

## Mapping routine malaria incidence at village level for targeted control in Papua New Guinea

Daniela Rodriguez-Rodriguez <sup>1,2,3</sup>, Seri Maraga <sup>3</sup>, Sharon Jamea-Maiasa <sup>3</sup>, Anthony Tandrapah <sup>3</sup>, Leo Makita <sup>4</sup>, Peter M. Siba <sup>3</sup>, Ivo Mueller <sup>5,6</sup>, Justin Pulford <sup>7</sup>, Manuel W. Hetzel <sup>1,2</sup>

<sup>1</sup> Swiss Tropical and Public Health Institute, Basel, Switzerland; <sup>2</sup> University of Basel, Basel, Switzerland; <sup>3</sup> Papua New Guinea Institute of Medical Research, Goroka and Madang, Papua New Guinea; <sup>4</sup> National Department of Health, Waigani, Papua New Guinea; <sup>5</sup> Walter and Eliza Hall Institute of Medical Research, Melbourne, Australia; <sup>6</sup> Institut Pasteur, Paris, France; <sup>7</sup> Liverpool School of Tropical Medicine and Hygiene, Liverpool, United Kingdom

**Acknowledgements:** The authors would like to thank all patients and communities who agreed to participate in this study. We are grateful to the study nurses who collected data and samples at the sentinel health facilities (Mrs. Naeobo Oren in East Cape, Mr. Paul Basanu in Karimui, Ms. Rhona Sao in Sausi, and Mrs. Lia Pitan in Lemakot), and to the authorities and organizations hosting them. We thank the PNGIMR census field teams, the data managers and data entry staff and acknowledge the support of the National Department of Health through Mr. Leo Makita and his team. We thank Mr. Andreas Weber and Point de Vue for their support and enthusiasm. Their collaboration made the audio-visualization of this paper possible.

**Correspondence:** Daniela Rodriguez-Rodriguez, Department of Epidemiology and Public Health, Swiss Tropical and Public Health Institute, Socinstrasse 57, 4002 Basel, Switzerland.

Tel: +41 61 284 86 00

Email: [daniela.rodriguez@swisstph.ch](mailto:daniela.rodriguez@swisstph.ch)

**Key words:** Surveillance, malaria, Papua New Guinea, malaria control, surveillance-response

Commented [A1]: Spell out in the acknowledgments

Field Code Changed

Formatted: French (Switzerland)

**Contributions:** MWH, JP, IM, LM and PMS conceived of the study. SM, SJM, AT, MWH and JP established and supervised data collection systems and LM facilitated access to the study sites. DRR analysed the data and drafted the manuscript and video concept with inputs from MWH. All authors reviewed and approved the final manuscript and video.

**Conflict of interests:** The authors declare no conflict of interest.

**Funding:** Funding for data collection was provided through a Global Fund to Fight AIDS, Tuberculosis and Malaria grant to PNG. DRR was supported by the Forlen Stiftung, Basel, and the R. Geigy Foundation, Basel, Switzerland.

**Video link:** <https://vimeo.com/32777871>; password: 12345Qwert

Commented [A2]: Spell out

1 **Abstract**

2 Malaria surveillance and response-systems are essential for identifying the areas most  
3 affected by malaria and for targeting resources. This study aimed to assess whether [the](#)  
4 [visualization of](#) routinely collected health facility data linked to village of residence  
5 provides [useful](#) evidence for targeting control interventions in four sentinel health-  
6 facilities (SHF) in Papua New Guinea. During the surveillance period a total of 8,173  
7 fever cases [within from the SHFs catchment areas](#) tested positive for malaria and were  
8 mapped [by village of residence within the SHFs catchment area](#). Despite limitations, this  
9 approach appeared useful in sites with very few remaining cases or with increasingly  
10 marked heterogeneity. Villages that could benefit from targeted interventions or  
11 investigations were identified.

