

Effectiveness of private sector malaria control at a large sugar facility in Southern Mozambique

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Abstract

Context: Incentivizing scaled-up private firms’ involvement in malaria control and elimination activities requires a clear and robust demonstration of its benefits, and a quantification of which benefits can be expected to return to the firm. However, most studies to date examine benefits of malaria control purely from the perspective of the public sector. **Methods:** We analyze 4 years of malaria control and worker health and absenteeism data from a large sugar processing facility in Mozambique. We estimate the effect of indoor residual spraying (IRS) on worker health and absenteeism. **Results:** We demonstrate that the firm’s engagement in malaria control activities is profitable, even if we only consider those benefits which directly affect the firm’s bottom line and ignore secondary and societal benefits.

Introduction

Malaria has a nearly unquantifiably large economic impact on endemic societies. By affecting saving, investment (Shretta, Avanceña, and Hatefi 2016), risk perception, productivity, absenteeism (Nonvignon et al. 2016), human capital accumulation (Castel-Branco 2014), mortality, and costs of care (Sachs and Malaney 2002), malaria likely has a negative effect on GDP and growth (McCarthy, Wolf, and Wu 2000) (Orem et al. 2012). Because of the relative affordability of most interventions and the enormous societal costs of malaria, most forms of malaria control are cost-effective from the perspective of the government (White et al. 2011) (Purdy et al. 2013) (Howard et al. 2017).

From the perspective of the private sector, however, investing in malaria control is not such a clear-cut case. Public health interventions targeting malaria - and their corresponding cost-effectiveness evaluations - most often pertain to the public sector. Accordingly, the analytical framework and metrics employed by these analyses most often focus on impact pertaining to the public good, such as an increase in life years adjusted for disability or quality (Goodman, Coleman, and Mills 1999) (Shretta, Avanceña, and Hatefi 2016) (Lee et al. 2017) (Hanson 2004). Though population-level health is certainly of importance to businesses, and improvements in health incidentally improve the economy at all levels (Brundtland 1999) (Bloom and Canning 2008) (Vecchi, Hellowell, and Gatti 2013), these improvements may be too disperse or long-term to incentivize private sector involvement in health campaigns.

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In the context of Mozambique, where a significant sector of the economy is dominated by a full large-scale foreign direct investment projects (Robbins and Perkins 2012), the role of the private sector in health generally, and malaria specifically, is unequivocally important. Large agriculture and extractive industry firms take up wide swaths of land and employ hundreds of thousands (German, Schoneveld, and Mwangi 2013). The Mozambican state has encouraged large-scale enterprise with the aim of general economic development (Buur, Tembe, and Baloi 2012). And where large firms exist, they often take on social roles such as housing and health care (Winkler 2013). At times, this role is necessary from a purely practical standpoint; in other cases, it is employed under the guise of “corporate social responsibility” (Azemar and Desbordes 2009). Regardless of the language used, it is clear that private industry plays an important role in public health in Mozambique (Robbins and Perkins 2012) (Castel-Branco 2014).

Several cases exist of foreign firms engaging in large-scale malaria control campaigns (Mouzin and al. 2011) (Han 2015) (Bennett et al. 2017) (Kaula, Buyungo, and Opigo 2017). But these studies generally consider population health as the outcome measure of interest, rather than worker absenteeism or productivity. Similarly, they often neglect to differentiate between those clinical costs which are absorbed by the local health system versus those which are absorbed by the firm itself. In the literature, making the “investment case” for malaria control or elimination generally implies that the investor is the public sector, and takes into account those costs and benefits which are applicable to the public sector (Shretta et al. 2017); though appropriate in most cases (the public sector being the primary malaria control agent in most locations), the findings of these studies are rarely applicable to the private sector. In the case of a private firm not interested in “corporate social responsibility”, it is not clear whether investing in malaria control would be profitable or not. This lack of certainty most likely discourages investment in many cases.

To address the question of the profitability of malaria control activities from the standpoint of a private firm, we analyze data during a 7 year period from a private sugar facility in Southern Mozambique. We assess the effect of indoor residual spraying (IRS) on the absenteeism and health of workers, and demonstrate that the firm’s engagement in malaria control not only improved worker health, but also generated a positive return on investment.

