familiarity								Interaction effect			Gaming familiarity								Interaction effect		
	High (n = 41)				Low (n = 53)				software/gender			High (n = 66)				Low (n = 47)				gaming/gender		
	Male (n=15)		Female (n=13)		Male (n=35)		Female (n=18)		F (3,137)	p	η_p^2	Male (n=54)		Female (n=9)		Male (n=22)		Female (n=24)		F (4,136)	p	η_p^2
M	SD	M	SD	M	SD	M	SD				M	SD	M	SD	M	SD	M	SD				
Representational fidelity	3.46	0.82	3.30	1.23	2.55	0.91	2.66	0.95	0.54	0.66	0.01	3.05	0.91	3.27	0.97	2.79	0.86	2.54	1.20	0.77	0.55	0.02
Immediacy of control	3.83	0.73	3.48	1.20	3.21	1.08	3.22	1.36	0.28	0.84	0.01	3.65	0.86	3.94	0.97	3.05	0.85	2.94	1.20	0.30	0.88	0.01
Perceived usefulness	3.69	0.81	3.21	1.33	2.97	1.00	3.09	1.20	0.75	0.53	0.02	3.34	0.87	3.69	0.70	3.09	0.81	2.80	1.30	0.75	0.56	0.02
Perceived ease of use	3.66	0.78	3.69	0.82	3.09	0.66	2.79	0.91	0.84	0.47	0.02	3.41	0.73	3.69	0.81	3.10	0.76	2.80	0.84	0.80	0.53	0.02
Motivation	3.11	0.90	2.81	0.99	2.85	0.83	2.93	0.98	0.79	0.79	0.01	3.00	0.83	3.06	0.63	2.84	0.88	2.68	0.98	0.26	0.90	0.01
Control and active learning	3.65	0.79	3.50	1.00	3.28	0.90	3.04	1.12	0.14	0.93	0.00	3.48	0.84	3.69	0.86	3.28	0.69	2.92	1.00	1.14	0.34	0.03
Reflective thinking	3.50	0.76	2.98	0.99	3.05	0.91	3.15	1.16	1.00	0.39	0.02	3.29	0.93	3.50	0.91	3.15	0.81	2.67	1.20	0.85	0.49	0.02
Perceived learning effectiveness	3.63	0.71	3.23	1.00	2.92	0.90	3.02	1.07	0.58	0.63	0.01	3.22	0.89	3.68	0.80	3.05	0.87	2.57	1.00	1.80	0.13	0.05
Satisfaction	3.57	0.81	3.13	1.13	2.88	0.86	2.92	1.17	0.46	0.71	0.01	3.23	0.91	3.50	0.83	3.16	0.75	2.58	1.00	1.87	0.12	0.05
Self-efficacy	3.84	0.82	3.28	0.84	3.25	0.85	3.01	0.96	0.32	0.81	0.01	3.52	0.81	3.74	0.69	3.33	0.92	2.72	0.83	1.26	0.29	0.04

total n for high and low software and gaming familiarity includes unspecified gender

440 3.2 *Qualitative analysis*

441 The results from our qualitative analysis of the two open-ended survey questions are reported in
442 Tables 6 and 7 that follow. With respect to the first open-ended question, “How was your learning
443 experience using this tool?”, almost 18% of participants reported that the tool was easy to use
444 while nearly 17% reported that the tool was difficult to use (Table 6). For example, two contrasting
445 participant comments are: “it was easy to navigate” and “I felt confused and overwhelmed on the
446 program almost the entire time I was using it...”. Related to useability, almost 11% of participants
447 indicated that the controls for using the tool are not intuitive, e.g.: “it was very frustrating to try
448 and rotate the compass to the right spot...”. Another 8% indicated that the tool had a high and
449 steep learning curve, e.g.: “firstly, I thought it is hard but then I got used to it”. Furthermore, about
450 12.5% of participants had performance issues such as lagging and crashing, e.g.: “it was a little
451 slow, as it did not respond immediately to my inputs...”.

452 Importantly, a little over 15% of participants reported that the tool has increased their interest in
453 learning the topic and 22% reported that they perceived the tool as effective for learning, while
454 only 6% reported that they did not perceive the tool to be effective for learning. For example, two
455 contrasting participant comments about the experience are: “...I felt like I was doing actual
456 work...” and “...I think that an in-person experience would be more effective to understand strike
457 and dip...”. Related to the latter example, 11% of participants indicated that they would prefer the
458 real environment to the virtual for learning about this topic.

459 Finally, our results show that 49% of the sampled population had an overall positive impression
460 of the tool whereas only 17% and 13% reported an overall negative or overall mixed impression,
461 respectively. Others did not express clear inclination.

462 A chi-square test of independence revealed that participants with low geographical software
463 familiarity had a much higher overall negative impression (29.5 %) compared to those with a high
464 geographical software familiarity (2.85 %), $\chi^2(1, N = 79) = 9.52, p < 0.01$. The post-hoc test with
465 Bonferroni correction was in agreement that negative impressions are significantly more common
466 for participants in the low geographical software familiarity category ($p < 0.01$). No other
467 significant differences between the geographical software familiarity categories or game
468 familiarity categories were observed.

Table 6. Qualitative analysis results for open-ended learning experience question

Code	Example	Cohen's Kappa	% of total	% of H-SF	% of L-SF	% of H-GF	% of L-GF
Easy to use	"...It was fun and easy to use all around..."	0.838	17.85	22.85	11.36	21.05	12.82
Difficult to use	"I felt confused and overwhelmed on the program almost the entire time I was using it"	0.968	16.96	5.70	27.27	14.03	17.94
Controls not intuitive	"...I struggled with getting everything in place each time and it was hard to fix my dip if I got that wrong..."	0.799	10.71	8.57	13.63	10.52	10.25
Performance issues	"...It was a little slow, as it did not respond immediately to my inputs..."	1.000	12.50	11.42	20.45	17.54	5.12
Caused high levels of interest	"...I feel like I actually understand what a strike and dip measurement is..."	0.184	15.17	17.14	9.00	15.78	7.69
Steep learning curve	"Initially I didn't know how to use it, so it was frustrating, but I looked at the short tutorial and it made it a lot easier"	0.936	8.03	5.71	6.81	8.77	5.12
Prefer the real environment	"...I prefer doing things in real life, than virtual..."	0.874	11.60	11.42	15.90	15.78	10.25
Perceived as effective for learning	"...I felt like I was doing actual work..."	0.858	22.32	34.28	15.90	26.30	20.51
Perceived as not effective for learning	"...was not effective in learning because I was only able to see one strike dip at a time and could not figure out how they related to each other spacially..."	0.918	6.25	2.85	6.81	0.00	10.25
Overall impression positive	"I feel like this helped me visualize and understand strike and dip and geologic mapping much better than before"	0.911	49.10	57.14	43.18	50.87	51.28
Overall impression negative	"...I don't like this class to begin with, and I this activity did not make me like this class any more than before..."	0.934	16.96	2.85	29.50	14.03	20.51
Overall impression mixed	"It was cool but frustrating"	0.769	13.39	14.30	15.90	17.50	5.12

