

1 **Harnessing alternative sources of antimicrobial resistance data to support**
2 **surveillance in low-resource settings**

3 Running title: Alternative sources of antimicrobial resistance data

4

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28

29 **Abstract**

30 One of the most pressing challenges to the global surveillance of antimicrobial
31 resistance is the generation, sharing, systematic analysis and dissemination of data in
32 low-resource settings. Numerous agencies and initiatives are working to support the
33 development of globally distributed microbiology capacity, but the routine generation
34 of a sustainable flow of reliable data will take time to establish and deliver clinical and
35 public health impact. By contrast, there are a large number of pharma and academia-
36 led initiatives that generate a wealth of data on antimicrobial resistance and drug-
37 resistant infections in low-resource settings, together with high volume data generation
38 by private laboratories. Here, we explore how untapped sources of data could provide
39 a short-term solution that bridges the gap between now and the time when routine
40 surveillance capacity has been established, and how this could continue to support
41 surveillance efforts in the future. We discuss the benefits and limitations of data
42 generated by these sources, the mechanisms and barriers to making this accessible, and
43 how academia and pharma might support the development of laboratory and analytical
44 capacity. We provide key actions that will be required to harness these data, including
45 a mapping exercise; creating mechanisms for data sharing; use of data to support
46 National Action Plans; facilitating access to, and use of data by the WHO Global
47 Antimicrobial Resistance Surveillance System; and innovation in data capture, analysis
48 and sharing.

49 **Introduction**

50 Surveillance is central to understanding the global burden of antimicrobial resistance
51 (AMR). The generation of surveillance data begins with appropriate sampling of
52 patients with a suspected infectious disease (diagnostic stewardship). Surveillance of
53 sepsis is one example of this, although other specimen types will be required for more
54 comprehensive surveillance. Culture and antimicrobial susceptibility testing of
55 pathogens can improve individual patient management through optimisation of drug
56 therapy, and supports their appropriate use (antibiotic stewardship). These data are
57 commonly collated to inform local empiric prescribing policies for patients presenting
58 with infectious disease syndromes. National data may then be collected by Ministries
59 of Health for the purposes of surveillance, evidence-based guidelines, programmes of
60 prevention and resource planning. Finally, national data may be submitted to global
61 surveillance initiatives that are used to document and track rates of resistance over time,
62 signal where and when interventions are needed, and identify countries that require
63 support to build capacity. The most prominent global initiative for the surveillance of
64 bacterial pathogens (excluding TB) is the WHO Global Antimicrobial Resistance
65 Surveillance System (GLASS),¹ which collects and reports data on resistance rates
66 aggregated by country.

67 This description of the generation, flow and analysis of AMR data represents an
68 ideal situation in which locally generated microbiological results move from a patient
69 care setting to a national or supranational network, but the reality is that these data are
70 fragmented and dispersed. A recent review commissioned by the Fleming Fund created
71 an inventory of supranational antimicrobial resistance surveillance networks in low-
72 and middle-income countries (LMIC) between January 2000 and August 2017.^{2,3} This
73 identified 72 supranational networks for AMR surveillance of bacteria, fungi, HIV,

74 tuberculosis and malaria, of which 34 are on-going.^{2,3} Their median duration was 6
75 years (range 1 to 70 years) and the median number of LMICs included in each network
76 was 8 (range 1 to 67). This scenario is not limited to the lowest resource settings. A
77 review of European AMR surveillance found similar fragmentation and heterogeneity,
78 with numerous local and national systems that lacked coordination, harmonization and
79 information sharing with international networks.⁴ There was also inadequate
80 standardization of epidemiological definitions, samples and data collected,
81 microbiological testing methods and data sharing policies.⁴

82 Categorization of the 72 LMIC networks identified in the review revealed that
83 the minority (n=26) were led by governments or the WHO, with the remainder being
84 associated with academia (n=24) or pharma (n=22).^{2,3} The number of networks that
85 provided unrestricted access to the data was low (n=3); the remainder were closed (no
86 access) (n=38), or categorised as ‘shared or unclear’ (n=38, shared meaning that data
87 sharing is restricted to specific groups or consortium members). The proportion of
88 networks identified for bacteria (excluding TB) and network provider is reproduced in
89 Figure 1. Although this does not equate to the proportion of data generated by each
90 network, it is notable that for bacterial pathogens the majority of networks are led by
91 researchers and pharma, none of which have the capability to upload their data to
92 GLASS. This represents a substantial untapped source of data from settings where the
93 need for surveillance data is greatest, and could provide a short-term solution that
94 bridges the gap between now and the time when routine surveillance capacity has been
95 established.

