

1 **Conference demographics and footprint changed by virtual platforms**

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20 **Abstract:**

21 Conferences disseminate research, grow professional networks, and train employees.  
22 Unfortunately, they also contribute to climate change and present significant barriers to  
23 achieving a socially sustainable work environment. Here, we analyze the recent impact of  
24 transforming in-person conferences (IPCs) into virtual conferences (VCs) on improving  
25 diversity, equity, and inclusion (DEI) in science and engineering conferences. Factors including  
26 cost, gender, career stage, and geographic location were evaluated. VCs demonstrated a clearly  
27 discernable and in some case orders of magnitude improvement across nearly all metrics. Based  
28 on participant survey results, this improvement may be attributed to a combination of reduced  
29 financial and personal-life burdens. However, despite this clear impact, further development of  
30 virtual networking features and poster sessions is necessary in order to achieve widespread  
31 adoption and acceptance of this new format.

38 Conferences fulfill a range of needs by facilitating dissemination of ideas, initiating  
39 collaborative relationships, and providing education, training, and career opportunities.  
40 Traditional in-person conferences (IPCs) have filled this role for centuries<sup>1</sup>, and these events cut  
41 across all sectors: academia, industry, and government. However, this format has been criticized  
42 as outdated and detrimental to the environment<sup>2-4</sup>. More recent, emerging evidence is also  
43 connecting this modality to social sustainability issues as well, notably poor retention of a  
44 diverse workforce. In this context, the two dominant contributors are the intrinsic power-  
45 imbalance in the workplace and an imbalance in home-life responsibilities<sup>5, 6</sup>.

46 Over the past two decades, the creation and sustainment of a diverse, equitable and  
47 inclusive (DEI) work environment in the scientific and engineering community has not kept pace  
48 with many other fields. In part, this can be attributed to career expectations revolving around  
49 conference travel and participation. Participation in conferences can be cost prohibitive for  
50 many, as the cumulative expenses can be thousands of dollars per person. International travel  
51 creates additional barriers<sup>7</sup> which are exacerbated by the frequent changes in document  
52 requirements and lengthy delays in obtaining visas. These financial and documentation barriers  
53 can also dissuade scientists that have difficulty securing funding to cover conference costs such  
54 as students, postdocs or scientists from historically underrepresented institutions. These factors  
55 can also exclude participants from countries that do not have very high research activity, such as  
56 nations that are not in the top 10 research countries as defined by the Nature Index (NI)<sup>8</sup>, NI>10.

57 However, even for those researchers who are able to travel, the time away from home  
58 necessitated by work-related travel is intrinsically exclusionary to care-givers, who are primarily  
59 women<sup>3, 7, 9</sup>. Yet, given how important conference attendance is to career advancement, this  
60 community is frequently faced with the decision of choosing between work and family. Lastly,  
61 despite conference organizers' attempts to solve accessibility concerns of the disabled  
62 community, many conferences still fall short of providing an equitable experience.

63 The recent surge in virtual events is forcing the scientific community to re-evaluate its  
64 long-held position against VCs. The initial anecdotal evidence indicated that VCs enabled a more  
65 diverse population to participate. But a quantitative analysis of the impact on DEI challenges has  
66 yet to be performed. Such analysis is critical to make decisions regarding the format of future  
67 events, potentially resulting in a paradigm shift in the field. Here, we evaluate several metrics,  
68 including cost, carbon footprint, impact of conference format, and attendee demographics. We  
69 collected historical data from three US-based IPCs of varying sizes and disciplines within  
70 STEM. These results were compared to the same three conferences after they transitioned to a  
71 VC format in 2020. These scientific conferences were among the first conferences to transition  
72 online in response to the COVID 19 pandemic and were chosen to investigate the impact of an  
73 abrupt transition from historically IPCs to a new virtual format.

74 The historically IPCs-turned-VCs analyzed here are the Annual International  
75 Conferences on Learning Representations (ICLR), the American Astronomical Society (AAS),  
76 and the North American Membrane Society (NAMS) conferences. Also analyzed here are  
77 several conference series that were originally designed for the VC environment including the  
78 Photonics Online Meetups (POM 1: January 2020, POM 2: June 2020) and the International  
79 Water Association Biofilms (IWA) conference. These conferences span five fields of science and  
80 engineering and range from small to large scale events. All have international audiences

81 We focused our analysis on the environmental, social and economic costs of VCs vs IPCs  
82 and accompanying demographic impacts (global participation), participation from women, early

83 career researchers and scientists from underrepresented institutions. We also assessed the  
84 challenges and benefits of the VC format.

## 85 RESULTS

86

87 **Demographic Impact** The elimination of the travel and cost burdens realized with the VC  
88 format resulted in a large increase in attendance at all events (**Figure 1**). The increase in  
89 attendance was particularly pronounced for international attendees. We propose that this trend  
90 may be related to the substantial decrease in costs as compared to IPCs as described below.

91 The cost of attending IPCs for international attendees was dominated by airfare (**Figure 1**  
92 **and Table S1, S2**). When compared to US attendees, the average researcher from Africa, Asia,  
93 Europe, the Middle East, and Oceania paid between 90% to 210% more to attend NAMS IPCs  
94 (**Table S3**). When placed in financial context, the cost of attendance for scientists from Africa to  
95 past ICLR (2018-2019), AAS (2016-2019), and NAMS (2015-2019) IPCs was on average  
96 between 80% to 250% of their country's annual per capita gross domestic product (GDP),  
97 compared to approximately 3% of per capita GDP for US participants (**Figure 1c and Table S4**).  
98 Cost of attendance for participants from Asia to past ICLR (2018-2019), AAS (2016-2019), and  
99 NAMS (2015-2019) IPCs was approximately 15% of their country's per capita GDP (**Figure 1c**  
100 **and Table S4**). However, it is important to note that many conferences not included in this  
101 analysis have registration fees in excess of \$700. For these events, registration fees can begin to  
102 compete with airfare as a significant contributing financial consideration.

