



**Article title:** Key opportunities and challenges for the use of big data in migration research and policy

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7

8 Dear Editorial Staff,

9

10 **Key opportunities and challenges for the use of big data in migration research**  
11 **and policy**

12 I am writing to submit the above manuscript for consideration for publication in UCL Open:  
13 Environment, as previously discussed with Ian Caswell and Professor Dan Osborn.  
14 Migration is high on the global political agenda, with one billion people recorded as having  
15 migrated in 2018 alone. Such levels of migration have great implications for global health,  
16 the global development agenda and political discourse at all levels of society. Meeting the  
17 needs of migrant populations requires accurate data on who is moving, where to and from,  
18 and what is driving their movements. However, current information on migrant populations is  
19 relatively scarce, leading to the UCL-Lancet Commission on Migration and Health to propose  
20 'big data' (from mobile phone data to satellite data) as a potential solution to help address  
21 these knowledge gaps.

22 In response to the Commission's call, we held a cross-disciplinary workshop in London, UK  
23 in July 2019, bringing together UN representatives, humanitarian NGOs, policymakers and  
24 academics to facilitate knowledge exchange and identify the key opportunities and  
25 challenges for the implementation of big data in migration research. This workshop was  
26 supported by a UCL Grand Challenges grant for projects focusing on migration and  
27 displacement.

28 Here, we provide a summary of key discussion points and conclusions identified in the  
29 workshop to assist migration experts in deciding whether the use of big data is appropriate  
30 for their work. We also aim to stimulate discussion about the potential of this approach in  
31 aiding future migration research and policy and the needs of migrant populations globally,  
32 especially given ongoing high-level political narratives about migration, environmental  
33 drivers, and migrant health. Furthermore, the workshop was a highly interdisciplinary event  
34 and we take the opportunity to discuss how to develop further cross-disciplinary solutions to  
35 migration research.

36 I confirm that I have agreement from all authors to submit this paper and that the manuscript  
37 is not submitted elsewhere and is original.

38 I look forward to your response in due course.

39

40 Yours faithfully,

41

42 Dr Lydia Franklinos

43 Key opportunities and challenges for the use of big data in  
44 migration research and policy

45

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69

70 **Abstract**

71 Migration is one of the defining issues of the 21st century. Better data is required to improve  
72 understanding about how and why people are moving, target interventions and support  
73 evidence-based migration policy. Big data, defined as large, complex data from diverse  
74 sources, is regularly proposed as a solution to help address current gaps in knowledge. The  
75 authors participated in a workshop held in London, UK, in July 2019, that brought together  
76 experts from the UN, humanitarian NGOs, policy and academia to develop a better  
77 understanding of how big data could be used for migration research and policy. We identified  
78 six key areas regarding the application of big data in migration research and policy: ethical and

79 security concerns around the use of big data; accessing and utilising data; integrating data  
80 sources and knowledge; understanding environmental drivers of migration; improving  
81 healthcare access for migrant populations; and addressing political narratives. We advocate  
82 the need for careful consideration of the challenges faced by the use of big data, as well as  
83 increased cross-disciplinary collaborations to advance the use of big data in migration  
84 research whilst safeguarding vulnerable migrant communities.

85

## 86 **Keywords**

87 Big data; migration; cross-disciplinary research; policy; humanitarian; environment;  
88 displacement; climate change; health; data security.

89

## 90 **Introduction**

91 With the number of global refugees reaching the highest levels since the second World War  
92 (1) and one billion migrants recorded in 2018 alone (2), human migration is high on the global  
93 political agenda. The UCL-Lancet Commission on Migration and Health (2) and the UN Global  
94 Compact on Migration (3) have called for improved data to understand drivers of migration,  
95 target interventions and support evidence-based migration policy. The application of big data  
96 in migration research and policymaking has been proposed as a possible solution to help  
97 address these knowledge gaps (2,4). Big data refers to large, complex data from varied  
98 sources, ranging from social media and mobile phone data (Figure 1), to electronic health  
99 records and satellite data, and has the potential to provide new sources of information for  
100 migration research (5,6). Previous studies have used big data to predict patterns of human  
101 movement during natural disasters (7) and track movement in near real-time (8), quantify  
102 migration at national scales (9–13), guide and evaluate humanitarian interventions (14) and  
103 examine the effects of human movement on disease transmission (15). In addition, satellite-  
104 based earth observation data has been used to map the relationship between environmental  
105 change and human movement (16,17), model subnational migration flows (18) and inform  
106 policy decisions (17,19,20). Despite the immense opportunity big data can provide for  
107 migration research and policy, several challenges have hindered its widespread  
108 implementation (2,4).

109

110 In response to the call for increased collaboration (2) and improved research on ways to utilise  
111 big data sources in the field of migration (4), we participated in a cross-disciplinary workshop  
112 in London, UK 3<sup>rd</sup> July 2019, bringing together UN representatives, humanitarian NGOs,  
113 policymakers and academics to facilitate knowledge-exchange and identify the key

114 opportunities and challenges for the implementation of big data in migration research. Here,  
115 we provide a summary of key discussion points identified in the workshop via presentations,  
116 panel discussions and break-out groups in which participants explored different topics and  
117 possible solutions. We provide major conclusions from the workshop supported by a review of  
118 relevant literature to assist migration experts in deciding whether the use of big data is  
119 appropriate for their work and to stimulate discussion about the potential of this approach in  
120 aiding migration research and policy, and the needs of migrant populations globally. In  
121 particular, the outcomes of this workshop may provide a timely resource for the recently  
122 launched Lancet Migration, a global collaboration of migration experts that aims to address  
123 evidence gaps and drive policy change in the field (21). Importantly, this workshop also tested  
124 a range of methods of collaborative working and identified challenges and opportunities in  
125 achieving a truly cross-disciplinary approach to migration research, an important but often  
126 neglected aspect in such a complex and politically charged topic as migration.