12 **Background Section**

13 Variation in the risk of malaria prevalence and incidence between villages in regions  
14 with on-going transmission has long been recognized (Bousema et al., 2012;  
15 Greenwood, 1989). Such variations become more evident in regions with moderate and  
16 low transmission, e.g. after scale-up of malaria control (Bousema et al., 2012). Malaria  
17 control efforts in Papua New Guinea (PNG) were re-intensified in 2004. Differences in  
18 malaria burden, transmission, and impact of interventions have subsequently been  
19 identified even between neighbouring villages (Hetzl et al., 2016, 2014), confirming  
20 earlier findings of small-area heterogeneity (Cattani et al., 1986). An analysis of routine  
21 health-facility data from seven sentinel sites found that reductions in malaria incidence  
22 were associated with [LLIN](#) distributions but the effect varied between sites (Rodríguez-  
23 Rodríguez et al., 2019). As malaria transmission decreases, and resources remain

**Commented [A3]:** Currently, the abstract reads too much like an abstract of a full research article rather than a vHealth publication. Suggesting that you slightly reframe your abstract, placing more emphasis on the visualisation/video (motivation, benefit) that comes with the paper.

**Commented [A4]:** Consider making this adjustment too in the video (introduction and conclusion)

**Commented [A5]:** Introduce abbreviation

24 limited, targeting of interventions becomes increasingly important (Bousema et al.,  
25 2012; Mendis et al., 2009).

26 Malaria surveillance and response-systems are essential for identifying the areas or  
27 population groups that are most affected by malaria and for targeting resources for  
28 maximum impact (World Health Organization, 2018). The Global Technical Strategy for  
29 Malaria 2016-2030 proposes the use of a comprehensive approach that includes vector  
30 control measures and early diagnosis and treatment, especially at the village level  
31 (World Health Organization, 2018, 2015). Identifying villages with on-going  
32 transmission and monitoring changes over time is therefore of utmost importance to  
33 effectively target interventions. Particularly in settings with weak health systems this  
34 information should be generated by simple approaches. Routine data collected at health  
35 facilities may provide a viable option if the relevant information is captured and readily  
36 analysed. However, research on how to best apply geospatial analyses of simple routine  
37 data to identify heterogeneity at village level and support targeting malaria  
38 interventions is scarce (Kelly et al., 2012).

39 This study aimed to assess whether [the visualizations of](#) health facility data linked to  
40 village of residence of patients provides [useful](#) evidence for targeting malaria control  
41 interventions. It used malaria incidence data linked to the self-reported village of  
42 residence of the patients, collected routinely in four sentinel health facilities (SHF) in  
43 PNG. If found to be a valid approach, the operational feasibility of targeting malaria  
44 control at district or sub-district level would have to be investigated within the frame of  
45 existing capacities and resources of the local health system.

46 Sentinel surveillance was established in the health centres of East Cape (Southern  
47 Region), Sausi (Momase), Karimui (Highland) and Lemakot (Islands) – one SHF per  
48 geographical region of PNG. Surveillance was established as part of the continuous  
49 independent evaluation of the National Malaria Control Program (NMCP) (Hetzl et al.,  
50 2015, 2014, 2012). Details of the sites are provided elsewhere (Hetzl et al., 2014;  
51 Rodríguez-Rodríguez et al., 2019).

52 All outpatient cases attending the SHF were routinely screened for a history of fever  
53 during the previous three days. A study nurse at the facility collected a capillary blood  
54 sample from all consenting fever patients for ~~1~~ diagnosis of malaria by Rapid  
55 Diagnostic Test (RDT). Demographic details including village of residence and self-  
56 reported mosquito-net use the previous night were recorded on paper case report  
57 forms alongside RDT results. Paper forms were then double entered at the PNGIMR. The  
58 study team ensured availability of RDTs throughout the surveillance period.

59 Data was collected from 2010 to 2014 to characterize annual incidence variations  
60 between villages in the catchment areas of the SHFs and identify patterns that could  
61 guide malaria control efforts. Recently, a paper-based “Malaria Register” has been  
62 implemented by the PNG National Department of Health. ~~In addition, an electronic~~  
63 ~~National Health Information System (eNHIS) is being piloted~~ (Rosewell et al., 2017). The  
64 register ~~and eNHIS~~ routinely collects ~~the same malaria indicators variables linked to the~~  
65 ~~village of residence of the patient, which is comparable to~~ ~~used for~~ this analysis,  
66 allowing the scale-up of ~~this the method approach~~ if it proves useful.