103 The 2020 ICLR, AAS and NAMS VC delegations were more geographically diverse,  
104 likely due to the elimination of these travel and registration costs as seen from responses to our  
105 surveys (Supplementary Information, SI). Notably, the audiences were approximately 40% to  
106 120% larger than the historical average for IPCs (**Table S5**). Attendance by scientists from  
107 NI>10 countries increased significantly from the historical average at ICLR, AAS, and NAMS  
108 IPCs to the 2020 ICLR, AAS, and NAMS VCs (**Figure 1d and Table S6**). The increased  
109 representation was more comparable to delegations seen at conferences originally designed for  
110 the virtual environment; specifically, 31% to 38% of attendees at the POM 1, POM 2, and IWA  
111 VCs were from NI>10 countries (**Figure S1**).

112 In this context, the environmental impact of international conferences can also be  
113 considered. In a collection of decarbonization pathways designed to limit global warming to 1.5°  
114 C with a small overshoot, the median global per capita carbon budget for the entire year of 2030  
115 was 3.26 tonnes of CO<sub>2</sub> equivalents (CO<sub>2</sub>e)<sup>10</sup>. The carbon footprint for a single international  
116 attendee to the 2019 ICLR, AAS, or NAMS IPCs approached this value. Conversely, the  
117 cumulative footprints of the more than 7000 attendees to 2020 ICLR, AAS, and NAMS VCs  
118 (1.07 tonnes CO<sub>2</sub>e) was comparable to the average footprint of a single attendee (combined  
119 average of domestic and international) to one of the analyzed 2019 IPCs as shown in **Figure S2**  
120 **and Table S7**, and discussed further in the SI.

121

122 **Participation of Women** The VC format also eliminated travel burdens that can act as a barrier  
123 to attendance for certain sociodemographic groups. This impact is likely reflected in changes in  
124 the gender makeup of VC delegations (**Table S8**) and supported by survey responses to a follow  
125 up survey sent separately to men and women attendees of NAMS 2020 (**Table S9**). Attendance  
126 by women increased between 60% to 260% at ICLR, AAS, and NAMS VCs compared to the  
127 IPC baselines (**Figure 2 and Table S10**). On average, women made up larger fractions of the

128 conference delegations at 2020 VCs as compared to IPCs (**Figure 2g and Table S11**). The  
129 increase in the number of female attendees is particularly significant considering that women  
130 make-up smaller portions of STEM fields compared to men. For example, women comprise only  
131 33% to 34% of STEM researchers in the countries that made up the delegations for historical  
132 ICLR, AAS, and NAMS IPCs (**Table S12, S13, S14**). Survey responses confirmed that the  
133 elimination of the travel requirement realized with VCs partially explain trends in attendance by  
134 gender. For example, about half (47%) of the 2020 NAMS VC survey respondents that did not  
135 plan on attending the originally scheduled 2020 NAMS IPC indicated that the primary reason for  
136 attending the VC was convenience (**Figure S3**).

137 Abstract submittals to the 2020 NAMS conference from before and after the decision to  
138 switch from an in-person to a virtual format also indicated an increase in interest and  
139 participation from female researchers for the VC. Approximately a quarter (26%) of abstracts  
140 submitted to the 2020 NAMS IPC were from female researchers, which was aligned with  
141 historical average attendance by women to 2015-2019 NAMS IPCs (**Figure S4**). After it was  
142 announced that the 2020 NAMS conference would be held online, 37% of submitted abstracts  
143 came from female scientists (**Figure S4**). The 2020 ICLR VC also saw an increase in attendance  
144 from gender queer and transexual scientists. On average, 2018-2019 ICLR IPCs were attended  
145 by 1 gender queer scientist and 0 transgender scientists. The 2020 ICLR VC was attended by 8  
146 gender queer scientists and 2 transgender scientists (**Figure 2a**). However, it should be noted that  
147 this increase in reported attendance by LGBTQ scientists could be the result of an increased  
148 willingness to identify as LGBTQ.

149  
150 **Participation of Students and Postdoctoral Researchers** High costs characteristic to IPCs can  
151 also be exclusionary to certain sociodemographic groups that may face challenges securing  
152 funding for travel, such as students and postdoctoral researchers. Cost of attendance to historical  
153 NAMS IPCs was on average \$1612 for students and \$2142 for postdocs. The shift to a virtual  
154 environment resulted in a significant growth in this population of attendees (**Figure 3a-c and**  
155 **Table S15**). Additionally, on average, for all conferences evaluated, the VC delegations had  
156 higher proportions of students (29% to 42%) and postdoctoral researchers (5% to 11%)  
157 compared to historical IPCs (**Figure 3d and Table S16, S17**). Additionally, the audiences of  
158 conferences designed for the virtual environment (POM 1, POM 2, and IWA) were all comprised  
159 of over 45% students and post-doctoral scholars, demonstrating the impact that virtual events can  
160 have on the careers of emerging scholars (**Figure S5**). The AAS conference surprisingly did not  
161 show much change in conference composition as seen from surveys (32% completion) (**Figure**  
162 **S6**). The role of cost on attendance was evident in survey responses, as 33% of respondents to  
163 NAMS surveys indicated that they were not planning on attending the scheduled 2020 NAMS  
164 IPC prior to the decision to move online (**Figure S7**). Of the respondents that were not planning  
165 on attending, 34% indicated that cost was the primary motivation for attending the 2020 NAMS  
166 VC (**Figure S3**).