H-SF=high software familiarity, L-SF=low software familiarity, H-GF=high gaming familiarity, L-GF=low gaming familiarity

470 With respect to the second question, “How did your experience ... change between ... mapping
471 activities?”, 62.5% of participants reported that their experience improved from the first to the
472 second mapping activity (Table 7). More than half of those who reported an improvement to their
473 experience explicitly mentioned that their experience was easier in the second mapping activity
474 because of practicing in the first mapping activity. About 20% of participants reported that their
475 experience remained the same, and 18% reported that their experience worsened from the first to
476 the second mapping activity. From those who reported that their experience worsened, 12.4%
477 stated that the second mapping activity was more difficult and almost 8% stated that they
478 experienced more lag in the second mapping activity. A chi-square test of independence revealed
479 no significant differences between geographical software familiarity categories or gaming
480 familiarity categories and the codes. In summary, the qualitative analysis of the second question
481 indicates that more exposure to the SaD tool improves the overall experience for users but the fact
482 that second activity is more demanding in terms of required graphic power resulted in more
483 performance issues.

484

Table 7. Qualitative analysis results for open-ended experience change across activities question

Code	Example	Cohen's Kappa	% of total	% of H-SF	% of L-SF	% of H-GF	% of L-GF
Improved	"...In the second activity I was more used to it and it was easier to take the measurements..."	0.911	62.5	68.1	64.2	66.6	66.6
	"It became easier for me to use the strike and dip technique"						
Worsened	"...The second map was harder because some of them didn't have strike or dip"	0.845	18	13.6	10.7	13.8	14.8
	"...The second map ended up lagging and ran slower and slower the longer I used it"						
Same	"...I noticed no major changes between mapping activities..."	0.916	20.83	18.1	28.5	19.4	22.2
	"It was a poor experience both times..."						

486 4 DISCUSSION AND CONCLUSIONS

487 Using the SaD tool, an entirely remote introductory field mapping exercise was successfully
488 completed by students during the COVID-19 pandemic. This field mapping exercise replicated
489 exactly, in the digital world, the tasks the students would have normally completed in an in-person
490 lab: measuring strike and dip of staged “outcrops”, using those data to assemble a map, and
491 interpreting the geologic history for that “region”. Using traditional aspects in a new way, this
492 environment not only taught students how to visualize the orientation of strike and dip on a rock
493 plane, but also how to correctly line up a compass using the RHR convention. It also challenged
494 students to conceptualize and infer overall geologic relationships using the measurements they
495 took at each individual 3D outcrop model. From a teaching perspective, the SaD tool also provides
496 three distinct advantages: 1) the time required to set up a staged beginner mapping area is
497 conserved, which in turn permits 2) multiple mapping environments to be explored by the students
498 (e.g. “practice” Map 1 followed by “real” Map 2) with different levels of challenge (e.g. *bumper*
499 *cubes vs bumper rocks*) available to facilitate individualized learning; as well as 3) the *bumper*
500 setting flagging incorrect measurements, providing the opportunity for self-correction. In a regular
501 face-to-face introductory mapping lab, there is realistically only time to set up one staged mapping
502 environment and during the exercise, the instructor is trying to assist individual students with a
503 wide range of issues from using their left hand, to holding the compass upside down, to having
504 made and mapped several incorrect measurements without realizing their error. SaD dramatically
505 increased efficient instruction through error flagging alone.

506

507 Both quantitative and qualitative results suggest that the students reacted overall positively to the
508 SaD tool. Further, qualitative results suggest that SaD was an effective learning instrument for the
509 mapping exercise, as participants reported an increase in understanding of strike and dip from Map
510 1 to Map 2. These findings are in agreement with those from the earlier pilot study (n=11) using
511 the same software (Bursztyn et al., 2021) and suggest that SaD can be considered an effective
512 learning instrument. The quantitative results indicate that students familiar with other geographical
513 software or gaming software had a much better experience in light of *representational fidelity*,
514 *immediacy of control*, *perceived usefulness*, *perceived ease of use*, *control and active learning*,

515 *perceived learning effectiveness, satisfaction, and self-efficacy* compared to those who were
516 unfamiliar. This is important as it suggests that introducing students to virtual learning
517 environments more frequently will have positive effects on their learning experience.

518
519 Although the results of the qualitative analysis are valuable on their own, when considering the
520 prior individual experiences of users in relation to their open-ended feedback, interesting themes
521 emerge. When comparing participants in the high geographical software familiarity group to those
522 in the low geographical software familiarity group, we see that those in the high familiarity group
523 perceived the tool to be much easier to use and controls to be more intuitive. Similarly, participants
524 in the high familiarity group experienced less performance issues and had a less steep learning
525 curve. It was also the case that participants in this group had a lower tendency to claim preference
526 for the real environment over the virtual one and these participants determined the tool to be
527 effective for learning at much higher rates than those in the low familiarity group. The high
528 geographical software familiarity grouping reported a much higher overall positive impression and
529 much lower overall negative impression of the tool. Finally, a very similar trend is seen when
530 comparing participants of high and low gaming familiarity. Apart from performance issues and
531 learning curve, in almost all the other metrics, participants in the high gaming familiarity group
532 reported a much better experience than those in the low gaming familiarity group. The qualitative
533 results align with the quantitative results, which further strengthens the conclusion that students
534 with higher geographical software familiarity and to some degree, gaming familiarity, gained more
535 cognitively and psychologically from their SaD experience. Our results corroborate observations
536 made in other experiments evaluating the importance and impact of prior familiarity with similar
537 software on the experiences and performance of learners in virtual environments (Bagher et al.,
538 under review). Importantly, the absence of effects of gender on the participants' experience and
539 learning metrics suggests an equitable learning experience across gender demographics.

540
541 To further explore some of the feedback received through the open-ended questions, we address
542 comments geared towards issues with usability, fidelity to real world environments, and limitations
543 with software.