67 The ~~size and population delineation~~ of the SHF catchment areas ~~were as was~~ defined by  
68 local health authorities ~~and differed between sites~~. ~~A population census conducted by~~

Commented [A6]: Specify age range

Commented [A7]: Brand name?

Commented [A8]: Introduce abbreviation

69 [the PNGIMR in the SHF catchment areas at baseline, during which all houses were](#)  
70 [identified and each household member was listed, was used as source of village](#)  
71 [coordinates and population denominator data. The geo-referenced year 2000 National](#)  
72 [Census database was used to complement the PNGIMR census, particularly to identify](#)  
73 [villages outside the catchment area.](#) During the surveillance period 25,097 fever cases  
74 were tested for malaria across all SHF, 38% (95% CI: 37.6-38.8) were RDT-positive.  
75 Table 1 details the number of malaria cases diagnosed at the SHF residing within and  
76 outside the SHF catchment area. The analysis includes only the cases within the  
77 catchment area. It is important to note that a large number of patients from outside the  
78 catchment area were diagnosed and treated in East Cape and Lemakot SHFs. Both HCs  
79 are located in areas of constant transit of people.

80 [Table 1 here]

81 [The maps in this visualization illustrate variations in incidence in space and time, using](#)  
82 [open source GIS software.](#) The video format ~~of the v-communication enabled~~ allowed us  
83 to [visualize the dynamics display over 40 maps in a convenient and in case incidence](#)  
84 [over time engaging way alongside together with images photographs- illustrating the](#)  
85 [context of the data collection that provide insights to of in](#) the study sites, ~~and the, often~~  
86 ~~unknown, geographical and cultural diversity of PNG. In addition, †~~The audio-visual  
87 [format is easily accessible to a great range of stake-holders and with the potential to](#)  
88 [could potentially better communicate complex geospatial relationships in an](#)  
89 [understandable format](#) (Krieger et al., 2012).

90 Aggregated annual malaria incidence was calculated for every village with at least one  
91 malaria case reported by the SHFs. Annual incidence in children under five years of age

**Commented [A9]:** Insert as reference: *PNG Med J* 57, 7–29.

**Commented [A10]:** Introduce abbreviation

**Commented [A11]:** Suggesting to move this sentence to after the next one. You may also slightly re-frame the sentence in order to highlight that this is a key aspect of your visualisation

92 was calculated and mapped as a proxy of local transmission since young children are  
93 the least immune age group and compared to adults are less likely to travel (Bousema et  
94 al., 2012). ~~A census conducted by the PNGIMR in the SHF catchment areas at baseline  
95 was used as source of village coordinates and population denominator data. The geo-  
96 referenced year 2000 National Census database was used to complement the PNGIMR  
97 census, particularly for villages outside the catchment area.~~ Self-reported mosquito-net  
98 use by all fever patients during the last year of surveillance (2014) was mapped by  
99 village and reported net use by all fever cases was graphed by year as an indicator of  
100 trends over time.

101 -Following a general declining trend, clear differences in incidence between villages  
102 were found in some sites. Mapping of village-level incidence appeared most useful in  
103 settings with very few malaria cases (Karimui and Sausi) or with pronounced spatial  
104 clustering of cases (Lemakot). In such settings, villages that could benefit from targeted  
105 interventions could be identified. -However, further investigations in some of the  
106 identified local foci are required to understand local heterogeneity. Unequal access [to](#)  
107 [health facilities](#), availability of other health care providers and treatment seeking  
108 behaviour may confound village-level incidence particularly if data is only originating  
109 from one facility. In addition, in some communities, village, hamlet and ward names may  
110 be used inconsistently by both patients and health workers. It is possible that cases in  
111 small communities are attributed to larger nearby villages. Furthermore, case-reporting  
112 becomes inaccurate in areas with constant transit of people. Uniform surveillance  
113 across all health facilities and a harmonized use of village names could optimize the  
114 current approach. Villages have been used as operational units for household-level net

115 distributions in PNG. Villages may also be the smallest feasible unit to target  
116 interventions making village level data highly relevant.