167  
168 **Participation of Historically Underrepresented Institutions** A unique and particularly  
169 challenging subset of researchers to engage are those from Primarily Undergraduate Institutions  
170 (PUIs) and High Research Activity (R2) Universities (as distinct from the Very High Research  
171 Activity Category – R1). Attendance from both groups increased at VCs. At the 2020 NAMS  
172 VC, attendance by researchers from PUIs and R2 Universities increased from the IPC baseline  
173 by 157% and 45%, respectively. Similarly, attendance at the 2020 AAS VC from PUIs and R2

174 Universities increased by 72% and 106%, respectively (**Figure 3e and Table S18**). Increasing  
175 participation of researchers from these historically excluded institutions will enhance their  
176 educational experiences and provide more research opportunities. Additionally, attending  
177 technical events will provide students with mentoring and networking opportunities, potentially  
178 increasing the likelihood that they pursue graduate degrees.

179  
180 **Effect of Time Zones and Conference Format** While VCs may eliminate many barriers to  
181 participation, the impact on international attendances seen in this work was strongly dependent  
182 on the VC format (**Figure S8 and Table S19**) with the primary variations being synchronous,  
183 asynchronous, or blended (both options available) content delivery. The 2020 NAMS VC was  
184 organized around synchronous live talks. Consequently, attendance from regions where the  
185 conference was held during normal work hours was significantly higher than in other regions. As  
186 a result, attendance from Europe and the Middle East increased by 102% and 76%, respectively,  
187 when compared to the 2015-2019 NAMS IPC average. Conversely, for Asia, where the 2020  
188 NAMS VC was held around or past midnight local time, attendance decreased by 62%. In the  
189 case of the 2020 AAS VC which was also synchronous, attendance increased for all regions  
190 compared to AAS IPCs (60% to 700% increase), and the largest percent increases came from  
191 Europe, Oceania, and Other Americas. Therefore, the dependence on working hours was not  
192 universally observed. However, it is important to note that some regions had very small  
193 participant numbers which could influence the analysis.

194 The 2020 ICLR VC was asynchronous, with only a few live events and most talks pre-  
195 recorded and released for consumption at the attendee's leisure. A live Q&A session was held for  
196 each keynote speaker after the video had been available for some time, thus affording the  
197 opportunity to interact with the speaker. As a result of this format, attendance at the 2020 ICLR  
198 VC increased for all regions (57% to 1700% increase), when compared to the 2018-2019 ICLR  
199 IPC average. Additionally, unlike the AAS and NAMS conferences, over 50 people attended the  
200 2020 ICLR VC from every region in the world, increasing confidence in the analysis. Based on  
201 these results, it is clear that to take full advantage of the virtual format and to make these events  
202 effective at disseminating science, it is necessary to offer content asynchronously or using a  
203 blended format. A similar blended approach was used by the IWA VC. At IWA, pre-recorded  
204 presentations were released at a specified time and presenters were available to answer questions  
205 during and after the video presentation.

206  
207 **Initial Attendee Perceptions of Virtual Conferences** The VC format, in general, was well  
208 received by attendees and helped to shift negative perceptions to more positive views towards  
209 this format. No major alterations in the type of content presented was observed between IPCs  
210 and VCs, as discussed in the SI. Attendees to 2020 VCs indicated via pre-conference surveys  
211 that they were initially skeptical about the efficacy of VC components, but overall felt that the  
212 format could possibly improve IPCs in some ways. When asked what they foresaw as the biggest  
213 challenge with the virtual format, networking and social interaction was the most common  
214 response for NAMS surveys (42% of respondents) and POM 2 surveys (25% of respondents)  
215 (**Figure S9**). Aversion to engaging with the virtual format was lowest among students, as  
216 indicated by the fact that only 25% of graduate students and no undergraduate students who  
217 submitted abstracts to the 2020 NAMS IPC elected to withdraw from the conference once it was  
218 moved online. Conversely, 37% of industry personnel and 39% of postdoctoral researchers who  
219 applied to the 2020 NAMS IPC elected not to attend the 2020 NAMS VC (**Figure S11**). NAMS

220 survey respondents indicated that they were looking forward to some aspects of the virtual  
221 format, particularly the opportunity to seamlessly transition between sessions and quickly access  
222 the internet to research unfamiliar concepts that arose during the conference.

223 Part of the success realized by VCs is related to the wide range of currently available  
224 virtual environments for hosting oral sessions. Oral sessions at analyzed conferences were either  
225 livestreamed via webinar (synchronous format) (**Figure S12**) or pre-recorded and released at a  
226 specified time (asynchronous format). They were popular among attendees, with 43% of NAMS  
227 survey respondents and 74% of POM 2 survey respondents indicating that they preferred the  
228 virtual format for oral sessions over the in-person format (**Figure S13 and S14**). Some  
229 presentations and Q&A sessions were recorded and made available indefinitely, eliciting  
230 persistent viewing after the conference ended. The ICLR platform drew 652,087 total pageviews  
231 during the scheduled conference days, and then views increased again by 74% (481,092  
232 additional views) in the three months following the conference, indicating a significant increase  
233 in exposure for presenters and sponsors compared to the in-person format (**Figure S15**).