127

## 128 **Opportunities and challenges relating to big data**

129 The aims of the workshop were to facilitate discussion on both the opportunities of big data  
130 but also the challenges, shortfalls and current ethical considerations surrounding big data and  
131 its uses. Whilst it is beyond the scope of a single workshop or article to find solutions to these  
132 major challenges, the workshop proved a valuable forum for raising awareness and nuance  
133 discussion around these challenges which are often not fully recognised or addressed by  
134 researchers and practitioners alike when collecting or using big data as it relates to migration.

135 The format of the workshop consisted of keynote speeches, three panel discussions on the  
136 topics of 1) the role of big data in understanding the environmental drivers of migration, 2) the  
137 opportunities and challenges of big data in healthcare, and 3) the ethical considerations  
138 surrounding the use of big data in migration research and policy. The panel discussions were  
139 followed by group exercises which aimed to conceive structured, actionable solutions to a set  
140 of questions raised during the panel discussions. The resultant discussions and ideas were  
141 grouped into six topics which we summarise here and in Table 1.

142

## 143 **Accessing and utilising big data**

144 The first topic focused on the access, awareness and expertise required for big data use. The  
145 application of big data is often hindered by the fact that many big data sources such as mobile  
146 phones, internet-based platforms and other digital devices are managed by private companies  
147 who collect the data for business purposes. Therefore, costs associated with accessing big

148 data and issues of ownership are significant barriers to its use (22). Big data generation will  
149 vary geographically and may be reduced in many high mobility contexts where infrastructure  
150 (i.e., cell towers, wi-fi connection and electronic bank transfer services) is less established.  
151 Specific to the use of mobile phone call detail record (CDR) data, the presence of multiple  
152 mobile network operators in a country mean that multiple data sharing agreements would be  
153 required (23). In addition, there are significant issues around the potential extraction of  
154 sensitive information contained in big data (24,25) and data sources are often fragmented  
155 across disciplines which reduces the awareness of available datasets (26). Accounting for  
156 multiple biases and the complex analyses required to interpret the data are further examples  
157 of methodological difficulties associated with the use of big data (6,27).

158

159 Workshop discussions highlighted the importance of understanding how, why and when the  
160 data were collected to identify potential gaps and biases, therefore ensuring it can be used  
161 effectively. There is great need for more centralised repositories of data, projects and  
162 publications such as The Humanitarian Data Exchange (28), to promote knowledge- sharing,  
163 collaborations and inform evidence-based programming. Increased partnerships between  
164 governments, international agencies, civil society, and the private sector are also required to  
165 improve data access and ensure the optimum exploitation of available data and technologies.  
166 Furthermore, capacity building in countries or organisations with an interest in big data analysis  
167 is needed to support cross-disciplinary research and improve specialist knowledge in certain  
168 regions. This could be achieved via collaborations with relevant partners and agencies such  
169 as demonstrated with the United Nations Economic Commission for Europe's (UNECE) Big  
170 Data Sandbox which provided a platform for statistical organisations to collaborate and learn  
171 to use big data analytics (29). However, there may be ethical considerations for private-public  
172 partnerships. For example, published commentaries have voiced fears over the partnership  
173 between the UN's World Food Programme and the data analytics company Palantir which may  
174 have serious consequences for the privacy and security of aid recipients due to the company's  
175 links to US intelligence agencies (30).

176

## 177 **Integrating data sources and knowledge**

178 The second topic concerned the integration of data and knowledge across disciplines. The  
179 main source of data for migration statistics originates from traditional methods such as  
180 household surveys recorded at local scales and national population estimates, as well as data  
181 on forced displacement collected through key informant networks (4) (Figure 2). Big data  
182 sources have the potential to complement traditional data and address significant spatial and  
183 temporal gaps via updating migration statistics in an accurate and low-cost way (4,10). For

184 example, analysis of CDR data can be used to replicate national internal migration statistics  
185 and complement outputs from censuses (10). However, integrating migration data from  
186 traditional methods with varied sources of big data requires new methodology that considers  
187 complex interactions over differing geographic and temporal scales. The quality of various data  
188 sets (e.g., demographic biases present in social media datasets) remains an unresolved challenge  
189 in teasing out comprehensive, policy-relevant results. Validating estimated migration using near-  
190 real-time big data is also problematic, with no trusted 'gold standard' currently available (31). The  
191 slow adoption of big data analyses in the humanitarian sector is partly due to a lack of expertise  
192 in how to apply these approaches in operational settings (32). Workshop participants  
193 discussed the need to bridge the gap between experts on the ground collecting the data via  
194 traditional methods and big data analysts via increased transdisciplinary training and  
195 collaborations. A recent workshop hosted by the International Organization for Migration (IOM)  
196 and the German Federal Foreign Office concluded that "greater cooperation and engagement  
197 among stakeholders" both within and external to the migration sector are required to inform  
198 decision making (33). If we are to integrate different data sources effectively, a collaborative  
199 cross-disciplinary approach is required to ensure we understand the data and how they can  
200 be used to deepen our understanding of the drivers and impacts of migration. This approach  
201 is practiced in "Data Collaboratives"; collaborative projects in which different sectors including  
202 private companies, research institutions, and government agencies collaborate to enable data  
203 exchange and help solve public problems (34). NetHope is an example of a Data Collaborative  
204 project which has helped to integrate data sources and produce maps of connectivity sites  
205 across Puerto Rico to assist in delivering aid in the aftermath of Hurricane Maria (35).

