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Human-animal interactions and bat coronavirus spillover potential among rural residents in Southern China

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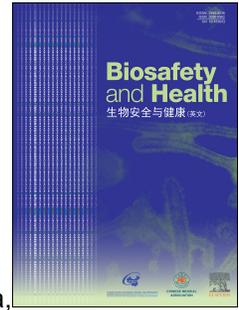
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1 **Human-animal interactions and bat coronavirus spillover potential among rural residents**  
2 **in Southern China**

3

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7

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25

26

27 **Abstract**

28 Human interaction with animals has been implicated as a primary risk factor for several high  
29 impact zoonoses, including many bat-origin viral diseases; however, the animal-to-human  
30 spillover events that lead to emerging diseases are rarely observed or clinically examined, and  
31 the link between specific interactions and spillover risk is poorly understood. To investigate this  
32 phenomenon, we conducted biological-behavioral surveillance among rural residents in the  
33 Yunnan, Guangxi, and Guangdong provinces of Southern China, where we have identified a  
34 number of SARS-related coronaviruses in bats. Serum samples were tested for four bat-borne  
35 coronaviruses using newly developed enzyme-linked immunosorbent assays (ELISA). Survey  
36 data were used to characterize associations between human-animal contact and bat  
37 coronavirus spillover risk. A total of 1,596 residents were enrolled in the study from 2015 to  
38 2017. Nine participants (0.6%) tested positive for bat coronaviruses. 265 (17%) participants  
39 reported severe acute respiratory infection (SARI) and/or influenza-like illness (ILI) symptoms in  
40 the past year, which were associated with poultry, carnivore, rodent/shrew, and bat contact, with  
41 variability by family income and province of residence. This study provides serological evidence  
42 of bat coronavirus spillover in rural communities in Southern China. The low seroprevalence  
43 observed in this study suggests that bat coronavirus spillover is a rare event. Nonetheless, this  
44 study highlights associations between human-animal interaction and zoonotic spillover risk.  
45 These findings can be used to support targeted biological behavioral surveillance in high-risk  
46 geographic areas in order to reduce the risk of zoonotic disease emergence.

47

48 **Key words**

49 Bat coronavirus, human-animal interaction, disease emergence, Southern China, rural  
50 community

**51 Highlights****52 Scientific question**

53 What are the behavioral risks in human-animal interactions that could lead to the emergence of  
54 bat coronaviruses in human population.

**55 Evidence before this study**

56 Bat borne coronaviruses have caused several emerging infectious disease outbreaks of global  
57 significance, including SARS. Novel SARS-related coronaviruses have been discovered in bat  
58 populations in South China, some of which have the capacity to infect human cells. Human-  
59 animal interactions are thought to be critical for the emergence of bat coronaviruses, however  
60 the specific interactions linked to animal-to-human spillover remain unknown.

**61 New Findings**

62 This study found serological evidence for bat-borne coronavirus transmission to people. Direct  
63 contact with bats was not identified as a risk factor. However, self-reported severe acute  
64 respiratory infection (SARI) and/or influenza-like illness (ILI) was linked to human interaction  
65 with other wildlife and livestock, suggesting that there may be other zoonotic exposures leading  
66 to clinical illness in these populations.

**67 Significance of the study**

68 Findings from this study suggested that an integrated biological and behavioral surveillance in  
69 healthy community settings can help identify potential zoonotic disease spillover events or target  
70 surveillance to at-risk populations. This approach represents a potential early-warning system  
71 that could be used under non-outbreak conditions to identify potential zoonotic emerging  
72 diseases prior to largescale outbreaks.

## 73 1. Introduction

74 In the highly biodiverse southern region of China, interactions among humans, wildlife, and  
75 livestock are likely to be common, and are hypothesized to be a risk factor in the emergence of  
76 zoonotic infectious diseases [1-3]. Human-animal interactions may pose a particular public  
77 health threat in rural communities where frequent contact with animals occurs and where  
78 disease prevention measures are likely less well-developed [4]. Although human-animal  
79 interactions are thought to be associated with zoonotic disease emergence, few studies have  
80 addressed the nature of specific interactions that occur between animals (particularly wild  
81 animals) and humans that lead to pathogen spillover.

82  
83 Bats (order Chiroptera) are reservoirs of a large number of zoonotic viruses, including  
84 coronaviruses (CoVs) that have caused disease outbreaks in human and livestock populations  
85 [5-13]: Severe Acute Respiratory Syndrome coronavirus (SARS-CoV), the causative agent of  
86 the SARS outbreak affecting 32 countries in 2002-3, infecting 8,096 people and causing 774  
87 deaths [14]; Middle East Respiratory Syndrome coronavirus (MERS-CoV), which has caused  
88 823 deaths from 2,374 human cases in 27 countries by the end of February 2019, and is  
89 thought to have originally spilled over from bats into camels, in which is it now endemic [15-18];  
90 and Severe acute diarrhea syndrome coronavirus (SADS-CoV) which emerged in the pig  
91 population of Southern China and caused the deaths of more than 20,000 piglets in 2017 and  
92 2018 [5].

93  
94 A large diversity of coronaviruses, including SARS-related Coronaviruses (SARSr-CoVs) have  
95 been discovered in bats, and phylogenetic and pathogenesis studies of these suggest a high  
96 capacity for transmission across species barriers [9, 11, 13, 18-22]. However, few studies have  
97 analyzed bat-to-human spillover events in non-outbreak conditions, likely due to the rarity of

98 these events, and difficulties in identifying at-risk populations or target geographies. Additionally,  
99 the symptoms of novel bat coronavirus infection in the human population may not be clinically  
100 recognized at the time of emergence due to lack of adequate surveillance, or confusion with  
101 other diseases. This represents a significant biosafety risk due to the large and increasing  
102 number of coronaviruses discovered in bats [23, 24] and the wide distribution of bat populations  
103 in rural regions such as Southern China [25].

104

105 In this paper we report on a study designed to characterize the bat coronavirus spillover  
106 potential associated with presumed high-risk human behavior in rural communities of Southern  
107 China [26]. We collected data from community serological and behavioral surveillance to  
108 understand the driving factors of bat coronavirus spillover provide evidence for community-  
109 based strategies to help prevent zoonotic disease emergence.

110

## 111 **2. Materials and Methods**

### 112 **2.1 Study Location and Target Population**

113 We conducted a cross-sectional study in the provinces of Yunnan, Guangxi, and Guangdong,  
114 China which are known for their high levels of animal biodiversity, active animal trade activity,  
115 and historic zoonotic disease emergence events [3, 5, 10, 14, 22, 24, 27]. Eight study sites were  
116 selected in areas where we have previously reported diverse coronaviruses in bat populations  
117 [24] close (within five kilometers) to human dwellings. The study targeted human populations  
118 that are highly exposed to bats and other wildlife, including people who visit or work around bat  
119 caves, work in local live animal markets, raise animals, or are involved in trading wild animals  
120 (e.g., wild animal harvest, trade, transportation, and preparation), as identified by previous  
121 exploratory ethnographic interviews.

