

Antibiotics and Activity Spaces:**Protocol of an Exploratory Study of Behaviour, Marginalisation, and Knowledge Diffusion**

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ABSTRACT

Background: Antimicrobial resistance (AMR) is a global health priority. Leading UK and global strategy papers to fight AMR recognise its social and behavioural dimensions, but current policy responses to improve the popular use of antimicrobials (e.g. antibiotics) are limited to education and awareness-raising campaigns. In response to conceptual, methodological, and empirical weaknesses of this approach, we study people's antibiotic-related health behaviour through three research questions:

RQ1. What are the manifestations and determinants of problematic antibiotic use in patients' healthcare-seeking pathways?

RQ2. Will people's exposure to antibiotic awareness activities entail changed behaviours that diffuse or dissipate within a network of competing healthcare practices?

RQ3. Which proxy indicators facilitate the detection of problematic antibiotic behaviours across and within communities?

Methods: We apply an interdisciplinary analytical framework that draws on the public health, medical anthropology, sociology, and development economics literature. Our research involves social surveys of treatment-seeking behaviour among rural dwellers in northern Thailand (Chiang Rai) and southern Lao PDR (Salavan). We sample approximately 4,800 adults to produce district-level representative and social network data. Additional 60 cognitive interviews facilitate survey instrument development and data interpretation. Our survey data analysis techniques include event sequence analysis (RQ1), multilevel regression (RQ1-3), social network analysis (RQ2), and latent class analysis (RQ3).

Discussion: Social research in AMR is nascent, but our unprecedentedly detailed data on micro-level treatment-seeking behaviour can contribute an understanding of behaviour beyond awareness and free

73 choice, highlighting for example decision-making constraints, problems of marginalisation and lacking
74 access to healthcare, and competing ideas about desirable behaviour.

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SUMMARY BOX

- Antimicrobial resistance (AMR) is a global health priority, and leading UK and global strategy papers recognise its social and behavioural dimensions.
- Behavioural elements of these strategy papers have conceptual, methodological, and empirical weaknesses.
- We will carry out social research to understand the nature of antibiotic-related treatment-seeking behaviour in rural Thailand and Lao PDR.
- We will conduct survey research with 4,800 adult villagers (yielding district-level representative and social network data), complemented with cognitive interviews and secondary administrative data.
- Our study will contribute to the nascent yet urgently needed social research in AMR.

BACKGROUND

Access to non-prescription antibiotics is a widespread phenomenon in low- and middle-income countries (LMICs) [1], contributing to inappropriate medicine use, to the development of antimicrobial resistance (AMR), and potentially to the subsequent spread of resistant bacteria across the world [2]. Leading global and UK policy papers aiming to deal with the over- and misuse of antibiotics among the general population focus thereby wholly on educational and awareness-raising campaigns to encourage positive behaviour change [3-6]. Awareness-raising is important [7], but as the sole global strategy focusing on people's healthcare-seeking behaviour (aside from public health interventions to prevent illness) it has three central weaknesses.

97 The first is conceptual: By limiting our attention to “awareness” as the main driver of people’s
98 antibiotic use, we are prone to neglecting determinants of health behaviour beyond information and
99 free choice such as economic constraint, social pressure, or local conceptions of illness [8-11].
100 However, little is known about how economic constraints, social discrimination, or spatial
101 marginalisation deprive people of choices and drive them into seemingly adverse antibiotic-related
102 behaviours, and whether and how interventions should address these constraints in the context of
103 global AMR policy.

104 The second is methodological: Quantitative community- and population-level analyses of antibiotic
105 usage disregard routinely that healthcare processes involve combinations of “no care,” “self-care,” and
106 healthcare from many different practitioners [12]. Although conceptually established and applied in
107 qualitative research [13 14], the sequential understanding of treatment-seeking behaviour has not yet
108 entered quantitative public health research. The majority of quantitative analyses of healthcare
109 behaviour in low- and middle-income countries instead adopt a single-stage approach, implying that a
110 patient “chooses” once from a portfolio of healthcare options, some of which may be more likely to
111 involve antibiotics than others [15 16]. This conventional analysis can be useful to measure rates of
112 antibiotic access, but their aggregate nature forgoes valuable information and obscures the factors
113 influencing antibiotic use throughout an illness, for example the use of information technology to
114 gather information [as we demonstrate in 17].