544 **4.1 Usability and fidelity to learning mapping in the real world**

545 Notably, most of the negative comments with the SaD tool are with regard to lag and frustration
546 of becoming familiar with the settings and controls (Tables 5 and 6) and not the sometimes
547 confusing aspect of taking and interpreting strike and dip measurements. Within this lab, the 3D
548 virtual outcrops presented had easy to determine strike planes. Because the RHR convention was
549 represented with a digital right hand that could be manipulated, users could easily determine dip
550 direction and therefore angle. Furthermore, because participants were using the tool with the
551 beginner *bumper* setting, they were alerted to any incorrect measurements instantaneously.

552 In the field without a perfectly staged 3D outcrop, it is sometimes difficult to determine the true
553 strike of a lithologic unit, and therefore easy to accidentally measure an apparent plane instead of
554 a true one. Although no “lag” time is associated with field mapping (except perhaps prolonged
555 snack breaks), good, easily determinable strike and dip outcrops are not always abundant. This
556 forces introductory students to learn and practice strike and dip on outcrops that are overly
557 complicated for new learners. For example, Appalachian State students must travel one to two
558 hours each way to the Valley and Ridge Province where they learn how to map in “sedimentary”
559 units that are, in reality, slightly metamorphosed meta-sedimentary rocks, and sometimes have
560 slight foliation or crystallization. Furthermore, the region is heavily deformed with outcrop-
561 regional sized folds and faults. Finding appropriate outcrops for introductory students is difficult
562 and those that are found are on steep terrain and therefore not wholly accessible.

563 Interestingly enough, most of the comments made about the reactiveness of the controls are
564 variations of comments heard as an instructor from students in the field. For example, “Initially I
565 didn't know how to use it, so it was frustrating...” and “...At first it was a bit overwhelming, but
566 with some instruction it became much easier and quicker to use...”. This is a common comment
567 from students at the end of the semester in a field methods course. Another comment, “It was very
568 frustrating to try and rotate the compass to the right spot...” or “...I struggled with getting
569 everything in place each time...”, is a staple in regards to placing the compass when students first
570 get into the field. The comment regarding only seeing one strike and dip measurement at a time
571 (“...was not effective in learning because I was only able to see one strike dip at a time and could
572 not figure out how they related to each other spatially...”) is also not an uncommon struggle in the

573 field. Most places do not have kilometer long outcrops in which to visualize the structures of the
574 whole area. One must actively map each individual strike and dip measurement one at a time, only
575 interpreting the structures once there are enough points across the map to put together the geologic
576 story. Similarly, SaD users may also view their “map” with the *World Map* feature (Fig. 2c) and
577 visualize the region in its entirety. Lastly, the comment “I felt confused and overwhelmed on the
578 program almost the entire time I was using it” is so common in the field that many instructors
579 address this as a known occurrence and the statement is frequently countered with some version
580 of ‘You may be lost the majority of the time, the key is to recognize when you are “found” and to
581 fill in the gaps.’

582 Despite the participants in this study having never actually mapped geology before, let alone in a
583 real-world environment, there were several confident comments that *in person experience would*
584 *be more effective for learning and alleviating confusion* than the SaD tool. These comments are
585 difficult to address with their “the grass is always greener” perspective. This type of perspective
586 was seen in a study by Stumpf et al., (2008), who found that students exposed to an in-person only
587 field trip claimed preference for the virtual version while students in the virtual field trip group
588 decreed the opposite.

589
590 The thought that nothing can compare with a real-world field trip is predominant among some
591 geologists but it is one that is exclusive and unimaginative. With the development of realistic
592 virtual desktop environments, iVR experiences, and platforms like Sketchfab and Open
593 Topography, along with public access to texture and material designers like Substance by Adobe,
594 it is becoming more possible and pertinent to develop virtual environments that mimic real world
595 structures, and therefore their value for replicating place- or discovery-based learning (e.g.
596 O’Connor and Domingo, 2017; Atit et al., 2020; Nesbit et al., 2020; Parong and Mayer, 2020;
597 Riquelme et al., 2020; Wu et al., 2020). With iVR, users can even navigate through and interact
598 with virtual environments in a very realistic way, which we suggest is also valuable in discovery-
599 based learning (e.g. Liu et al., 2020; Parong and Mayer, 2020; Wu et al., 2020; Métois et al., 2021).

600 The results of this study point to a mix of positive evaluation and room for improvements of the
601 SaD tool. Considering that SaD is still evolving, it is expected to receive comments related to
602 usability issues from the participants. Such comments can help us better identify the shortcomings

603 of this tool and plan for future improvements. It is important to emphasize that our results also
604 indicate that a high number of participants perceived the tool as useful for their learning and the
605 overall impression of the tool is positive.

606

607 ***4.2 Limitations and future work***

608 *4.2.1. Procedural limitations*

609 For this study, SaD was used in a single lab session following an earlier workbook-style mapping
610 exercise. Although all students were assigned the earlier mapping exercise, only those who
611 completed it had it returned and available for their reference during the subsequent SaD lab
612 activity. Furthermore, it is unclear how many students, if any, referred back to this exercise for
613 reminders or guidance during the SaD exercise. Because the SaD lab was administered during a
614 single lab session, Map 1 (the practice map) was made optional to alleviate the pressure of potential
615 time constraints. Consequently, not all students completed the practice map prior to the main
616 assignment (Map 2). The small-scale pilot study (Bursztyn et al., 2021) built in two work and
617 submission sessions to the exercise with instructor feedback following the first “practice” mapping
618 activity. We were unable to follow this procedure due to curriculum scheduling complications for
619 the present study and this limitation resulted in students either opting to not complete the first
620 mapping exercise, or completing both with the pressure of time-constraints.

621

622 Within the dVR experience itself, participants were limited to using the SaD tool restricted to the
623 beginner *bumper cubes* setting. With *bumper cubes* incorrect measurements are flagged, but
624 students do not know why they are wrong or how to correct themselves. It will be important to
625 develop the SaD tool to include adaptive interventions such as individualized embedded hints and
626 mapping guidance that would facilitate the learning experience of beginner mappers using the
627 *bumper* settings. In VR environments it is feasible to implement adaptive learning strategies, such
628 as adaptive interventions, hints, and feedback (Peirce and Wade, 2010; Zaharias et al., 2012), in
629 addition to more dynamic strategies in the form of difficulty and learning content adjustments
630 within the learning experience (Hocine et al., 2015; Streicher and Smeddinck, 2016). Such
631 strategies can support personalized experiences for learners exhibiting different levels of abilities

632 and competencies in relation to the learning experience. In the case of our *bumper* settings,
633 adaptive interventions might provide feedback on the nature of the error the user has made. It will
634 also be important to study the effect of including such adaptive interventions into the learning
635 environment, both on student learning and on user experience.