122

## 123 **2.2 Recruitment and Sampling**

124 Prior to recruitment and sampling, project staff who received human subject research training  
125 visited each participating site to introduce the project to the community with assistance from  
126 officials from provincial and city-level Centers for Disease Control and Prevention. To generate  
127 interest and develop recruitment strategies, project staff held meetings with village committees  
128 to discuss topics relevant to their daily contact with animals and any health issues in the  
129 community that were particularly concerning for them. With permissions from local authorities,  
130 community leaders conducted house visits and broadcast announcements a week before data  
131 collection took place to inform community residents about the study and its recruitment plan. All  
132 information was communicated in local dialects using simple language to convey the study  
133 purpose, eligibility and inclusion guidelines, potential risks and benefits of participation, and the  
134 time and locations at which the study would take place.

135  
136 We aimed to obtain a minimum sample size of 400 participants from each of the three provinces  
137 (Yunnan, Guangxi, and Guangdong), for a total sample size of over 1,200 participants. A  
138 snowball sampling method was used because the population size at selected sites and the  
139 people who were highly exposed to wild animals were difficult to elucidate [28]. During each  
140 house visit, we requested information about potential eligible participants from the residents'  
141 networks, and we then followed their referrals to recruit from the community. Only one person  
142 per household was recruited to participate in this study, and no participants were recruited from  
143 clinics or healthcare settings. We made every effort to include participants across a range of  
144 demographic indices including gender, age, and socioeconomic status, as well as to ensure that  
145 any contribution was voluntary and involved minimal risk to the participants.

146

## 147 **2.3 Data collection and management**

148 Following the completion of the informed consent process, a standardized Mandarin  
149 questionnaire was administered by study staff in local dialects. The interview was conducted in  
150 a private environment where confidentiality was maintained, and interviewers and participants  
151 were paired by sex. Children aged 10 to 18 years were interviewed with the permission and in  
152 the presence of a parent or guardian.

153  
154 The questionnaire included five sections consisting of demographics, living circumstances and  
155 livelihood, travel, and types of contact with animals, as well as unusual illness symptoms in the  
156 past 12 months. The survey assessed symptoms including fever with cough and shortness of  
157 breath or difficulty breathing (severe acute respiratory infection [SARI] symptoms) and fever with  
158 muscle aches, cough, or sore throat (influenza-like illness [ILI] symptoms) (Appendices). SARI  
159 and ILI symptoms were included in the survey in anticipation of potentially low coronavirus sero-  
160 positivity rates. These symptoms are commonly used as metrics in emerging infectious  
161 respiratory disease surveillance and are known to be associated with coronavirus infections  
162 (e.g., MERS-CoV, SARS-CoV) [29]. Therefore, SARI and ILI symptom histories can be  
163 analyzed in addition to serological testing to maximize our understanding of bat coronavirus  
164 spillover risk.

165  
166 After the questionnaire interview, participants were asked to provide a blood sample (2.5-5 mL  
167 stored in a serum-separating tube) and an oropharyngeal swab (stored in a cryotube with viral  
168 transport medium). Samples were collected by study staff from local clinics. All samples were  
169 stored in liquid nitrogen immediately after collection and transferred to an ultralow (-80°C)  
170 freezer within 48 hours.

171  
172 A unique alphanumeric identification code was assigned to each questionnaire and biological  
173 specimen collected from each participant. No personal identifying information was collected.

174 Only authorized study personnel who received human subject research training were allowed  
175 access to the questionnaire and biological data.

176

#### 177 **2.4 Serological testing**

178 Serum samples collected from study participants were analyzed using newly developed IgG  
179 enzyme-linked immunosorbent assays (ELISA) based on selected nucleocapsid proteins (NP)  
180 expressed and purified in *E. coli* for four specific coronaviruses: SARSr-CoV (DQ071615, Bat  
181 SARS coronavirus Rp3, NP), HKU10-CoV (sample 3740, NP), HKU9-CoV (MG762674,  
182 BatCoV\_HKU9-2202, NP), and MERS-CoV (JX869059, Human betacoronavirus 2c EMC/2012,  
183 NP). Micro-titer plates were coated with recombinant and purified NP (100ng/well); samples  
184 were tested at 1:20 dilution; and an anti-Human IgG-HRP conjugated monoclonal antibody  
185 (Kyab Biotech Co., Ltd, Wuhan, China) was used as the secondary antibody with different  
186 dilution ratios for different coronaviruses. 100 serum samples collected from healthy people in  
187 Wuhan were tested using this ELISA kit to set up the cutoff value, and positive test results were  
188 determined by the cut-off value in each run for each of the four coronaviruses, as the product of  
189 the mean of all serum samples' optical density (OD) values plus three standard deviations, and  
190 confirmed by Western blot test [30].

191

#### 192 **2.5 Questionnaire data analysis**

193 Questionnaire data were entered into an Excel database with quality control for data cleaning  
194 and validation. The glmnet package in R version 3.6.0 was used to fit a least absolute shrinkage  
195 and selection operator (LASSO) regression to characterize associations between animal contact  
196 and SARI and/or ILI symptoms in the preceding 12 months [31, 32]. The bat coronavirus  
197 serology testing outcome was not analyzed in the LASSO due to low rates of sero-positivity.  
198 The LASSO regression is an adaptation of the generalized linear model (GLM) and was  
199 selected because it is effective at minimizing prediction error for datasets with many predictor

200 variables. The model identifies subsets of predictors that are associated with the outcome of  
201 interest by applying a shrinkage operation to regression coefficients and shrinking some  
202 coefficients to exactly 0. The LASSO is often utilized for its variable selection capabilities for  
203 sparse datasets including surveys and questionnaires. Demographic variables (age, gender,  
204 province, and income) were included in the model as independent and interaction terms in order  
205 to account for potential confounding. Because the LASSO does not generate confidence  
206 intervals, we repeated the model using bootstrapping to instead calculate bootstrap support, i.e.,  
207 the proportion of times a predictor variable is selected into the model [33-36]

208

209 Chi-Square and fisher exact tests were also conducted to explore the associations between  
210 potential risk factors in local demographics, behaviors, and attitudes (independent variables)  
211 and bat CoV serological evidence (dependent variables), with effect size examined. However,  
212 due to the low positivity rate (9/1,497), the results were not robust and are not reported in this  
213 paper.

### 214 **3. Results**

215 From October 2015 to July 2017, a total of 1,596 residents from eight sites in Yunnan (n=761),  
216 Guangxi (n=412), and Guangdong (n=423) provinces were enrolled in this study. Of these,  
217 1,585 participants completed the questionnaires and 11 participants withdrew from the  
218 questionnaire interview due to scheduling reasons. After the interviews, 1,497 participants  
219 provided biological samples for lab analysis.