Methods

In collaboration with the sugar processing facility, Maragra Açúcar SA (henceforth referred to as “Maragra”), we collected data for the period from January 2010 through December 2016. Data came from four sources: (i) the Human Resources’ roster of worker details and absences, (ii) the facility’s on-site clinic’s medical and laboratory records, (iii) the facility’s on-site malaria control program’s records pertaining to the dates, chemicals, and location of IRS activities, and (iv) interviews with company employees pertaining to costs, data limitations, etc. Digitization and collection of data took place during the period from March 2016 through May 2017. Supplementary data pertaining to worker characteristics was obtained from through the Centro de Investigação em Saude de Manhica’s (CISM) demographic census, which covered workers from the district, but not those who migrated from other parts of the country (Nhacolo et al. 2006).

Data pertaining to district-wide malaria incidence was obtained from Mozambique’s Boletim Epidemiológico Semanal (BES), which is the system by which the National Malaria Control Program monitors incidence at the district level throughout the entire country, and reports the number of confirmed weekly malaria cases at government health facilities. Using these case numbers, combined with population estimates from the National Statistical Institute (INE), we estimate each day’s annualized weekly malaria incidence rate (cases per 1000 population at risk), interpolated from the weekly figures. We retrieved weather data for all Mozambican stations from NOAA. We estimated the meteorological conditions at the centroid of Manhica using a simple interpolation method whereby the district’s weather conditions were estimated to be a function of all Mozambican weather stations’ reported conditions, inversely weighted by kilometers from district centroid.

Maragra regularly employs IRS at on-site worker households in order to reduce those workers’ (and their families’) risk of malaria infection. By using the CISM’s census, we were able to identify from the record of

houses sprayed (maintained by the company) who lived in those houses, as well as who lived off-site (and therefore did not receive IRS from Maragra). Then, using company HR and clinical records, we were able to identify absences and episodes of clinical malaria among all workers, as well as identify the time since the most recent IRS episode before the onset of absence or illness.

Worker characteristics, illness and absenteeism data, along with IRS activity data, were systematically stored, collected, and used at the individual level by Maragra, and therefore of generally high quality. We sought to understand the effect of IRS on individual workers' likelihood of absence from work as well as their likelihood of clinical malaria. To estimate this effect, we estimated separate models for absence and illness. We employed interrupted time series (Lopez Bernal, Cummins, and Gasparrini 2016) and a linear probability approach using the following econometric model.

$$\hat{Y}_{it} = \beta_0 + (\beta_1)(\text{Season}_t) + (\beta_2 \text{IRS} * \beta_3 \text{IRS}_t) + \dots + \epsilon$$

\hat{Y} is the rate of absence. β_1 represents the clinical malaria incidence at that time in the entire district of Manhica. Our demographic confounders (represented by ...) are sex, age, and worker department (field, factory, or administrative). Our intervention was not a simple yes/no, but rather the product of whether the residence of the worker in question was treated in the last year, and, if so, the time since treatment (represented above as the interaction term, where t represents time elapsed since commencement of the most recent IRS campaign). We define the malaria season as any time during which the clinical incidence of malaria in the district of Manhica was at or greater than the median clinical incidence of malaria for the entire study period. These weeks are flagged as red in Figure 1, Panel A. By using clinical incidence of the area of residence of the workers (as opposed to more typical proxies for malaria risk, such as only rainy vs. non rainy season), our seasonality estimate is a closer approximation of true malaria risk, incorporating lagged effects such as the incubation period of the parasite, as well as any inherent non-linear effects of weather. In addition, we adjust for daily precipitation

Because cost data was less systematically collected by Maragra, and because many costs could not be precisely quantified due to the abundance of in-kind and cross-departmental expenditures, we had to rely on rough estimations based on a mix of interviews, receipts, and interpolations. Since our program cost data is not as reliable as our worker characteristic and outcome data, we were conservative in our estimates, and generally tried to err on the side of program activities and materials costing *more* than what was reported, when doubt was aired. Cost data consisted of three types: (i) wages of malaria control employees, (ii) transportation and vehicle costs for IRS teams, and (iii) acquisition costs of purchasing IRS chemicals for fumigation (ACT and DDT), the latter two being combined into malaria control "programme" costs. We define the IRS program as "profitable" if return on investment is greater than 100%, ie if the savings associated with the estimated effect of IRS is greater than the costs of the program's administration.