96 The purpose of this Personal View is to explore how such information
97 generated in countries with weak AMR surveillance systems could be harnessed for
98 patient and public health, including consideration of the strengths and weaknesses of

99 these data, mechanisms to increase their standardization, harmonization and sharing,
100 and the benefits that could be derived by investing in innovation.

101

102 **Alternative sources of AMR data generated in LMIC**

103 *Pharma*

104 Pharmaceutical companies generate a large volume of high quality bacterial
105 susceptibility data before and after new agents are marketed to fulfil regulatory
106 requirements. These data are largely undisclosed, but some companies are now
107 providing aggregated data, including Pfizer who have developed ATLAS
108 (Antimicrobial Testing Leadership and Surveillance), a searchable database on
109 resistance to Pfizer anti-infective agents.⁵ The SENTRY Antimicrobial Surveillance
110 Program is a notable commercial surveillance programme run by JMI Laboratories
111 since 1997 that collects data from more than 200 sites worldwide,⁶ largely from the US
112 and Europe. Findings based on aggregated data of specific species or genera are
113 published, but the dataset is proprietary and not available for wider use. GSK began the
114 Survey of Antibiotic Resistance (SOAR) study in 2002. This focuses on the
115 effectiveness of antibiotics in the treatment of community-acquired respiratory tract
116 infections.⁷ This concentrates in particular on countries and regions for which little
117 other susceptibility data is available, findings from which are regularly reported in the
118 published literature.^{8,9}

119 Pharma data have several strengths. Isolates are obtained from a global
120 distribution. Organism identification and susceptibility testing procedures are strictly
121 standardised and quality controlled according to international standards. Isolates are
122 transported and re-tested in a central accredited laboratory, ensuring reliability and
123 reproducibility of results. However, there are several notable disadvantages. There are

124 often no meta-data (clinical presentation and outcomes or demographic information)
125 associated with the isolates. Organism sampling fulfils the requirements of the
126 particular pharma project rather than being representative or generalizable to the local
127 population, and centralised testing means that quality controlled test data are not
128 available to guide individual patient care, nor do local laboratories benefit from
129 improved quality management as a result of participation. Locations perceived to
130 represent small markets are typically under-represented. Furthermore, there is no
131 information on denominator data and so the results may be biased and may not reflect
132 the true burden of resistance in the area tested.

133

134 *Academia*

135 Researchers generate a wealth of data on antimicrobial resistance in numerous LMIC.
136 The reasons for data generation include the study of infectious disease aetiology and
137 associated antimicrobial susceptibility, therapeutic drug trials, and studies on disease
138 pathogenesis and the molecular biology of bacteria. Some research laboratories have
139 become embedded within district hospitals or other healthcare facilities in LMIC where
140 they provide the only source of ongoing diagnostic microbiology, adopting a model in
141 which research and the provision of microbiological services work in partnership.
142 Compared with pharma data, these often provide greater depth of information for
143 specific populations (e.g. all patients treated in a particular hospital) and many have
144 been operating for several years or even decades.

145 Research data has several potential strengths. Laboratories may be subject to
146 good laboratory practice (GLP) when the information generated is performed to specific
147 standards, supported by laboratories and methods that are rigorously evaluated through
148 robust Quality Assurance and Quality Control programmes. The methods used to

149 generate bacterial susceptibility data are notoriously error prone and data generated
150 outside of quality controlled laboratories may be of sub-optimal quality,^{10,11} and the
151 inclusion of quality-assured research data in national and global databases could
152 increase the proportion of accurate data points. Patient information may also be
153 collected on clinical presentation, duration of hospital stay, antibiotic treatment,
154 complications and outcome. Furthermore, understanding the trajectory for resistance
155 often requires evaluation of susceptibility data over long time-scales, and newer
156 national surveillance programmes may not yet have sufficient retrospective information
157 to make the most of new data.