220

#### 221 **3.1 Demographics**

222 More female (62%) than male (38%) community members participated in this study. Most  
223 participants were adults over 45 years old (69%) and had been living in the community for more  
224 than 5 years (97%) with their family members (95%). A majority relied on a comparatively low

225 family annual per capita income less than 10,000 RMB (86%), which is below the national mean  
226 of per capita disposable income of rural households from 2015 to 2017 (11,422 - 13,432 RMB)  
227 [37]. Most participants (98%) had not received a college education and were making a living on  
228 crop production (76%). 9% of participants frequently traveled outside the county as migrant  
229 laborers. Some participants were working in sectors where frequent human-animal contacts  
230 occur, such as the animal production business (1.7%), wild animal trade (0.5%),  
231 slaughterhouses or abattoirs (0.5%), protected nature reserve rangers (0.4%) or in wildlife  
232 restaurants (0.3%). It was common for participants to have multiple part-time jobs as income  
233 sources (Table 1)

234

Variable		Total	
		N	Valid %
<b>Gender</b> (n= 1,574)	Female	968	61.5
	Male	605	38.4
	Other	1	0.1
<b>Age</b> (n=1,582)	Under 18 years	71	4.5
	18 to 44 years	420	26.5
	45 to 64 years	780	49.3
	Age 65 or older	311	19.7
<b>Province</b> (n=1,585)	Guang Dong	420	26.5
	Guang Xi	412	26.0
	Yun Nan	753	47.5
<b>Residence time</b> (n=1,568)	< 1 month	4	0.3
	1 month – 1 year	12	0.8
	1 year – 5 years	26	1.7
	> 5 years	1,526	97.3
<b>Family annual PCI</b> (n=1,565)	<1000 yuan	271	17.3
	1001-10000 yuan	1067	68.2
	>10000 yuan	227	14.5
<b>Livelihood since last year</b>	Extraction of minerals, gas, oil, timber (n=1,566)	5	0.3
	Crop production (n=1,569)	1,196	76.2
	Wildlife restaurant business (n=1,564)	5	0.3
	Wild/exotic animal trade/market business (n=1,566)	8	0.5
	Rancher/farmer animal production business (n=1,566)	27	1.7
	Meat processing, slaughterhouse, abattoir (n=1,567)	8	0.5
	Zoo/sanctuary animal health care (n=1,565)	1	0.1
	Protected area worker (n=1,567)	7	0.4
	Hunter/trapper/fisher (n=1,565)	3	0.2
	Forager/gatherer/non-timber forest product collector (n=1,566)	4	0.3
	Migrant laborer (n=1,567)	144	9.2
	Nurse, doctor, healer, community health worker (n=1567)	7	0.4
	Construction (n=1,564)	41	2.6
	Other (n=1,568)	293	18.7
<b>Education</b> (n=1,570)	None	428	27.3
	Primary School	632	40.3
	Secondary school/Polytechnic school	479	30.5
	College/university/professional	31	2.0
<b>Live with family</b> (n=1,564)	No	73	4.7
	Yes	1491	95.3

235 Table 1 Demographics of study participants. Total counts differ due to missing responses.

### 236 3.2 Animal contact and exposure to bat coronaviruses

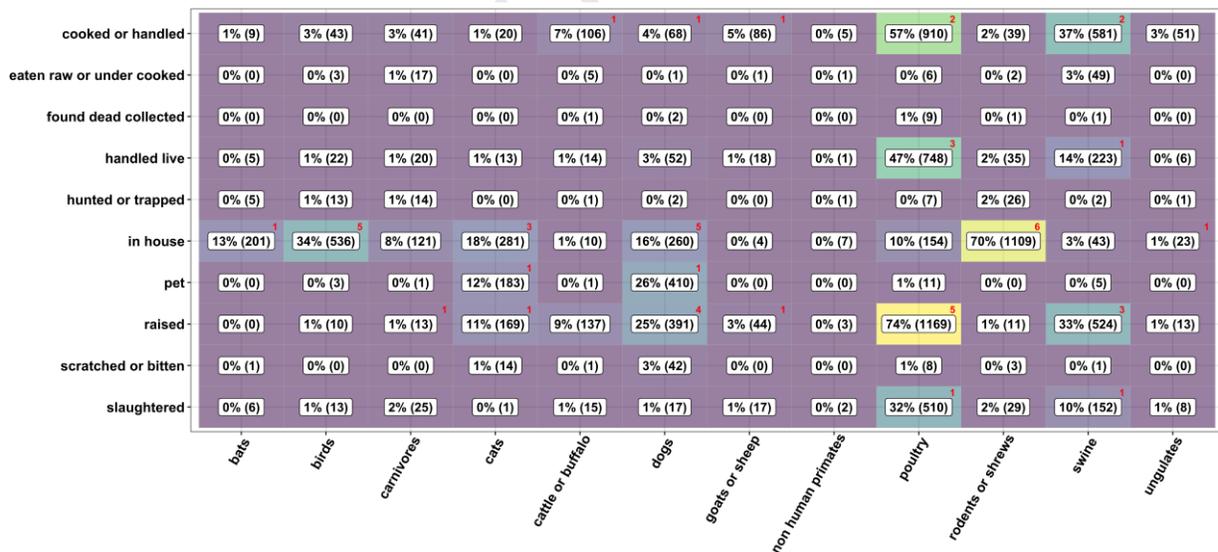
237 Serological testing of serum samples from 1,497 local residents revealed that 9 individuals  
 238 (0.6%) [in four study sites](#) were positive for bat coronaviruses, indicating exposure at some point  
 239 in their life to bat origin SARSr-CoVs (n=7, Yunnan), HKU10 CoV (n=2, Guangxi), or other

240 coronaviruses that are phylogenetically closely related to these. All individuals who tested  
 241 positive (male=6, female=3) were over 45 years old, and most (n=8) were making a living from  
 242 crop production. None of those participants reported any symptoms in the 12 months preceding  
 243 the interview.

244

245 Due to the low rate of sero-positivity, we did not obtain robust results from the statistical  
 246 comparisons of animal-contact behavior by coronavirus outcome. Figure 12 shows animal  
 247 contact rates in the previous 12 months among the survey population (n= 1,585) and among  
 248 seropositive individuals (n=9). Participants reported common contact with poultry and  
 249 rodents/shrews, and most animal contact occurred in domestic settings through animal raising  
 250 or food preparation activities.

