

### **Query to the Author**

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2 DEPARTMENT: VISUALIZATION VIEWPOINTS

3 **The Next Billion Users of Visualization**

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14 *We argue that visualization research has overwhelmingly focused on users from the*  
15 *economically developed world. However, billions of people around the world are*  
16 *rapidly emerging as new users of information technology. Most of the next billion*  
17 *users of visualization technologies will come from parts of the world that are*  
18 *extremely populous but historically ignored by the visualization research*  
19 *community. Their needs may be different to the types of users that researchers have*  
20 *targeted in the past, but, at the same time, they may have even more to gain in*  
21 *terms of access to data potentially affecting their quality of life. We propose a call*  
22 *to action for the visualization community to identify opportunities and use cases*  
23 *where users can benefit from visualization; develop universal design principles;*  
24 *extend evaluations by including the general population; and engage with a wider*  
25 *global population.*

26 **D**ata visualization is arguably a mature and  
27 respected field of research by many stand-  
28 ards, having existed as a recognized topic in  
29 academia for several decades. It is a research “success  
30 story” in terms of the degree to which ideas originating  
31 in academic research have made their way into com-  
32 mercial software (from the likes of Tableau and Micro-  
33 soft) and popular media (for example, the *New York*  
34 *Times* now famously has an information graphics  
35 department). These commercial interests have made  
36 further contributions, popularizing and making visuali-  
37 zation successful in their respective markets. Some of  
38 this success may be attributed to firm research

foundations, such as rigor around experimental meth- 39  
40 odologies, integration of theory from human-computer  
41 interaction and perceptual psychology, and technologi-  
42 cal tool-building. However, these foundations are lim-  
43 ited by a skewed authorship from universities and  
44 industry in highly developed countries (especially the  
45 U.S. and Europe). Furthermore, the foundations are  
46 built upon studies with an inherent selection bias of  
47 participants from a highly educated subset of the pop-  
48 ulations of these highly privileged nations who have a  
49 high level of graphic and numeric literacy and access  
50 to the latest information communication technologies  
51 (ICT).

52 However, the divisions between the technological  
53 have- and have-nots are breaking down across the  
54 world, at least in terms of access to the Internet and  
55 basic mobile technologies. At the same time, we are  
56 seeing more and more examples of the relevance of

57 data to the lives of every citizen of our planet. The  
58 COVID-19 pandemic is the most obvious, focusing the  
59 world's attention on time-series graphs like never  
60 before; but there are a host of other pressing issues  
61 that should be viewed from a data-centric perspective  
62 by a truly global audience.

63 In this article, we argue that data visualization  
64 researchers need to reconsider their assumptions  
65 about the audiences for visualization to include these  
66 emergent users of ICT. This new access to technology  
67 can bring many positives to peoples' lives, none the  
68 least of which is the potential to access information.  
69 However, as we have seen in recent times, information  
70 may be disseminated to people through media that  
71 are potentially destructive (Google bubbles, Facebook  
72 echo chambers, and so on). Visualization has an  
73 important role to play here in being a tool that allows  
74 people to explore data for themselves, rather than  
75 making them passive recipients of information.

76 But is our field of data visualization research able to  
77 provide or support the development of data communi-  
78 cation and exploration tools that are suitable for emer-  
79 gent ICT users? Do the assumptions about the end  
80 users of visualization, in place throughout the develop-  
81 ment of our field, still apply to these new users? The  
82 numbers of people gaining access to basic Internet-  
83 enabled devices in the developing world is staggering.  
84 Furthermore, there is great potential for visualization  
85 to profoundly affect these peoples' lives providing  
86 (potentially) access to information and data in a form  
87 that may cross cultural, educational, geographical, and  
88 accessibility barriers. But there are as many research  
89 questions as there are opportunities. In this article, by  
90 an international and interdisciplinary team of visualiza-  
91 tion, design, and inclusive technology researchers, we  
92 reflect on the development of data visualization, and  
93 we compare the needs of emergent users in economi-  
94 cally developing countries (primarily India) versus users  
95 (existing but also emerging) in developed countries.  
96 From this reflection, we call for action on a number of  
97 research but also organizational fronts.