206

207

### 208 **Understanding environmental drivers of migration**

209

210 The third topic considered the use of big data in understanding and identifying environmental  
211 drivers of migration such as natural hazards and climate change (36) for example via remote  
212 sensing. Currently, there is no internationally agreed definition for "environmental migrant"  
213 despite it being required to collect long-term data and guide the policies of governments and  
214 international agencies. The IOM have proposed a broad working definition (37) which  
215 importantly considers that environmental migration might be triggered both by sudden-onset  
216 events (38), such as earthquakes and cyclones, and slower environmental change processes,  
217 such as desertification and sea-level rise. In the context of slow-onset events and gradual  
218 environmental change, the effects on migration are often difficult to quantify since it can be  
219 hidden behind more immediate socioeconomic drivers such as poverty or political processes  
220 (39). There are opportunities for the application of big data to unpick these different drivers and to

221 help to define the term “environmental migrant”. A further application could be in helping to quantify  
222 the effect of environmental change on the trapping of individuals or communities, usually due to  
223 rising poverty barriers which impede mobility. Human mobility can improve resilience and  
224 attenuate the negative outcomes of environmental degradation, but poverty, disability and  
225 social exclusion may limit people's ability to resort to migration as an adaptation strategy (40).

226

227

228 Despite the significant attention that environment-induced human migration has received (41),  
229 evidence on the relationship between migration and environmental change is limited. When  
230 used in combination with traditional datasets, big data has the potential to identify hot spots of  
231 environmental change and exposed populations that may be affected by the change and  
232 therefore liable to migrate, or to become trapped. Satellite data is a particularly valuable  
233 resource in analysis of environmental drivers as it enables the systematic, consistent and  
234 accurate monitoring of areas (even if remote or inaccessible) that are affected by anthropic or  
235 natural hazards. Indeed, satellite-based technologies are key to analysing observable effects  
236 of climate change effects and even predicting environmental-led migration (42-47).  
237 Furthermore, a study combined satellite data and CDR data to quantify the incidence, direction  
238 and duration of flood-driven migration, revealing important short-term (hours–weeks) aspects of  
239 migration associated with extreme weather events that are often not captured with traditional  
240 survey-based research (17). There is great potential for big data to help in understanding and  
241 addressing environment-related displacement and to inform policies that will improve resilience  
242 to environmental change and support migration that is required to improve the health and  
243 livelihoods of vulnerable people. In such examples, improved data on both environmental  
244 change (as a driver of migration) and of migration itself (such as displacement following a natural  
245 hazard) cannot ‘solve the problem’ of forced environmental migration but can inform  
246 interventions such as aid as well as discussions between affected communities and  
247 stakeholders in devising context-appropriate solutions for the future.

248

249 One of the most valuable aspects of satellite-based analyses is the capability for retrospective  
250 analysis which is required to detect changing patterns across space and time and to inform  
251 predictions. However, a recent review stated that current initiatives do not exploit the full  
252 possibilities of satellite-based earth observation in migration with a lack of services offering the  
253 systematic flow of detailed information to researchers, managers and migration analysts (48).  
254 One of the main gaps identified was that currently consolidated satellite-based monitoring  
255 systems work at regional scales which it is often too coarse to understand the specificities of  
256 how particular communities are affected, thus unable to inform sub-national policies design,  
257 implementation and monitoring. Indeed, reported discrepancies between the assumed



258 environmental drivers of migration perceived from satellite data analyses and the self-reported  
259 drivers of migration (e.g., see 49) underline the importance of including information on the  
260 lived experiences of migrants to inform actions.

261

## 262 **Improving healthcare access for migrant populations**

263 The fourth area of discussion focused on the potential for big data to enhance migrant health  
264 via improved disease outbreak preparedness, identification of vulnerable groups, increased  
265 access to healthcare and by informing evidence-based health interventions. Many studies  
266 have demonstrated the potential for big data to understand the impact of migration on the  
267 transmission of infectious diseases such as dengue (15), malaria (18,50) and cholera (51) at  
268 a national scale. These analyses highlight the potential benefit of big data use to improve  
269 preparedness and mitigation efforts for disease outbreaks. This may be particularly useful  
270 when supporting on-the-ground activities by helping to predict potential disease outbreaks for  
271 displaced persons. A recent collaboration between IOM and the mobile operator data  
272 analytics organisation Flowminder, is combining IOM Flow Monitoring Registry surveys with  
273 CDR data to gather anonymous information about people on the move at key transit points to  
274 inform public health interventions for the COVID-19 pandemic (52). A further application of  
275 big data in humanitarian settings is the use of satellite data to map refugee settlements (20)  
276 which can be used to ensure healthcare access for displaced persons.

