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Is It Scientific? Viewer Perceptions of Storm Surge Visualizations

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1 Is it scientific? Perceptions of hybrid
2 Landscape-Data-Visualizations of storm
3 surge.

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16 Abstract

17 Scientists and coastal and risk managers are using semi-realistic visualizations of storm-
18 surge connected to hydrodynamic models. These visualizations may make depictions of
19 forecast impacts more engaging and accessible. However, they do not fit within established
20 frameworks for visualizing risk because they add representational detail that exceeds current
21 guidance. This study explores how audiences regard these visualizations in relationship to
22 perceived representational norms. Survey respondents were asked about characteristics that
23 make a representation “scientific.” Results suggest that audiences may perceive semi-realistic
24 visualizations of real places as scientific providing some conventions are met. It demonstrates
25 that the persons and institutions behind the visualization may influence perceptions of
26 legitimacy more than the style of the visualization. Although this opens new representational
27 possibilities, it also may increase the potential of visualizations to be misleading and may foster
28 perceptions that scientists are engaged in advocacy.

29 Keywords

30 storm surge; visualization; risk communication; visual rhetoric; argumentation; data
31 visualization

32 1.0 Introduction

33 Coastal communities are subject to increasing but uncertain risks from storm surge,
34 wind and precipitation (Romero and Emanuel 2017; Rosowsky 2018; Woodruff et al. 2013).
35 Emergency managers and risk communicators fear that people underestimate these hazards.
36 (Bostrom et al. 2018; Morrow et al. 2015). Although current emphasis is placed on map based
37 representations created with Geographic Information Systems (GIS) (NOAA 2013), scientists
38 and researchers are developing novel visualization technologies to better communicate risks to
39 policy makers and the public (Fenech et al. 2017; Spaulding et al. 2016).

40 This paper considers one such approach, model-driven semi-realistic 3D visualizations of
41 recognizable places (Figure 1). These visualizations combine aspects of Landscape Visualization
42 and Data Visualization. Landscape Visualization “represents actual places and on-the-ground
43 conditions in 3D perspective views, often with fairly high realism” (Sheppard and Salter 2004).
44 Data visualization is concerned with “the representation and presentation of data to facilitate
45 understanding.” (Kirk 2016). Although this definition of data visualization does not directly
46 exclude the notion of representing landscapes, norms of data visualization exhibit a preference
47 for emphasizing essential aspects of the data (e.g., Harold et al. 2016; Tufte and Weise Moeller
48 1997). The level of detail and scenography included in many landscape visualizations

49 contradicts this norm. We thus define these visualizations as a hybrid for purposes of
50 exploration and discussion.

51 Landscape Data Visualizations (LDVs) are here defined as: realistic and semi-realistic 3D
52 visualizations of real places distinguished by their integration of numerical models (e.g.,
53 ADvanced CIRCulation model) and 3D visualization platforms (e.g., game engines, custom
54 software) that make visualization outcomes a direct product of underlying modelling, such as
55 the implementation of fragility curves and damage functions (e.g., Spaulding et al. 2016). Some
56 Landscape Data Visualizations have the capacity to display results from forecast models in near
57 real-time, and real-time use is being considered (Stempel et al. 2018).

58 Development and use of LDVs by ocean scientists and planners coincides with a broader
59 recognition that visualizations in multiple arenas (e.g., species diversity) are playing increasingly
60 important role in communicating between scientists, policymakers and the public (McInerney et
61 al. 2014). Guidance for such visualizations (e.g., McInerney et. al. 2014, Harold et. al., 2016)
62 rejects presumptions that scientific graphics are necessarily plain, unadorned or inscrutable,
63 but maintains a preference for emphasizing essential aspects of the underlying data and
64 avoiding extraneous stylistic elements that may distract or obscure underlying meanings
65 (Smallman and John 2005). The boundary of when a visualization is a product of science or a
66 product of art or journalism, however, is not explicit.

67 LDVs arguably seek to leverage a more dramatic rhetorical style while maintaining
68 claims of legitimacy that stem from their basis scientific and technical processes. Although
69 experts tend to overestimate the role of simulations or technical processes in evaluations of
70 legitimacy (Fogg and Tseng 1999), the explicit (as opposed to tacit) use of persuasive imagery
71 directly challenges norms that favor essential rhetorical styles and presumptions on the part of
72 some experts and audiences that scientific graphics are somehow devoid of argumentation
73 (Muehlenhaus 2012; Walsh 2014).

74 LDVs of storm surge thus form a potent case to explore the perceived boundaries of
75 representational norms because they obviously do not fit within accepted paradigms for risk
76 communication or scientific data visualization. This research thus uses these visualizations as a
77 basis to explore characteristics that contribute to whether a visualization is perceived as being
78 “scientific”. Respondents to an online survey evaluating semi-realistic visualizations (n=735)
79 were asked:

80 **“What characteristics make a graphic or visualization scientific?”.**

81 Although the question can hypothetically stand on its own and be asked in the abstract,
82 responses are grounded in the context of the survey and the nature of visualizations being
83 tested therein. They thus provide insight into the ways in which audiences evaluate LDVs,
84 providing a starting point for further research into the use of these visualizations for risk
85 communication.

86 [1.1 Paradigms for using semi-realistic visualizations](#)

87 Semi-realistic visualizations of storm surge (even without the model driven component)
88 intended for direct use with the public currently fall into a neither-world outside of both

89 disciplinary frameworks (e.g., cartographic frameworks for visualizing risk such as Kostelnick et.
90 al., 2013), and transdisciplinary paradigms that use combined models and visualizations in
91 climate communication or landscape and urban planning (e.g., Schroth, Pond, & Sheppard
92 2011).