Our formula for return on investment can be described in a straightforward fashion. . .

$$R = \frac{P_w - S_{wa} - S_{wc}}{P_w}$$

... where R is the return on investment, P is the malaria control program's total operating cost, w refers to costs at the per-worker level, a refers to savings through avoided absences, and c refers to savings through avoided clinical encounters.

All data processing and analysis were carried out in R (R Core Team 2017) and all analysis code is freely available online (Brew 2017). Ethical approval for this project was obtained from the Institutional Ethics Review Board for Health at the CISM prior to data collection.

Results

In Southern Mozambique, malaria peaks during the summer months (December through March) most years (Figure 1, panel A), and worker absenteeism rates track malaria incidence closely, following the same seasonal

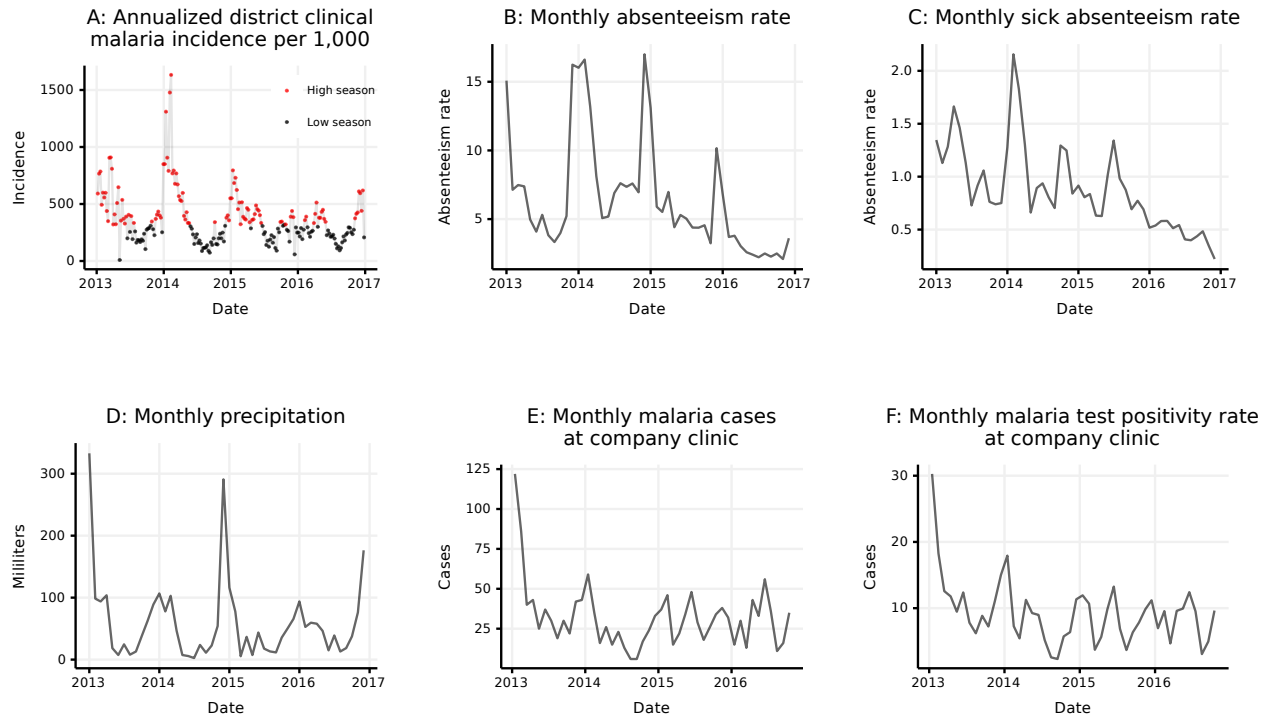


Figure 1: Clinical malaria (district of Manhica), all-cause absenteeism among Maragra workers, sick absenteeism among Maragra workers, estimated rainfall, positive cases at company clinic, and test positivity rate at company clinic

patterns (Figure 1, panel B). Both all-cause absenteeism and sick absenteeism have declined in recent years at Maragra (Figure 1, panel C), with the latter declining at a faster rate than the former. The fact that the rate of confirmed cases at the company clinic is largely non-seasonal (Figure 1, panels E and F) suggests that a significant portion of workers either seek care for malaria elsewhere (for example, government health posts, of which several are nearby and in some many cases closer to workers' residence than the company clinic) or do not seek care during malaria infection. Accordingly, we focus our analysis on all-cause absenteeism rather than sickness absenteeism or malaria diagnostics, with the assumption that much of illness is captured by absenteeism but not by the clinical data.