251



252

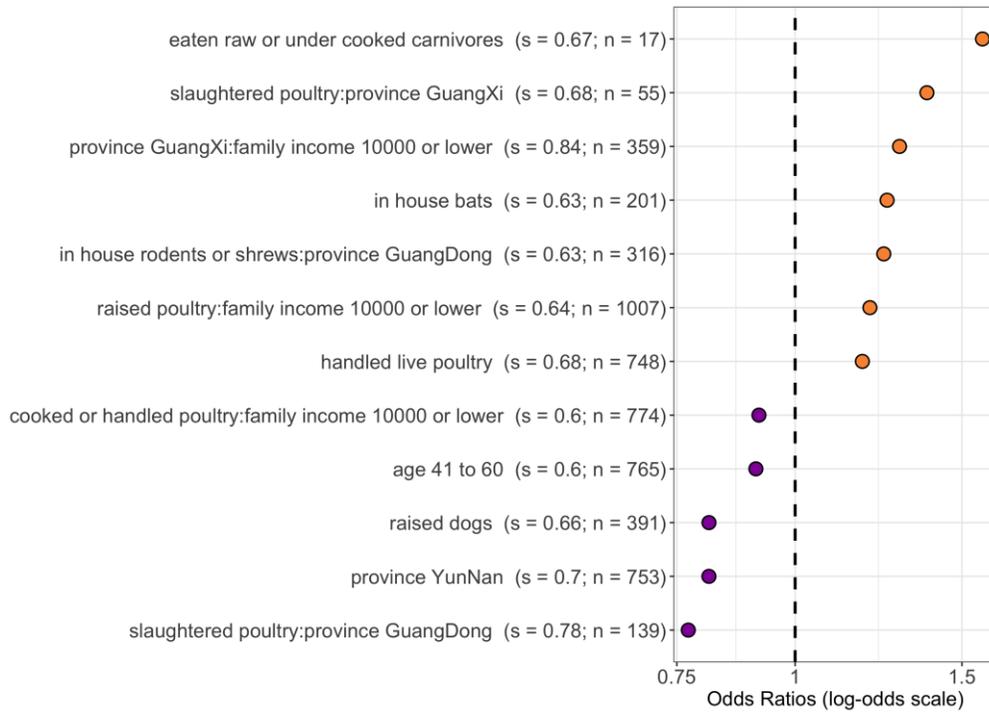
253 Figure 12. Animal contact by taxa and activities. Values and shading represent the survey  
 254 population; red numbers in the upper-right corners of the cells indicate the number of  
 255 seropositive individuals with the given contact.

256

### 257 3.3 Self-report SARI/ILI symptoms and animal contact

258 Among the 1,585 participants who responded, 265 (17%) reported experiencing SARI (n = 73)  
259 and/or ILI (n = 227) symptoms in the last year. The LASSO regression showed that eating raw  
260 or undercooked carnivores in the preceding 12 months was the most salient predictor of self-  
261 reported SARI and/or ILI symptoms over the same time period (odds ratio [OR] = 1.6; bootstrap  
262 support = 0.67). Additional salient predictors were slaughtering poultry as a resident of Guangxi  
263 province (OR = 1.4; support = 0.68), having an income below 10,000 RMB as a resident of  
264 Guangxi province (OR = 1.3; support = 0.84), domestic contact with bats (OR = 1.3; support =  
265 0.63) and domestic contact with rodents or shrews as a resident of Guangdong province (OR =  
266 1.2; support = 0.63) (Figure 23).

267  
268 Some demographic variables were associated with self-reported SARI and/or ILI symptoms as  
269 either independent or interactive terms. For example, respondents aged 41 to 60 and residents  
270 of the YunNan province were less likely to report symptoms. Slaughtering poultry was positively  
271 associated with the outcome only in GuangXi residents, whereas the association was negative  
272 in GuangDong residents. Family income also showed interactions, with family income less than  
273 10,000 RMB being positively associated with the outcome in respondents who raised poultry but  
274 negatively associated in respondents who cooked or handled poultry. Gender was not found to  
275 be salient in either direction.



276

277 | Figure 23. Most salient predictors of self-reported ILI and/or SARI symptoms in the last year (s =  
 278 bootstrap support; n = count positive out of 1,585 respondents). Bootstrap support values  $\geq 0.6$   
 279 are demonstrated here, meaning they were identified as associated with the outcome for 60% or  
 280 more of the bootstrap iterations. Odds ratios  $> 1$  (orange) are positively associated with the  
 281 outcome, and odds ratios  $< 1$  (purple) are negatively associated with the outcome.

282

### 283 3.4 Attitudes towards zoonotic diseases emergence

284 When asked about animals and disease transmission, more than half of the study participants  
 285 believed that animals could spread disease (n=871, 56%) and were worried about disease  
 286 emergence from animals at wet markets (n=810, 52%). Of those worried about disease  
 287 emergence, 46% (n=370) still purchased animals from wet markets in the past 12 months.  
 288 Among all participants who purchased animals from wet markets in the past 12 months (n=502,  
 289 32%), some (n=194, 39%) took protection measures or strategies such as washing hands,  
 290 purchasing live animals less often (n=153, 30%), or purchasing meat at supermarkets instead of

291 live animal markets (n=148, 29%). Very few participants considered wearing a mask (n=7, 1%)  
292 or gloves (n=7, 1%) while visiting the markets.

293

#### 294 **4. Discussion**

295 We used a novel human surveillance approach to integrate serological and behavioral data to  
296 characterize associations between human-animal contact and zoonotic disease spillover risk in  
297 Southern China. This study provides the first serological evidence of bat-origin SARSr-CoVs  
298 and HKU10 CoV transmission to people and highlights potential spillover pathways through  
299 animal contact. Given the high diversity and recombination rate of bat coronaviruses, and close  
300 relationship of SARSr-CoVs to SARS-CoV, it is possible that exposure to these coronaviruses  
301 may lead to disease emergence in human populations. Continuous surveillance of both human  
302 and bat populations, as well as further pathogenesis studies of these viruses, are important to  
303 determine the extent of the disease risk.

304

305 Contact with animals was prevalent among the survey population. Raising poultry and having  
306 rodents/shrews in the house were the most common types of contact. Correspondingly, contact  
307 with poultry and rodents/shrews, as well as with carnivores, was identified in the LASSO  
308 regression as being associated with self-reported ILI and/or SARI symptoms, with results  
309 varying by income and province. It's important to note that the questionnaire used broad  
310 classification of the type of animals for these exposures due to the presumed variability in  
311 respondent's capacity to identify species or genera of wildlife. It is likely that the most significant  
312 exposure we identified (to carnivores) reflects animals as diverse as civets, porcupines, ferret  
313 badgers and animals that respondents identified as non-rodent and non-shrew. This study also  
314 assessed health risks from human interaction activities for each study participant in the survey  
315 based on their travel history and the health history of people who they lived with, to minimize the

316 possibility of human-to-human transmission of other pathogens causing ILI and/or SARI  
317 symptoms. We did not find evidence supporting a direct relationship between bat contact and  
318 bat coronavirus sero-positivity in the human population. However, there is frequent contact with  
319 domestic animals in these communities and it is known that other bat-origin viruses have been  
320 transmitted to humans via livestock (e.g. henipaviruses and filoviruses) [38-41]. It is possible that  
321 these findings reflect indirect exposure to bat CoVs via these pathway, or fomites and future  
322 surveillance may benefit from including a wide range of livestock and peri-domestic animals in  
323 viral and serological studies to identify potential spillover pathways [42-45].

324

325 While it is known that bias can occur in self-reported illness data, this approach has been widely  
326 used in previous disease surveillance and risk factor studies [46-49]. It may be particularly  
327 useful as an early warning system during non-outbreak conditions to assess broad categories of  
328 high-risk within communities for further longitudinal surveillance. This may be particularly  
329 important in rural communities, where people have high levels of contact with domestic and wild  
330 animals but may not seek diagnosis or treatment in a timely fashion, slowing early detection and  
331 response.