636

637 Finally, from an individual differences perspective, the exclusion of ethnicity from the participant
638 questionnaire survey was an oversight not realized until too late in the procedure to be corrected.
639 Critically examining individual differences in the context of the learning experience will continue
640 to be of utmost importance moving forward. Furthermore, in this study, we did not collect the
641 scores from the student work. Individual differences are not only important to consider for the
642 useability of the instrument, but also critical to examine the effect the tool has on student learning.

643

644 *4.2.2 Technical limitations*

645 Several students experienced technical difficulties including their computers crashing, the SaD
646 tool lagging, and difficulty maneuvering within the virtual environment. Between the pilot study
647 (Bursztyn et al., 2021) and this study we tried to address the lagging concern, knowing that many
648 students would not have access to gaming computers with high-powered video cards. Visual lag
649 can be reduced by minimizing the complexity of the 3D rock models through reducing the number
650 of polygons for each 3D model. However, the tradeoff in this regard is that the 3D models with
651 reduced polygons will at some point become no longer recognizable as particular rock types. We
652 have since been exploring other avenues such as applying detailed texture maps over simplified
653 geometries. Through the use of programs such as Adobe Substance highly detailed textures can be
654 created that give the appearance of complex 3D geometry, many of which are digital twins for
655 diagnostic rock textures. These textures can then be applied to 3D models with simple geometry
656 (such as cubes) while retaining the visual appearance and detail of highly complex 3D models but
657 without creating lag.

658 **5 OUTLOOK: ADVANCING INCLUSIVITY, ACCESSIBILITY, AND REALISM**

659 Beyond the students' technical difficulties, we also recognize that the interaction fidelity of
660 maneuvering in a two-dimensional dVR environment representing a complex 3D natural

661 environment is limited. Navigation within such an environment is complex and requires training
662 (key combinations, mouse and/or trackpad to maneuver and position the compass vs walking up
663 to a surface and using one's hands). On the other hand, it has been shown that virtual environments,
664 especially developed for web-based distribution and mobile devices, can remove barriers to
665 accessibility and create a culture of inclusion in geoscience classrooms (O'Sullivan and Kearney,
666 2018; Chenrai and Jitmahantakul, 2019). In recent years, field experiences have been critically
667 looked at from different perspectives. To name some of the prominent challenges: field trips pose
668 troubling accessibility issues excluding students with disabilities but also students who cannot
669 afford to participate due to time or financial constraints. Field work is further challenged by an
670 increasing awareness of harassment that is happening in the field, which is often targeting women
671 and minority students and faculty who do not conform to the stereotypical mainstream conceptions
672 of fieldwork, that is, it is a white, male-dominated domain. Marín-Spiotta et al. (2020) call out this
673 issue, comparing it to the Vegas Rule, criticizing the understanding that “what happens in the field,
674 stays in the field”. For the diversity of students who self-select out of geoscience programs to avoid
675 the physical and/or emotional burden of required field mapping experiences, the promise of virtual
676 mapping with digital twin environments such as provided by SaD may provide a solution that
677 facilitates their access, safety, and also retention.

678

679 In light of the new openness to virtual experiences, it is essential to critically look at the
680 opportunities (i.e., breaking down long standing barriers of accessibility and inclusion) and
681 challenges that remote learning offers to Earth educators. To establish remote learning
682 opportunities as alternative pathways in geoscience education, we need tools as well as empirical
683 studies that critically examine the opportunities, the challenges, and the feasibility of virtual
684 learning experiences. Many studies remain anecdotal (e.g., Marshall et al., under review) but it is
685 time to establish research frameworks and to connect place-based education with established
686 assessments and practices in virtual and immersive learning (Klippel et al. 2020; Petersen et al.
687 2020). Immersive virtual reality (iVR) is inherently a three-dimensional (3D), spatial medium
688 (Maceachren and Brewer, 2004) and therefore offers a natural interface to all representations of
689 data that, too, are three-dimensional in nature.

690

691 The COVID-19 pandemic has imposed an increased need for remote and online education. The
692 infrastructure, however, to equip every student with a headset to experience iVR was not in place,
693 nor would it have been feasible with rapid implementation of massive remote learning and abiding
694 by physical distancing restrictions. Though our research goals are ultimately to address the
695 advancement of the science of immersive experiences, there are still technological constraints
696 which we addressed by seizing the opportunity to conduct an exploratory study with a web-based
697 desktop virtual environment. We believe that with immersive VR technology becoming widely
698 accessible, we can achieve both: accessibility and natural interactivity. Immersive VR offers 3D-
699 in-3D interfaces which are ideal for representing the 3D data of geological structures as well as
700 realizing the 3D interactions of measuring them (e.g., positioning a compass on a planar surface).
701 The iVR interface of SaD has been developed this spring and we intend to leverage this version of
702 the tool to evaluate place-based learning and 3D interactions within that environment in the coming
703 fall semester.

704
705 Research on virtual learning environments has shown that the immersive, interactive, and 3D
706 nature of iVR can potentially reduce the performance gap between students with high and low
707 spatial abilities (Simpson et al., 2017; Lages and Bowman, 2018) which have been shown critical
708 for STEM education (Newcombe, 2010). Immersive 3D visualizations can demonstrate the extent
709 of landscapes and geological features in a form that is beneficial for students to develop spatial
710 thinking, since they closely mirror everyday perceptual experience (Simpson, 2020). This
711 mirroring capability is important in the context of the current study where students expressed
712 preference for a more real-world experience. In addition, current iVR technology allows for the
713 integration of high-fidelity perceptual information (e.g., position, orientation, shape, size, or
714 motion) and additional abstract information (e.g., video, graphs, and text) into a single virtual
715 environment, which would enable the teaching of complex geological concepts through
716 understandable visual demonstrations (Bowman et al., 2003). Such explicit graphical presentations
717 might act as a “cognitive prosthetic” for students with lower spatial ability (Mayer and Sims, 1994;
718 Höffler and Leutner, 2011; Jamieson et al., 2014; Zhao et al., 2020); that is, low ability learners
719 could gain a particular benefit from accessing an information-rich iVR environment as they have
720 difficulty mentally constructing their own representation when learning about geological features
721 and processes from the textbook or a traditional field trip alone. This is also important for the

722 present study as it has been shown that low spatial ability learners can benefit more from a desktop
723 VR application in comparison with high spatial ability learners (Lee et al., 2009). Future empirical
724 evaluations of SaD comparing immersive versus non-immersive instances will include a stronger
725 focus on spatial abilities.