158 Researcher-defined infectious disease aetiology and common susceptibility
159 patterns, even if performed intermittently, supports empiric prescribing in settings
160 where the treatment of febrile illness is based on clinical features and there is no funding
161 to offer routine testing to patients. Empiric prescribing may result in over use of
162 antibiotics and increased rates of resistance. In the longer term the ideal would be to
163 have a global surveillance programme that promotes laboratory testing for better patient
164 care and directed antibiotic therapy associated with antibiotic stewardship. In the short
165 term, however, empiric prescribing is an essential approach that saves lives, provided
166 that data are sufficiently contemporaneous. Although not always the case, research-
167 driven AMR data may not be generated with sufficient speed to provide information
168 that guides individual patient treatment.

169 When research laboratories are embedded in district hospitals in rural areas (as
170 they often are), inclusion of their data in global databases could also go some way to
171 balancing selection bias that can arise. For example, WHO GLASS requires
172 participating countries to establish at least one surveillance site and then extend the
173 number over time. In LMIC where diagnostic microbiology laboratories are scarce,

174 these are most likely to be situated in tertiary hospitals. Bacteria isolated at tertiary
175 hospitals in any part of the world are more likely to be associated with patients with
176 more severe or complex disease, patients who have received numerous courses of
177 antibiotics, patients with prolonged hospitalisation, and those transferred from other
178 hospitals with hospital acquired infection, all of which will lead to over-representation
179 of bacterial resistance compared with patients at district hospitals or the wider
180 community.

181 Research-generated data also has several potential limitations. Data may be
182 biased, including ascertainment and sampling biases.¹² For example, the study design
183 may target patient sub-sets that do not reflect the wider population with infectious
184 diseases, such as sampling of patients within cohorts that have better access to care, or
185 patients with the most severe infection syndromes. Since a proportion of bloodstream
186 infection will be hospital-acquired, studies of severe invasive disease may inflate rates
187 of resistance and may not capture milder forms of community-associated infection in
188 patients treated as out-patients, which may be caused by organisms with lower
189 antimicrobial resistance. Research data may also include duplicate samples. Six main
190 types of potential bias that may influence the validity or interpretation of surveillance
191 data have been identified, which provides a framework for reviewing the use of research
192 data in AMR surveillance (use of inadequate or inappropriate denominator data; case
193 definitions; case ascertainment; sampling bias; failure to deal with multiple
194 occurrences; and biases related to laboratory practice and procedures).¹³

195

196 ***Private laboratories***

197 A source of susceptibility data that may remain unseen by national surveillance
198 systems, particularly in many LMIC settings, is laboratories in private hospitals that

199 generate data for patient care.¹⁴⁻¹⁶ The quality of data generated by private laboratories
200 varies considerably, but those that are accredited and perform quality assured services
201 produce data of similar or better quality than public laboratories. In India, almost all
202 medical laboratories accredited by the National Accreditation Board for Testing and
203 Calibration Laboratories (NABL) are in the private sector, and in South Africa more
204 than 80% of SANAS (South African National Accreditation System)-accredited
205 medical laboratories are in the private sector.¹⁵ This has led to calls to utilise these data,
206 and the inclusion of such data by initiatives such as ResistanceMap.¹⁷ This displays
207 antimicrobial resistance data on twelve bacterial species isolated in 49 countries,
208 collected between 1999 to 2015 (depending on the country), together with antibiotic
209 consumption data from 75 countries between 2000 and 2014. The primary sources of
210 data are public and private laboratory networks that routinely collect susceptibility
211 results, but data from India comes exclusively from the private sector.

212 Private laboratories can provide extensive datasets for populations for whom
213 there is a very limited supply of reliable AMR data from alternative sources, but again
214 can suffer from the types of bias already described for research data. Furthermore, these
215 laboratories often serve a sub-set of more affluent people, including medical tourists
216 and members of the expatriate community, which may not provide an accurate
217 representation of rates of resistance for similar types of infection in the wider
218 population.