277

278 A perennial issue in equitable healthcare access is identifying and addressing the needs of  
279 invisible communities, such as migrants, particularly those with an undocumented status (2).  
280 Shortfalls with traditional datasets and data collection processes such as semi-structured  
281 interviews and surveys means there is limited information on undocumented migrants and  
282 vulnerable groups (i.e., unaccompanied children, people with disabilities and members of the  
283 LGBTI community). There is great potential for big data address the paucity of information on  
284 these groups, their health, access to healthcare and differing healthcare needs. In addition, it  
285 was suggested that the healthcare needs of migrants settled in countries such as the UK could  
286 also be improved by big data analysis, for example via a general migrant longitudinal study  
287 such as the cohort studies performed by the UK Economic and Social Research Council (53).  
288 There are also vast applications for big data in implementing evidence-based health  
289 interventions that need to be explored, specifically in identifying differences in patient  
290 responses to treatments and tailoring healthcare to the specific needs of individuals (54).

291

292 The COVID-19 pandemic has received widespread support for the use of big data in disease  
293 surveillance systems globally (55). Such approaches have had various levels of success at

294 curbing infection rates, particularly crucial for vulnerable persons (which could include some  
295 migrant communities). However, if the use of such big data approaches in healthcare became  
296 common place, it may cost many the protection that invisibility currently offers, with many  
297 communities such as undocumented migrants fearing disproportionate effects of state  
298 surveillance (56). Such ethical conundrums remain unresolved and often underexplored. At  
299 best, it is likely that big data should not be seen as a perfect solution but as one optional tool  
300 within a wider social toolkit which retains traditional and non-digital interventions.

301

302

### 303 **Ethical, privacy and security concerns**

304 The fifth topic focused on ethical, privacy and security concerns regarding the use of big data  
305 in migration research. This topic proved to be cross-cutting, with themes re-emerging across  
306 other topics. Principally, a recurrent question arose about whether it is appropriate to collect  
307 and/or use sensitive data in the pursuit of greater understanding for researchers and  
308 policymakers. The collection of personal data including migrant status is a contentious issue.  
309 There are concerns that information on personal migration status may create or increase  
310 existing discriminatory practices in society such as provision of healthcare and access to  
311 state funds, or that mobile tracking devices may be used against a migrant to forcibly return them  
312 to a previous location (2). Additionally, various sources of big data (such as social  
313 media) are consumed differently based on geography, demography and access. As  
314 such, any analysis of these datasets will carry these biases, which may follow  
315 through into policies. This may result in policies which propagate stereotypes and  
316 discriminatory practices, or else continue to underserve invisible groups (e.g.,  
317 those not engaged with social media or with smaller social networks) (31).

318

319 Ethics in the context of big data in migration may be considered in several ways. First, it may  
320 relate to the way in which the research is conducted and whether there has been  
321 consideration for data privacy and security. Secondly, it could refer to decision-making  
322 regarding migrants with consideration for their lived experience, especially in humanitarian  
323 situations. A recent report on migration noted that discussions on ethics often focus on the  
324 legal aspects of data protection rather than understanding how the results of analyses may  
325 detrimentally impact affected populations and counter the humanitarian principles to “do  
326 no harm” (33). Furthermore, data protection measures are often focused on personal data  
327 (e.g., General Data Protection Regulation in the EU) and do not consider group data  
328 protection needed to work with vulnerable groups (57).

329

330 It is important to consider who benefits from the use of big data sources in migration research.  
331 At the individual level, migrants may not wish for additional data to be gathered about them  
332 and may perceive no benefits of the process (58). However, at the community level, such data  
333 and analysis may help to address the health needs of migrants more generally. Certainly,  
334 there will be benefits to the academic community seeking to study migrant health needs and  
335 to decision-makers seeking evidence-based solutions. Pursuit of these research and policy  
336 goals can result in overlooking the individual rights and raise ethical issues for many  
337 vulnerable people (58). Furthermore, forcibly displaced people fleeing persecution may have  
338 little trust in authorities and therefore be less willing to seek healthcare or consent to having  
339 their data collected. This creates a barrier for healthcare professionals, humanitarian workers  
340 and researchers who wish to respect the rights of the individual, whilst deriving a better  
341 understanding of migration pathways and healthcare needs. Workshop discussions  
342 highlighted the power imbalance between various parties involved; those seeking data  
343 including governments and academics often from the Global North with inherent biases and  
344 power, and those the data is being sought from who are often vulnerable persons in  
345 precarious or dangerous situations, many originating from the Global South (59). Evidence  
346 of this power imbalance is seen with 'high-risk experiments' with new technologies such as  
347 the use of Canada's automated decision-making technology in immigration and refugee  
348 applications (60) which often lack regulation. Even after applying advanced safeguarding  
349 practices and aggregated outputs, researchers may still be reluctant to apply big data  
350 analysis for migration research as policy-makers often have their own agendas and may use  
351 the methods and deliverables in ways not intended or anticipated by the research authors.

352

353 Another established concern regarding the ethical deployment of big data in migration  
354 research is the focus on pattern-based analyses, rather than on achieving a conceptual  
355 interpretation and critical analysis for why behavioural trends might emerge and under what  
356 set of circumstances and assumptions (61,62). Indeed, the limitations and bias of big data  
357 remains an under-explored aspect in the discourse around ethical implications of big data.  
358 The COVID-19 pandemic provides a pertinent example of how the lack of data from  
359 vulnerable communities such as refugees or people on the move has led to the  
360 underrepresentation of these communities in the narrative and political responses of the  
361 pandemic (63).