115 The third is empirical: Studies of awareness campaign effectiveness focus on knowledge gains but
116 disregard the social mechanisms of information diffusion [see e.g. 18]. Awareness campaigns often
117 expect information to spread within communities, but these communities are not always collaborative
118 [19]. In addition, as a potential solution in a healthcare-seeking problem, new information about
119 antibiotic use also competes with other ideas, some of which may represent dominant healthcare

120 strategies from the individual's perspective [20].¹ We do not understand sufficiently how these
121 interdependencies play out during the diffusion process of antibiotic knowledge and practice. It
122 therefore appears risky to confine our behavioural tactics to the single mechanism of awareness raising
123 that is merely believed to function.

124 In response to these conceptual, methodological, and empirical challenges, we intend to improve the
125 understanding of patients' antibiotic-related behaviour to support creative thinking about targeted and
126 unconventional AMR interventions in LMICs. Three research questions will guide our enquiry:

127 RQ1. What are the manifestations and determinants of problematic antibiotic use in patients'
128 healthcare-seeking pathways?

129 RQ2. Will people's exposure to antibiotic awareness activities entail changed behaviours that diffuse
130 or dissipate within a network of competing healthcare practices?

131 RQ3. Which proxy indicators facilitate the detection of problematic antibiotic behaviours across and
132 within communities?

133 We adhere to conventions in the field of public health when using the contentious language of
134 "problematic" and "appropriate" behaviour. However, from a behavioural perspective, a conventional
135 clinical definition of "problematic behaviour" is impractical to pursue because it would involve claims
136 on the mis-/match between a patient's condition (e.g. being caused by a particular micro-organism)
137 and the type, dosage, duration, and affordability of the administered drugs. Patients are not necessarily

¹ Information is also unlikely to remain static during the diffusion process; utilisation and (re-)interpretation of educational messages can alter nature, meaning, and value of this information over time and across people. Our focus on healthcare-seeking behaviour and attitudes towards antibiotics can capture unintended behavioural responses to a limited extent, but our survey is unable to study the implications and mechanisms underlying such potential transformations comprehensively (for which qualitative and ideally ethnographic research would be better suited). We thank an anonymous reviewer for pointing this out.

138 able to diagnose themselves, decide whether an antibiotic is needed, and then select the clinically most
139 suitable course of treatment. Indeed, many illnesses do not involve a doctor at all. Considering that
140 “problematic behaviour” is subjective and context specific, we instead record patients’ behavioural
141 trajectories during an illness and apply different evaluative criteria to make judgements of
142 “appropriateness” flexibly and transparently. In consultation with the social anthropologists, medical
143 practitioners, and local field staff in our study team, we will categorise individually as well as socially
144 “appropriate” behavioural sequences that go beyond binary assessments of healthcare access or
145 antibiotic use. For example, individually rational bypassing of referral systems could entail healthcare
146 resource misallocation from a public health perspective, and individual antibiotic use can entail
147 negative externalities on the societal level through potential contributions to antibiotic resistance. Our
148 interest in human behaviour thereby does not intend to attribute blame to patients for patterns of
149 antibiotic usage that might contribute to AMR, but rather to explore decisions and decision-making
150 constraints on the healthcare demand side. The provision of the raw behavioural sequences will allow
151 other researchers to make their own evaluations of health behaviours depending on their specific
152 assessment criteria and interests.

153 **METHODS**

154 **Theoretical Framework**

155 Our study departs from conventional policy assumptions that antibiotic misuse among patients stems
156 from a lack of knowledge regarding appropriate medicine use. Instead, we frame antibiotic use as one
157 among multiple solutions in people’s healthcare “activity space.”