219

220 **Barriers to using alternative sources of AMR data from LMIC**

221 Despite the obvious utility of placing AMR surveillance data generated by academia,
222 pharma or private laboratories into the public domain, very little of these data generated
223 in LMIC is utilised by organisations involved in regional or global surveillance. There

224 are several barriers that prevent this from happening. Data are held in numerous silos
225 with highly restricted access. Academics generate data that usually remains private until
226 published in peer-review journals, when individual patient-level data may not be
227 released and may be delayed by several years from the point of collection because of
228 the time taken to analyse, write and publish. Pharma companies have to jump through
229 several legal hoops before they release their data into the public domain. Even if
230 researchers and pharma companies are keen to deposit data towards global analyses, its
231 aggregation is hampered by a lack of harmonisation in data collection, a lack of tools
232 that allow easy data deposition, and a framework that prevents publication of their data
233 by unscrupulous competitors. Furthermore, GLASS collects and reports data on
234 resistance rates aggregated at national level by Ministries of Health, and cannot
235 currently accept information generated by research activities or pharma. In general,
236 national programmes take ownership of surveillance activities in-country, and
237 agreement may not be reached for direct data deposition to WHO GLASS by non-
238 governmental groups. Furthermore, AMR surveillance data can represent potentially
239 sensitive data, particularly when these describe high rates of resistance or the
240 emergence of a novel resistance mechanism to a key antibiotic.

241

242 **Mechanisms to unlock AMR data**

243 *Incentivising access to data from academia and pharma*

244 ‘Bottom-up’ research and pharma activity that generates AMR data is not public health
245 surveillance in the strict sense. Furthermore, the majority of researchers and pharma-
246 employed scientists would be quick to highlight that public health surveillance is
247 neither their responsibility nor area of interest. Agreeing on the principle that
248 researchers and pharma companies could make a major contribution to global

249 surveillance should be aligned with the recognition that this is not their primary
250 purpose, and will be associated with a financial cost. Debate is required about incentives
251 to support the additional workload associated with sharing data to national programmes
252 or other repositories and who should coordinate this. This could draw on experience
253 gained on academic incentives during the development of WWARN (the WorldWide
254 Antimalarial Resistance Network platform).¹⁸ Any investment should not detract from
255 funding that provides improved data sources for patient care, surveillance and
256 prevention of antimicrobial resistance.

257 The flow of research-generated data to global initiatives could be facilitated by
258 funders, who could develop guidelines on sharing of specific data sets, which could
259 become an integral component of a successful funding award. This is already the case
260 for some forms of data, examples being the submission of all sequence data generated
261 by the Wellcome Trust Sanger Institute into public databases, and funding by the
262 Wellcome Trust linked to an open access publication policy. Such changes would
263 require a clear plan for formatting and destination of data deposition. Journals and
264 publishers could also develop guidelines on data deposition for publications on drug-
265 resistant infection, and could make this a necessary part of submission. Data released
266 into public databases by researchers would need to be protected by data access
267 committees or through other mechanisms, but this is not insurmountable since solutions
268 are already in place for numerous types of data. There are also examples of training and
269 data sharing/open access agreements having been developed that are contextualised and
270 locally acceptable.^{19, 20}

271 The Wellcome Trust have begun to address access to untapped sources of global
272 surveillance data held by pharma through a recently funded project conducted by the
273 Open Data Institute. This has created a mechanism to bring together leaders from public

274 health and the pharmaceutical industry, who are collaborating to explore how value
275 could be added by the re-use of available data. An evaluation of the mechanisms and
276 barriers to making this open access has been completed and detailed in a post-project
277 report.²¹ One proposal is to create a public-private partnership to improve local
278 laboratory capacity. Another is to suggest that relevant metadata and denominator data
279 are also collected, fulfilling the objectives of the pharma project whilst also informing
280 local AMR prevalence, which while not informing individual patient care could inform
281 empiric prescribing protocols.