234 Analyzed VCs had poster authors publish their posters via twitter, using a web-based  
235 iPoster sharing platform, or by uploading a 5-minute pre-recorded presentation to the conference  
236 website. The poster presentations had high view counts (NAMS iPosters had on average 142  
237 views) (**Figure S16**), but presenters could not tell how many attendees were viewing their  
238 posters and features for communicating with poster viewers were not effective. In contrast,  
239 Twitter-based poster sessions are increasing in frequency and allow asynchronous  
240 communication. However, Twitter is not available in every country, limiting access.  
241 Consequently, virtual posters were less popular, with 85% of NAMS survey respondents and  
242 43% of POM 2 survey respondents indicating that they preferred in-person poster sessions to  
243 virtual poster sessions (**S13 and S14**).

244 Analyzed VCs attempted to facilitate networking by employing a variety of social media,  
245 messaging, video chat, and virtual reality features (**Table S20**). However, survey respondents  
246 indicated that the interactions felt inauthentic and contrived. As a result, 75% of POM 2 survey  
247 respondents and 96% of NAMS survey respondents indicated that they preferred in-person  
248 networking to virtual networking (**Figure S13 and S14**). In response to this feedback, VCs that  
249 occurred later in 2020 and in early 2021 took advantage of improvements in virtual networking  
250 technology. These features included robust central chat and discussion board features, as well as  
251 Gather.town, an app that allowed participants to navigate a virtual room with an avatar and video  
252 chat with other avatars in close proximity. The January 2021 POM used Gather.town to hold a  
253 virtual job fair and poster session among other networking events. Gather.town was also used at  
254 the 2020 IWA VC and was popular with attendees, as all 56 survey respondents indicated that  
255 they would like the Gather.town Interactive Lounge feature to be included in future IWA VCs.

256

## 257 **DISCUSSION**

258 Our findings reveal that VCs reduce the environmental impact of conferences, the  
259 financial burden, and the social cost. In the VC format, researchers are much more likely to be  
260 able to overcome economic and travel related barriers that are intrinsic to IPCs and that  
261 ultimately discourage participation from institutions and countries with limited resources,  
262 women, disabled scientists, and early career researchers and practitioners (e.g., students,  
263 postdocs). These factors are discussed further in the SI. Thus, virtual formats can provide an  
264 excellent avenue to address DEI challenges stemming from barriers to participation and

265 representation at IPCs and other professional events. However, despite these clear benefits, the  
266 difficulties networking in a virtual environment are routinely emphasized as a limitation.

267 Seventy-five percent of POM 2 survey respondents and 96% of NAMS survey  
268 respondents indicated that they preferred in-person networking to virtual networking (**Figure**  
269 **S13 and S14**). Analyzed VCs experimented with incorporating social media and organizing  
270 virtual breakout rooms to facilitate networking with some success. However, survey respondents  
271 indicated that the interactions felt inauthentic and contrived. Therefore, while virtual networking  
272 technology has improved considerably, there is substantial need for further development of these  
273 features as well as research into their efficacy.

274 One approach to overcome this challenge and increase in-person interaction without  
275 increasing cost or travel was piloted during POM 1 by creating locally organized viewing and  
276 networking sites (POM-hubs). This “conference within a conference” approach allowed for  
277 reduced cost and travel, increased local and regional networking, and created an international  
278 conference. Notably, approximately half of the POM 1 attendees participated in the conference  
279 from a local hub-site.<sup>1</sup> This hybrid hub approach pioneered by POM 1<sup>1</sup> is a promising solution to  
280 this challenge that warrants further study. A hybrid format could allow communities to realize  
281 many of the advantages identified by this analysis of COVID VCs, while still offering the option  
282 of a traditional IPC experience. It would be ideal for post-pandemic conferences to utilize the  
283 rich knowledge gained on the benefits of expanding inclusion using virtual tools. The resultant  
284 conferences could facilitate networking and effective dissemination of scientific knowledge to  
285 diverse audiences in an environmentally sustainable manner, moving toward more equitable  
286 environments and opportunities. Innovative VC strategies and platforms used to administer oral  
287 and poster sessions and virtual networking are further discussed in the SI along with additional  
288 discussion on organization recommendations.

289 Our study is characterized by one important limitation. While nearly all interactions made  
290 the abrupt shift from in-person to virtual, our analysis is focused on STEM subjects. In some  
291 ways, the demographic and financial sensitivities of this population are distinct from other  
292 academic communities or an industry or government audience. However, they do share several  
293 similarities, particularly for global industry consortiums. Notably, all groups are sensitive to  
294 international politics and visa policies, fluctuations in currencies and the financial markets, and  
295 gender inequities. However, the attendees at scientific events tend to be highly educated (BS  
296 degree or higher in a STEM field) and speak English as a primary or secondary language. These  
297 limits do not adversely affect our conclusions, as we are focusing on STEM. However, to extend  
298 our conclusions outside of higher education and STEM fields specifically, a broader population  
299 analysis should be performed with appropriate benchmarking. Such an analysis will require  
300 engaging conference organizers in other areas including humanities, commerce, business as well  
301 as related industry, nonprofits, and government organizations.

302 In addition to extending the analysis outside of STEM, the present research findings  
303 motivate several new areas of investigation. A few examples include: (1) developing strategies  
304 for improving virtual networking, (2) role of organization type on the impact of travel (small vs.  
305 large business, domestic vs. global), (3) policy development by technical/scientific societies,  
306 funding agencies, and universities, and (4) longitudinal study tracking travel and career  
307 progression. These topics are discussed further in the SI. In this context, we consider the present  
308 conclusions to be a significant first step in understanding the positive impact of VCs, paving the  
309 way for future policy decisions and reducing DEI challenges in the workplace.