158 Contrary to existing applications of activity space frameworks in areas like disability and mobility [21-
159 23] and social geography [24-26], we do not adhere to spatial conceptualisations through which e.g.
160 experienced space is linked to health outcomes like obesity or HIV [24 26]. Instead, we draw on

161 theoretical strands and techniques from the disciplines of public health, medical anthropology,
162 sociology, and development economics, which suggest that healthcare behaviour takes place within a
163 physical and social space populated by various healthcare providers (including drug vendors), and that
164 this space is defined by the difficulty and the perceived and dynamically changing value of utilising
165 any of these providers during an illness. Difficulty is determined by the tools and solutions at the
166 patients' disposal (e.g. social support networks, cars, communication technology), but not every
167 solution affects access to different providers equally. In addition, patients might not be aware of every
168 provider in their vicinity (they therefore do not enter the activity space), and some providers signal
169 better healthcare value than others, depending on type and severity of the illness. Moreover, the activity
170 space overlaps across patients and it is thus a shared space. These characteristics lead us to identify
171 three key elements of antibiotic usage in a healthcare activity space: (a) the emergence of pathways
172 through the health system during the course of an illness; (b) the co-existence of multiple solutions for
173 the health problem, the value of which changes dynamically; and (c) cooperation, competition, and
174 exclusion in a shared social space. As a result, our definition of activity spaces can be likened to
175 "markets" in the strategic marketing literature [27], where markets can be delineated by different
176 customer groups (in the case of health, e.g. different socioeconomic groups), the customer function to
177 be served (e.g. curative care), and the "alternative solutions" available to fulfil this function (e.g.
178 antibiotic use at home, care from primary health centre, sick leave).

179 The breadth of the activity space framework allows us to consider multiple, and otherwise conceptually
180 more restricted, explanatory approaches for treatment-seeking behaviour side-by-side [e.g. transaction
181 cost approaches alongside the information deficit arguments that often underlie policy narratives; 28
182 29]. The conceptualisation as a shared social space also permits us to go beyond conventional
183 individualistic treatment-seeking models in order to explore new forms of health-related collective
184 action problems. In addition, our framework permits us to examine the determinants of problematic
185 behaviour and the positive as well as negative outcomes of technology use, rather than merely

186 articulating the enabling conditions of desired behaviour change as is common in the public health
187 literature [30-32]. Activity spaces are therefore not a theory per se, but a useful analytical domain to
188 guide our research.

189 **Research design**

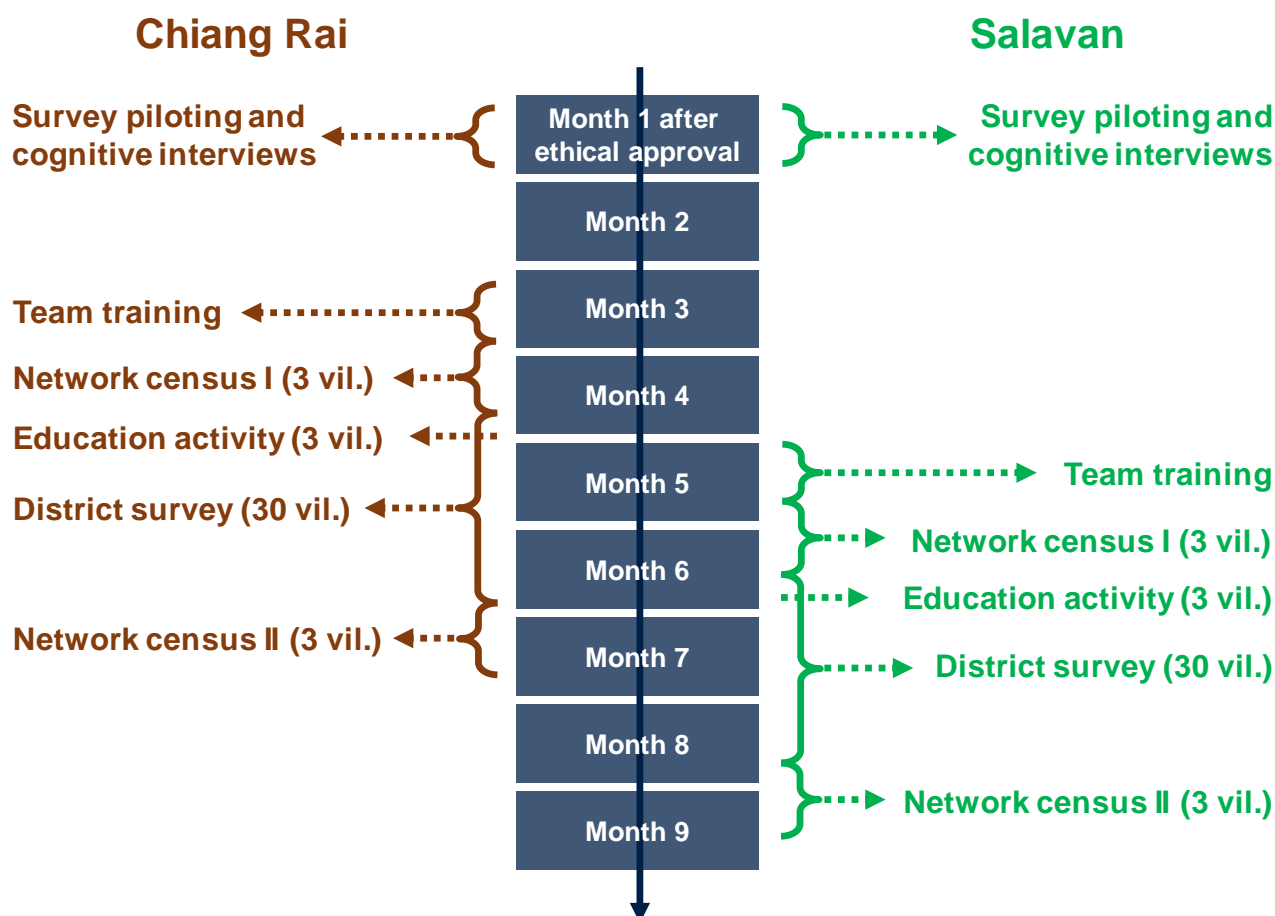
190 We will carry out population- and community-level health behaviour surveys in rural Chiang Rai
191 (Thailand) and Salavan (Lao PDR), and we collect complementary qualitative data. This will result in
192 two survey data sets: the first contains district-level representative treatment-seeking behaviour of
193 approx. 1,200 adults across 30 rural communities per country; the second comprises social network
194 censuses of approx. 400 adults each in three rural communities per country. Within the sampled
195 villages, we will complete checklists about existing formal and informal healthcare facilities and gather
196 patient load data from primary care units catering to the respective villages. As part of the questionnaire
197 testing process, we will conduct (and collect as primary data) cognitive interviews to improve the
198 survey tool and to interpret our data.