282

283 *Supporting Ministries of Health to access data*

284 Ministries of Health could collaborate with research institutions where this is not the
285 case, or a public-private partnership could be forged so that data generated by research,
286 pharma or private laboratories efforts can be submitted to GLASS as national data.
287 There are examples of research units in Asia and Africa that have already developed
288 close and sustainable working relationships with the relevant Ministry of Health, who
289 use the information provided to shape national prescribing policies. In this way,
290 countries can be empowered to utilize data generated in their own country, with local
291 researchers undertaking in-depth analyses using a range of sources, encouraging
292 comparisons of incidence and prevalence rates of drug-resistant infection between
293 different areas in the country to monitor disease burden and impact of action plans in
294 each area. This also represents an important training opportunity for government staff,
295 who can develop the technical capability to analyse data that is ultimately generated
296 through country-led capacity building.

297

298 *The need for specialist networks*

299 Having argued that the development of new initiatives that effectively replicate WHO
300 GLASS and that fragment data are generally to be discouraged, there are some
301 circumstances when additional networks add vital new information. A notable example
302 is the Institute for Health Metrics and Evaluation, which has recently been funded by a
303 joint award from the Wellcome, the UK Fleming Fund and the Bill and Melinda Gates
304 Foundation to gather, map and analyse disease and death caused by drug-resistant
305 infections. This will be used to quantify the global burden of disease (GBD) from drug-
306 resistant infections compared with other diseases and causes of death, and so inform
307 policy and decision-making. Estimating GBD for AMR faces numerous challenges,
308 including difficulties in linking surveillance data with clinical or outcome data and
309 causal attribution, and cannot be regarded as a routine surveillance activity at present.

310

311 **Investing in innovation**

312 Investment is needed to promote innovation in AMR surveillance. For example,
313 harnessing emerging technologies relating to Big Data and Artificial Intelligence could
314 lead to more effective mechanisms of AMR data capture, sharing and analysis tools.
315 An innovative system to support automatic data harmonisation between different
316 laboratories and institutions could achieve numerous objectives, including an inbuilt
317 system to standardise data analysis and quality tests for data from multiple sources;
318 capture of patient outcome data to underpin calculations of GBD; and rapid relay of
319 information to treating clinicians e.g. via electronic decision support algorithms. Data
320 could be automatically linked to national agencies and international data repositories.
321 Innovation in data capture would benefit from early involvement of experts in the social
322 sciences so that the behaviour change required to support buy-in is an integral part of
323 planning and development. Mapping of data sources may also require consideration of

324 the regulatory environment in some settings. New innovation needs to be linked to
325 effective translation, scale up and integration, and assessed in terms of impact on
326 policy.²² Any alternative system developed will need to be either fully inter-operable
327 with WHO GLASS, or able to generate data in a format that can be uploaded.

328

329 **Conclusions and next steps**

330 Our understanding of the global burden of disease from drug-resistant infection in
331 LMICs is rudimentary, and data from academia, pharma and private laboratories could
332 make an important and rapid contribution. A dialogue is required to determine how data
333 generated in LMICs might flow from these bodies to national and global AMR
334 surveillance networks, and how they might support the development of laboratory and
335 analytical capacity, including robust quality management systems, for prospective data
336 collection. This should build on current initiatives such as the Fleming Fund, which is
337 providing regional grants to collect existing AMR and antimicrobial use data from all
338 possible sources. Table 1 summarises proposed changes that could help to bring this
339 into effect.

340

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345

346 **Transparency declaration**

347 Nothing to declare

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404 **FIGURE LEGENDS**

405 **Figure 1. AMR surveillance networks since 2000**

406 Sunburst chart representing 44 supranational networks performing AMR surveillance
407 in bacteria (not including TB) categorised according to their lead organisation type
408 (Pharma, academia, WHO/governmental). Adapted from reference 1.