Fumigations: During the period from January 1st, 2012 through December 31st, 2016, the Maragra Malaria Control Unit carried out 11,567 episodes of fumigation of residential “agregados” (household combinations), for a total of 13,937 building-fumigation combinations. The total number of unique agregados sprayed during this period was 4,045. Among the 3,362 workers for whom we have reliable absenteeism and residential data, 561 had their homes fumigated at least once (the majority of workers live off of the facility).

Absences: We observed 1,759,100 unique worker-days among the 3,362 workers. The all-period average absenteeism rate was 5.56%, though this rate varied widely as a function of worker department, sex, residence, and season (table 1).

Costs: The malaria control program at Maragra has an annual operating budget of approximately XX, which includes the purchase of insecticide, the wages of IRS sprayers and drivers, transportation, record-keeping, and general administrative costs. Assuming linearity in costs, the program spends approximately XX per agregado sprayed. With each agregado containing an average of 2.2 workers, this translates to a cost of XX per worker protected per season. Much of the benefit of IRS goes to non-worker residents of sprayed agregados (who constitute a majority), but this benefit is purposefully ignored for this analysis.

Given the likelihood that clinical data does not fully capture all malaria cases, we do not quantify the costs of malaria infection to the company. Rather, we first estimate the reduction in absenteeism attributable

	2013	2014	2015	2016
Season: low	5.19%	6.97%	5.39%	3.3%
high	8.24%	12.79%	6.29%	2.65%
Department: Administrative	10.9%	11.01%	11.03%	10.11%
Factory	10.51%	13.3%	11.74%	9.31%
Field	4.36%	7.6%	3.88%	1.73%
Sex: F	3.98%	8.07%	4.36%	1.91%
M	8.12%	10.01%	6.54%	3.69%
Residence: Off site	6.6%	9.27%	5.59%	2.72%
On site	12.39%	13.69%	13.65%	13.47%
Precipitation: Dry	5.37%	7.72%	4.8%	2.42%
Rainy	7.88%	10.55%	6.96%	3.27%