726

727 The SaD tool continues to be developed and evolve with each iteration into becoming a more
728 realistic digital twin for teaching field geology technique. The next steps for this tool are mapped
729 out, focused on creating 3D models that mirror real world lithologic features (including, but not
730 limited to, individual sand grains, identifiable fossils, foliation and crystalline textures). As a
731 community, we are ever closer to creating complete, realistic virtual environments for an inclusive
732 and accessible geology field class with world class “outcrops” that mimic those one sees in the
733 classic geology field camps and trips hosted in the Western United States.

734

735 **CODE/DATA AVAILABILITY**

736 The SaD tool is located at <https://sites.psu.edu/virtualfieldtrips/strike-and-dip/>. Additionally, a
737 developer log is located at <https://sites.psu.edu/bartonmasters/sad-strike-and-dip-links/> and is
738 maintained by Bart Masters.

739 **AUTHOR CONTRIBUTIONS**

740 NB, HR and AK conceived the experiment in collaboration with PS and BM who developed and
741 programmed the SaD software. NB created the lab exercise. Analyses were conducted by PS, JH,
742 JZ and JOW in collaboration with AK. Manuscript was written by NB, PS, HR, AK, and JH. HR
743 created the figures. All authors discussed the results and manuscript narrative, and all contributed
744 substantially to the editing process.

745 **COMPETING INTERESTS**

746 The authors declare that they have no conflict of interest.

747 **SPECIAL ISSUE STATEMENT**

748 This article is part of the special issue “Virtual geoscience education resources” in Geoscience
749 Communication Letters.

750 **ACKNOWLEDGMENTS**

751 The authors would like to thank Dr. Peter La Femina of Penn State University and his teaching
752 assistants for performing this lab in his Geosc 001 - Physical Geology laboratory. We would also
753 like to thank Dr. Mahda Bagher for creating the survey used to gather the data used in this paper.

754 **REFERENCES**

755 Alles, M., and Riggs, E. M.: Developing a process model for visual penetrative ability, in Feig,
756 A.D., and Stokes, A., (Eds.), *Qualitative Inquiry in Geoscience Education Research*,
757 Geological Society of America Special Paper 474, p. 63–80, 2011.

758 Almquist, H., Stanley, G., Blank, L., Hendrix, M., Rosenblatt, M., Hanfling, S., and Crews, J.: An
759 Integrated Field-Based Approach to Building Teachers' Geoscience Skills, *Journal of*
760 *Geoscience Education*, Vol. 59, No. 1, pp. 31–40, 2018.

761 Andrews, G. D., Labishak, G., Brown, S., Isom, S. L., Pettus, H. D., and Byers, T.: Teaching with
762 Digital 3D Models of Minerals and Rocks. *GSA Today*, Vol. 30, pp. 42-43, 2020.

763 Atit, K., Uttal, D. H., and Stieff, M., Situating space: using a discipline-focused lens to examine
764 spatial thinking skills, *Cognitive research: principles and implications*, Vol. 5, No. 1, p. 19,
765 2020.

766 Bagher, M., Sajjadi, P., Wallgrün, J.O., La Femina, P.C., Klippel, A., under review. Move The
767 Object or Move The User: The Role of Interaction Techniques on Embodied Learning in
768 VR. *Frontiers in Virtual Reality*, 2021.

769 Bangera, G., and Brownell, S. E.: Course-based undergraduate research experiences can make
770 scientific research more inclusive, *CBE Life Sciences Education*, Vol. 13, No. 4, pp. 602–
771 606, 2014.

772 Barab, S. and Dede, C.: Games and Immersive Participatory Simulations for Science Education:
773 An Emerging Type of Curricula, *Journal of Science Education and Technology*, Vol. 16,
774 No. 1, pp. 1–3, 2007.

775 Bethune, K.: Changing Trends and Rethinking Geoscience Education in the Context of a Global
776 Crisis. *Geoscience Canada*, Vol. 47, No. 4, pp. 167–169,
777 <https://doi.org/10.12789/geocanj.2020.47.164>, 2020.

778 Bowman, D. A., North, C., Chen, J., Polys, N. F., Pyla, P. S., and Yilmaz, U.: Information-rich
779 virtual environments, *Proceedings of the ACM Symposium on Virtual Reality Software*
780 *and Technology*, ACM, New York, NY, p. 81., 2003.

781 Bowman, D. A. and McMahan, R. P.: Virtual reality: How much immersion is enough?, *Computer*,
782 Vol. 40, No. 7, pp. 36–43, 2007.

783 Bursztyn, N., Goode, R., and McDonough, C.: “I Felt Like a Scientist!”: Accessing America’s
784 National Parks on Every Campus, in Thompson, J. and Houseal, A., (Eds.) *America's*
785 *Largest Classroom: What We Learn from Our National Parks*, Berkeley, University of
786 California Press, pp. 151-166. <https://doi.org/10.1525/9780520974555-019>, 2020.

787

788 Bursztyn, N., Shelton, B., Walker, A., and Pederson, J.: Increasing Undergraduate Interest to Learn
789 Geoscience with GPS-based Augmented Reality Field Trips on Students' Own
790 Smartphones, *GSA Today*, pp. 4–10, 2017.

791 Bursztyn, N., Riegel, H., Sajjadi, P., Masters, B., Zhao, J., Huang, J., Bagher, M., Wallgrun, J.,
792 and Klippel, A. Fostering Geological Thinking Through Virtual Strike and Dip
793 Measurements. 2021 IEEE Conference on Virtual Reality and 3D User Interfaces Abstracts
794 and Workshops (VRW), pp. 303-308, doi: 10.1109/VRW52623.2021.00061, 2021.

795 Cawood, A. J., and Bond, C. E.: eRock: An Open-Access Repository of Virtual Outcrops for
796 Geoscience Education, *GSA Today*, <https://doi.org/10.1130/GSATG373GW.1>, 2018.

797 Chenrai, P., and Jitmahantakul, S.: Applying Virtual Reality Technology to Geoscience
798 Classrooms, *Review of International Geographical Education Online*, Vol. 9, No. 3, pp
799 577-590, 2019.

800 Cunningham, T.D., and Lansiquot, R.D.: Modeling Interdisciplinary Place-Based Learning in
801 Virtual Worlds: Lessons Learned and Suggestions for the Future, in Lansiquot, R.D., and
802 MacDonald S.P. (Eds.), *Interdisciplinary Perspectives on Virtual Place-Based Learning*,
803 Springer International Publishing, Cham, pp. 133–145, 2019.