93 As it pertains to risk communication frameworks, there are acknowledged research gaps
94 as to how realistic visualizations of probabilistic risk are perceived (Bostrom et al. 2008). The
95 use of realism (or even 3D graphics) for risk communication is thus discouraged because
96 detailed realistic imagery may increase affective (instantaneous, subconscious, emotional)
97 responses that are difficult to account for (Bostrom et al. 2008; Kostelnick et al. 2013). Although
98 more recent research suggests that the effects of dual-process theory (distinct affective and
99 cognitive pathways) may be overstated (Kahan 2012), the crystalizing effects of detailed
100 imagery and the possibility that graphic features distract from meaning persist. Detailed
101 imagery overstates the resolution of the underlying data and makes outcomes appear more
102 certain than they are (Kostelnick et al. 2013). This creates the impression that there is greater
103 knowledge than exists. Moreover, the use of realism and other dramatic elements can distract
104 from the intended meaning of the communication (Smallman and John 2005).

105 The case for using visualizations with varying degrees of realism to communicate risks
106 has largely been made in the context of landscape and urban planning and climate
107 communication (Sheppard 2005). Realistic visualizations of future climate impacts such as sea
108 level rise have a unique capacity to engage the public by contextualizing information in
109 immediately recognizable and relatable contexts (Sheppard 2015; Sheppard et al. 2008).
110 Depictions of recognizable contexts may further stimulate feelings of place attachment and
111 *potentially* increase risk perception (Sheppard 2005). The question of how one appropriately
112 calibrates these visualizations such that they are perceived by the viewer as being salient,
113 credible, and legitimate has lead researchers to emphasize reflexive processes in which the
114 audience assists in shaping the physical and temporal scope of what is visualized (Schroth et al.
115 2011; White et al. 2010). This may include providing inputs to predictive models, such as data
116 derived from expert stakeholders (Schroth et al. 2011). For instance, impact visualizations that
117 incorporate qualitative data from emergency managers (e.g. Witkop et al. In Press) , or data for
118 hydrological models (e.g. White et al. 2010).

119 Although these practices are highly evolved (Sheppard et al. 2013), they primarily
120 address limited audiences such as persons involved in local visioning workshops (e.g., Becker
121 2017). As previously discussed, the question of how uninitiated audiences perceive these
122 visualizations (e.g., effects on risk perception, perceived legitimacy) outside of managed
123 processes is largely unanswered (Edsall and Deitrick 2009). Moreover, research addressing
124 graphics and visualizations used in risk communication suggests proponents of using semi-
125 realistic visualizations should be concerned with how visualizations may be dislocated in time as
126 they are shared, decontextualized, and recast for other purposes such as political advocacy.
127 (Bica et al. 2019).

128 **1.2 Use of imagery to communicate risk**

129 Imagery of storm impacts stimulates individual’s ability to call to mind exemplars of an
130 event (the availability heuristic). Images of flood damages accompanying flood projections
131 (Keller et al. 2006), or of storm surge impacts accompanying hurricane forecasts enhance risk
132 perception (Rickard et al. 2017). For example, in a comparison of hypothetical forecast images,
133 photographs of impacts showed the greatest effects on risk perception as compared to maps
134 depicting inundation (Rickard et al. 2017).

135 It is unlikely that LDVs that include qualities of both photographs and maps operate in
136 the same way. Photographs, despite advances in fakery, demonstrate that something exists,
137 and have a difficult time proving something does not exist (Messaris 1994). In contrast, the
138 notion of artifice and potential for disbelief is inherent to a visualization, and has been a long
139 preoccupation of visualization proponents (Orland et al. 2001). The perceived status of semi-
140 realistic visualizations, whether they are regarded as representations of underlying scientific or
141 technical processes, is thus highly relevant to whether other benefits such as making impacts
142 relatable in context are relevant. Research that shows a legitimacy bias favoring stripped down
143 or unadorned visualizations as somehow being more “scientific” (e.g. Walsh 2014) reinforces
144 the concern that there may be a ‘style penalty’ for using advanced semi-realistic visualizations.

145 Exploring how these visualizations are understood, and the characteristics that make a
146 visualization appear to be “scientific” is thus highly relevant to any use of semi-realistic 3D
147 visualizations for risk communication. To the extent that boundaries are malleable enough to
148 admit the use of these visualizations it may be possible to realize benefits such as making
149 impacts more relatable to local contexts and thus more salient to audiences (Lewis and
150 Sheppard 2006). Conversely, it’s also possible that scientists and experts considering use of
151 these visualizations are simply falling prey to a false presumption that advanced visualizations
152 are necessarily more effective (Harold et al. 2016; Smallman and John 2005). In the worst case,
153 well intentioned efforts to engage the public may be misleading or perceived as fear appeals
154 and undermine credibility of scientists and planners (O’Neill and Nicholson-Cole 2009).

155 **2.0 Methods**

156 The question, **“What characteristics make a graphic or visualization scientific?”**, was
157 incorporated into a larger survey evaluating semi-realistic visualizations in order to assess how
158 audiences perceive the status of model-driven semi-realistic visualizations. The survey was
159 distributed between June and August of 2017 and was open to all persons in the United States
160 over the age of 18. The study presented minimal risk and was anonymous, and thus was
161 granted “exempt” status by the University of Rhode Island Institutional Review Board.
162 Participants provided online consent at the start of the survey.