## EXPERIENCES OF EMERGENT ICT USERS IN INDIA

100 We are conscious that there are radical and rapid  
101 changes in ICT use occurring or about to occur in  
102 many places around the world. Our lived experience is  
103 of India, which is an archetypal example of a develop-  
104 ing country with a large, emergent population of ICT  
105 users. 50% of India's population of 1.4 billion people is  
106 now connected to the Internet. With 1 billion mobile  
107 connections, close to 900 million of which are via

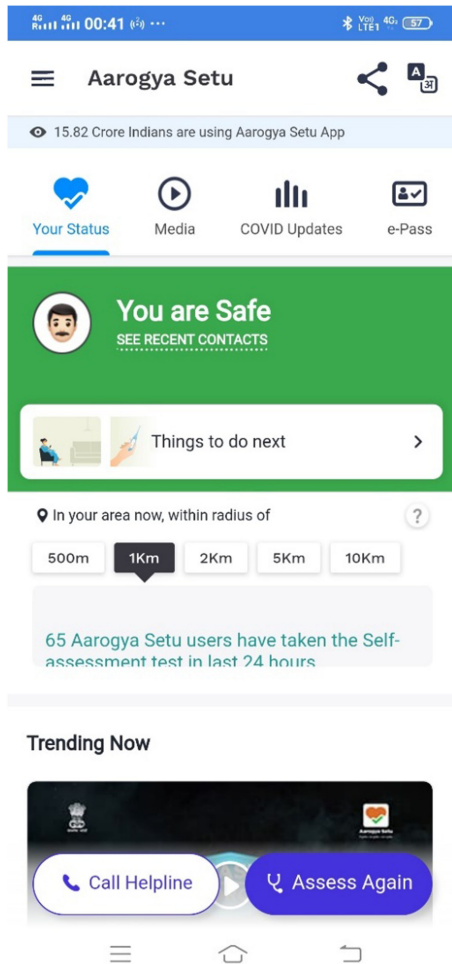
108 smartphones, the majority of the new Internet users  
109 (emergent users) access information, services, and  
110 entertainment through a multitude of apps that are  
111 designed for users who are not like them. Less than  
112 10%<sup>12</sup> of the Indian population can read and transact  
113 in the English language, yet nearly all apps are  
114 designed in English.<sup>13</sup> Language options in devices  
115 and automatic translation tools alleviate this problem  
116 to some extent, primarily at the user interface (UI)  
117 layer, but more sophisticated aspects of the apps  
118 remain inaccessible to the users because they are not  
119 designed with these users in mind.

120 Visualizations that enable users to solve complex  
121 numerical or spatial problems expeditiously and accu-  
122 rately are one such aspect we increasingly see in  
123 apps. For example, a banking app might allow an  
124 emergent user to perform simple transactions such as  
125 transferring money to a family member, paying a bill,  
126 or checking the balance. However, a set of visualiza-  
127 tions in the app that could help analyze her finances  
128 would remain unused because the user does not  
129 know how to interpret charts. We have come across  
130 several instances of delivery persons from e-com-  
131 merce vendors having little difficulty in picking up  
132 orders on delivery platforms such as Zomato or  
133 Swiggy, but who are unable to locate the customers'  
134 address using the navigational maps integrated with  
135 the shopping orders in their apps. We have seen  
136 parents unable to understand school report cards  
137 that presented visual analysis of their child's academic  
138 performance. School report cards using visualizations  
139 like the bar, line, pie charts were difficult to compre-  
140 hend by the parents.

141 These examples point to the fact that access is a  
142 multilayered problem. The lack of formal education/  
143 training in numeracy and graphicacy for emergent  
144 users is the primary reason for their inability to com-  
145 prehend and benefit from visualizations. However,  
146 computer-mediated visualizations that are personal,  
147 customized, adaptive, and progressively complex pres-  
148 ent us with an opportunity to address their needs.

149 Industry is already moving to deliver data-centric  
150 apps to this enormous new user base. One important  
151 example is Google's India-first payment app Tez, which  
152 was launched in September 2017. As an illustration of  
153 its success, over 22 million people and businesses  
154 used Tez to make over 750 million transactions that  
155 are collectively worth over USD 30 billion annually.<sup>5</sup>  
156 Now Tez has been taken beyond India and available as  
157 Google Pay worldwide, unifying all of Google's pay-  
158 ment offerings globally.