804 Dalgarno, B.: The potential of 3D virtual learning environments: A constructivist analysis.
805 *Electronic Journal of Instructional Science and Technology*, Vol. 5, No. 2, pp. 3-6, 2002.

806 Dalgarno, B., Hedberg, J., and Harper, B.: The contribution of 3D environments to conceptual
807 understanding. In: Paper presented at the ASCILITE 2002, Auckland, New Zealand, 2002.

808 Dede, C.: Immersive interfaces for engagement and learning, *Science*, Vol. 323, No. 5910, pp. 66–
809 69, 2009.

810 Dolphin, G., Dutchak, A., Karchewski, B., and Cooper, J.: Virtual field experiences in introductory
811 geology: Addressing a capacity problem, but finding a pedagogical one, *Journal of*
812 *Geoscience Education*, Vol. 67, No. 2, pp. 114–130, 2019.

813 Elkins, J., and Elkins, N. M. L.: Teaching geology in the field: Significant geoscience concept
814 gains in entirely field-based introductory geology courses, *Journal of Geoscience*
815 *Education*, Vol. 55, No. 2, 2017.

816 Gagnier, K. M., Atit, K., Ormand, C. J., Shipley, T. F.: Comprehending 3D Diagrams: Sketching
817 to Support Spatial Reasoning, *Topics in Cognitive Science*, Vol. 9, No. 4, pp. 883–901,
818 2017.

819 Giles, S., Jackson, C., and Stephen, N.: Barriers to fieldwork in undergraduate geoscience degrees,
820 Nature Reviews Earth & Environment, Vol. 1, No. 2, pp. 77–78, 2020.

821 Hall, T., Healey, M., and Harrison, M.: Fieldwork and disabled students: Discourses of exclusion
822 and inclusion, Journal of Geography in Higher Education, Vol. 28, No. 2, pp. 255–280,
823 2004.

824 Hocine, N., Gouaich, A., and Cerri, S.A.: Dynamic Difficulty Adaptation in Serious Games for
825 Motor Rehabilitation, International Conference on Serious Games, pp. 115–128, 2015.

826 Höffler, T.N., and Leutner, D.: The role of spatial ability in learning from instructional animations
827 - Evidence for an ability-as-compensator hypothesis, Computers in Human Behavior, Vol.
828 27, No. 1, pp. 209–216, 2011.

829 IJsselsteijn, W.A., and Riva, G.: Being there: The experience of presence in mediated
830 environments, in IJsselsteijn, W.A., Riva, G., and Davide, F., (Eds.), Being there:
831 Concepts, effects and measurements of user presence in synthetic environments, IOS,
832 Amsterdam, Oxford, pp. 3–16, 2003.

833 Jamieson, M., Cullen, B., McGee-Lennon, M., Brewster, S., and Evans, J. J.: The efficacy of
834 cognitive prosthetic technology for people with memory impairments: a systematic review
835 and meta-analysis, Neuropsychological rehabilitation, Vol. 24, No. 3-4, pp. 419–444,
836 2014.

837 Johnson-Glenberg, Mina C.: Immersive VR and Education. Embodied design principles that
838 include gesture and hand controls. In Front. Robot. AI 5, p. 27. DOI:
839 10.3389/frobt.2018.00081, 2018.

840 Klingenberg, S., Jørgensen, M. L., Dandanell, G., Skriver, K., Mottelson, A., and Makransky, G.:
841 Investigating the effect of teaching as a generative learning strategy when learning through
842 desktop and immersive VR: A media and methods experiment. British Journal of
843 Educational Technology, Vol. 51, No. 6, pp. 2115-2138, 2020.

844 Klippel, A., Zhao, J., Jackson, K. L., La Femina, P., Stubbs, C., and Oprean, D.: Transforming
845 earth science education through immersive experiences - delivering on a long held promise.
846 In Journal of Educational Computing Research, Vol. 57, No. 7, pp. 1745–1771. DOI:
847 10.1177/0735633119854025, 2019.

848 Klippel, A., Zhao, J., Oprean, D., Wallgrün, J. O., Stubbs, C., La Femina, P., and Jackson, K. L.:
849 The value of being there: toward a science of immersive virtual field trips. In *Virtual*
850 *Reality* 24, pp. 753–770. DOI: 10.1007/s10055-019-00418-5, 2020.

851 Klippel, A., Zhao, J., Sajjadi, P., Wallgrün, J. O., Bagher, M. M., and Oprean, D.: Immersive place-
852 based learning – An extended research framework. In : 2020 IEEE Conference on Virtual
853 Reality and 3D User Interfaces Abstracts and Workshops (VRW). Piscataway, NJ: IEEE,
854 pp. 449–454. Available online at [https://conferences.computer.org/vr-](https://conferences.computer.org/vr-tvcg/2020/pdfs/VRW2020-4a2sylvMzvhjhioY0A33wsS/653200a449/653200a449.pdf)
855 [tvcg/2020/pdfs/VRW2020-4a2sylvMzvhjhioY0A33wsS/653200a449/653200a449.pdf](https://conferences.computer.org/vr-tvcg/2020/pdfs/VRW2020-4a2sylvMzvhjhioY0A33wsS/653200a449/653200a449.pdf),
856 2020.

857 Kolb, David A.: *Experiential learning. Experience as the source of learning and development.*
858 Second edition. Upper Saddle River, New Jersey: Pearson Education Inc., 2014.

859 Lages, W. S. and Bowman, D. A.: Move the Object or Move Myself?: Walking vs. Manipulation
860 for the Examination of 3D Scientific Data, *Frontiers in ICT*, Vol. 5, p. 236., 2018.

861 Lansiquot, R. D. and MacDonald S. P. (Eds.): *Interdisciplinary Perspectives on Virtual Place-*
862 *Based Learning*, Springer International Publishing, Cham., 2019.

863 Lee, E.A.-L., Wong, K. W., and Fung, C. C.: Educational values of virtual reality: The case of
864 spatial ability, *Proceedings of World Academy of Science, Engineering and Technology.*
865 Paris, France, 2009.

866 Liben, L.S., and Titus, S.J.: The importance of spatial thinking for geoscience education: Insights
867 from the crossroads of geoscience and cognitive science, in Kastens, K.A., Manduca, C.A.
868 (Eds.), *Earth and Mind II*, Geological Society of America, Boulder, Colorado, 2012.

869 Liu, R., Wang, L., Lei, J., Wang, Q., and Ren, Y.: Effects of an immersive virtual reality-based
870 classroom on students’ learning performance in science lessons, *British Journal of*
871 *Educational Technology*, 2020.