199 We will carry out the district-level village survey in one round, and the village-level social network
200 censuses in two rounds (see Figure 1). Between the two village social network censuses, we will carry
201 out education activities in the selected villages as part of antibiotic- and medicine-use-related public
202 engagement. Developed after a year of qualitative health behaviour research across Southeast Asia,
203 these small-scale activities aim to help villagers learn more about drug resistance and to help the social
204 and medical research communities to better appreciate local people's access to healthcare and medicine
205 conceptions and constraints. The activities will take place after the network surveys in each of the
206 network villages, lasting one to two days, involving approximately 30 villagers each, and including
207 interactive sessions like trading games, poster making, storytelling, and role plays.² The district survey

² While we do not expect regular social networks to evolve as a result of our educational activity, we will be able to ascertain whether people activate different social contacts when seeking treatment before and after our activity. In addition, we will

208 will take place after the education activities, and we will subsequently re-survey all adults in the
 209 network villages (i.e. two to three months after the first network village survey round).

210



211

212 Figure 1. Study Design and Timeline.

213 Source: Authors.

214 **Study participants**

215 Groups included in this research comprise adults (aged 18 years and above) in rural Salavan and rural
 216 Chiang Rai. We focus on Thailand and Lao PDR because they are situated in a region that experiences
 217 high rates of antibiotic use and increasing resistance [4 33 34]. Compared to Lao PDR, Thailand

also learn whether respondents who did not participate in the activity learned about it through their regular social network or from other contacts.