163 The survey included two primary instruments. The first, an expert survey, was
164 distributed to respondents using the email lists used by experts in the region such as the Rhode
165 Island Shoreline Change Special Area Management Plan, the Department of Homeland Security
166 Center of Excellence at the Coastal Resilience Center at the University of North Carolina, and an
167 internal mailing list for the Rhode Island Emergency Management Agency / Federal Emergency
168 Management Agency Integrated Emergency Management Course. The expert survey contained

169 questions regarding the depiction of probability and the appropriateness of using visualizations
170 for risk communication. The second, a public survey, was distributed via email and social media
171 to explore the broader perceptions of these visualizations.

172 There were a total of 115 responses to the expert survey and 620 responses to the
173 public survey instrument. 87 expert survey respondents and 362 public survey respondents
174 answered the question: **What characteristics make a graphic or visualization scientific?** (76%
175 and 58% respectively). As the sample for the public responses was distributed via a shareable
176 link, the survey sample of not statistically representative of the broader population. Moreover,
177 the respondents to the “public” survey had disproportionately high levels of education and
178 income, and in some cases were self-reported experts.

179 All surveys included three visualizations made for the Coastal and Environmental Risk
180 Index (CERI) (Spaulding et al. 2016). These visualizations depicted three communities in coastal
181 Southern Rhode Island USA and incorporate depictions of storm surge and of projected
182 structural damages (Spaulding et al. 2016) (Figure 2). Damage estimates are based on functions
183 developed by the US Army Corps of Engineers North Atlantic Coast Comprehensive Study
184 (NACCS) (Coulbourne et al. 2015). CERI models combine models for inundation, wave, and
185 erosion (Spaulding et al. 2016). Subsequent variations of CERI have been modified to depict
186 wind and a more generalized quantification of risk. The expert survey included an additional
187 visualization depicting inundation of coastal port infrastructure made for Federal Emergency
188 Management Agency Integrated Emergency Management Training Course (Stempel et al.
189 2018). This visualization depicted inundation (maximum envelope of water – MEOW) only
190 (Figure 1).

191 The following introductory statement was included:

192 “On the next three pages, you will see visualizations that show both the extent
193 of storm surge and the potential impact to houses in coastal Rhode Island
194 communities. The projected surge and damage to houses are based on computer
195 simulations that incorporate flooding, waves, and erosion. The names of the
196 communities are omitted from the visualizations so that we can test whether
197 they are recognizable to people who are familiar with them.

198 Damage to structures is represented as a percent of damage to the structure
199 between 1% and 100%. Structures that are colored red are destroyed. Structures
200 that are colored green are not damaged by storm surge. There is also an
201 indication of the likelihood of the depicted storm surge and damage event. In all
202 examples used in this survey, this is a 1% chance in any single year at present sea
203 levels. The style and position of the labels may vary slightly. Enlarged examples
204 of the labels are shown below. When you press continue, you will be taken to
205 the first visualization to evaluate.”

206 The question regarding characteristics that make visualizations scientific was included in
207 a summary page following the randomized evaluation of individual visualizations. Large

208 numbers of responses directly referencing the visualizations confirm that the question was
209 understood in relationship to the visualizations being tested.

210 Responses to the question were organized into a spreadsheet and inductively coded by
211 the research team (Thomas 2006). This process was made more straightforward by the strength
212 of scientific conventions of representation that clearly shaped responses. Answers often
213 included repeated phrases such as, "citation of data and sources". Initial groupings were based
214 on obvious similarities, taking care to discern when the intent of a phrase was altered by other
215 aspects of the text. From these groupings a set of four major themes was identified. To validate
216 the coding, codes were applied to a random subset of the data (n = 100) by an independent
217 coder. That coded sample was then compared and found to be 84% in agreement with the
218 coded data.

219 3.0 Results

220 The hybrid nature of LDVs, having both characteristics of maps and of realistic
221 scenography, is evident in the responses. A small number of the respondents used the term
222 "picture". Many answers, however, suggest that respondents place the visualizations in the
223 category of maps and other geographic information products they are familiar with:

224 "I am a commercial fisherman and rely on satellite images and have found them
225 invaluable in my industry and in protecting my property on the coast" (author's
226 emphasis).

227 "Whether the map is depicting possible events....." (author's emphasis).

228 One respondent was critical of the notion that these visualizations were scientific or necessary:

229 "Pretty pictures belie the science beneath them; Science data traditionally is
230 shown in a less aesthetic manner. "Being pretty" doesn't help sell the data."

231 Most respondents, however, exhibited a high degree of flexibility as to the style of the
232 visualizations presented and necessary characteristics provided that the sources of the data
233 were appropriately cited (103 mentions), background or methods elaborated (37 mentions),
234 and from a reputable source such as a research university, National Oceanic and Atmospheric
235 Administration (NOAA) or National Weather Service (NWS) (47 mentions).

236 "The fact that they are created by the University of Rhode Island and by qualified
237 scientists."

238 "A clear description of the data that were used and the process followed to
239 generate the visualization. Preferably, these visualizations would be peer
240 reviewed and published in a scientific journal."

241 "I would want to see something from NWS/NOAH [sic] blessing the program. I'm
242 not going to trust a graphic because of how it looks."