872 Lopes, R., and Bidarra, R.: Adaptivity challenges in games and simulations: A survey, *IEEE*
873 *Transactions on Computational Intelligence and AI in Games*, Vol. 3, No. 2, pp. 85–99,
874 2011.

875 Maceachren, A.M., and Brewer, I.: Developing a conceptual framework for visually-enabled
876 geocollaboration, *Int. J. Geographical Information Science*, Vol. 18, No. 1, pp. 1-34, DOI:
877 10.1080/13658810310001596094, 2004.

878 Marín-Spiotta, E., Barnes, R. T., Berhe, A. A., Hastings, M. G., Mattheis, A., Schneider, B., and
879 Williams, B. M.: Hostile climates are barriers to diversifying the geosciences. In *Adv.*
880 *Geosci.* 53, pp. 117–127. DOI: 10.5194/adgeo-53-117-2020, 2020.

881 Marshall, M. S. and Higley, M. C.: Multi-scale virtual field experience, Grand Ledge, Michigan,
882 USA, *Geosci. Commun. Discuss.* [preprint], <https://doi.org/10.5194/gc-2021-10>, in
883 review, 2021.

884 Mayer, R. E. and Sims, V. K.: For whom is a picture worth a thousand words? Extensions of a
885 dual-coding theory of multimedia learning, *Journal of Educational Psychology*, Vol. 86,
886 No. 3, pp. 389–401, 1994.

887 McComas W.F. (Ed.): *The Language of Science Education: An Expanded Glossary of Key Terms*
888 *and Concepts in Science Teaching and Learning*, SensePublishers and Imprint:
889 SensePublishers, Rotterda, 2014.

890 Mead, C., Buxner, S., Bruce, G., Taylor, W., Semken, S., and Anbar, A.D.: Immersive, interactive
891 virtual field trips promote science learning, *Journal of Geoscience Education*, Vol. 67, No.
892 2, pp. 131–142. DOI: 10.1080/10899995.2019.1565285, 2019.

893 Métois, M., Martelat, J. E., Billant, J., Andreani, M., Escartin, J., & Leclerc, F.: Deep oceanic
894 submarine fieldwork with undergraduate students, an exceptional immersive experience
895 (Minerve software). *Solid Earth Discussions*, pp.1-17, 2021.

896 Morales, N., O’Connell, K. B., McNulty, S., Berkowitz, A., Bowser, G., Giamellaro, M., and
897 Miriti, M. N.: Promoting inclusion in ecological field experiences: Examining and
898 overcoming barriers to a professional rite of passage, *The Bulletin of the Ecological*
899 *Society of America*, Vol. 101, No. 4, p. 11., 2020.

900 Moysey, S. M. J., and Lazar, K. B.: Using virtual reality as a tool for field-based learning in the
901 earth sciences, in Lansiquot, R.D., and MacDonald, S.P. (Eds.), *Interdisciplinary*
902 *Perspectives on Virtual Place-Based Learning*, Springer International Publishing, Cham.,
903 2019.

904 Moysey, S., Maas, B., Lazar, K., Klippel, A., and Bursztyn, N.: The whys and hows of
905 implementing virtual and augmented reality in Earth Science Classrooms. *Earth Educators*
906 *Rendezvous workshop*, July 13-17, 2020.

907 Nesbit, P. R., Boulding, A. D., Hugenholtz, C. H., Durkin, P. R., & Hubbard, S. M.: Visualization
908 and sharing of 3D digital outcrop models to promote open science. *GSA Today*, Vol. 30,
909 No. 6, pp.4-10, 2020.

910 Newcombe, N. S.: *Picture This: Increasing math and science learning by improving spatial*
911 *thinking*, American Educator, 2010.

912 O'Connor, E.A. and Domingo, J.: A Practical Guide, With Theoretical Underpinnings, for
913 Creating Effective Virtual Reality Learning Environments, *Journal of Educational*
914 *Technology Systems*, Vol. 45, No. 3, pp. 343–364, 2017.

915 Orion, N., and Hofstein, A.: Factors that influence learning during a scientific field trip in a natural
916 environment, *Journal of Research in Science Teaching*, Vol. 31, No. 10, pp. 1097–1119,
917 2014.

918 Ormand, C. J., Shipley, T. F., Tikoff, B., Harwood, C. L., Atit, K., and Boone, A. P.: Evaluating
919 Geoscience Students' Spatial Thinking Skills in a Multi-Institutional Classroom Study,
920 *Journal of Geoscience Education*, Vol. 62, No. 1, pp. 146–154, 2014.

921 O'Sullivan, M., and Kearney, G.: Virtual Reality (VR) Technology: Empowering Managers to
922 Reduce and Eliminate Accessibility Barriers for People with Autism Spectrum Disorders,
923 *Studies in Health Technology and Informatics*, Vol. 256, pp. 253–261, 2018.

924 Parong, J., and Mayer, R. E.: Cognitive and affective processes for learning science in immersive
925 virtual reality, *Journal of Computer Assisted Learning*, 2020.

926 Peirce, N., and Wade, V.: Personalised learning for casual games: The 'language trap' online
927 language learning game, *Leading Issues in Games Based Learning*, Vol. 159, 2019.

928 Petersen, G. B., Klingenberg, S., Mayer, R. E., and Makransky, G.: The virtual field trip:
929 Investigating how to optimize immersive virtual learning in climate change education. In
930 *Br J Educ Technol*. DOI: 10.1111/bjet.12991, 2020.

931 Powers, A. L.: An evaluation of four place-based education programs, *The Journal of*
932 *Environmental Education*, Vol. 35, No. 4, pp. 17–32, 2004.

933 President's Council of Advisors on Science and Technology (PCAST): *Engage to Excel:*
934 *Producing One Million Additional College Graduates with Degrees in Science,*
935 *Technology, Engineering and Mathematics*, U.S. Government Office of Science and
936 *Technology*, Washington, DC, 2012.

937 Pyle, E. J.: The evaluation of field course experiences: A framework for development,
938 improvement, and reporting, in Whitmeyer, S.J., Mogk, D.W., and Pyle, E.J. (Eds.), *Field*
939 *Geology Education: Historical perspectives and modern approaches: GSA Special Paper*
940 *461*, Geological Society of America, Boulder, CO, pp. 341–356, 2009.

941 Riquelme, A., Pastor, J. L., Cano, M., Tomás, R., Benavente, D., & Jordá, L.: Digitalisation of
942 rock specimens and outcrops for training. In *ISRM International Symposium-EUROCK*
943 *June, 2020*.