159 In terms of digital civics, Aarogya Setu (see  
160 Figure 1) is a mobile application launched by the



**FIGURE 1.** Aarogya Setu app launched by Government of India for Covid-19 contact tracing. It uses Bluetooth and GPS to show active cases near the user.

161 Government of India as a response to COVID-19 to  
 162 connect essential health services with the people of  
 163 India. It is available in 11 different languages. Aarogya  
 164 Setu uses contact tracing to record details of all the  
 165 people one may have come in contact with, as the per-  
 166 son goes about normal activities. If any one of them, at  
 167 a later point in time, tests positive for COVID-19, the  
 168 user is immediately informed and proactive medical  
 169 intervention is arranged for them.

## ASSUMPTIONS ABOUT VISUALIZATION USERS

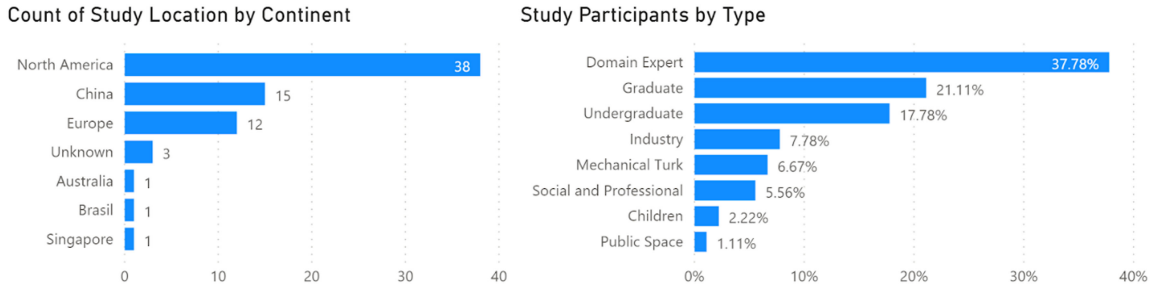
170 At present, the data visualization community has not  
 171 considered the needs of emergent ICT users like those  
 172 in India. Figure 2 shows an analysis of recent locations  
 173 and types of participants studies reported at the IEEE  
 174 VIS conference (InfoVis and VAST) 2019.<sup>17</sup> We did not  
 175  
 176

177 consider SciVis as our focus was on papers with a  
 178 user study component. As per the figure caption,  
 179 the studies are almost exclusively conducted in  
 180 developed countries, and the vast majority of par-  
 181 ticipants are highly educated. In 2010, Henrich  
 182 *et al.*<sup>6</sup> criticized behavioral science researchers for  
 183 their disproportionate reliance on WEIRD (Western,  
 184 Educated, Industrialized, Rich, and Democratic) par-  
 185 ticipants in studies and the implicit but unwar-  
 186 ranted assumption that findings from this group  
 187 generalize to all populations. This criticism has also  
 188 been leveled at human-computer interaction (HCI)  
 189 research.<sup>11</sup> It seems this reliance on WEIRD partici-  
 190 pants is also true for data visualization research.

191 It seems very unlikely that findings from data visu-  
 192 alization studies using WEIRD participants will gener-  
 193 alize to non-WEIRD populations. We know from  
 194 comparative studies that susceptibility to visual illu-  
 195 sions such as the Mueller-Lyer Illusion<sup>6</sup> and visual  
 196 preferences for websites<sup>8</sup> significantly vary between  
 197 WEIRD and non-WEIRD participants. Thus, what is  
 198 regarded as best practice in data visualization design  
 199 may well only apply to Western developed countries.