243 "Explanation of the data it is based on; understandable legends; university logos;
244 references to "behind the scenes" work to create model (i.e. scientific papers)."

245 These responses were coded into the larger theme of **“Transparency of sources, data and**
246 **methods” (181 mentions)**. Validation and integrity of the data based on replication, Quality
247 Assurance and Quality Control, and peer review were regarded as a component of this theme
248 (36 mentions).

249 “Does the study conform to previous estimates by other organizations not
250 associated with study creators. Do goals of the study reflect a common good or
251 is it the insurance industry and/or government trying to influence a program. e.g.
252 NFIP” (National Flood Insurance Program).

253 Historic storms were mentioned as means of validation or a basis for damage
254 assessments (16 mentions), which is logical given the extent to which comparisons to historic
255 events are used to evaluate model performance. **Error! Reference source not found.** presents
256 the major themes that were identified.

257 The **“Ability to discern underlying data and outcomes” (74 mentions)** was regarded as
258 a separate theme because it pertained to representation. Respondents cited the presence of
259 labelling, legends, and use of color gradients (51 mentions), and the ability to discern
260 quantifiable values (e.g., percent of damage) or detailed outcomes (26 mentions). The
261 juxtaposition of specific outcomes, for instance an undamaged structure depicted next to a
262 destroyed structure or not apparently aligned with the level of inundation) (23 mentions) was
263 cited as making the visualizations seem less scientific (Figure 3). Those with specific experience
264 of storm surge, however, were more likely to cite the juxtaposition positively.

265 “I did not understand how houses could be green and unthreatened while they
266 were directly next to red houses which were threatened. This made me feel the
267 visualizations were not totally accurate.”

268 “. . . Inland public at large may believe storm surge hits the coast with uniform
269 damage to all houses adjacent to body of water, unless elevation is readily
270 apparent.”

271 “Is it believable - there are always survivor structures, and they appear
272 appropriate to terrain.”

273 “Consistency--for example. in several of the visualizations. some of the houses in
274 the "surge zone" or "danger zone" were shown as being likely to suffer heavy
275 damage. while another next door to it was shown as being likely to suffer no
276 damage at all. This does not make sense.”

277 **“Style and Quality” (87 mentions)** also emerged as an important theme. Visualization
278 style (38 mentions) included the degree of abstraction, illustrative quality, and nature of the
279 colors chosen.

280 “It's an illustration not an actual picture.”

281 “The color coding and graphic visualization of the potential impact with 3D
282 structures really helps.”

283 “It obviously took time to make, it's not some quick journalist work. The 3D
284 models and the continuous color scale look like there's some real work (math,
285 computer simulation) behind this. It looked scary but didn't use flashy things to
286 increase the "alarming" feeling.”

287 “Lack of superfluous anesthetics [sic]. Inclusion of topology [sic].”

288 Comparatively few respondents cited the visualization style with a negative valence, for
289 instance being “too pretty” for the subject or being cartoonish:

290 “The more realistic it looks the better. The second example looked too much like
291 a cartoon.” (referring to the Misquamicut visualization).

292 Only one respondent specifically excluded visualizations from the notion of being scientific in
293 their response, preferring graphs and charts over visualizations:

294 “Graphs, charts. and numbers with source data listed at the bottom. The
295 visualizations and rainbow colors seem more like a computer generation from a
296 movie.”

297 Related to visualization style, the quality and professionalism of the visualizations was cited
298 positively (16 mentions) as was the use of geographic information (e.g., air photos, LiDAR) and
299 accurate depiction of geographic areas in terms of appearance of features and scale (25
300 mentions).

301 Although the survey overall reflects the high degrees of expertise present in both the
302 expert and public cohorts, some respondents were forthright about the influence of their own
303 personal experience, enough so that **“Personal Knowledge or experience” (40 mentions)**
304 became the fourth and smallest theme. In a few cases this included respondents who had direct
305 knowledge of the project or similar data (4 mentions). More frequently, however, respondents
306 evaluated stated that their evaluation depended on whether or not is seemed believable (15
307 mentions), or conformed to their personal experience of the place (23 mentions).

308 “My personal experiences as an aquaculturalist [sic] and living for 37 years in a
309 house within site [sic] of a shoreline evacuated 13 times for ‘hurricanes’ thus
310 witnessing first hand tidal surges. My home is also 200 years old.”

311 “Knowledge of damages along NJ coast for storms.”

312 Given the level of expertise, surprisingly very few respondents mentioned the
313 quantification or inclusion of uncertainty as being a characteristic of a visualization being
314 scientific. (16 mentions). Moreover, there is some evidence that at least some respondents
315 assumed analyses conducted were more complex than they were (e.g., including modelling of
316 soil types in relationship to structural damages).

317 Lastly, as compared to the number of respondents who sought transparency of data and
318 methods and the ability to decipher and engage with that data through the visualizations,
319 comparatively few respondents cited the use of data in and of itself (16 mentions), the use of
320 computer generated models (18 mentions), and notions of facts and objectivity (11 mentions).

321 The brevity of the answers associated with these mentions and other contextual information in
322 the answers did not support development of a theme.

323 3.1 Limitations

324 The primary limitation of the findings is the disproportionately high level of education
325 and wealth (as compared to census data) of survey respondents to the “public” survey
326 instrument, the data gathering method for this cohort was not rigorous enough to be regarded
327 as anything other than exploratory. Biases of wealth and expertise exist in the expert cohort;
328 these biases accord with the nature of the group surveyed.