332

333 While the majority of survey respondents believed that animals could spread disease and were  
334 worried about disease emergence from animals at wet markets, many did not take measures to  
335 protect themselves from exposure. Further work on what drives these local attitudes to risk may  
336 help in developing risk-mitigation behavior change programs. A number of affordable and  
337 readily adaptable measures could be targeted to these at-risk populations, including use gloves  
338 and masks while killing or butchering animals, and handwashing.

339

340 The low levels of sero-positivity found in the study could reflect a number of factors: 1) the rarity  
341 of spillover and bat-to-human transmission, as has been reported for other virus-host systems

342 [50-54]; 2) the use of a snowball technique for sample selection that could have biased the  
343 population sampled; 3) the limited diversity of CoVs that this study tested for; 4) the possibility  
344 that these infections cause high mortality rates and therefore the number of survivors and  
345 number of seropositive people is low, although this seems unlikely because the mortality rate  
346 from SARS was >10% during an outbreak that included hospital exposure and therefore likely  
347 high infectious doses [55, 56]; and 5) that antibodies to these viruses wane rapidly in humans.  
348 The latter hypothesis is supported by findings that antibodies to SARS decline rapidly (2-3  
349 years) after illness [57]. Expanding this approach to a larger cohort of subjects, using a  
350 longitudinal (repeated sampling) approach, and targeting selection to people who are in the  
351 higher risk categories we have identified may provide a larger number of sero-positives and  
352 more critical information on what drives spillover risk. However, despite the small sample sizes,  
353 this study suggests that there are a substantial number of people in rural Southern China who  
354 are exposed to bat-origin viruses, and that this exposure is likely within normal practices for  
355 rural communities, rather than specific high risk groups (e.g. wet market workers). Considering  
356 the proven potential of some SARSr-CoVs currently circulating in bats in southern China, to  
357 infect human cells, cause clinical signs in humanized mouse models, and lead to infections that  
358 cannot be treated with monoclonal therapies effective against SARS-CoV [58-60], this  
359 represents a clear and present danger to our biosafety and public health. Further studies to  
360 determine the relationship between SARSr-CoV and HKU10-CoV exposure and illness in  
361 people may help elucidate this risk and provide critical mitigation strategies.

362

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364 HY, and AC designed the study, developed the research tools, and obtained ethical approval;  
365 SL, HY, HH, and GZ implemented the field data collection; WZ and NW conducted the  
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368

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373

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375

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## 380 References

- 381 [1] K. E. Jones, N. G. Patel, M. A. Levy, A. Storeygard, D. Balk, J. L. Gittleman and P. Daszak,  
382 Global trends in emerging infectious diseases. *Nature*. 451 (2008) 990-993.  
383 <https://doi.org/10.1038/nature06536>
- 384 [2] R. K. Plowright, S. H. Sokolow, M. E. Gorman, P. Daszak and J. E. Foley, Causal inference  
385 in disease ecology: investigating ecological drivers of disease emergence. *Frontiers in*  
386 *Ecology and the Environment*. 6 (2008) 420-429. <https://doi.org/10.1890/070086>
- 387 [3] US Centers for Disease Control and Prevention, Asian Lineage Avian Influenza A(H7N9)  
388 Virus. 2018 <https://www.cdc.gov/flu/avianflu/h7n9-virus.htm> (2019 February 1)
- 389 [4] National Bureau of Statistics of China, China Census Data. 2018  
390 <http://data.stats.gov.cn/easyquery.htm?cn=C01&zb=A0A01&sj=2018> (2019 March 01)
- 391 [5] P. Zhou, H. Fan, T. Lan, X.-L. Yang, W.-F. Shi, W. Zhang, Y. Zhu, Y.-W. Zhang, Q.-M. Xie,  
392 S. Mani, X.-S. Zheng, B. Li, J.-M. Li, H. Guo, G.-Q. Pei, X.-P. An, J.-W. Chen, L. Zhou, K.-J.  
393 Mai, Z.-X. Wu, D. Li, D. E. Anderson, L.-B. Zhang, S.-Y. Li, Z.-Q. Mi, T.-T. He, F. Cong, P.-J.  
394 Guo, R. Huang, Y. Luo, X.-L. Liu, J. Chen, Y. Huang, Q. Sun, X.-L.-L. Zhang, Y.-Y. Wang,  
395 S.-Z. Xing, Y.-S. Chen, Y. Sun, J. Li, P. Daszak, L.-F. Wang, Z.-L. Shi, Y.-G. Tong and J.-Y.  
396 Ma, Fatal swine acute diarrhoea syndrome caused by an HKU2-related coronavirus of bat  
397 origin. *Nature*. 556 (2018) 255. <https://doi.org/10.1038/s41586-018-0010-9>
- 398 [6] X. Y. Ge, J. L. Li, X. L. Yang, A. A. Chmura, G. Zhu, J. H. Epstein, J. K. Mazet, B. Hu, W.  
399 Zhang, C. Peng, Y. J. Zhang, C. M. Luo, B. Tan, N. Wang, Y. Zhu, G. Cramer, S. Y. Zhang,  
400 L. F. Wang, P. Daszak and Z. L. Shi, Isolation and characterization of a bat SARS-like  
401 coronavirus that uses the ACE2 receptor. *Nature*. 503 (2013) 535-538.  
402 [10.1038/nature12711](https://doi.org/10.1038/nature12711)
- 403 [7] J. Huynh, S. Li, B. Yount, A. Smith, L. Sturges, J. C. Olsen, J. Nagel, J. B. Johnson, S.  
404 Agnihothram, J. E. Gates, M. B. Frieman, R. S. Baric and E. F. Donaldson, Evidence