310

## 311 **METHODS**

### 312 **Data**

313 Registration, digital platform and survey data were collected from three IPCs-turned-VCs  
314 and are presented in **Table S21**. The three analyzed IPCs-turned-VCs include the Annual ICLR  
315 (~2300 historical average attendees), the AAS Summer Meeting (~700 historical average  
316 attendees), and the NAMS Annual Conference (~450 historical average attendees).  
317 Complementing this is data from the POMs (~1000 attendees) and the IWA conference (~350  
318 attendees), conference series that were specifically designed for the virtual ecosystem. These  
319 conferences represent varying fields and community sizes and allow for comparisons across a  
320 range of STEM backgrounds. Data for IPCs-turned-VCs were collected for 2020 VCs and for  
321 historical IPCs. POM and IWA data provided a control for an always VC, while the baseline data  
322 for historically IPCs allowed for the elimination of effects from other variables, facilitating direct  
323 analysis of the impact that virtual components had on conference performance.

324 Specific data collected include registration and abstract information, spanning information  
325 such as the number and type of participants (e.g., students, industry personnel), geographic  
326 participation, institution, and gender. For IPCs-turned-VCs, this data was collected for  
327 registrations accrued before and after moving online. Carbon footprint and cost of attendance  
328 were estimated based on attendee work locations and conference destinations. Descriptive  
329 statistics<sup>11</sup> and thematic mapping<sup>12</sup> were applied to understand changing sociodemographics  
330 realized in the shift to a virtual format. Additional data collected on webinar attendance and  
331 virtual platform activity were used to assess the efficacy with which the VCs distributed content  
332 to attendees. Qualitative data was collected by asking participants to fill out polls as well as pre-  
333 and-post conference surveys designed to interrogate the participant experience and field  
334 suggestions for improvement. Surveys were also used to investigate the impact of travel burden  
335 and cost barriers for female versus male NAMS attendees. Survey questions included multiple  
336 choice and open-ended questions about specific conference components and the participant  
337 experience. The surveys were produced by the authors for the conferences that they organized.  
338 Survey and polling questions underwent IRB review receiving and exempt status (Protocol 2020-  
339 05-0026) at the University of Texas at Austin.

340 Sociodemographic data was provided by conference organizers and filled in as necessary.  
341 Attendee countries were manually categorized by region for analysis. Job type data (i.e. Graduate  
342 Student, Industry Personnel) was provided by conference organizers via registration or survey  
343 data. Data that included specific job titles (i.e. Operations Director, Research Scientist) for  
344 attendees were categorized manually by job type. Gender data was provided by organizers for  
345 some conferences via voluntary surveys. Gender data for the NAMS conference was manually  
346 assigned based on author familiarity with the participants and through internet search of attendee  
347 names. The Gender API<sup>13</sup> was also employed to assign gender to attendee names for NAMS and  
348 AAS conference attendees. Due to confidence in the accuracy of manually assigned names for  
349 NAMS attendees, discrepancies in the genders assigned to NAMS attendees by the manual  
350 process and the Gender API indicated that the Gender API was less accurate than the manual  
351 process (**Table S8**). Consequently, the Gender API was only applied to assign gender to AAS  
352 participants. Attendee academic institutions were manually categorized according to databases of  
353 institution types. Minority Serving Institutions were defined according to the 2007 U.S.  
354 Department of Education database<sup>14</sup>. High Research Institutions (R2) were defined as any  
355 institution that was included in the 2018 Carnegie Classification of R2 Universities<sup>15</sup>. Primarily  
356 Undergraduate Institutions (PUI) were defined as any university that awarded 20 or fewer PhD

357 degrees in NSF-supported fields during the combined previous two academic years<sup>16</sup> as reported  
358 by the U.S. National Science Foundation (NSF) records on PhD degrees for major science and  
359 engineering fields awarded by universities during 2017 and 2018<sup>17, 18</sup>. Non-research-intensive  
360 countries were defined as countries that were not in the top 10 countries for scientific research as  
361 defined by the Nature Index that measured top countries in terms of contributions to papers  
362 published in 82 leading journals during 2019 (NI>10)<sup>19</sup>.

363

### 364 **Travel Distance**

365 Attendee travel distance, carbon footprint, and cost were calculated via python scripts using  
366 attendee origin location data provided by conference organizers. NAMS and AAS registrant  
367 origin locations were provided by organizers via registration data as a list of attendees with  
368 attendee-specific locations. If location for an attendee was not included, origin location was  
369 determined via internet search of the attendee name. ICLR and POM registrant origin location  
370 data was provided by conference organizers and comprised a list of countries in attendance and  
371 the number of attendees from each country. While the sample size of data for single ICLR  
372 conferences varied by data type (i.e. origin country, gender, job title), origin country was the  
373 largest dataset for all ICLR conferences, and was thus assumed to be the true size of the  
374 conference delegations.

375 Conference city and attendee origin coordinates were determined by querying the Google  
376 Maps API<sup>20</sup> with the location names. If a city-specific attendee origin was not recognized by the  
377 API, the attendee origin was set to the attendee's origin country name. Google Maps API queries  
378 of only country name return coordinates for the geographical center of the country. Travel  
379 distance between attendee origin and conference location were calculated as the great circle  
380 distance (great\_circle python package).