329 A subsequent survey more narrowly designed and distributed to a statistically
330 representative sample of participants is necessary to make any inferences regarding the
331 perceptions of the lay public. The response rates to the question also do not provide any insight
332 as to why the question was skipped (e.g., survey fatigue, feeling the question is inappropriate).
333 It is therefore possible that a disproportionate number of people who felt positively about the
334 visualizations answered the question.

335 Despite these limitations, the survey does have a comparably high number of
336 respondents and provides valuable insight into the attitudes of persons likely to be engaged in
337 activities around risk communication, coastal resilience and other related topics that may
338 employ visualizations like those tested here. It similarly provides insight into how this audience
339 regards and evaluates such visualizations. A significant number of people who responded
340 resided in coastal communities depicted, in total. 131 respondents reported recognizing
341 Matunuck, RI, USA, 187 reported recognizing Charlestown, RI USA, and 168 reported
342 recognizing Misquamicut, RI USA in the visualizations. Numerous personal testimonies are
343 evidenced in the responses. Moreover, the extent to which the survey cohort (especially the
344 expert cohort) is directly engaged in risk assessment and communication provides valuable
345 insight as to how experts may perceive the use of advanced visualizations.

346 4.0 Discussion

347 Visual rhetoric such as maps all aspire to persuade even if the object of argumentation is
348 simply the veracity of the map and the cultural context that created it (Harley 1989). Minimalist
349 displays with spare high contrast graphics popularized in the 20th century may have aspired to
350 universality, but are nonetheless still transformed by the skills, interests, and interpretations of
351 diversifying audiences (Kostelnick 2008). These issues affect even the most narrowly configured
352 representations and are here revealed by the extent to which audiences brought their own
353 interpretation. For instance, differences in how the juxtaposition of damages was questioned
354 by respondents to this survey demonstrate likely differences between coastal and inland
355 residents based on their experience with storm damages. References to other maps and
356 graphics that audiences are familiar with further remind us that these interpretations are
357 shaped by expectations set by other graphics depicting storm surge, sea level rise, or
358 geographic data with which the audience is familiar. Whether intended or not, these graphics
359 set expectations and form the basis of conventions and norms (Kostelnick and Hassett 2003).

360 LDVs, however, likely stretch beyond these issues into territories of persuasion similar to
361 persuasive maps—maps that have a defined persuasive purpose (Muehlenhaus 2012, 2013): in
362 the case of LDVs, convincing coastal residents of the potential severity of damages in the
363 context of their coastal communities. LDV's differ from many persuasive maps in that they do
364 not use omission of data for their persuasive purposes (Muehlenhaus 2013), but rather rely on
365 portrayal of damages in context as means to enhance risk perception. Whether this is
366 warranted or not, largely depends on the communication objective.

367 This complicates the management of these graphics as compared to other graphics that
368 are used to depict storms in real-time or near-real-time. Hurricane track diagrams, for instance,
369 can be optimized based on cognitive factors, such as the salience of relevant information
370 (Hegarty 2011), the “naturalness” with which graphic features are mapped to dimensions of the
371 data (Boone et al. 2018), and the relative efficiency (e.g., ink to information ratio, Tufte and
372 Graves-Morris 1983). These optimizations, however, are only possible where the variables and
373 complexities are suitably controlled and use-case clear (Hegarty 2011).

374 Yet these restrained visualizations do not communicate the potential severity of
375 damages. The desire to more effectively communicate damages has led researchers to explore
376 the effects of supplemental imagery on risk perception (e.g., Keller et al. 2006; Rickard et al.
377 2017). LDVs potentially used in real-time arguably combine the role of forecasting tools with
378 supplemental imagery. The use of LDVs thus strains the conceit of limiting persuasion to
379 effective communication of data that governs more typical risk communication tools.

380 It is arguably the use of persuasion (e.g., promoting behavior change, Sheppard 2005)
381 related to climate change that lead to evolving paradigms for visualization use in landscape and
382 urban planning and climate communication. The term “permissible drama”, for instance, was
383 used to address the extent to which landscape visualizations incorporated some degree of
384 speculation and dramatization (Sheppard 2001; Sheppard et al. 2008). The uncertainty of future
385 conditions (Sheppard et al. 2008) and potential bias of experts (MacFarlane et al. 2005),
386 however, ultimately lead practitioners to increasingly emphasize skilled local visualizers such as
387 landscape architects who also served to relate science to the cultural context (Sheppard 2015).
388 This conforms to the larger drive in the context of climate communication to engage in co-
389 creation of outputs by communities and experts (Moser 2016).

390 The integration of model and visualization in LDVs attenuates these practices that
391 evolved to support the use of landscape visualizations. Even if the initial programming of these
392 systems is onerous, the development and use of these systems foreshadows more direct access
393 to increasingly dramatic forms of persuasive visual rhetoric by scientists.

394 This research into how these visualizations are perceived by audiences makes several
395 things clear.

- 396 1. **The style of a visualization is less important than the people that stand behind it.**
397 Audiences seem willing to accept LDVs as being “scientific” (or representations of
398 science) providing that certain conventions are adhered to such as transparency of data
399 and sources, and that those sources are reputable. Scientists who employ LDVs as tools

400 to communicate their work should thus not presume that the visualization will be
401 regarded differently than other maps or representations they stand behind. Moreover,
402 in media situations where visualizations are available, they are likely to become the
403 emblems of any project. For instance, newspaper coverage of the Coastal Environmental
404 Risk Index emphasized the most extreme and dramatic visualizations made available
405 (e.g., Kuffner 2016). This emphasis of the dramatic not only risks demotivating
406 audiences, it likely undermines the credibility of the team (O'Neill and Nicholson-Cole
407 2009).