200 One reason for this is cultural difference. An exam-  
 201 ple is the difference in the significance of colors, for  
 202 instance the color used for mourning is not consistent  
 203 around the world. In Western cultures black is used; in  
 204 India, it is white; in much of Asia, red, in South Africa  
 205 and Egypt, it is yellow, and purple in Thailand. So, color  
 206 coding may be interpreted differently depending on  
 207 the culture of the users. In contrast, cultures can have  
 208 shared color meaning, for example, warning signs  
 209 around the world use a common color with red indi-  
 210 cating stop or danger. In addition to color, icons have  
 211 been used as an effective method for communication  
 212 to bridge language and cultural barriers, but these are  
 213 only effective if the objects and concepts are familiar  
 214 and compatible across cultures.<sup>19</sup>

215 Probably, however, the most important reason for  
 216 differences are different levels of familiarity with the  
 217 information graphics used in data visualization. In  
 218 Western developed countries, children are explicitly  
 219 taught graphic literacy as an integrated part of the  
 220 curriculum. For example, in Australia, maps are taught  
 221 from the first year of school, with more sophisticated  
 222 concepts like grid references taught in the fourth and  
 223 fifth years of school. Graphs are progressively intro-  
 224 duced with column graphs in the third year of school,  
 225 pie graphs in the fourth year, line graphs in the sev-  
 226 enth year, and scatter plots are not introduced until  
 227 the eleventh year of schooling.<sup>18</sup> In addition, informa-  
 228 tion graphics are common in educational materials  
 229 and in the popular media, such as newspapers and



**FIGURE 2.** Analysis of the full papers published at the IEEE VIS 2019 conference (InfoVis and VAST)<sup>17</sup> that included human studies reveals the vast majority of studies are conducted in North America, Western Europe, or China. For three studies the location was not specified. Study participants were almost entirely domain experts or university students or staff. Less than a quarter could be considered to be representative of a broader public.

230 magazines. Thus, most citizens in Western developed  
231 countries have a high level of graphic literacy.

232 This was not always true. In the European Renaissance,  
233 most city maps used a birds-eye view. It was  
234 only later that people became accustomed to the use  
235 of top-down planimetric views. When William Playfair  
236 introduced bar charts and used them to show expendi-  
237 ture, he felt obliged to justify and explain that he  
238 was using a visual metaphor in which the bars in a bar  
239 chart represented piles of guineas. It was only in the  
240 late 20th century that educators began to realize that  
241 graphic literacy was also an important part of general  
242 education.<sup>1</sup>

In many developing countries, poorer people leave  
243 school at an early age and may not receive formal  
244 training in the use of graphics. Reflecting this, popular  
245 media designed for less educated audiences does not  
246 make use of data graphics. Thus, many emergent ICT  
247 users in such countries lack the knowledge or experi-  
248 ence to comprehend and benefit from visualizations.  
249

**EMERGENT USERS OF COMPUTER-MEDIATED VISUALIZATIONS**

250  
251  
252 In India, ICTs, in particular mobile phones, have  
253 reached beyond the traditional tech-savvy English edu-  
254 cated users and have acted as an enabler toward

Population, mobile phone and internet user statistics				Internet users aged 16 to 64 who own each kind of device			
In 2020	Australia (in million)	India (in million)	India (in Australias)	In 2020	Australia (in million)	India (in million)	India (in Australias)
Total Population	25.35	1370	54	Mobile Phone of any type	15.13	815.82	54
Mobile Phone Connections	32.89	1060	32	Smart Phone	14.97	806.85	54
Internet Users	22.31	687.60	31	Non-smart Phone	0.71	125.51	177
Population between age 16 to 64	16.10	896.50	56	Laptop	13.37	555.83	42

**FIGURE 3.** Comparative analysis of India and Australia’s population, mobile phone connections, Internet users, and population between age 16 to 64. We present the data on the number of Internet users aged between 16 to 64 who own mobile phones of any type, smart and nonsmart phones, and laptops. There are many users who own both a smart and nonsmart phone. To highlight the massive difference in size between the two countries, the right-most column shows statistics for India in units of the entire population of Australia.

255 human development at large by reaching new users.  
 256 This is true in many parts of the developing world, for  
 257 example, Avle *et al.*<sup>20</sup> described the effect of mobile  
 258 phone adoption in “the Global South.” Africa shows a  
 259 similar mobile phone adoption trend to India.<sup>23</sup> In India,  
 260 these new users of ICTs include people who may have  
 261 been educated in an Indian vernacular language, work  
 262 in low-income professions like farmers, are small busi-  
 263 ness owners, daily wage laborers, urban poor, and cul-  
 264 turally diverse, and may not have reached college.<sup>4</sup>