Table 1: Absenteeism rate by year and worker characteristics

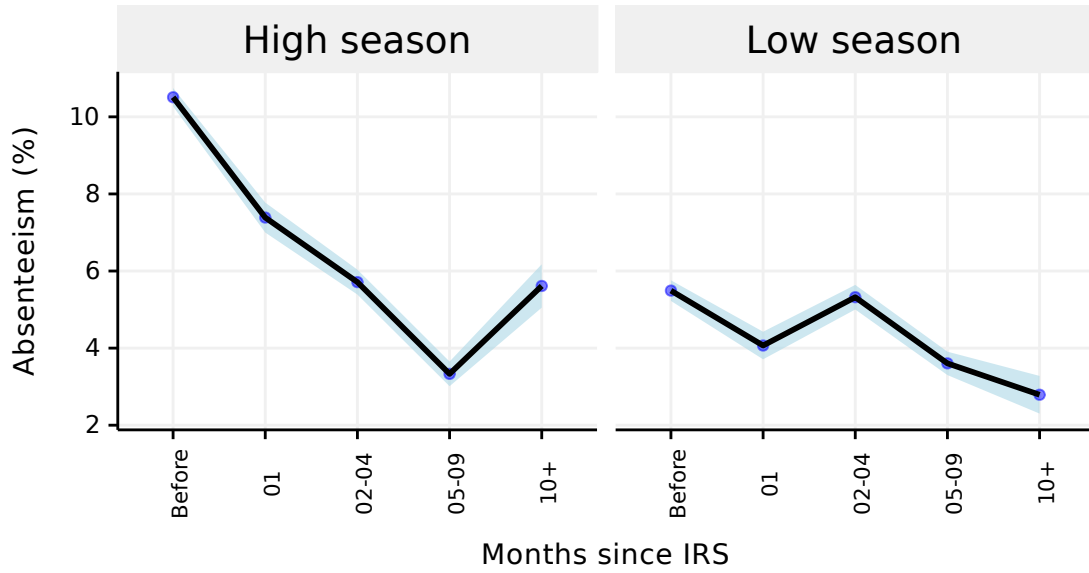


Figure 2: Estimated effect of IRS on absenteeism by season

to IRS, and then quantify the savings associated with prevented absences. Additionally, we calculate the clinical savings of IRS by first estimating the share of absences which are associated with an episode of clinical malaria, and then applying the clinical cost per case to the equivalent share of prevented absences. We intentionally ignore the savings accrued by the public health system, as well as the likely utility gains in secondary realms such as school absenteeism, productivity, etc.

Effect of IRS on absenteeism: IRS is associated with a year-long, significant reduction in absenteeism during the malaria season (figure 2). As one would expect if the mechanism by which IRS reduces absence is through reduced malaria infection, the effect of IRS during the low transmission season is significant, but far less substantial in effect size.

Details on full model here. Bla bla bla. See table 3.

	Term	Model 1	Model 2	Model 3	Model 4
1	(Intercept)	4.688 (P = < 0.001)	2.783 (P = < 0.001)	9.376 (P = < 0.001)	10.521 (P = < 0.001)
2	Malaria season	1.651 (P = < 0.001)	1.559 (P = < 0.001)	1.243 (P = < 0.001)	0.692 (P = < 0.001)
3	Months since IRS: Before	0.804 (P = < 0.001)	0.794 (P = < 0.001)	-0.162 (P = 0.2399)	-0.881 (P = < 0.001)
4	Months since IRS: 01	-0.621 (P = < 0.001)	-0.372 (P = 0.0433)	-0.536 (P = 0.0033)	-0.43 (P = 0.0167)
5	Months since IRS: 02-04	0.633 (P = < 0.001)	0.804 (P = < 0.001)	0.703 (P = < 0.001)	0.857 (P = < 0.001)
6	Months since IRS: 05-09	-1.084 (P = < 0.001)	-0.882 (P = < 0.001)	-1.085 (P = < 0.001)	-1.081 (P = < 0.001)
7	Months since IRS: 10+	-1.898 (P = < 0.001)	-1.567 (P = < 0.001)	-1.485 (P = < 0.001)	-1.488 (P = < 0.001)
8	Male		2.784 (P = < 0.001)	0.786 (P = < 0.001)	-0.177 (P = < 0.001)
9	Department: Factory			0.355 (P = < 0.001)	0.855 (P = < 0.001)
10	Department: Field			-6.993 (P = < 0.001)	-0.56 (P = < 0.001)
11	Rainy day				1.3 (P = < 0.001)
12	Temp contract				-9.548 (P = < 0.001)
13	Malaria season:Months since IRS: Before	3.366 (P = < 0.001)	3.42 (P = < 0.001)	3.217 (P = < 0.001)	2.616 (P = < 0.001)
14	Malaria season:Months since IRS: 01	1.669 (P = < 0.001)	1.623 (P = < 0.001)	1.32 (P = < 0.001)	0.499 (P = 0.0605)
15	Malaria season:Months since IRS: 02-04	-1.259 (P = < 0.001)	-1.195 (P = < 0.001)	-1.484 (P = < 0.001)	-2.04 (P = < 0.001)
16	Malaria season:Months since IRS: 05-09	-1.924 (P = < 0.001)	-1.834 (P = < 0.001)	-1.439 (P = < 0.001)	-1.059 (P = < 0.001)
17	Malaria season:Months since IRS: 10+	1.172 (P = 0.0022)	1.149 (P = 0.0026)	1.211 (P = 0.0014)	1.416 (P = < 0.001)