408 2. **Realism and 3D are not equivalent.** The distorting effects of realism are not unique to
409 3D representations, and have been observed in 2D representations (e.g., using air
410 photos)(Zanola et al. 2009). It's notable that one respondent stated "the fewer graphic
411 'enhancements' applied. the more scientific it appears (no animated waves, etc.)"
412 (Several semi-realistic systems use animated water surfaces). This comment had less to
413 do with three dimensionality or perspective than the type of 2D texturing strategy used.
414 Moreover, audiences are sensitive to seemingly minor graphic differences of style or
415 coloration in LDVs that are categorically similar in terms of being perspectival and 3D.
416 Colors in some cases were perceived as appropriate, in others cartoonish. This suggests
417 that three-dimensionality in the landscape may be less important than other graphic
418 treatments.

419 3. **Overstatement.** As has been elsewhere observed, sophistication was associated with
420 professionalism and legitimacy (Kostelnick 2008). Some audience members presumed
421 the underlying models were more sophisticated than they were. One respondent, for
422 instance, presumed that structural damage models accounted for the soil type in
423 calculating damages (damages were calculated with fragility curves based on type of
424 construction), and erosion calculations were based on projected shoreline change
425 (Spaulding et al. 2016). This suggests that some criticisms of realistic 3D visualizations
426 (e.g., Kostelnick et al. 2013) are well placed but likely applicable to other sophisticated
427 visual rhetoric or 2D visualizations.

428 Although these observations would seem to disqualify these visualizations from certain
429 uses, they are also indicative of the limitations of categorical distinctions and the need to re-
430 think how we organize the discussion of visualizations more generally. The ability to orient
431 audiences and persuade them to evacuate, for instance, may be aided by the judicious use of
432 3D landscapes and landmarks that make the potential extent of a hazard less abstract. As it
433 pertains to geographic contexts there may be a very good argument for inclusion of realistic 3D
434 elements to contextualize data. The extent to which these features make information engaging
435 and salient do not necessarily contradict guidance for more restrained, cognitively justified
436 visualizations (e.g., Hegarty 2011).

437 In the preceding example, however, the level of detail included in some of the
438 visualizations tested here (e.g., damage calculations), is irrelevant to this purpose. One
439 respondent made exactly this point regarding the elaborate nature of the modeling shown.
440 Showing high levels of detail or complex outcomes may be secondary to displaying a
441 generalized indication of the hazard in a recognizable context. Moreover, providing indication
442 of wind or other hazard may be equally important so as not to understate risks faced in

443 adjacent areas. If the real-time capabilities of LDVs are to be relevant, these use-cases should
444 be identified in advance and appropriate design processes initiated to shape the visualizations.

445 Those processes will result in a different set of 3D depictions than those used in this
446 study. The visualizations tested here conform more closely to the parameters of visualizations
447 used in Disaster Risk Reduction or climate adaptation processes, for which guidance has been
448 established (e.g., Schroth et. al. 2011), and the rapid real-time aspects are irrelevant. LDVs in
449 this context are likely effective. For instance, the extent to which juxtaposition of outcomes was
450 associated with both increases and decreases in perceived credibility of modelled outcomes
451 suggests the discernible model details fostered engagement. Had these visualizations been
452 coupled with more in-depth explanations of the effects of building elevation (as they would be
453 in a DRR or climate communication process), they could leverage this curiosity regarding
454 unexpected effects into an educational opportunity (Kahan et al. 2017). The challenge then
455 becomes to convince ocean scientists and persons who are not DRR or Climate Communication
456 practitioners to employ climate communication practices (Moser 2016).

457 One logical conclusion of these critiques is that the capacity to produce LDVs is a
458 technology in search of a purpose; like other preceding visualization advancements, the
459 development of the technology has likely outpaced our understanding of its best use (Lovett et
460 al. 2015; Sheppard and Cizek 2009). The logic of this conclusion, however, is subject to changing
461 expectations and evolving norms. It's reasonable to speculate that the elasticity of norms
462 observed in this study reflects the robustness of scientific conventions and the familiarity of the
463 respondents with those conventions (Kostelnick and Hassett 2003). However, the apparent
464 acceptability of LDVs within these conventions may also reflect the evolution of norms based
465 on the increasing use of advanced visualizations across the breadth of the sciences. Scientists,
466 planners, and policy makers are shaping audiences (Kostelnick and Hassett 2003).

467 5.0 Conclusion

468 This research suggests that the boundaries of what is perceived to be scientific are malleable,
469 providing certain conventions for disclosure and transparency are met. These shifting boundaries are
470 indicative of the larger set of value judgements that scientists and consumers of scientific graphics make
471 every day (Walsh 2017). LDVs surface these judgements because they include elements such as light and
472 shadow and perspective that are clearly extraneous to the underlying data but may nonetheless
473 contribute to the ergonomics of the presentation and the impression it creates. This forefronts a range
474 of decisions, color choice, emphasis, that can make even the simplest presentation of data into a
475 dramatic and iconic image (Schneider 2016). It is therefore understandable that the use of semi-realistic
476 visualizations would be roundly discouraged in the context of existing frameworks for risk
477 communication.