265 Today the majority of mobile phone and Internet  
 266 users in India are Indian language users, and in 2020  
 267 this number stands at 688 million (see Figure 3). These  
 268 emergent users are increasingly being exposed to visu-  
 269 alizations through electronic and print media,  
 270 embedded in various mobile apps, political campaigns,  
 271 and through in-match sports visualizations, specifi-  
 272 cally cricket. But such visualizations are not useful  
 273 unless they support these new users with varied visu-  
 274 alization literacy, diversity in culture and language  
 275 which changes every few kilometers, various usage  
 276 contexts like mobile phone as a shared resource in  
 277 the household, and different mental models.<sup>16</sup>

278 Emergent users are not limited to developing  
 279 countries such as India. Developed nations also play  
 280 host to a subset of underconsidered visualization  
 281 users. For instance, Peck *et al.*<sup>21</sup> find personal differ-  
 282 ences in the perception and use of data visualizations by  
 283 residents of rural Pennsylvania (USA). We are also  
 284 aware of differences in individuals’ ability to under-  
 285 stand visualizations in our part of the developed world  
 286 (Australia). While numeracy and graphicacy are now  
 287 formally supported within the Australian education  
 288 system, there are still groups of users who may not  
 289 have had the opportunity to develop meaningful visu-  
 290 alization literacy. Considering the elderly, visualization  
 291 literacy would be expected to have been obtained  
 292 through continued exposure, rather than formal edu-  
 293 cation. Potentially more compromised, immigrants  
 294 and refugees to developed nations such as Australia  
 295 have widely varied education backgrounds as well as  
 296 the possibility of limited exposure to visualizations  
 297 more generally. In Australia, a significant proportion of  
 298 the Australian Indigenous community resides in  
 299 remote places where education opportunities may be  
 300 more restricted, and a comprehensive education in  
 301 numeracy and graphicacy may not be a given. Finally,  
 302 for people with disabilities, education curriculum and  
 303 materials are often tailored to best support the needs  
 304 of the person in question. This again can lead to  
 305 uncertainty regarding the exact nature of the visuali-  
 306 zation literacy they have had the opportunity to  
 307 develop.

308 Although the numbers of affected people in a  
 309 developed nation may be smaller relative to a develop-  
 310 ing nation such as India, assumptions regarding visual-  
 311 izations and their use can still be damaging. Not only is  
 312 there compromised access, but a sense of isolation  
 313 may emerge, of living in an information society and  
 314 not being able to access that information. As such,  
 315 the impact on these affected groups can be of major  
 316 significance.

### 317 *Characteristics of Emergent Users of* 318 *Computer-Mediated Visualization*

319 Considering both developing and developed nations,  
 320 characteristics emerge that may define an emergent  
 321 user. These include:

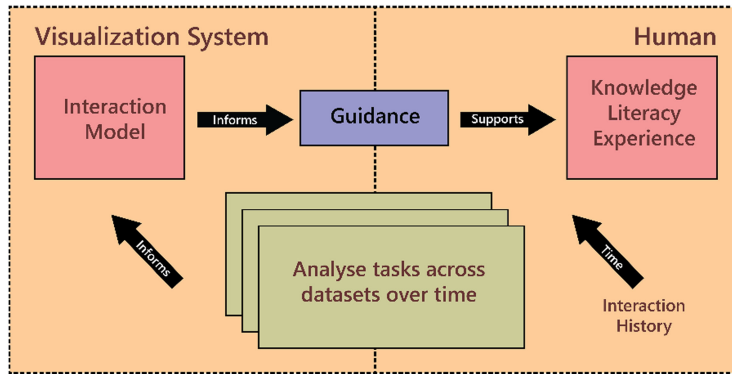
- 322 ▶ lack of education opportunity, in particular with  
 323 relation to numeracy and/or graphicacy;
- 324 ▶ cultural backgrounds “outside the norm” for  
 325 visualizations;
- 326 ▶ limited prior exposure to visualizations;
- 327 ▶ limited access to technology; and
- 328 ▶ diverse specific needs, such as those arising  
 329 from disability.