Table 2: Regression outputs

Discussion

Some text goes here

References

- Azemar, C., and R. Desbordes. 2009. “Public Governance, Health and Foreign Direct Investment in Sub-Saharan Africa.” *Journal of African Economies* 18 (4). Oxford University Press (OUP): 667–709. doi:10.1093/jae/ejn028.
- Bennett, Adam, Anton L. V. Avanceña, Jennifer Wegbreit, Chris Cotter, Kathryn Roberts, and Roly Gosling. 2017. “Engaging the Private Sector in Malaria Surveillance: A Review of Strategies and Recommendations for Elimination Settings.” *Malaria Journal* 16 (1). Springer Nature. doi:10.1186/s12936-017-1901-1.
- Bloom, David, and David Canning. 2008. “Population Health and Economic Growth,” 1–25.
- Brew, J.R. 2017. “Malaria and Sugar: An in-Depth Examination of the Effect of Malaria Control Activities on the Health and Productivity of Maragra Sugarcane Factory Workers.” *GitHub Repository*. <https://github.com/joebrew/maragra>; GitHub.
- Brundtland, Gro Harlem. 1999. “WHO on Health and Economic Productivity” 25 (2): 396–402.
- Buur, Lars, Carlota Mondlane Tembe, and Obede Baloi. 2012. “The White Gold: The Role of Government and State in Rehabilitating the Sugar Industry in Mozambique.” *Journal of Development Studies* 48 (3). Informa UK Limited: 349–62. doi:10.1080/00220388.2011.635200.
- Castel-Branco, Carlos Nuno. 2014. “Growth, Capital Accumulation and Economic Porosity in Mozambique: Social Losses, Private Gains.” *Review of African Political Economy* 41 (sup1). Informa UK Limited: S26–S48. doi:10.1080/03056244.2014.976363.
- German, Laura, George Schoneveld, and Esther Mwangi. 2013. “Contemporary Processes of Large-Scale Land Acquisition in Sub-Saharan Africa: Legal Deficiency or Elite Capture of the Rule of Law?” *World Development* 48 (August). Elsevier BV: 1–18. doi:10.1016/j.worlddev.2013.03.006.
- Goodman, CA, PG Coleman, and AJ Mills. 1999. “Cost-Effectiveness of Malaria Control in Sub-Saharan Africa.” *The Lancet* 354 (9176). Elsevier BV: 378–85. doi:10.1016/s0140-6736(99)02141-8.
- Han, Lily. 2015. “Malaria in Mozambique: trialling payment by results.” <http://www.theguardian.com/global-development-professionals-network/2014/mar/31/malaria-control-payment-by-results>.
- Hanson, K. 2004. “Public and Private Roles in Malaria Control: The Contributions of Economic Analysis.” *The American Journal of Tropical Medicine and Hygiene* 71 (9176). The American Society of Tropical Medicine; Hygiene: 168–73. http://www.ajtmh.org/docserver/fulltext/14761645/71/2_suppl/0700168.pdf

expires=1503070750&id=id&accname=guest&checksum=F969ACD7BE7E3A5D929D87EEA8A6EA2B.

Howard, Natasha, Lorna Guinness, Mark Rowland, Naeem Durrani, and Kristian S. Hansen. 2017. "Cost-Effectiveness of Adding Indoor Residual Spraying to Case Management in Afghan Refugee Settlements in North-west Pakistan During a Prolonged Malaria Epidemic." Edited by Guilherme S. Editor Ribeiro. *PLOS Neglected Tropical Diseases* 11 (10). Public Library of Science (PLoS): e0005935. doi:10.1371/journal.pntd.0005935.

Kaula, Henry, Peter Buyungo, and Jimmy Opigo. 2017. "Private Sector Role, Readiness and Performance for Malaria Case Management in Uganda, 2015." *Malaria Journal* 16 (1). Springer Nature. doi:10.1186/s12936-017-1824-x.

Lee, Bruce Y., Eli Zenkov, Chandrani Chatterjee, Baltazar Candrinho, Shufang Zhang, James Colborn, Olivier J. T. Briët, et al. 2017. "The Economic Value of Long-Lasting Insecticidal Nets and Indoor Residual Spraying Implementation in Mozambique." *The American Journal of Tropical Medicine and Hygiene* 96 (6). American Society of Tropical Medicine; Hygiene: 1430–40. doi:10.4269/ajtmh.16-0744.