381

### 382 **Carbon Footprint of Attendance**

383 The carbon footprint of conference attendees was calculated for all IPCs-turned-VCs as  
384 the cumulative emissions associated with the flight and hotel stay. The air travel carbon footprint  
385 was calculated according to the methodology for the myclimate air travel emissions calculator<sup>21</sup>.  
386 The myclimate calculator computes air travel footprint by adding 95 km to the great circle  
387 distance to account for flightpath inefficiencies and calculating greenhouse gas (GHG) emissions  
388 associated with the fuel burn and life cycle footprint of the airplane and associated aviation  
389 infrastructure. The GHG emissions are then converted to CO<sub>2e</sub>. It was assumed that all  
390 conference attendees flew economy class. If city-specific attendee origin data was available and  
391 the attendee was local (<= 100 km from the conference city) it was assumed that the attendee did  
392 not fly to the conference city, and their travel CO<sub>2e</sub> was set to 0. If registrant origin coordinates  
393 were not found, the attendee travel distance and travel footprint were set to the average for that  
394 conference.

395 The carbon footprint per night for the attendee hotel stay was determined using the Hotel  
396 Carbon Measurement Initiative (HCMI) rooms footprint per occupied room from the Hotel  
397 Sustainability Benchmarking Tool published by the Cornell Center of Hospitality Research<sup>22</sup>.  
398 The tool provides city-specific and country-specific footprint data. If data was not available in  
399 the Hotel Sustainability Benchmarking Tool for the conference city, then the footprint per night  
400 was set to the country average in the tool. If no data was available for the country in which the  
401 conference was held, the footprint was set to the value that was closest to the conference location  
402 geographically. Student hotel footprint calculations were adjusted to assume shared hotel rooms,

403 i.e. footprint per night was divided by two. If attendee specific job title (student vs. non-student)  
404 information was not available, percent students as defined by the voluntary survey data was  
405 multiplied by the number of attendees from each country to estimate the number of students from  
406 each country. When computing total hotel footprint, it was assumed that attendees stayed for all  
407 but one night of the conference (i.e. for a four-day conference, nightly hotel footprint was  
408 multiplied by 3). If the attendee was local, the hotel footprint was set to 0. If the attendee origin  
409 was not near the conference city and their job title (student vs. non-student) was not known, the  
410 attendee hotel footprint was set to the conference average.

411

### 412 **Cost of Attendance**

413 Cost of attendance for individual attendees was computed for historically IPCs-turned-  
414 VCs by calculating their cost of travel based on air travel distance and summing with the  
415 estimated cost of the hotel, food, and conference registration fees. Travel cost was calculated as  
416 the one-way air travel distance multiplied by the cost distance for air travel defined in <sup>23</sup>, and  
417 doubled to represent the cost of a round trip flight. If the registrant was local, their travel cost  
418 was set to 0. If the registrant origin was not known, the travel cost was set to the average  
419 conference travel distance and converted to cost using<sup>23</sup>. To account for a potential overestimate  
420 of travel cost, a sensitivity analysis where the one-way flight cost is multiplied by 1.5 instead of  
421 2 was conducted and is presented in **Table S1**.

422 NAMS hotel cost was taken from NAMS records. 2020 ICLR hotel cost was set to the  
423 average of hotel options provided by the ICLR website. For 2018-2019 ICLR and all AAS  
424 conferences, the cost of U.S. hotels was set to the U.S. General Services Administration lodging  
425 max per diem for the conference city. For 2018-2019 ICLR the cost of all hotels outside of the  
426 United States was set to the U.S. State Department lodging max per diem for the conference city.  
427 Nightly hotel costs were divided by two for students to assume shared rooms. If attendee specific  
428 job title (student vs. non-student) information was not available, percent students as defined by  
429 the voluntary survey data was multiplied by the number of attendees from each country to  
430 estimate the number of students from each country. ICLR 2020 student hotel cost data was taken  
431 from “double room rate” and ICLR 2020 non-student hotel cost data was taken from the “single  
432 room rate” cost on the ICLR website. When computing total hotel cost, it was assumed that  
433 attendees stayed for all but one night of the conference (i.e. for a four-day conference, nightly  
434 hotel cost was multiplied by 3). If the attendee was local, the hotel cost was set to 0. If the  
435 attendee was not local, but their job title (student vs. non-student) was not known, the hotel cost  
436 was set to the conference average.

437 Food cost for conferences held in U.S. cities was taken from U.S. General Services  
438 Administration city-specific per diem rates for breakfast, lunch, and dinner. For NAMS, one  
439 dinner is subtracted from the total cost to account for the banquet dinner provided by NAMS.  
440 Food cost for conference cities outside of the U.S. was taken from U.S. State Department city-  
441 specific Meals and Incidental Expenses (M & EI) per diem. Attendees were assumed to stay for  
442 all but one night of the conference. If the attendee was local, food cost was set to 0. If the  
443 attendee origin was not known, the food cost was set to the conference average.

444 Registration costs for historical NAMS IPCs was set to the recorded registration fee per  
445 registrant. Fees for the sponsor and exhibitor registration types, where sponsors made their  
446 contributions via the registration fee, at historical NAMS conferences were set to conference  
447 average of that year (these registration types are excluded from the average).

448 Hypothetical registration fees for a 2020 NAMS IPC were assigned to attendees to the  
449 2020 NAMS VC. 2020 NAMS attendees with Registrant Type “Student” were assigned a  
450 hypothetical 2020 NAMS IPC registration fee equal to the average fee for students at the 2015-  
451 2019 NAMS IPCs (average based on Title Category, with “Unknown/Other” title category  
452 excluded from the average). 2020 NAMS VC attendees with Registrant Type  
453 “Professional/Academic” were assigned a registration fee equal to the average fee for non-  
454 students at the 2015-2019 NAMS IPCs (average based on Title Category, “Unknown/Other”  
455 excluded).