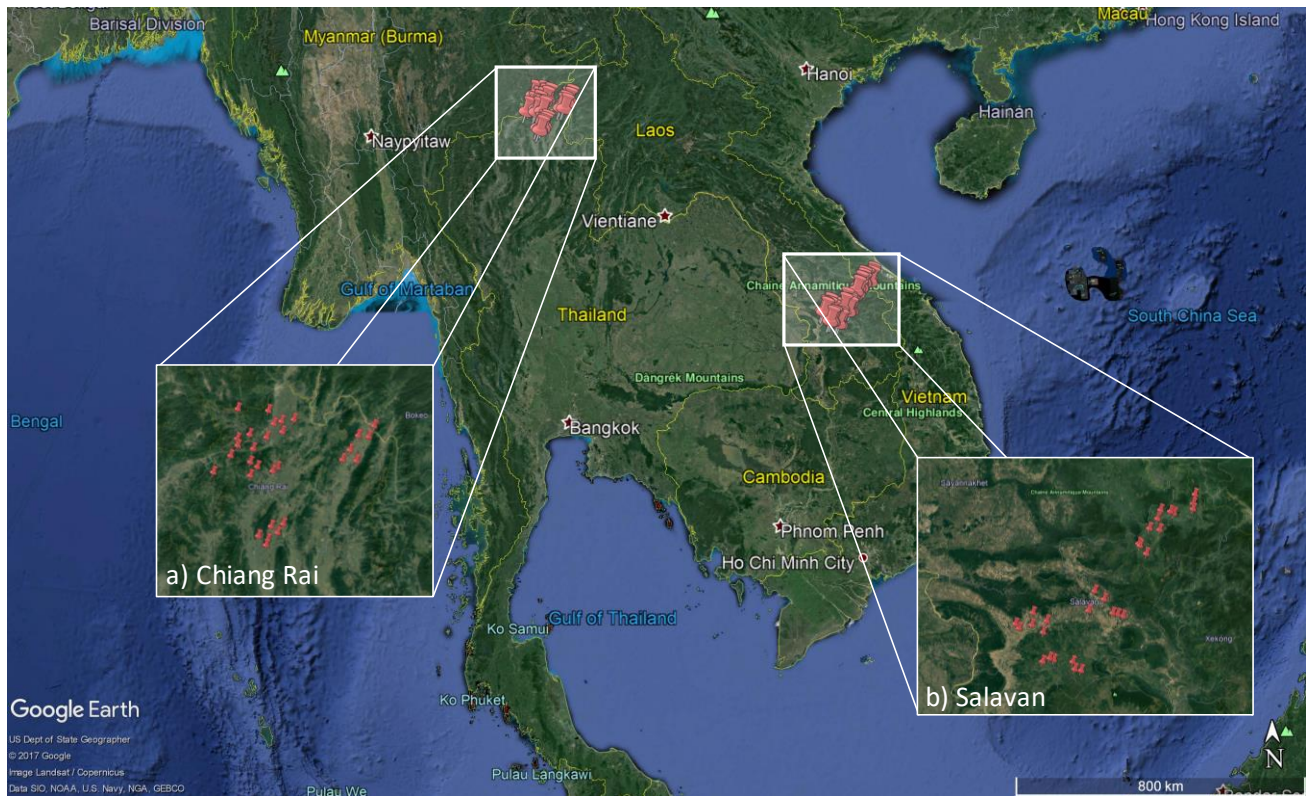
218 exhibits a more advanced economic and health system context and more established AMR
219 stewardship. A comparative study of these low- (Lao PDR) and middle-income (Thailand) contexts
220 therefore offers interesting lessons for domestic and global antibiotic policy. We focus on adults
221 because they account for much of the popular antibiotic demand and typically acquire and administer
222 antibiotics on behalf of children [35]. Lastly, we focus on rural areas of Chiang Rai and Salavan
223 because their formal and informal health systems face relatively high infrastructural, human resource,
224 financial, and regulatory constraints; while their inhabitants are more often characterised by economic,
225 social, and spatial marginalisation. This does not automatically mean that our entire study population
226 qualifies as “marginalised” since we define marginalisation in relative terms of wealth (e.g. bottom
227 quintiles of household asset and amenity indices), social position (e.g. within village social networks),
228 and geography (e.g. distance to urban centres). However, it is important to note that our study
229 implications will speak to rural areas with their specific constraints and patterns of marginalisation,
230 which are systematically different from urban settings.