- 405 Supporting a Zoonotic Origin of Human Coronavirus Strain NL63. *Journal of virology*. 86  
406 (2012) 12816-12825. <https://doi.org/10.1128/JVI.00906-12>
- 407 [8] V. M. Corman, H. J. Baldwin, A. F. Tateno, R. M. Zerbinati, A. Annan, M. Owusu, E. E.  
408 Nkrumah, G. D. Maganga, S. Oppong, Y. Adu-Sarkodie, P. Vallo, L. V. da Silva Filho, E. M.  
409 Leroy, V. Thiel, L. van der Hoek, L. L. Poon, M. Tschapka, C. Drosten and J. F. Drexler,  
410 Evidence for an Ancestral Association of Human Coronavirus 229E with Bats. *J Virol*. 89  
411 (2015) 11858-11870. 10.1128/JVI.01755-15
- 412 [9] W. Li, Z. Shi, M. Yu, W. Ren, C. Smith, J. H. Epstein, H. Wang, G. Cramer, Z. Hu, H.  
413 Zhang, J. Zhang, J. McEachern, H. Field, P. Daszak, B. T. Eaton, S. Zhang and L.-F. Wang,  
414 Bats Are Natural Reservoirs of SARS-Like Coronaviruses. *Science*. 310 (2005) 676-679.  
415 <https://doi.org/10.1126/science.1118391>
- 416 [10] L. F. Wang, Z. Shi, S. Zhang, H. Field, P. Daszak and B. T. Eaton, Review of bats and  
417 SARS. *Emerg Infect Dis*. 12 (2006) 1834-1840. <https://doi.org/10.3201/eid1212.060401>
- 418 [11] L. F. Wang and D. E. Anderson, Viruses in bats and potential spillover to animals and  
419 humans. *Curr Opin Virol*. 34 (2019) 79-89. 10.1016/j.coviro.2018.12.007
- 420 [12] K. J. Olival, P. R. Hosseini, C. Zambrana-Torrel, N. Ross, T. L. Bogich and P. Daszak,  
421 Host and viral traits predict zoonotic spillover from mammals. *Nature*. 546 (2017) 646-650.
- 422 [13] B. Hu, L. P. Zeng, X. L. Yang, X. Y. Ge, W. Zhang, B. Li, J. Z. Xie, X. R. Shen, Y. Z.  
423 Zhang, N. Wang, D. S. Luo, X. S. Zheng, M. N. Wang, P. Daszak, L. F. Wang, J. Cui and Z.  
424 L. Shi, Discovery of a rich gene pool of bat SARS-related coronaviruses provides new  
425 insights into the origin of SARS coronavirus. *PLoS pathogens*. 13 (2017)  
426 10.1371/journal.ppat.1006698
- 427 [14] World Health Organization, Summary of probable SARS cases with onset of illness from  
428 1 November 2002 to 31 July 2003. 2004  
429 [https://www.who.int/csr/sars/country/table2004\\_04\\_21/en/](https://www.who.int/csr/sars/country/table2004_04_21/en/) (2019 February 1)
- 430 [15] World Health Organization, Middle East respiratory syndrome coronavirus (MERS-CoV).  
431 2019 <https://www.who.int/emergencies/mers-cov/en/> (2019 March 15)
- 432 [16] Z. A. Memish, N. Mishra, K. J. Olival, S. F. Fagbo, V. Kapoor, J. H. Epstein, R.  
433 AlHakeem, A. Durosinioun, M. A. Asmari, A. Islam, A. Kapoor, T. Briese, P. Daszak, A. A. A.  
434 Rabeeah and W. I. Lipkin, Middle East Respiratory Syndrome Coronavirus in Bats, Saudi  
435 Arabia. *Emerging Infectious Diseases*. 19 (2013) 10.3201/eid1911.131172
- 436 [17] Q. Wang, J. Qi, Y. Yuan, Y. Xuan, P. Han, Y. Wan, W. Ji, Y. Li, Y. Wu, J. Wang, A.  
437 Iwamoto, P. C. Woo, K. Y. Yuen, J. Yan, G. Lu and G. F. Gao, Bat origins of MERS-CoV  
438 supported by bat coronavirus HKU4 usage of human receptor CD26. *Cell Host Microbe*. 16  
439 (2014) 328-337. 10.1016/j.chom.2014.08.009
- 440 [18] S. J. Anthony, K. Gilardi, V. D. Menachery, T. Goldstein, B. Ssebide, R. Mbabazi, I.  
441 Navarrete-Macias, E. Liang, H. Wells, A. Hicks, A. Petrosov, D. K. Byarugaba, K. Debbink,  
442 K. H. Dinnon, T. Scobey, S. H. Randell, B. L. Yount, M. Cranfield, C. K. Johnson, R. S.  
443 Baric, W. I. Lipkin and J. A. Mazet, Further Evidence for Bats as the Evolutionary Source of  
444 Middle East Respiratory Syndrome Coronavirus. *MBio*. 8 (2017) 10.1128/mBio.00373-17
- 445 [19] A. Roberts, D. Deming, C. D. Paddock, A. Cheng, B. Yount, L. Vogel, B. D. Herman, T.  
446 Sheahan, M. Heise, G. L. Genrich, S. R. Zaki, R. Baric and K. Subbarao, A Mouse-Adapted  
447 SARS-Coronavirus Causes Disease and Mortality in BALB/c Mice. *PLoS pathogens*. 3  
448 (2007) e5. <https://doi.org/10.1371/journal.ppat.0030005>
- 449 [20] R. L. Graham and R. S. Baric, Recombination, Reservoirs, and the Modular Spike:  
450 Mechanisms of Coronavirus Cross-Species Transmission. *Journal of virology*. 84 (2010)  
451 3134-3146. <https://doi.org/10.1128/JVI.01394-09>
- 452 [21] J. Cui, F. Li and Z. L. Shi, Origin and evolution of pathogenic coronaviruses. *Nat Rev*  
453 *Microbiol*. 17 (2019) 181-192. 10.1038/s41579-018-0118-9
- 454 [22] Y. Fan, K. Zhao, Z. L. Shi and P. Zhou, Bat Coronaviruses in China. *Viruses*. 11 (2019)  
455 10.3390/v11030210