362

### 363 **Addressing political narratives**

364 The final topic concerned the role of big data in high level political narratives around migration.  
365 With ongoing antimigrant rhetoric existing at all levels of government and society, migrant

366 research has the opportunity and mandate to address such political narratives. There are  
367 many examples of authorities treating migrants as political pawns or as statistical figures (64).  
368 Therefore, it is imperative that big data is not used to further discriminate against migrant  
369 communities or to target certain populations, but rather to support inclusive and fair migration  
370 governance. This can be particularly problematic with big datasets, where assumptions have  
371 been made throughout the data collection and analysis process. For example, in 2017 the UK  
372 data analytics company CGI together with the Dutch statistical agency CBS, conducted a study  
373 into migration forecasting of people in Syria using Twitter data (62). The study required a set  
374 series of assumptions to be made about Twitter content (more specifically its English  
375 translation) and was unable to consider the context within which tweets were posted. This case  
376 study demonstrates how the promotion of big data can reify such assumptions and distort the  
377 original meaning or intent behind an individual's migration decisioning, as well as pervert the  
378 aggregate narrative behind migration patterns (62). Conversely, an positive example discussed  
379 within the workshop was the Sentinel project; an NGO that works to gather and  
380 disseminate 'trusted' information to local people and governments in order to counter the  
381 spread of misinformation, antimigrant rhetoric and to prevent resultant hate crime and  
382 genocides (65). Participants also deliberated whether the increased evidence provided by big  
383 data would be instructive and powerful enough to overcome political and social biases  
384 associated with the topic. Given the highly political nature and high stakes of migration policy  
385 for migrants as well as for governments and the international community, more evidence  
386 may alone be insufficient to achieve multilateral, progressive action. It is worth  
387 therefore considering other factors contributing to the political discourse and designing  
388 strong, cross-disciplinary communication tactics to support maximum impact of evidence.  
389 Furthermore, it is worth considering how paucity in migration data has helped to shape  
390 public and media perceptions of global migration patterns to date, and whether big data  
391 could be used to address these perceptions.

392

## 393 **Discussion**

394

395 The application of big data in migrant research shows much promise in addressing the current  
396 gaps in knowledge. Big data sources can help to update internal migration statistics by  
397 addressing the significant gaps in quantity and quality of data collected from traditional  
398 methods (4,10). When combined with field-level data derived from household surveys and  
399 key-informant networks, big data can be used to detect how sudden onset natural hazards  
400 and gradual environmental change (e.g., desertification and sea-level rise associated with  
401 climate change) impact migration patterns. This can help to inform planning and scenario

402 building, as well as contributing to a more comprehensive definition of “environmental migrant”  
403 which is critical for migration policy within the context of ongoing environmental change. In  
404 addition, it has a potential application in considering the differing healthcare needs of migrants  
405 as well as identifying vulnerable populations unable to migrate due to environmental change.  
406 There is also scope for big data to inform evidence-based health interventions for migrant  
407 populations in everyday and emergency (including displacement) settings. Yet despite the  
408 vast opportunities that big data present, there are some important areas to consider before  
409 using these varied and complex data sources. Increased cross-disciplinary partnerships are  
410 required to improve data access, knowledge-sharing and capacity building across sectors and  
411 regions. In addition, a collaborative cross-disciplinary approach is required to ensure the  
412 different datasets are understood and to develop new methodologies to integrate data sources  
413 and identify complex interactions that influence how and why people are moving. Furthermore,  
414 the reported lack of agreement within the humanitarian sector on how migration modelling  
415 should be applied needs to be addressed so analyses can be effective (33). It is important  
416 to challenge the assumption that big data is always a suitable and insightful tool to use in  
417 research for migration policy. Whilst tempting to consider big data the magic bullet, it may not  
418 always be appropriate or may need to be used in conjunction with traditional methods.

419

420 International legislation is required to sufficiently address how migrant data should be  
421 collected and used to ensure ethical conduct by data gatherers and owners and the  
422 safeguarding of human rights, even in sensitive migration contexts (66). The United Nations  
423 Development Group (66) and Office for the Coordination of Humanitarian Affairs (67) provide  
424 guidelines for safe and ethical data management in humanitarian situations however, there is  
425 no legal enforcement of these practices. Although researchers would like quicker and easier  
426 access to data, workshop discussions challenged whether the process should be  
427 hastened, suggesting that administrative obstacles force researchers to duly consider  
428 whether additional data is necessary and beneficial to the current state of knowledge, given  
429 the risks and trade-offs that must be made. A key output of the workshop was a consensus that  
430 researchers and decision-makers must first ask why they require additional data and whether  
431 this is what all parties, particularly migrants, would want. As well as data protection issues,  
432 it is also imperative to understand the potential harmful impact of analyses on vulnerable  
433 migrant populations (33). It is especially important that big data is not used to further  
434 discriminate or target migrant populations considering current antimigrant political narratives.