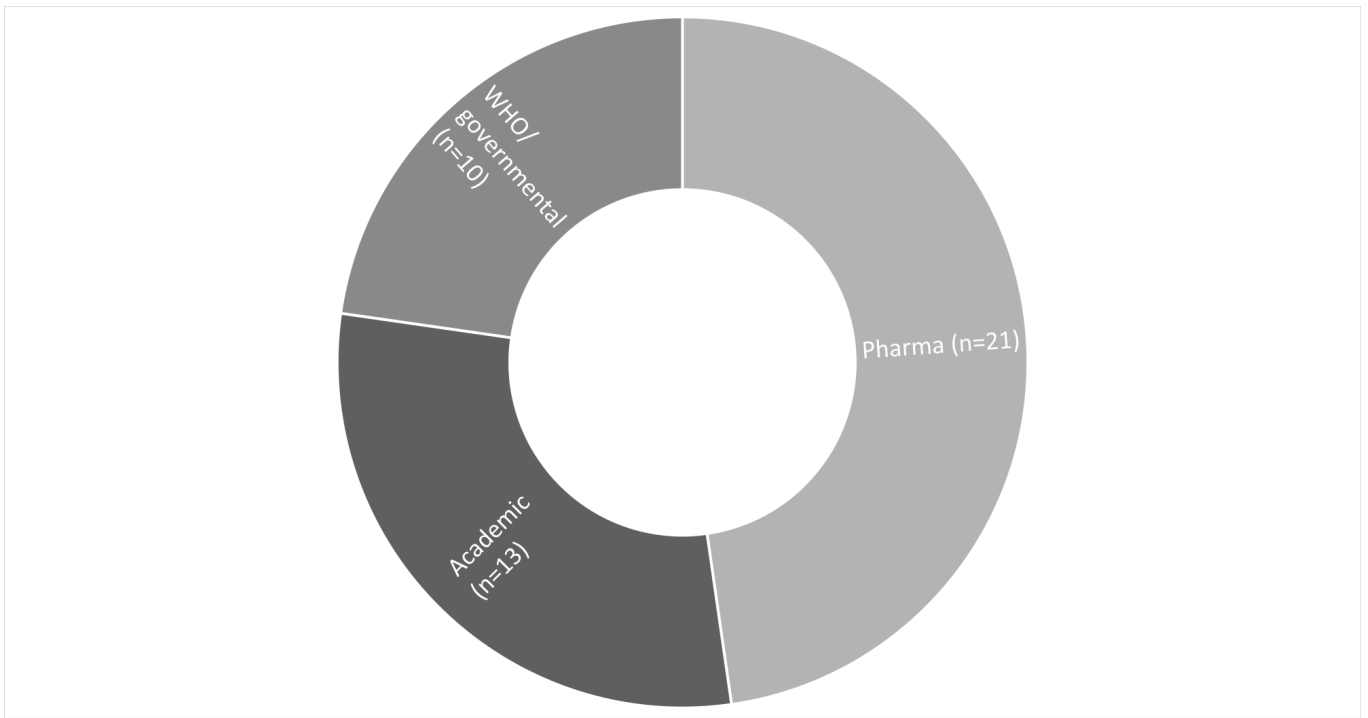
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Table 1. Key actions to harness AMR data from alternatives sources in LMIC

Objectives	Actions
Map data	<ul style="list-style-type: none">• Map and evaluate quality/utility of data held and generated• Determine how to enhance these resources, i.e. through the addition of patient outcome data• Determine how data can contribute to the global burden of disease due to AMR
Create mechanisms for data sharing and capacity building	<ul style="list-style-type: none">• Identify incentives that promote the contribution of data from academic, pharma and private laboratories• Agree the basis for data sharing, including ownership, ethical and legal considerations• Develop mechanisms for data harmonisation, collation and analysis• Promote private – public partnerships to build capacity in local laboratories for patient care and surveillance
Facilitate update of data nationally and internationally	<ul style="list-style-type: none">• Use data to support National Action Plans• Seek mechanisms and create funding opportunities to support uptake of academia/pharma/private lab data by WHO GLASS and other data sharing initiatives
Innovation in data capture, analysis and sharing	Create a data collection interface that supports: <ul style="list-style-type: none">• Case-based surveillance• Quality assurance and control and a universal reporting standard for patient data• Automated linkage to national agencies and international data repositories

411

412 **Figure 1**



413