## 330 SUPPORTING EMERGENT USERS

331 As a visualization community, we need to assume the  
 332 responsibility for designing and developing visual user  
 333 interfaces that are inclusive and accessible to all  
 334 potential end users, not only a select subset. While  
 335 much existing research about low-level visual percep-  
 336 tion may hold across a broad range of end users, we  
 337 argue that higher level considerations such as user  
 338 experiences, task goals, and application contexts may  
 339 diverge between emergent and proficient end users.  
 340 We see several key areas in which visualization  
 341 researchers can drive inclusivity and accessibility of  
 342 visualization for emergent users.

### 343 Education for Emergent Users to 344 Develop Graphical Literacy

345 While not solely being responsible for visualization  
 346 education, we can contribute to graphical literacy  
 347 through dedicated activities such as short courses  
 348 and workshops that target specifically visualization  
 349 education for emergent users. Currently, data visuali-  
 350 zation courses are often done in niche academic com-  
 351 munities (e.g., major conferences such as VIS and CHI)  
 352 and are unlikely to be accessible to people from most  
 353 emergent user backgrounds. Not only is it important  
 354 to teach how to understand graphical conventions



**FIGURE 4.** Guidance systems act as a responsive support to human capability as they interact with a visualization system.<sup>2,3</sup> We propose that guidance systems be informed by a rich longitudinal model of interaction by a particular user or population of users that fit a certain profile. The resulting visualization system will be able to provide guidance, and also potentially adapt styles of visualization or levels of detail, appropriate to their level of visual and numerical literacy, that will develop over time as they learn.

355 but also to discuss ethics and the need to critically  
356 evaluate data sources and presentation choices.<sup>22</sup>

357 **Visualization and User Interface Design**  
358 **Inclusive of Emergent Users**

359 We need to be more open minded in regard to the  
360 questions that we ask ourselves about how visualiza-  
361 tion and user interface design may influence users  
362 that have little to no graphical literacy. We need to  
363 consider a breadth of cultural and educational back-  
364 grounds that may potentially affect people’s interpre-  
365 tation of and interaction with classic data  
366 visualization idioms and user interface paradigms. Is  
367 the way emergent users associate values with  
368 lengths/areas/color/etc. different to experienced  
369 users? Do emergent users prefer different interaction  
370 techniques as compared to experienced users? Dedi-  
371 cated user evaluations inclusive of emergent users  
372 may answer some of these questions and lead us to  
373 more inclusive user models.

374 **Guidance for Emergent Users Capable**  
375 **of Longitudinal Support for a Wider**  
376 **Range of Users**

377 We argue that carefully guiding emergent users  
378 through a visual analysis process is of critical impor-  
379 tance. Dedicated user models of emergent users as  
380 well as adaptive systems that learn from user interac-  
381 tions may play an instrumental role here, but past sug-  
382 gestions of such systems<sup>2,3</sup> (see Figure 4) have  
383 focused on short-term use by domain experts such as  
384 scientists or analysts. Not assuming that emergent  
385 users have the same learning curve as more

experienced users is among many aspects that need  
to be taken into account when we aim to design inclu-  
sive systems. We need to be able to meet emergent  
users at their current skill level and, where appropri-  
ate, encourage them to use more complex visualiza-  
tions with the aim to build graphical literacy and  
interaction skills.

**Ensure Visualization and Interaction**  
**Techniques are Compatible With the**  
**Devices Used by Emergent Users**

While it is certainly appropriate for research to explore  
the capabilities of emerging (and expensive) technolo-  
gies (such as large displays and mixed-reality), we  
must not neglect advancing what can be done with  
“low-end” devices such as mobile phones with small  
screens that are not necessarily either high resolution  
or particularly responsive. Can we do more with less?

**CALL FOR ACTION**

We have charted how visualizations have developed  
as being tools for experts who are graphically literate  
and have access to the latest computer technologies.  
Yet this can be exclusionary of emergent users, or  
indeed, the vast majority of the world’s population. We  
therefore argue that an important next step is to take  
the rich history of visualization work and build upon it  
so it becomes relevant for the masses, including  
underrepresented minorities. Bringing this about  
involves a range of practical steps, that we propose  
the community adopt going forward.