435

436 In pursuit of cross-disciplinary collaboration, the workshop brought together a range of  
437 representatives from the UN, government, humanitarian agents and academics from a range  
438 of backgrounds. Cross-sector engagement in the workshop was difficult to achieve which may

439 be due to differences in the objectives of different sectors, as well as the language of  
440 engagement used. For instance, humanitarian organisations were particularly difficult to  
441 engage, and it is thought that this was due to both a shortage of networks linking academia  
442 and humanitarian organisations as well as differences in short, medium and long-term needs  
443 and objectives of the two sectors. We trialed different methods to stimulate interdisciplinary  
444 work including the use of business canvases (68) to explore and present solutions to  
445 questions. This approach was useful for stimulating debate within the groups and producing  
446 well-considered outputs. However, future interdisciplinary events would benefit from the  
447 development of methods that consider the language styles and information sharing techniques  
448 of different disciplines and thus facilitate effective communication and knowledge-sharing  
449 (4,33,69,70). The workshop also elucidated the extent to which the direction of conversation  
450 is steered by which actors are engaging in the conversation. For example, actors engaging  
451 with migrants directly (i.e., service providers) focus on practical implications. This serves as a  
452 pertinent reminder of the importance for academics to create and utilise diverse networks as  
453 well as the need for actors with gatekeeping power to exercise due diligence and engage a  
454 wide range of stakeholders and interest groups in discussions. Overall, the workshop  
455 highlighted the benefits of cross-disciplinary work, enabling the identification of key topics from  
456 a variety of angles and providing meaningful and effective outputs. Furthermore, we hope this  
457 workshop assists in cultivating a future transdisciplinary approach to migration research,  
458 whereby there is a move beyond the collaboration of individual disciplinary perspectives to  
459 develop curriculum integration that organizes knowledge production in the context of real-world  
460 problems (71).

461

462 **Table 1.** Key questions, challenges, opportunities and solutions for the use of big data in migration research and policy.

463

<b>Topics</b>	<b>Key research and policy questions</b>	<b>Potential challenges</b>	<b>Potential opportunities and solutions</b>
Accessing and utilising big data	<ul style="list-style-type: none"> <li>● How can we improve access to big data sources?</li> <li>● How can we enhance awareness of available data?</li> <li>● How can we develop the expertise required to use big data across disciplines?</li> </ul>	<ul style="list-style-type: none"> <li>● Issues of ownership and costs in accessing big data are significant barriers.</li> <li>● If multiple mobile network operators are operating in a country, multiple data sharing agreements would be required (23).</li> <li>● Geographic biases in the type of data available.</li> <li>● Poor awareness of fragmented data sources.</li> <li>● Issues around the potential extraction of sensitive information contained in big data, therefore access arguably should be difficult for these data.</li> </ul>	<ul style="list-style-type: none"> <li>● Opportunities for the development of centralised repositories of data such as The Humanitarian Data Exchange (28) to promote collaboration and knowledge-sharing.</li> <li>● Opportunities for partnerships across different sectors to improve access to available data and technologies.</li> <li>● Capacity building projects across disciplines such as the UNECE Big Data Sandbox (29) are needed to support big data use for migration research.</li> </ul>
Integrating data sources and knowledge	<ul style="list-style-type: none"> <li>● How can we produce more detailed and recent migration statistics using big data?</li> <li>● How can we best integrate data from different sources?</li> <li>● How do we manage fragmented</li> </ul>	<ul style="list-style-type: none"> <li>● Complex analyses are required to account for multiple biases in different datasets.</li> <li>● Lack of awareness and big data expertise in humanitarian sector has led to its slow adoption.</li> </ul>	<ul style="list-style-type: none"> <li>● The use of big data sources combined with traditional survey-based research can reveal important aspects of migration that are often not captured (e.g., CDRs reveal short-term migration patterns) (52).</li> <li>● Potential for methodological innovation for the</li> </ul>

	<p>data sources across varied spatial and temporal scales?</p> <ul style="list-style-type: none"> <li>• How can we develop a collaborative cross-disciplinary approach to address the challenges in the field of migration?</li> </ul>		<p>integration of different data and the development of a 'gold standard' to estimate migration using near-real-time big data.</p> <ul style="list-style-type: none"> <li>• Opportunities for cross-disciplinary "Data Collaboratives" (34) to enable data exchange and help address complex problems.</li> </ul>
Understanding environmental drivers of migration	<ul style="list-style-type: none"> <li>• How can big data be used to assess the ongoing impact of climate change on migration?</li> <li>• Can big data help to identify populations that are vulnerable to environmental change?</li> <li>• How can big data be used to predict mass migration events due to environmental change?</li> </ul>	<ul style="list-style-type: none"> <li>• Satellite-based environmental data is often at coarse geographic scales which are not suitable to inform sub-national policies and actions.</li> <li>• There may be discrepancies between the assumed environmental drivers of migration perceived from satellite data analyses and the self-reported drivers of migration.</li> </ul>	<ul style="list-style-type: none"> <li>• Opportunity to address the lack of evidence on the relationship between migration and environmental change and to help to define the term "environmental migrant" that is needed for policy action.</li> <li>• Potential to reveal important aspects of migration associated with extreme weather events that are often not captured with traditional data.</li> <li>• Potential to unpick socioeconomic factors that may be limiting people's ability to migrate as an adaptation strategy to environmental change.</li> <li>• Opportunities to inform policies that will improve resilience and support migration associated with environmental change.</li> </ul>
Improving healthcare access for migrant populations	<ul style="list-style-type: none"> <li>• How can big data be used to address the immediate health needs of displaced persons in camps?</li> </ul>	<ul style="list-style-type: none"> <li>• Big data approaches in healthcare risk jeopardise the protection from state surveillance that is currently provided to vulnerable communities such as</li> </ul>	<ul style="list-style-type: none"> <li>• Big data can help to understand the impact of migration on the transmission of infectious diseases.</li> <li>• Potential to support on-the-ground activities,</li> </ul>