478 As it pertains to the presentation of data that is spatially relevant to specific audiences,
479 however, there is an at least reasonable case for the consideration of perspectival presentations and
480 inclusion of recognizable landmarks that make outcomes less abstract (Lewis and Sheppard 2006). These
481 uses, however, require new and more careful designs. Use-cases such as communicating the effects of
482 building elevation where these visualizations may be highly effective are possible within existing
483 paradigms of climate communication.

484 Further research is necessary on three counts.

- 485 1. A visualization that a scientist may perceive as being accessory, or ‘just an illustration’ may be
486 nonetheless be perceived as the primary means of interface by audiences. A more refined
487 version of this research question should be repeated with a statistically representative sample
488 to determine whether the elasticity observed in this exploratory research is confined to highly
489 educated expert audiences, or if it is more widespread.
- 490 2. The role of visualizations should be clarified with the scientists seeking to use them such that
491 the precise use case can be tested. This begins by surveying those that would seek to use
492 visualizations and determining their specific objectives, and similarly requires surveying or
493 querying intended audiences such that visualizations can be developed not based on the
494 emergence of the technology but based on the purposes set forth and needs of the intended
495 audience. The use of 3D representations of recognizable places may be suited to some use
496 cases.
- 497 3. Categorical distinctions that place realistic and semi-realistic visualizations apart from other
498 sophisticated visualizations or 2D visualizations employing realism likely create false distinctions.
499 Highly diagrammatic 3D visualizations, for instance, may have more in common with 2D data
500 visualizations than is currently acknowledged by the ways in which visualizations are
501 categorized. More testing with real audiences is thus warranted.

502 The development of semi-realistic and realistic visualizations in real-time connection with storm
503 models follows a larger pattern of technology driving representational decisions (Lovett et al. 2015;
504 Sheppard and Cizek 2009). The desire to use the best and most advanced tools, even before the
505 evidence exists to support their use, is understandable given the desire of many scientists to connect to
506 audiences. Doing this blindly, however, risks distracting or misleading the public. The extent to which
507 reputation is a factor in assessments should give scientists pause, lest they undermine their own
508 credibility (O'Neill and Nicholson-Cole 2009). This study is a modest step to inverting the technology first
509 paradigm and better directing the development of these visualizations.

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656 Tables

657 Table 1, themes identified in response to: What characteristics make a graphic or visualization
658 scientific?

Theme	Total
Transparency of sources, data and methods.	181
Ability to discern underlying data and outcomes.	74
Style and quality.	87
Personal knowledge or experience.	40