944 Rotzien, J. R., Sincavage, R., Pellowski, C., Gavillot, Y., Cooper, S., Shannon, J., Sawyer, J.F.,
945 Yildiz, U., Filkorn, H., and Uzunlar, N.: Field-based geoscience education during the
946 COVID-19 pandemic: Planning and execution, Part II: Geological Society of America,
947 *GSA 2020 Connects Online*, abstract no. 359659. [https://doi.org/10.1130/abs/2020AM-](https://doi.org/10.1130/abs/2020AM-359659)
948 [359659](https://doi.org/10.1130/abs/2020AM-359659)., 2020.

949 Rotzein, J. R., Sincavage, R., Pellowski, C., Gavillot, Y., Filkorn, C., Cooper, S., Shannon, J., Yildiz,
950 U., Sawyer, F., and Uzunlar, N.: Field-Based Geoscience Education during the COVID-19
951 Pandemic: Planning, Execution, Outcomes, and Forecasts, *GSA Today*, v. 31, pp. 4-10,
952 doi.org/10.1130/GSATG483A.1. CC-BY-NC, 2021.

953 Sajjadi, P., van Broeckhoven, F., and de Troyer, O.: Dynamically adaptive educational games: A
954 new perspective", *International Conference on Serious Games*, pp. 71–76, 2014

955 Sajjadi, P., Zhao, J., Wallgrün, J. O., Fatemi, A., Zidik, Z., La Femina, P., Fuhrman, T., and
956 Klippel, A.: The effect of virtual agent gender and embodiment on the experiences and
957 performance of students in Virtual Field Trips. In : *2020 IEEE International Conference on*
958 *Engineering, Technology and Education*. Piscataway, New Jersey: IEEE, pp. 1–8, 2020.

959 Schreier, M.: *Qualitative content analysis in practice*. Sage Publications, 2012.

960 Semken, S., Ward, E.G., Moosavi, S., and Chinn, P.W.U., 2018. Place-based education in
961 geoscience: Theory, research, practice, and assessment: Theory, Research, Practice, and
962 Assessment, *Journal of Geoscience Education*, Vol. 65, No. 4, pp. 542–562.

963 Shipley, T. F., Tikoff, B., Ormand, C., and Manduca, C.: Structural geology practice and learning,
964 from the perspective of cognitive science, *Journal of Structural Geology*, Vol. 54, pp. 72–
965 84, 2013.

966 Shute, V.J., Wang, L., Greiff, S., Zhao, W., and Moore, G.: Measuring problem solving skills via
967 stealth assessment in an engaging video game, *Computers in Human Behavior*, Vol. 63,
968 pp. 106–117, 2016.

969 Simpson, M., Zhao, J., and Klippel, A.: Take a walk: Evaluating movement types for data
970 visualization in immersive virtual reality, *Workshop on Immersive Analytics (IA)*, at IEEE
971 VIS, Phoenix, Arizona, USA, October 1st, 2017.

972 Simpson, M.: *Scale and Space: Representations in Immersive Virtual Reality*, Ph.D., University
973 Park, PA USA, 2020.

974 Slater, M.: Measuring Presence: A Response to the Witmer and Singer Presence Questionnaire,
975 *Presence: Teleoperators and Virtual Environments*, Vol. 8, No. 5, pp. 560–565, 1999.

976 Streicher, A., and Smeddinck, J. D.: Personalized and Adaptive Serious Games, *Entertainment
977 Computing and Serious Games*, pp. 332–377., 2016.

978 Stumpf II, R. J. ,Douglass, J., and Dorn, R. I.: Learning Desert Geomorphology Virtually versus
979 in the Field, *Journal of Geography in Higher Education*, Vol. 32, No.3, pp. 387-399, DOI:
980 [10.1080/03098260802221140](https://doi.org/10.1080/03098260802221140), 2008.

981 Tibaldi, A., Bonali, F. L., Vitello, F., Delage, E., Nomikou, P., Antoniou, V. E., Becciani, U., Van
982 Wyk de Vreis, B., Krokos, M., and Whitworth, M.: Real world–based immersive Virtual
983 Reality for research, teaching and communication in volcanology. *Bull Volcanol.* Vol. 82,
984 No. 38, <https://doi.org/10.1007/s00445-020-01376-6>, 2020.

985 Titus, S., and Horsman, E.: Characterizing and Improving Spatial Visualization Skills", *Journal of
986 Geoscience Education*, Vol. 57, No. 4, pp. 242–254, 2009.

987 Uttal, D.H. and Cohen, C.A.: Spatial Thinking and STEM Education, Vol. 57, pp. 147–181, 2012.

988 Vandewaetere, M., Cornillie, F., Clarebout, G., and Desmet, P.: Adaptivity in Educational Games:
989 Including Player and Gameplay Characteristics, *International Journal of Higher Education*,
990 Vol. 2, No. 2, 2013.

991 Whitmeyer, S.J. and Dordevic, M.: Creating virtual geologic mapping exercises in a changing
992 world. *Geosphere* 2020; Vol. 17, No.1, pp. 226–243, <https://doi.org/10.1130/GES02308.1>,
993 2021.

994 Winn, W.: A conceptual basis for educational applications of virtual reality. Technical Publication
995 R-93-9, Human Interface Technology Laboratory of the Washington Technology Center,
996 Seattle: University of Washington, 1993.

- 997 Witmer, B.G. and Singer, M.J.: Measuring presence in virtual environments: A presence
998 questionnaire, *Presence*, Vol. 7, No. 3, pp. 225–240, 1998.
- 999 Wu, B., Yu, X., and Gu, X.: Effectiveness of immersive virtual reality using head-mounted
1000 displays on learning performance: A meta-analysis, *British Journal of Educational*
1001 *Technology*, Vo. 51, Iss. 6, pp. 1991-2005, <https://doi.org/10.1111/bjet.13023>, 2020.
- 1002 Zaharias, P., Mehlenbacher, B., Law, E.L.-C., and Sun, X.: Evaluating user experience of adaptive
1003 digital educational games with Activity Theory, *International Journal of Human-Computer*
1004 *Studies*, Vol. 70, No. 7, pp. 478–497, 2012.
- 1005 Zhao, J., and Klippel, A.: Scale-unexplored opportunities for immersive technologies in place-
1006 based learning, 2019 IEEE Conference on Virtual Reality and 3D User Interfaces (VR),
1007 pp. 155–162, 2019.
- 1008 Zhao, J., Simpson, M., Wallgrün, J.O., Sajjadi, P., and Klippel, A.: Exploring the Effects of
1009 Geographic Scale on Spatial Learning, *Cognitive research: principles and implications*,
1010 Vol. 5, No. 1, p. 14, 2020.