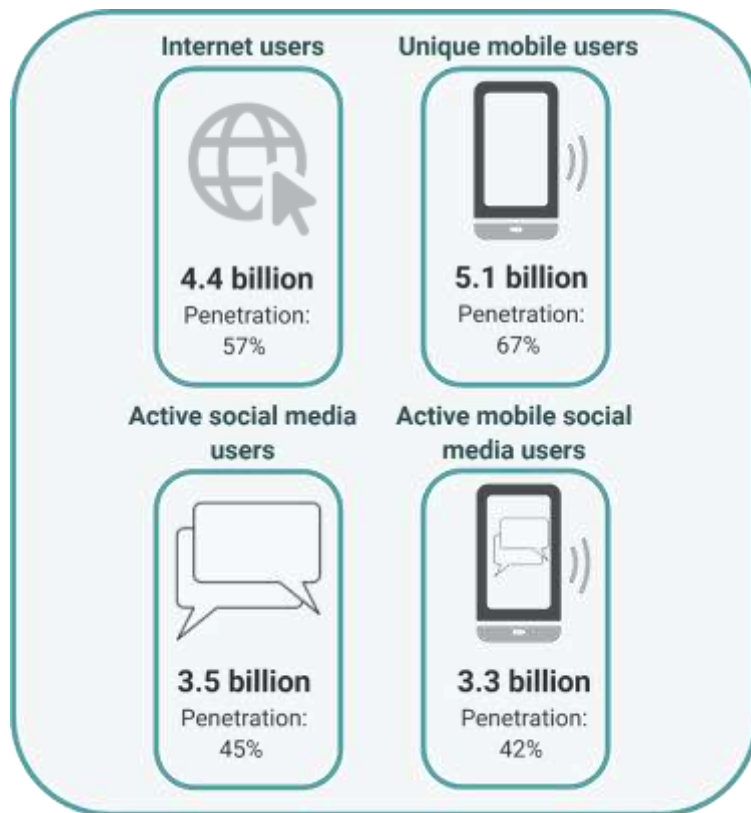


	<ul style="list-style-type: none"> <li>• How can big data help us learn more about undocumented migrants, their health and healthcare needs?</li> <li>• How can big data be used to implement evidence-based health interventions?</li> </ul>	undocumented migrants.	<p>helping to address the immediate health needs of displaced persons and to predict disease outbreaks.</p> <ul style="list-style-type: none"> <li>• Potential for big data to improve equitable healthcare access for underrepresented communities, such as undocumented migrants and vulnerable groups.</li> <li>• Opportunities to improve the specific healthcare needs of migrants with settled status.</li> <li>• Applications for big data in identifying differences in patient responses to treatments and tailoring healthcare to the specific needs of individuals.</li> </ul>
Ethical, privacy and security concerns	<ul style="list-style-type: none"> <li>• What is meant by ethics in the context of big data in migration research?</li> <li>• Who benefits from the use of big data (migrants at individual or community level, academic community, policymakers)?</li> <li>• How do power imbalances influence the use of big data?</li> <li>• How can we achieve ethical data usage?</li> </ul>	<ul style="list-style-type: none"> <li>• Big data analytics may lead to the introduction of stakeholders which have varied motivations.</li> <li>• Biases in big data may follow through into policies, propagating stereotypes and discriminatory practices, or else continuing to underserve invisible groups.</li> <li>• There is often focus on the legal aspects of data protection for personal data rather than the potential negative impacts on affected vulnerable groups.</li> </ul>	<ul style="list-style-type: none"> <li>• Ethical considerations and safeguarding practices are required when involving varied stakeholders in big data analytics</li> <li>• The use of big data in migrant research requires the clear stating of assumptions and methods to correct for biases to prevent the propagation of biases in policy.</li> <li>• Ethical considerations and safeguarding practices are required when involving varied stakeholders in big data analytics.</li> <li>• Researchers and decision-makers must</li> </ul>