456 Student and non-student registration fees for 2018-2019 ICLR IPCs were set to early  
457 registration fees from the conference website. The registration fees for the 2020 ICLR VC were  
458 set to the 2018-2019 ICLR IPC average fees. As attendee specific job title (student vs. non-  
459 student) information was not available, percent students as defined by the voluntary survey data  
460 was multiplied by the number of attendees from each country to estimate the number of students  
461 from each country (i.e. Total student registration fees by country = % students from job title data  
462 \* total attendees from country \* student registration fee).

463 Registration fees for 2016-2019 AAS IPCs were set to the early registration fees for “Full  
464 Member / Educator / International Affiliate”, “Graduate Student Member”, “Undergraduate  
465 Student Member”, “Emeritus Member”, and “Amateur Affiliate” from the 2020 Winter Meeting  
466 website. As attendee specific job title information was not available, percentages on attendee job  
467 title as defined by the voluntary survey data was multiplied by the number of attendees to  
468 estimate the number of each job type in attendance. The total registration fee for each conference  
469 was calculated accordingly. The total registration fees were then divided by the number of  
470 attendees and the average registration fee was assigned to each registrant.

471 VC registration fees for ICLR and NAMS were set to \$50 for students and \$100 for non-  
472 students. VC registration fees for AAS were set to the full meeting fees for “Full Member / LAD  
473 Member”, “Graduate Student”, “Undergraduate Student / High School Student”, “Emeritus  
474 Member”, and “Amateur Affiliate” from the 2020 VC website.

475

## 476 **World Map Figures**

477 Attendee origin coordinates and conference city coordinates were converted to great  
478 circle distance paths and saved in .kml files using the lxml and geographiclib.geodesic python  
479 packages. World maps were plotted using Tableau and MapBox.

480

## 481 **Global Annual per Capita Carbon Budget for 2030 and 2050**

482 Median global carbon budget calculated in terms of Kyoto GHG as CO<sub>2</sub>e for 2030 and  
483 2050 were taken from a set of decarbonization pathways as outlined in the Intergovernmental  
484 Panel on Climate Change report on Mitigation Pathways Compatible with 1.5° C in the Context  
485 of Sustainable Development<sup>10</sup>. The global carbon budget was divided by the medium variant of  
486 global population projections for 2030 and 2050 produced by the United Nations Department of  
487 Economic and Social Affairs<sup>24</sup>.

488

## 489 **Car Travel Footprint**

490 Car travel footprint per mile was taken from U.S. EPA estimates for average passenger  
491 vehicles<sup>25</sup>.

492

## 493 **Virtual Conference Carbon Footprint**

494 VC footprints were estimated based on emissions for YouTube video streaming  
495 multiplied by the projected duration of conference webinar and video streaming by attendees<sup>4</sup>.  
496

#### 497 **Regional Average Cost/Regional Per Capita GDP**

498 Country specific GDP per capita was defined as the 2019 GDP per capita in the attendee  
499 country's national currency converted to USD and divided by the total country population as  
500 calculated in the World Economic Outlook Database<sup>26</sup>. Total representative GDP per capita for  
501 conference attendees from each region was calculated as the sum of GDP per capita for all the  
502 countries in each region multiplied by the number of conference attendees from each country in  
503 the region. Total cost of attendance for each region was calculated as the sum of the cost of  
504 attendance for all the participants from each region. The regional average cost divided by the  
505 regional per capita GDP was calculated by dividing the total cost of attendance for all the  
506 attendees from each region by the total representative GDP for the attendees from each region.  
507

#### 508 **Gender Makeup of STEM Researchers from Conference Attendee's Countries**

509 Country-specific percent women data is taken from "Female researchers as a percentage  
510 of total researchers (Full-Time Equivalents) – Natural sciences and engineering (sub-total)"  
511 published by<sup>27</sup> with the exception of the US which is not included in that dataset. US percent  
512 women is derived from women as a percent of MS and PhD graduates employed in Science and  
513 Engineering occupations<sup>28</sup>. Overall percent women in STEM for the countries represented in the  
514 conference delegations was calculated with percent values from each country represented at the  
515 conference, weighted by the number of attendees from each country.  
516

#### 517 **Data availability**

518 The data that support the plots within this paper and other findings of this study have  
519 been deposited on Github<sup>29</sup> (DOI: [doi.org/10.5281/zenodo.5567764](https://doi.org/10.5281/zenodo.5567764)). Source data files for main  
520 text figures are also available.  
521

#### 522 **Code availability**

523 The custom code used to process and analyze the data for this study has been deposited on  
524 Github<sup>29</sup> (DOI: [doi.org/10.5281/zenodo.5567764](https://doi.org/10.5281/zenodo.5567764)).  
525

#### 526 **Additional information**

527 Correspondence and requests for materials should be addressed to M.K.  
528

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541

#### 542 **Contributions**

543 M.K., K.F and M.S. conceived the idea. M.K and M.S. collected data. M.S. and E.Y. analyzed  
544 data. O.R., M.L.L, P.P.C, R.N., A.R., and A.A. provided access to data and provided insights on  
545 data. M.K., K.F., A.A. and M.S. wrote the manuscript.