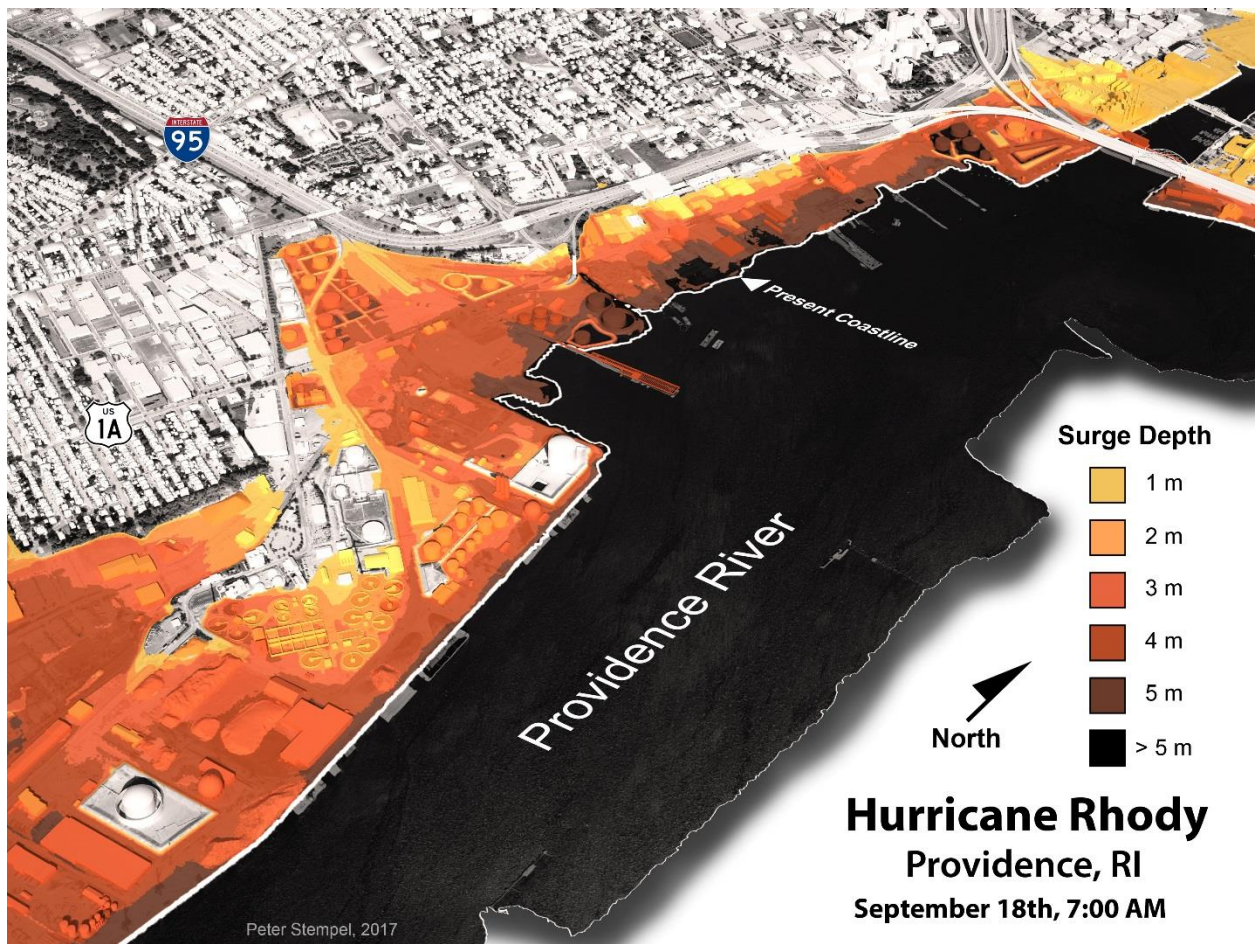
659

660 **Figure Captions**

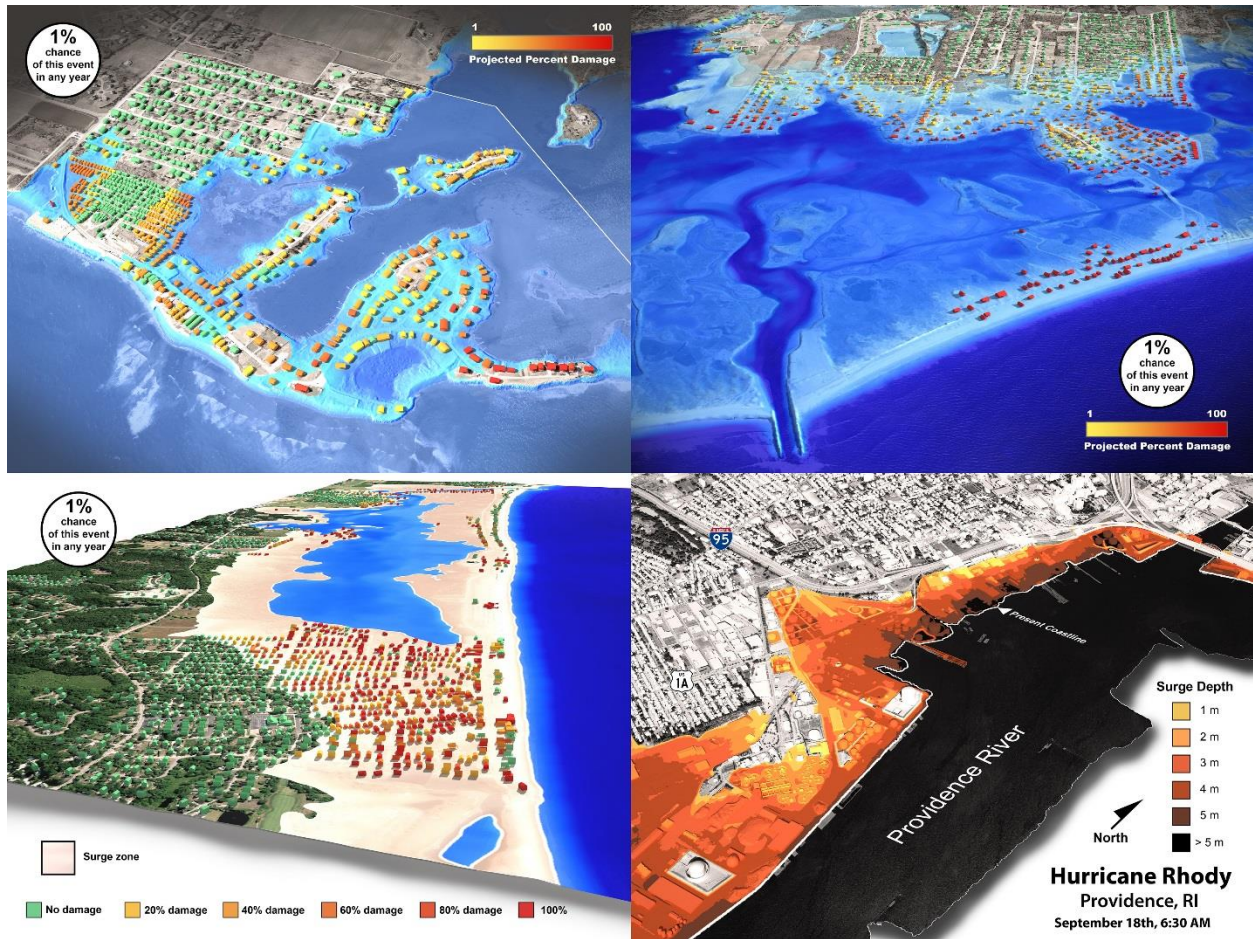
661 Figure 1, a model driven, semi-realistic depiction of the Port of Providence during an extreme
662 hurricane event. Image: Authors

663 Figure 2, four visualizations that were developed by the author and used in the survey.
664 Clockwise from lower left: Misquamicut, Westerly, RI USA, Matunuck, South Kingston, RI USA,
665 Charlestown, RI USA, and Providence, RI USA. The Providence visualization was only included in
666 the expert survey. Each visualization exhibited different stylistic characteristics such as the
667 distance at which the view was framed, and the color schema used.

668 Figure 3, respondents raised questions regarding apparent discrepancies in damage between
669 adjacent structures in this visualization of Matunuck, Rhode Island, USA. Image: Authors.



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