- 456 [23] H.-J. Han, H.-I. Wen, C.-M. Zhou, F.-F. Chen, L.-M. Luo, J.-w. Liu and X.-J. Yu, Bats as  
457 reservoirs of severe emerging infectious diseases. *Virus research*. 205 (2015) 1-6.  
458 <https://doi.org/10.1016/j.virusres.2015.05.006>
- 459 [24] B. Hu, L.-P. Zeng, X.-L. Yang, X.-Y. Ge, W. Zhang, B. Li, J.-Z. Xie, X.-R. Shen, Y.-Z.  
460 Zhang, N. Wang, D.-S. Luo, X.-S. Zheng, M.-N. Wang, P. Daszak, L.-F. Wang, J. Cui and  
461 Z.-L. Shi, Discovery of a rich gene pool of bat SARS-related coronaviruses provides new  
462 insights into the origin of SARS coronavirus. *PLoS pathogens*. 13 (2017) e1006698.  
463 <https://doi.org/10.1371/journal.ppat.1006698>
- 464 [25] J. Luo, T. Jiang, G. Lu, L. Wang, J. Wang and J. Feng, Bat conservation in China:  
465 should protection of subterranean habitats be a priority? *Oryx*. 47 (2013) 526-531.  
466 <https://doi.org/10.1017/s0030605311001505>
- 467 [26] M. Miller and E. Hagan, Integrated biological-behavioural surveillance in pandemic-  
468 threat warning systems. *Bulletin of the World Health Organization*. 95 (2017) 62.  
469 <https://doi.org/10.2471/blt.16.175984>
- 470 [27] P. Yu, B. Hu, Z. L. Shi and J. Cui, Geographical structure of bat SARS-related  
471 coronaviruses. *Infection, genetics and evolution : journal of molecular epidemiology and*  
472 *evolutionary genetics in infectious diseases*. 69 (2019) 224-229.  
473 [10.1016/j.meegid.2019.02.001](https://doi.org/10.1016/j.meegid.2019.02.001)
- 474 [28] R. Atkinson and J. Flint, Accessing hidden and hard-to-reach populations: Snowball  
475 research strategies. *Social research update*. 33 (2001) 1-4.
- 476 [29] J. A. Al-Tawfiq, A. Zumla, P. Gautret, G. C. Gray, D. S. Hui, A. A. Al-Rabeeah and Z. A.  
477 Memish, Surveillance for emerging respiratory viruses. *Lancet Infect Dis*. 14 (2014) 992-  
478 1000. [https://doi.org/10.1016/S1473-3099\(14\)70840-0](https://doi.org/10.1016/S1473-3099(14)70840-0)
- 479 [30] N. Wang, S. Y. Li, X. L. Yang, H. M. Huang, Y. J. Zhang, H. Guo, C. M. Luo, M. Miller,  
480 G. Zhu, A. A. Chmura, E. Hagan, J. H. Zhou, Y. Z. Zhang, L. F. Wang, P. Daszak and Z. L.  
481 Shi, Serological Evidence of Bat SARS-Related Coronavirus Infection in Humans, China.  
482 *Virology*. 33 (2018) 104-107. [10.1007/s12250-018-0012-7](https://doi.org/10.1007/s12250-018-0012-7)
- 483 [31] R. C. Team, A language and environment for statistical computing. Vienna, Austria: R  
484 Foundation for Statistical Computing; 2012. <https://www.R-project.org>. (2019)
- 485 [32] J. Friedman, T. Hastie and R. Tibshirani, Regularization paths for generalized linear  
486 models via coordinate descent. *Journal of statistical software*. 33 (2010) 1.  
487 <https://doi.org/10.18637/jss.v033.i01>
- 488 [33] R. Tibshirani, Regression shrinkage and selection via the lasso. *Journal of the Royal*  
489 *Statistical Society: Series B (Methodological)*. 58 (1996) 267-288.
- 490 [34] C. S. Signorino and A. Kirchner, Using LASSO to Model Interactions and Nonlinearities  
491 in Survey Data. *Survey Practice*. 11 (2018) 2716. <https://doi.org/10.29115/SP-2018-0005>
- 492 [35] H. Lin, C. Wang, P. Liu and D. J. Holtkamp, Construction of disease risk scoring systems  
493 using logistic group lasso: application to porcine reproductive and respiratory syndrome  
494 survey data. *Journal of Applied Statistics*. 40 (2013) 736-746.  
495 <https://doi.org/10.1080/02664763.2012.752449>
- 496 [36] T.-F. Lee, P.-J. Chao, H.-M. Ting, L. Chang, Y.-J. Huang, J.-M. Wu, H.-Y. Wang, M.-F.  
497 Horng, C.-M. Chang and J.-H. Lan, Using multivariate regression model with least absolute  
498 shrinkage and selection operator (LASSO) to predict the incidence of xerostomia after  
499 intensity-modulated radiotherapy for head and neck cancer. *PloS one*. 9 (2014) e89700.  
500 <https://doi.org/10.1371/journal.pone.0089700>
- 501 [37] National Bureau of Statistics of China, 农村居民人均可支配收入. 2014-2018  
502 <http://data.stats.gov.cn/easyquery.htm?cn=C01&zb=A0A0C&sj=2018> (2019 February 1)
- 503 [38] G. C. Paola Katrina, R. Vikki Carr de los, S. Maria Nemia, T. Enrique, C.-V. Alah Baby,  
504 F. M. Fedelino, C. B. Gilbert, S. James, E. Debbie, P. Geoffrey, D. Erica, K. Yoshihiro, M.  
505 Shigeru, K. Makoto, A. M. Glenn, M. Sam and A. R. Foxwell, Outbreak of Henipavirus

- 506 Infection, Philippines, 2014. *Emerging Infectious Disease journal*. 21 (2015) 328.  
507 10.3201/eid2102.141433
- 508 [39] Z. Wu, L. Yang, F. Yang, X. Ren, J. Jiang, J. Dong, L. Sun, Y. Zhu, H. Zhou and Q. Jin,  
509 Novel Henipa-like virus, Mojiang Paramyxovirus, in rats, China, 2012. *Emerg Infect Dis*. 20  
510 (2014) 1064-1066. 10.3201/eid2006.131022
- 511 [40] Sukanta Chowdhury, Salah Uddin Khan, Gary Crameri, Jonathan H. Epstein,  
512 Christopher C. Broder, Ausraful Islam, Alison J. Peel, Jennifer Barr, Peter Daszak, Lin-Fa  
513 Wang and S. P. Luby, Serological Evidence of Henipavirus Exposure in Cattle, Goats and  
514 Pigs in Bangladesh. *Plos Neglect. Trop. Dis.* (2014)
- 515 [41] R. W. Barrette, S. A. Metwally, J. M. Rowland, L. Z. Xu, S. R. Zaki, S. T. Nichol, P. E.  
516 Rollin, J. S. Towner, W. J. Shieh, B. Batten, T. K. Sealy, C. Carrillo, K. E. Moran, A. J.  
517 Bracht, G. A. Mayr, M. Sirios-Cruz, D. P. Catbagan, E. A. Lautner, T. G. Ksiazek, W. R.  
518 White and M. T. McIntosh, Discovery of Swine as a Host for the Reston ebolavirus. *Science*.  
519 325 (2009) 204-206. 10.1126/science.1172705
- 520 [42] D. Middleton, J. Pallister, R. Klein, Y.-R. Feng, J. Haining, R. Arkininstall, L. Frazer, J.-A.  
521 Huang, N. Edwards and M. Wareing, Hendra virus vaccine, a one health approach to  
522 protecting horse, human, and environmental health. *Emerging infectious diseases*. 20  
523 (2014) 372. <https://doi.org/10.3201/eid2003.131159>
- 524 [43] E. E. Glennon, O. Restif, S. R. Sbarbaro, R. Garnier, A. A. Cunningham, R. D. Suu-Ire,  
525 R. Osei-Amponsah, J. L. Wood and A. J. Peel, Domesticated animals as hosts of  
526 henipaviruses and filoviruses: A systematic review. *The Veterinary Journal*. 233 (2018) 25-  
527 34. <https://doi.org/10.1016/j.tvjl.2017.12.024Get>
- 528 [44] L. Joffrin, M. Dietrich, P. Mavingui and C. Lebarbenchon, Bat pathogens hit the road: But  
529 which one? *PLoS pathogens*. 14 (2018) e1007134.  
530 <https://doi.org/10.1371/journal.ppat.1007134>
- 531 [45] A. P. Dobson, What links bats to emerging infectious diseases? *Science*. 310 (2005)  
532 628-629. <https://doi.org/10.1126/science.1120872>
- 533 [46] G. C. Gray, T. McCarthy, A. W. Capuano, S. F. Setterquist, C. W. Olsen, M. C. Alavanja  
534 and C. F. Lynch, Swine workers and swine influenza virus infections. *Emerging infectious  
535 diseases*. 13 (2007) 1871. <https://doi.org/10.3201/eid1312.061323>
- 536 [47] B. W. Soltis, J. W. Sanders, S. D. Putnam, D. R. Tribble and M. S. Riddle, Self reported  
537 incidence and morbidity of acute respiratory illness among deployed US military in Iraq and  
538 Afghanistan. *PLoS One*. 4 (2009) e6177. <https://doi.org/10.1371/journal.pone.0006177>
- 539 [48] A. M. Barbara, M. Loeb, L. Dolovich, K. Brazil and M. Russell, Agreement between self-  
540 report and medical records on signs and symptoms of respiratory illness. *Primary Care  
541 Respiratory Journal*. 21 (2012) 145. <https://doi.org/10.4104/pcrj.2011.00098>
- 542 [49] Centers for Disease Control and Prevention, Self-reported influenza-like illness during  
543 the 2009 H1N1 influenza pandemic--United States, September 2009-March 2010. *MMWR.  
544 Morbidity and mortality weekly report*. 60 (2011) 37.
- 545 [50] T. L. Goldberg, C. A. Chapman, K. Cameron, T. Saj, W. B. Karesh, N. Wolfe, S. W.  
546 Wong, M. E. Dubois and M. K. Slifka, Serologic evidence for novel poxvirus in endangered  
547 red colobus monkeys, western Uganda. *Emerging Infectious Disease*. 14 (2008) 801-803.
- 548 [51] G. Grard, J. N. Fair, D. Lee, E. Slikas, I. Steffen, J. J. Muyembe, T. Sittler, N.  
549 Veeraraghavan, J. G. Ruby, C. Wang, M. Makuwa, P. Mulembakani, R. B. Tesh, J. Mazet,  
550 A. W. Rimoin, T. Taylor, B. S. Schneider, G. Simmons, E. Delwart, N. D. Wolfe, C. Y. Chiu  
551 and E. M. Leroy, A novel rhabdovirus associated with acute hemorrhagic Fever in central  
552 Africa. *PLoS pathogens*. 8 (2012) e1002924. [10.1371/journal.ppat.1002924](https://doi.org/10.1371/journal.ppat.1002924)
- 553 [52] [PPATHOGENS-D-12-01590 \[pii\]](https://doi.org/10.1371/journal.ppat.1002924)
- 554 [53] O. Pernet, B. S. Schneider, S. M. Beaty, M. LeBreton, T. E. Yun, A. Park, T. T.  
555 Zachariah, T. A. Bowden, P. Hitchens, C. M. Ramirez, P. Daszak, J. Mazet, A. N. Freiberg,