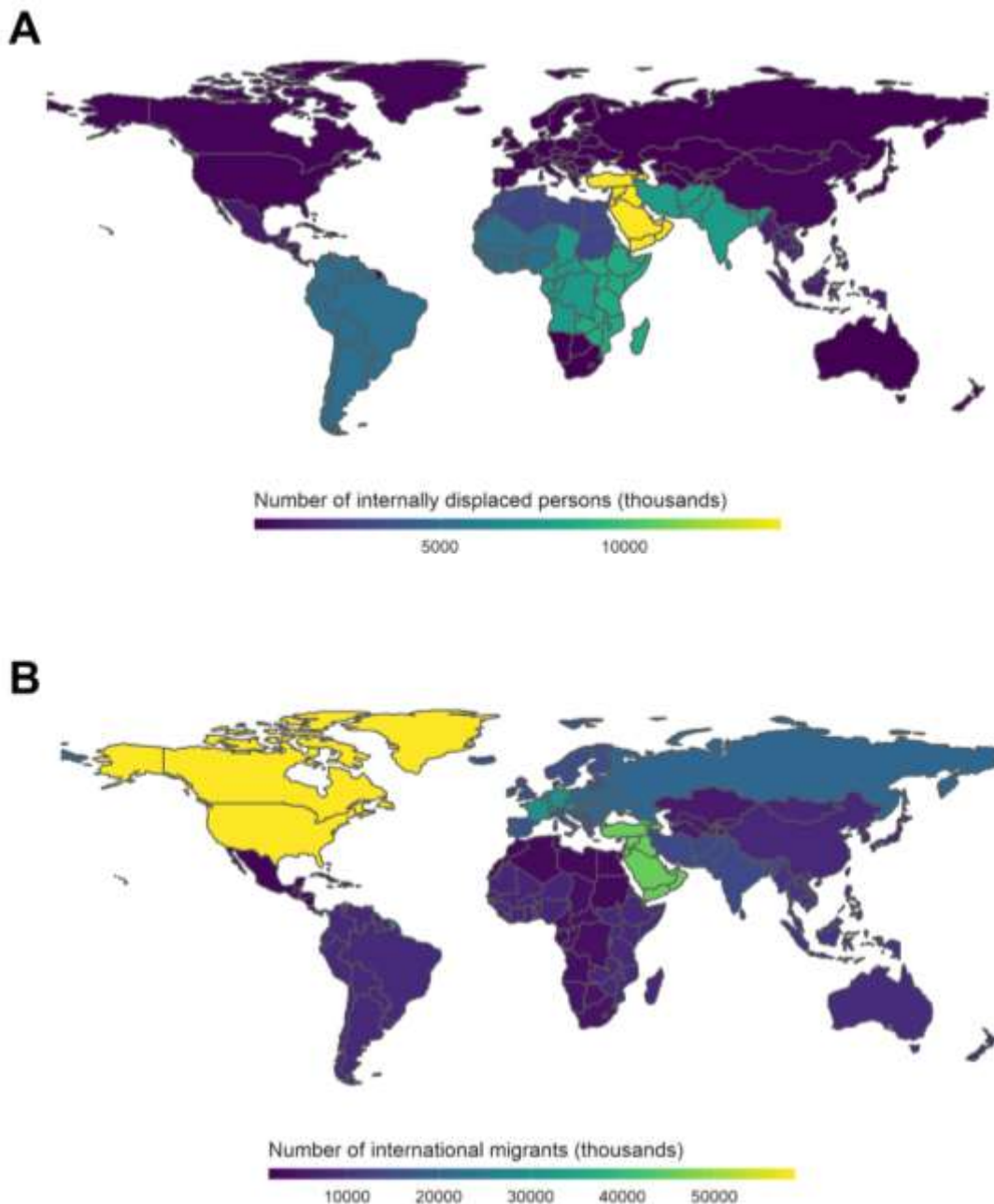
		<ul style="list-style-type: none"> <li>● Big data use in migration can promote the power imbalance between those seeking data and those the data is being extracted from.</li> <li>● There is no legal enforcement of guidelines for safe and ethical data management in humanitarian situations.</li> </ul>	critically examine and justify why they require the use of big data and whether this is what all parties, particularly migrants, would want.
Addressing political narratives	<ul style="list-style-type: none"> <li>● How can we prevent the use of big data for the discrimination of certain populations?</li> <li>● What role could big data have in addressing the negative political narratives around migration?</li> </ul>	<ul style="list-style-type: none"> <li>● Big data use may lead to lack on nuance behind the motivations for migration, potentially distorting the narrative behind migration patterns.</li> <li>● Ongoing anti-immigration rhetoric means it is imperative that big data is not used to further discriminate against migrant communities or to target certain populations.</li> </ul>	<ul style="list-style-type: none"> <li>● Opportunity to help to address negative political narratives and to support inclusive and fair migration governance as seen with the Sentinel project which works to counter the spread of misinformation and antimigrant rhetoric (65).</li> <li>● Designing strong, cross-disciplinary communication tactics to support maximum impact of evidence.</li> <li>● Opportunity to help to challenge public and media perceptions of global migration which has been shaped by misinformation due to paucity in migration data.</li> </ul>

465 **Figure 1. Key statistical indicators for global big data use.** The number of users and  
466 portion of the population that has access (penetration) to the internet, mobile phones, social  
467 media and mobile social media. Data were accessed via the Global Digital Report 2019 (72).

468  
469



470 **Figure 2. Global human migration by subregion.** A) Number of internally displaced  
471 persons (from conflicts and disasters) in 2020 ranging from 22 in the Eastern Europe (dark  
472 blue) to 14,200,000 in Western Asia (yellow); data were accessed via the Internal  
473 Displacement Monitoring Centre' s Global Internal Displacement database (73). B) Number  
474 of international migrants in 2019 ranging from 1,542,000 in  
475 the Caribbean (dark blue) to 58,647,000 in North America (yellow); data were accessed via  
476 the UN Department of Economic and Social Affairs, Population Division database (74). The  
477 legends for both graphs show the number of migrants in thousands.  
478



480

481

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487

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489     LF and RP led the workshop and took the lead role in writing the report. All authors participated  
490     in discussion sessions at the workshop and contributed to the report. Authors 3–10 are listed  
491     alphabetically.

492

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494     All other authors declare no conflicts of interest in connection to this article.

495

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497     No further data were used in addition to referenced works.

498

499

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