117 Since effective management of malaria [programmes](#) requires geo-spatial components to  
118 inform response-systems a next step could devise a simple standardized approach to  
119 generate spatial data on malaria risk that can be easily translated into response action  
120 can complement universal coverage campaigns in a meaningful way. [Since 2015, eNHIS](#)  
121 [has been piloted in 184 health facilities in PNG. The platform includes a geo-](#)  
122 [referenceding feature for mapping malaria cases ~~in~~at a village level and automated data](#)  
123 [analysis, reporting and identification of outbreaks \(Rosewell et al., 2017\). If proven](#)  
124 [successful eNHIS could considerably strengthen malaria surveillance in PNG.](#) A similar  
125 Spatial Decision Support System building upon and extending existing data collection  
126 systems and exploiting current geo-spatial tools has been validated in nearby Vanuatu  
127 and Solomon Islands in areas of very low transmission (Kelly et al., 2012). Areas in PNG  
128 that have reached a low level of transmission with clear foci may benefit from a similar  
129 approach. However, local (sub-national) capacity to further investigate local foci and  
130 implement targeted response action would be required as much as sufficient and  
131 sustained funding for these activities.

## 132 "Outlook"

- 133 • Age-specific malaria data collected routinely at health facilities and linked to the  
134 village of residence of patients may direct programmes to local foci of  
135 transmission.
- 136 • This approach appears most useful in settings with few cases or marked  
137 heterogeneity, where it may direct further investigations or (complementary)  
138 interventions.

**Commented [A12]:** I am missing the 'visualisation' aspect here. Please replace 1-2 of the bullet pointers provided with an outlook that links to the visualisation that you have produced



139 • A Malaria Register introduced in health facilities in PNG [and the eNHIS currently](#)  
140 [piloted in selected provinces](#) that records the village of residence of ~~test-~~  
141 ~~confirmed~~ malaria cases ~~is are an~~ [opportunities](#) to validate this approach at  
142 larger scale.

143 • [A simple tool for calculating and mapping malaria case incidence at district or](#)  
144 [sub-district level, as is currently included in the eNHIS](#), is required to  
145 operationalize the approach, [along with the capacity, policies, and mechanisms](#)  
146 [required to implement targeted response action at the respective operational](#)  
147 [level](#).

#### 148 **Box 1 ~~e~~Overall aim**

149 To assess whether [the simple visualization of](#) health information (malaria incidence, net  
150 use and residence of patients) extracted from a routinely implemented surveillance  
151 system can inform local malaria control programs to better target interventions.

152 Malaria surveillance systems are crucial for identifying the areas that are most affected  
153 by malaria. The proposed approach adds a geospatial component to health facility data  
154 in order to understand differences in malaria burden between villages and identify  
155 communities that would benefit from targeted interventions.

#### 156 **Box 2 Software used**

157 Maps were generated using the open-source software QGIS (version 3.0 Girona). The  
158 video was edited using Adobe Premier Pro CC (version 13.0.2, Adobe Systems  
159 Incorporated, San Jose, CA, USA)

#### References

160 Bousema, T., Griffin, J.T., Sauerwein, R.W., Smith, D.L., Churcher, T.S., Takken, W., Ghani,  
161 A., Drakeley, C., Gosling, R., 2012. Hitting hotspots: Spatial targeting of malaria for

162 control and elimination. PLoS Medicine 9, 1–7.

163 Cattani, J.A., Moir, J.S., Gibson, F.D., Ginny, M., Paino, J., Davidson, W., Alpers, M.P., 1986.

164 Small-Area Variations in the Epidemiology of Malaria in Madang Province. PNG

165 Med J 29, 11–17.

166 Greenwood, B.M., 1989. The microepidemiology of malaria and its importance to

167 malaria control. Trans R Soc Trop Med Hyg 83, 25–29.

168 Hetzel, M.W., Gideon, G., Lote, N., Makita, L., Siba, P.M., Mueller, I., 2012. Ownership and

169 usage of mosquito nets after four years of large-scale free distribution in Papua

170 New Guinea. Malar J 11, 192.

171 Hetzel, M.W., Morris, H., Tarongka, N., Barnadas, C., Pulford, J., Makita, L., Siba, P.M.,

172 Mueller, I., 2015. Prevalence of malaria across Papua New Guinea after initial roll-

173 out of insecticide-treated mosquito nets. Trop Med Int Health 20, 1745–1755.