**FIGURE 5.** Geographic distribution of the 76 members of the 2020 InfoVis Conference Program Committee by continent (top-left), by country (bottom), and the continental make-up of the 2010 committee for historical comparison. The continent of Africa (1.216 billion people) is entirely unrepresented, as is India (1.353 billion), Southeast Asia (655 million) or the Middle East (411 million), Eastern Europe (293 million) and—apart from 1 member in Brazil—South America (423 million).

## 415 Identifying Opportunities and Use 416 Cases

417 Ensuring that both underrepresented minorities and  
418 “non-experts” are included requires a focus on identifying the scenarios and purposes for which they might  
419 rely on visualizations. This is an issue that requires  
420 active investigation, to ensure that we do not overlook  
421 unusual or important use cases that apply in the real  
422 world. It also requires being active and ambitious, by  
423 considering novel (and hitherto unidentified) ways in  
424 which visualizations might be used in the future and  
425 ensuring that these opportunities can be rapidly  
426 reacted to as and when they arise (with COVID-19 perhaps being a striking example of this).  
427  
428

## 429 Developing and Evidencing Universal 430 Design Principles

431 Universal design is about ensuring that systems are  
432 usable by a diverse range of people, including those  
433 with disabilities and emergent users. Yet there is no  
434 body of principles that explains how visualizations  
435 can be optimally designed to be fully inclusive of a  
436 wider constituency. These principles will especially  
437 need to address the lack of literacy and numeracy  
438 in many of these communities, as well as ensuring  
439 that inappropriate cultural assumptions are not  
440 made (e.g., color coding can have different interpretations in different societies). Addressing this might  
441 even mean taking a step back to the time of William  
442 Playfair (as mentioned above) and making graphs  
443 inherently more intuitive, rather than assuming any  
444

445 understanding on the part of users establishing this  
446 body of principles and practices is an important and  
447 necessary step for supporting everyone in benefiting  
448 from visualizations.

## 449 Asking the Right Questions (and Doing 450 the Right Evaluations)

451 Visualization research has largely proceeded on the  
452 basis of an assumption of an idealized “expert” user  
453 who is WEIRD (Western, Educated, Industrialized, Rich,  
454 and Democratic). The trouble with this approach is  
455 that it is implicitly biased toward a minority of expert  
456 users, rather than the general population. This means  
457 that many of the presumptions and principles that  
458 have been built up over time are unlikely to generalize  
459 to most users, and may even lead toward biased systems  
460 that are easier to interact with by some groups  
461 compared with others. This is an important issue that  
462 other related academic communities have been grappling  
463 with—perhaps most notably “FATE” (Fairness, Accountability, Transparency, and Ethics) with respect  
464 to Artificial Intelligence (AI)—and with which the visualization  
465 research community should be engaging.  
466

## 467 Engaging More Widely

468 The academic visualization community is not representative  
469 of the global population: while this might be somewhat  
470 improving, the community still remains heavily centered  
471 on North America (see Figure 5). We therefore need to  
472 find a way to involve a wider constituency in charting  
473 the path of visualization research



474 going forward, to ensure that important concerns are  
 475 not overlooked when shaping visualization research  
 476 goals. As a starting point, it would be worth investigat-  
 477 ing the range of barriers that may exclude people from  
 478 underrepresented groups from engaging in our com-  
 479 munity, be they geographic, language related, or dis-  
 480 ability related (a common challenge in academic  
 481 circles<sup>7,10</sup>). We would add that it is not just a matter of  
 482 “balancing” academic committees, but ensuring that  
 483 end users are engaged and providing a full range of  
 484 opportunities for emergent users to become  
 485 stakeholders.

**CONCLUSION**

486 Addressing the issues raised in this article could have  
 487 a profound and positive impact on the future, being  
 488 both transformative to our research but also transfor-  
 489 mative for the lives of emergent users of visualiza-  
 490 tions. The four recommendations set out above, while  
 491 only a beginning, are important first steps for our  
 492 research community once we agree upon the impor-  
 493 tance of serving the entire world rather than a privi-  
 494 leged few. We welcome wider debate in the  
 495 community: this is the start, not the end. At the same  
 496 time, we encourage visualization researchers to con-  
 497 nect to other communities with similar goals, e.g., Fair  
 498 AI and HCI, and see this as an opportunity to put data  
 499 visualization at the forefront of systems used by most  
 500 people.🌍  
 501

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