231 **Data collection**

232 The district-level representative survey will be conducted in a three-stage stratified cluster random
233 sampling design. A cluster sample is necessary to ensure the logistical and financial feasibility of the
234 survey, and we aim to reduce its negative implications for the effective sample size through
235 stratification, which helps to increase the information contained in each cluster [36]. The first stage
236 involves the random selection of 30 villages (clusters) across five purposively selected districts in each
237 site, stratified by their distance to the nearest urban centre [using data from 37]. Figure 2 depicts the
238 resulting village samples. The second stage enumerates all residential buildings within the selected
239 villages using satellite imagery from *Google Maps* and *Bing Maps*, of which we sample 5% of the
240 buildings (but at least 30 houses) in a stratified interval sampling approach to ensure spatial

241 representativeness. During the survey implementation, the third sampling stage will select randomly
 242 one respondent for every five adults in each chosen house.

243



244

245 Figure 2. Field Sites and Sampled Villages in Thailand and Lao PDR

246 Source: Authors, adapted from Google Inc. [38]. 2017 map data from Landsat / Copernicus.

247

248 The sampling strategy for the community-level social network census surveys involves the purposive
 249 selection of three comparable villages in both countries. Selected in consultation with local
 250 stakeholders, guiding criteria for selection were (a) village size and structure, (b) remoteness and road
 251 accessibility, (c) economic status as approximated by village-level infrastructure and facilities, (d)
 252 ethnic composition, and (e) number and location of health facilities within a 2km radius. The villages
 253 are estimated to have a size of 100 to 200 households with 2.9 adults per rural household in Lao PDR
 254 (ranging from 1.8 to 4.6 per village) and 2.4 in Thailand on average [ranging from 1.6 to 4.3 per village;

255 39 40]. Within the selected communities, all households will be approached, their adult members
256 enumerated, and invited to participate.

257 Our survey instrument will be a 45-minute questionnaire that captures people's complex healthcare-
258 seeking pathways and their medicine use therein. An important feature of this instrument is the
259 collection of self-reported sequential healthcare pathway data for acute illnesses and accidents that
260 occurred in the two months prior to the interview. As shown in Figure 3, we will subdivide an illness
261 into discrete steps of activities and record their type, duration, and location; with whom the patient
262 interacted during the healthcare activity; whether the patient used any medicines during the step
263 (elicited using a "drug card" containing the most common local medicines), their source, and how
264 long, often, and at what dosage they were taken; and whether, why, and by whom any kind of health-
265 related mobile phone, Internet, media, or vehicle use took place.³ We also collect data on the social
266 and economic background of the respondents as well as information about people's existing
267 conceptions of and attitudes towards antibiotics. In the case of the social network censuses, we will
268 ask additional questions to construct four different kinds of (health-related) social networks:

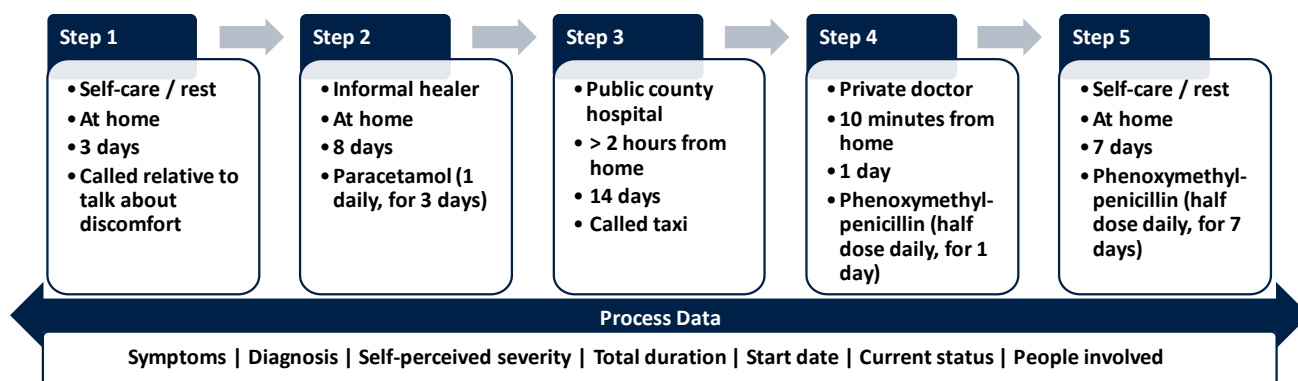
- 269 a) (Health) communication networks: People within and outside the village with whom the
270 respondent interacts and talks about health (elicited in first network survey round).
- 271 b) Incidence network: Places where the respondent typically interacts with other villagers (elicited
272 in first network survey round).
- 273 c) Help network: Contacts who are activated during an illness (elicited in both network survey
274 rounds).