174 Hetzel, M.W., Pulford, J., Maraga, S., Barnadas, C., Reimer, L.J., Tavul, L., Jamea-Maiasa, S.,

175 Tandrapah, T., Maalsen, A., Makita, L., Siba, P.M., Mueller, I., 2014. Evaluation of the

176 Global Fund-supported National Malaria Control Program in Papua New Guinea,

177 2009-2014. PNG Med J 57, 7–29.

178 Hetzel, M.W., Reimer, L.J., Gideon, G., Koimbu, G., Barnadas, C., Makita, L., Siba, P.M.,

179 Mueller, I., 2016. Changes in malaria burden and transmission in sentinel sites after

180 the roll-out of long-lasting insecticidal nets in Papua New Guinea. Parasites and

181 Vectors 9, 340.

182 Kelly, G.C., Tanner, M., Vallely, A., Clements, A.C.A., 2012. Malaria elimination: Moving

183 forward with spatial decision support systems. Trends Parasitol 28, 297–304.

184 Krieger, G.R., Bouchard, M.A., de Sa, I.M., Paris, I., Balge, Z., Williams, D., Singer, B.H.,  
185 Winkler, M.S., Utzinger, J., 2012. Enhancing impact: Visualization of an integrated  
186 impact assessment strategy. *Geospatial Health* 6, 303–306.

187 Mendis, K., Rietveld, A., Warsame, M., Bosman, A., Greenwood, B.M., Wernsdorfer, W.H.,  
188 2009. From malaria control to eradication: The WHO perspective. *Trop Med Int*  
189 *Health* 14, 802–809.

190 Rodríguez-Rodríguez, D., Marga, S., Lorry, L., Robinson, L.J., Siba, P.M., Mueller, I.,  
191 Pulford, J., Ross, A., Hetzel, M.W., 2019. (Submitted) Repeated mosquito net  
192 distributions, improved treatment, and trends in malaria cases in sentinel health  
193 facilities in Papua New Guinea. *Malar J.*

194 Rosewell, A., Makita, L., Muscatello, D., John, L.N., Bieb, S., Hutton, R., Ramamurthy, S.,  
195 Shearman, P., 2017. Health information system strengthening and malaria  
196 elimination in Papua New Guinea. *Malaria Journal* 16, 1–10.

197 World Health Organization, 2018. *World Malaria Report 2018*. Geneva.

198 World Health Organization, 2015. *Global Technical Strategy for Malaria 2016-2030*.  
199 Geneva.

200

**Table 1** Total number of malaria cases (all ages) and malaria cases in children under five years of age residing within and outside the catchment area

Site	Malaria cases residing in the catchment area		Malaria cases residing outside the catchment area		Total N (%)
	N (%)		N (%)		
<b>East Cape</b>					
Total cases (all ages)	3,265 (93)		250 (7)		3,515 (100)
Cases in children <5 years	1,076 (93)		82 (7)		1,158 (100)
<b>Sausi</b>					
Total	1,532 (99)		13 (1)		1,545 (100)
Children <5 years	305 (98)		6 (2)		311 (100)
<b>Karimui</b>					
Total	545 (99)		8 (1)		553 (100)
Children <5 years	260 (98)		5 (2)		265 (100)
<b>Lemakot</b>					
Total	2,831 (71)		1,142 (29)		3,973 (100)
Children <5 years	818 (73)		299 (27)		1,117 (100)
<b>Total cases</b>	<b>8,173 (85)</b>		<b>1,413 (15)</b>		<b>9,586 (100)</b>

**Commented [A13]:** Should this not rather be 6-59 months...? If your answer to my question is yes, you should also adjust this in the video

**Commented [A14]:** The formatting of this table is not ideal. 'Total...' and 'Children...' are not a 'Site'. Suggesting to insert an additional column that features the age categories

**Commented [A15]:** Consider to add this addition also in the following rows

**Commented [A16]:** See comment above

**Commented [A17]:** See comment above

**Commented [A18]:** See comment above

**Commented [A19]:** See comment above

Total cases in	2,459 (86)	392 (14)	2,851 (100)
children <5 years			

---

Commented [A20]: See comment above