546

#### 547 **Competing interests**

548 The authors the following competing interests: M.K. and M.L.L. were organizers of NAMS.  
549 A.R. was an organizer of ICLR. P.P.C. and R.N. were organizers of IWA. A.A. and O.R. were  
550 organizers of POM.

551 **Fig 1 | VCs increase overall attendance and geographical diversity while reducing costs.** (a) The  
552 delegation for the 2019 ICLR IPC located in the US was global but concentrated in the United States  
553 (n=2584). (b) The delegation for the 2020 ICLR, which was originally scheduled to occur in Ethiopia but  
554 transitioned online, was larger (n=4980) and more geographically diverse. (c) Regional average cost of  
555 attendance to IPCs as a percent of attendee country's GDP per capita for ICLR (n=2), AAS (n=4), and  
556 NAMS (n=4) conferences was significantly higher for African participants, and very low for US  
557 participants. \*Error bars are not included for AAS Middle East because n<3. (d) The delegations for 2020  
558 ICLR (n=1), AAS (n=1), and NAMS (n=1) VCs generally represented more countries that were not in the  
559 top ten research countries as defined by the Nature Index<sup>19</sup> (NI>10) and included a higher number of  
560 attendees from those countries compared to the average delegations from IPCs. (e) Average registration,  
561 food, hotel, and travel costs for a single attendee to past ICLR (n=2), AAS (n=4), and NAMS (n=4) IPCs  
562 totaled thousands of USD, compared to less than 200 USD for 2020 ICLR (n=1), AAS (n=1), and NAMS  
563 (n=1) VCs. Error bars are the propagated uncertainty for Food, Registration, Hotel, and Travel costs.  
564 \*Error bars in all panels are defined as standard deviation and are not included for ICLR IPC data  
565 because n<3.

566

567 **Fig. 2 | VCs increase gender diversity.** (a) The 2020 ICLR VC (n=1) was attended by more scientists of  
568 all genders compared to the 2018-2019 ICLR IPCs (n=2). (b) A positive percent change in attendance for  
569 all genders was observed between the 2018-2019 ICLR IPCs and the 2020 ICLR VC, with the highest  
570 percent increase in attendance observed for Gender Queer scientists and scientists that identified as a  
571 gender that was not included in the survey. (c) The 2020 AAS VC (n=1) was attended by more male and  
572 female scientists compared to the 2016-2019 AAS IPCs (n=4). (d) A positive percent change in  
573 attendance for males and females was observed between the 2016-2019 AAS IPCs and the 2020 AAS  
574 VC, with a larger percent increase for female scientists. (e) The 2020 NAMS VC (n=1) was attended by  
575 more male and female scientists compared to the 2015-2019 NAMS IPCs (n=4). (f) A positive percent  
576 change in attendance for males and females was observed between the 2015-2019 NAMS IPCs and the  
577 2020 NAMS VC, with a larger percent increase for female scientists. (g) The female fractions of the  
578 delegations at the ICLR (n=1), AAS (n=1), and NAMS (n=1) VCs were larger than at historical ICLR (n=2),  
579 AAS (n=4), and NAMS (n=4) IPCs and were more comparable to the delegation-specific STEM average,  
580 with female fractions in STEM calculated as a weighted average of females in STEM for the origin  
581 countries of conference attendees<sup>27, 28</sup>. For panels a and b: Female=Red, Gender Queer=Orange,  
582 Male=Yellow, Other=Green, Prefer not to answer=Blue, Trans=Black. For panels c, d, e, and f:  
583 Female=Red, Male=Yellow, Unknown=Purple. For panel g: STEM Average=dark grey, IPC Average=light  
584 gray, 2020 VCs=magenta \*Error bars in all panels are defined as standard deviation and are not included  
585 for ICLR IPC data because n<3.

586

587 **Fig. 3 | VCs increase participation by early career scientists (students and postdocs) and from**  
588 **non-research-intensive institutions.** (a) The 2020 NAMS VC (n=1) was attended by substantially more  
589 students and postdoctoral researchers than the 2015-2019 NAMS IPCs (n=4), while attendance by other  
590 job types remained fairly constant. (b) A positive percent change in attendance for all categories was

591 observed between 2015-2019 NAMS IPCs and the 2020 NAMS VC, and percent increase in attendance  
 592 by students and postdoctoral researchers was very high. \*Error bar for Undergrad Student is too large to  
 593 be included. (c) Students and postdoctoral researchers made up a larger percentage and industry  
 594 personnel and academic scientists represented smaller fractions of both the 2020 ICLR VC delegation  
 595 (n=1) compared to the 2019 ICLR IPC (n=1) and the 2020 NAMS VC delegation (n=1) compared to the  
 596 2015-2019 NAMS IPCs (n=4). (d) On average, postdoctoral researchers and students made up smaller  
 597 fractions of the delegations at historical IPCs (total n=6: ICLR (n=1), AAS (n=1), and NAMS (n=4))  
 598 compared to the fractions they represented at analyzed 2020 VCs (total n=6: ICLR (n=1), AAS (n=1),  
 599 NAMS (n=1), POM (n=2), and IWA (n=1)). (e) A positive percent change in attendance by persons from  
 600 PUIs and R2 Universities was observed at the 2020 NAMS (n=1) and AAS (n=1) VCs compared to the  
 601 2015-2019 NAMS IPCs (n=4) and 2016-2019 AAS IPCs (n=4), while attendance from minority serving  
 602 institutions decreased, but this is likely a result of small sample sizes (attendees from minority serving  
 603 institutions<10). \*Error bars in all panels are defined as standard deviation.  
 604

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