- 556 N. D. Wolfe and B. Lee, Evidence for henipavirus spillover into human populations in Africa.  
557 Nature Communications. 5 (2014) 10.1038/ncomms6342
- 558 [54] N. D. Wolfe, P. Daszak and A. M. Kilpatrick, Bushmeat Hunting, Deforestation, and  
559 Prediction of Zoonotic Disease Emergence. Emerging infectious diseases. 11 (2005) 1822-  
560 1827.
- 561 [55] N. D. Wolfe, W. M. Switzer, J. K. Carr, V. B. Bhullar, V. Shanmugam, U. Tamoufe, A. T.  
562 Prosser, J. N. Torimiro, A. Wright, E. Mpoudi-Ngole, F. E. McCutchan, D. L. Birx, T. M.  
563 Folks, D. S. Burke and W. Heneine, Naturally acquired simian retrovirus infections in central  
564 African hunters. Lancet. 363 (2004) 932-937.
- 565 [56] R. M. Anderson, C. Fraser, A. C. Ghani, C. A. Donnelly, S. Riley, N. M. Ferguson, G. M.  
566 Leung, T. H. Lam and A. J. Hedley, Epidemiology, transmission dynamics and control of  
567 SARS: the 2002-2003 epidemic. Philosophical Transactions of the Royal Society of London  
568 Series B-Biological Sciences. 359 (2004) 1091-1105.
- 569 [57] L. P. Wu, N. C. Wang, Y. H. Chang, X. Y. Tian, D. Y. Na, L. Y. Zhang, L. Zheng, T. Lan,  
570 L. F. Wang and G. D. Liang, Duration of antibody responses after severe acute respiratory  
571 syndrome. Emerging infectious Diseases. 13 (2007) 1562-1564.
- 572 [58] M. M. Becker, R. L. Graham, E. F. Donaldson, B. Rockx, A. C. Sims, T. Sheahan, R. J.  
573 Pickles, D. Corti, R. E. Johnston, R. S. Baric and M. R. Denison, Synthetic recombinant bat  
574 SARS-like coronavirus is infectious in cultured cells and in mice. Proceedings of the National  
575 Academy of Sciences of the United States of America. 105 (2008) 19944-19949.  
576 10.1073/pnas.0808116105
- 577 [59] V. D. Menachery, B. L. Yount, Jr., K. Debbink, S. Agnihothram, L. E. Gralinski, J. A.  
578 Plante, R. L. Graham, T. Scobey, X. Y. Ge, E. F. Donaldson, S. H. Randell, A.  
579 Lanzavecchia, W. A. Marasco, Z. L. Shi and R. S. Baric, A SARS-like cluster of circulating  
580 bat coronaviruses shows potential for human emergence. Nature Medicine. 21 (2015) 1508-  
581 1513. 10.1038/nm.3985
- 582 [60] V. D. Menachery, B. L. Yount, A. C. Sims, K. Debbink, S. S. Agnihothram, L. E.  
583 Gralinski, R. L. Graham, T. Scobey, J. A. Plante, S. R. Royal, J. Swanstrom, T. P. Sheahan,  
584 R. J. Pickles, D. Corti, S. H. Randell, A. Lanzavecchia, W. A. Marasco and R. S. Baric,  
585 SARS-like WIV1-CoV poised for human emergence. Proceedings of the National Academy  
586 of Sciences of the United States of America. 113 (2016) 3048-3053.  
587 10.1073/pnas.1517719113  
588

## Highlights

### Scientific question

What are the behavioral risks in human-animal interactions that could lead to the emergence of bat coronaviruses in human population.

### Evidence before this study

Bat borne coronaviruses have caused several emerging infectious disease outbreaks of global significance, including SARS. Novel SARS-related coronaviruses have been discovered in bat populations in South China, some of which have the capacity to infect human cells. Human-animal interactions are thought to be critical for the emergence of bat coronaviruses, however the specific interactions linked to animal-to-human spillover remain unknown.

### New Findings

This study found serological evidence for bat-borne coronavirus transmission to people. Direct contact with bats was not identified as a risk factor. However, self-reported severe acute respiratory infection (SARI) and/or influenza-like illness (ILI) was linked to human interaction with other wildlife and livestock, suggesting that there may be other zoonotic exposures leading to clinical illness in these populations.

### Significance of the study

Findings from this study suggested that an integrated biological and behavioral surveillance in healthy community settings can help identify potential zoonotic disease spillover events or target surveillance to at-risk populations. This approach represents a potential early-warning system that could be used under non-outbreak conditions to identify potential zoonotic emerging diseases prior to largescale outbreaks.