³ Due to language ambiguities in the term "antibiotic," we do not ask people directly whether they took an antibiotic, but rather which medicine they received and used at each step in the treatment-seeking process. Pilot interviews suggested that a part of the rural population is able to describe medicine unambiguously (e.g. using colloquial or technical terms for antibiotics), while other groups do not know what medicine they received. In the latter case, we triangulate their responses with their descriptions of antibiotics (e.g. as "sore throat medicine," which we also capture in the questionnaire), and we ask for descriptions of these medicines (e.g. "a shot," "a green-blue capsule"). We then code the medicines for sensitivity analysis into categories "likely to be an antibiotic," "unlikely to be an antibiotic," and "cannot ascertain type of medicine" according to common medicines used in the field sites.

275 d) Information network: People with whom the respondent talked about the public engagement
 276 activity (elicited in second network survey round).

277 The questionnaire will be administered through tablet-based electronic data collection by locally
 278 recruited survey teams comprising six enumerators and two survey supervisors per country. The survey
 279 period will be between November 2017 and April 2018, which is the post-rice-harvest dry season in
 280 both field sites. This season was chosen for village accessibility (landslides and floods are common
 281 during the rainy season) and the availability of villagers for interviews (villagers often reside
 282 temporarily near their rice fields during the planting and harvest seasons). Due to the temporal focus
 283 on one season, our survey is therefore not able to capture seasonal change affecting the epidemiological
 284 environment, internal migration, healthcare-seeking patterns, and interactions within social networks.

285



286

287 Figure 3. Example of Treatment-Seeking Pathway Data Collected in the Survey.

288 Source: Authors, adapted from [41].

289

290 We will pilot the questionnaire to identify respondents' understanding of the survey questions and the
 291 range of possible answers. Between 10 and 30 one-hour cognitive interviews per site will support this
 292 process and enable insights into how respondents understand survey questions and how they arrive at
 293 their answers [42]. The qualitative data generated through the cognitive interviews will also facilitate
 294 the interpretation of the quantitative survey results.

295 Two further sources of data will complement our survey data. First, in order to understand antibiotic-
296 seeking behaviour in the local health systems, separate checklists will help us to gather observational
297 information about the location of formal and informal healthcare providers in each village. Secondly,
298 we will estimate patient load and peak demand for public healthcare services by accessing secondary
299 administrative data from public primary care facilities that cater to the sampled villages.

300 **Analysis**

301 The data analysis techniques to inform our research questions include:

- 302 • RQ1: Sequence analysis to describe and understand linear series of events [41 43 44],
- 303 • RQ1-3: Multilevel regression analysis to test the relationship between antibiotic use as a
304 dependent variable, and a range of established determinants of treatment-seeking behaviour as
305 independent variables [45],
- 306 • RQ2: Social network analysis (network-based event history and relational event sequence
307 analysis) to examine how behaviours and beliefs across a social network relate to individual
308 behaviours and beliefs, and how this relationship persists over time [46-48], and
- 309 • RQ3: Latent class analysis to identify (a) common symptoms associated with problematic
310 antibiotic usage, (b) the characteristics of populations who are likely to exhibit problematic
311 antibiotic behaviours, and (c) contextual conditions predicting adverse behaviours, all of which
312 may help guide future interventions and policies [49].

313 Related in particular to Research Questions 1 and 2, we will further examine six hypotheses about
314 antibiotic use among the general population:

- 315 H1. Marginalised groups have fewer means to access formal treatment, which increases their
316 likelihood to rely on over-the-counter medicines including antibiotics as an alternative solution.

- 317 H2. Technology use increases access to formal healthcare providers but is directed towards those
318 who are more inclined to prescribe antibiotics.⁴
- 319 H3. Awareness about “rational antibiotic use” alone has only a minor influence on antibiotic usage
320 behaviour if patients are economically, socially, or spatially marginalised.
- 321 H4. In the absence of competing healthcare practices, new antibiotic-related behaviours diffuse
322 through social networks.
- 323 H5. Pre-existing competing practices subdue the spread of new antibiotic-related behaviours within
324 the community network if no “critical mass” can be achieved.
- 325 H6. Peak demand for scarce high-quality healthcare providers drives less competitive (i.e. more
326 marginalised) patients into behaviours that are more likely to involve adverse antibiotic use.

327 Note that these hypotheses do not intend to declare the behaviour of marginalised groups to be
328 “irrational.” Rather, we hypothesise that the behaviour of marginalised groups is subject to greater
329 healthcare access constraints, owing to which antibiotic use might be more likely. Whether this is
330 indeed the case, and whether these behaviours are less appropriate than otherwise, are empirical
331 questions that we hope to inform through our survey.

332 **DISCUSSION**

333 **Ethical Considerations**

334 **Informed Consent**

335 We received a waiver for written consent requirements in order to not unfairly exclude illiterate
336 population sub-groups in our field sites [50], and to ensure trust between the researcher and the rural

⁴ The interest in technology use is rooted in our activity space framework, according to which different healthcare providers (e.g. private doctors) are more responsive to patients’ use of e.g. mobile phones, which can affect treatment-seeking pathways and, potentially, antibiotic use.

337 respondents [51]. Instead of participant-dated signature, we follow a verbal consent process in which
338 (1) we seek permission from village leaders to carry out our survey in their villages; (2) the survey
339 fieldworker reads out (and records on audio tape) an oral consent script to the potential respondent and
340 provides them with a printed copy of the Participant Information Sheet; (3) the survey fieldworker asks
341 the respondent to state her or his consent, name, and date on audio record; and (4) the survey
342 fieldworker personally signs and dates a written record of oral consent. We provide a detailed
343 justification and explanation of this verbal consent process in Appendix 1.

344 Privacy and Confidentiality

345 Further ethical considerations in this study relate to privacy and confidentiality. The data collected in
346 this study include self-reported health and economic information. Personal contact details will be
347 stored separately from the data sets in order to match data from repeated network survey rounds. Any
348 identifying information will be deleted from the analytical data set or coded into anonymous
349 respondent numbers for the social network census survey data set. Geographical data allowing
350 household identification will be translated into distance measures and a village-centred metric
351 coordinate system (similar to the Universal Transverse Mercator system). Should village layouts prove
352 idiosyncratic so that the metric coordinate system enables identification, we will withdraw these data
353 from the data sets.

354 **Proposed Impact**

355 The academic impact of our study pertains to antibiotic-related behaviour and its relationship to
356 marginalisation, technology, and social relationships. Our innovations therein are theoretical
357 (development of the activity space framework to conceptualise and situate people's antibiotic access
358 and use during illness), methodological (sequence analysis for healthcare pathways), and empirical
359 (novel insights into the impact of marginalisation, technology, and knowledge on antibiotic usage). In
360 addition, we will build capacity for social research in AMR for instance through four internships for

361 local candidates from Southeast Asia who consider interdisciplinary academic careers, and our project
362 aims to influence the global health discourse about AMR for example by hosting four eight-week
363 student placements with the MSc International Health and Tropical Medicine (IHTM; a global health
364 degree at the University of Oxford).

365 In conclusion, social research in AMR is nascent, but our unprecedentedly detailed data on micro-level
366 treatment-seeking behaviour promises to contribute to understanding behaviour beyond awareness and
367 free choice, highlighting for example decision-making constraints, problems of marginalisation and
368 lacking access to healthcare, and competing ideas about desirable behaviour.

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376 AUTHOR CONTRIBUTIONS

377 MJH conceptualised the study, its design, and its theoretical framing. All authors provided inputs into
378 the study design. MJH and NC developed the study protocol. MJH drafted the manuscript. All authors
379 approved the protocol.

380 COMPETING INTERESTS

381 We declare that no conflict of interest – financial or otherwise – exists.

382 ETHICS APPROVAL

383 The protocol has been reviewed and approved by the University of Oxford Tropical Research Ethics
384 Committee (Ref. OxTREC 528-17), and it received local ethical approval in Thailand from the Mae
385 Fah Luang University Research Ethics Committee on Human Research (Ref. REH 60099), and in Lao
386 PDR from the National Ethics Committee for Health Research (Ref. NEHCR 074). Written
387 permissions for the use of secondary administrative data from public primary healthcare facilities will
388 be obtained during the course of the study.

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