Lopez Bernal, James, Steven Cummins, and Antonio Gasparrini. 2016. "Interrupted Time Series Regression for the Evaluation of Public Health Interventions: A Tutorial." *International Journal of Epidemiology*, June. Oxford University Press (OUP), dyw098. doi:10.1093/ije/dyw098.

McCarthy, Desmond, Holger Wolf, and Yi Wu. 2000. "The Growth Costs of Malaria," February. National Bureau of Economic Research. doi:10.3386/w7541.

Mouzin, Eric, and Et al. 2011. "Business Investing in Malaria Control: Economic Returns and a Healthy Workforce for Africa." *Progress & Impact Series*, no. 6.

Nhacolo, Ariel Q, Delino A Nhalungo, Charfudin N Saco, John J Aponte, Ricardo Thompson, and Pedro Alonso. 2006. "Levels and Trends of Demographic Indices in Southern Rural Mozambique: Evidence from Demographic Surveillance in Manhica District." *BMC Public Health* 6 (1). Springer Nature. doi:10.1186/1471-2458-6-291.

Nonvignon, Justice, Genevieve Cecilia Aryeetey, Keziah L. Malm, Samuel Agyei Agyemang, Vivian N. A. Aubyn, Nana Yaw Peprah, Constance N. Bart-Plange, and Moses Aikins. 2016. "Economic Burden of Malaria on Businesses in Ghana: A Case for Private Sector Investment in Malaria Control." *Malaria Journal* 15 (1). Springer Nature. doi:10.1186/s12936-016-1506-0.

Orem, Juliet, Joses Kirigia, Robert Azairwe, Ibrahim Kasirye, and Oladapo Walker. 2012. "Impact of Malaria Morbidity on Gross Domestic Product in Uganda." *International Archives of Medicine* 5 (1). Springer Nature: 12. doi:10.1186/1755-7682-5-12.

Purdy, Mark, David Rublin, Kuangyi Wei, and Matthew Robinson. 2013. "The Economic Case for Combating Malaria." *The American Journal of Tropical Medicine and Hygiene* 89 (5). American Society of Tropical Medicine; Hygiene: 819–23. doi:10.4269/ajtmh.12-0689.

R Core Team. 2017. *R: A Language and Environment for Statistical Computing*. Vienna, Austria: R Foundation for Statistical Computing. <https://www.R-project.org/>.

Robbins, Glen, and David Perkins. 2012. "Mining Fdi and Infrastructure Development on Africa's East Coast: Examining the Recent Experience of Tanzania and Mozambique." *Journal of International Development* 24 (2). Wiley-Blackwell: 220–36. doi:10.1002/jid.2817.

Sachs, Jeffrey, and Pia Malaney. 2002. "The Economic and Social Burden of Malaria." *Nature* 415 (6872). Springer Nature: 680–85. doi:10.1038/415680a.

Shretta, Rima, Anton L. V. Avanceña, and Arian Hatefi. 2016. "The Economics of Malaria Control and Elimination: A Systematic Review." *Malaria Journal* 15 (1). Springer Nature. doi:10.1186/s12936-016-1635-5.

Shretta, Rima, Ranju Baral, Anton L. V. Avanceña, Katie Fox, Asoka Premasiri Dannoruwa, Ravindra Jayanetti, Arumainayagam Jeyakumaran, Rasiike Hasanthan, Lalanthika Peris, and Risintha Premaratne. 2017. "An Investment Case to Prevent the Reintroduction of Malaria in Sri Lanka." *The American Journal of Tropical Medicine and Hygiene*, January. American Society of Tropical Medicine; Hygiene, 16–0209.

doi:10.4269/ajtmh.16-0209.

Vecchi, Veronica, Mark Hellowell, and Stefano Gatti. 2013. “Does the Private Sector Receive an Excessive Return from Investments in Health Care Infrastructure Projects? Evidence from the Uk.” *Health Policy* 110 (2-3). Elsevier BV: 243–70. doi:10.1016/j.healthpol.2012.12.010.

White, Michael T, Lesong Conteh, Richard Cibulskis, and Azra C Ghani. 2011. “Costs and Cost-Effectiveness of Malaria Control Interventions - a Systematic Review.” *Malaria Journal* 10 (1). Springer Nature: 337. doi:10.1186/1475-2875-10-337.

Winkler, Deborah. 2013. “Potential and Actual FDI Spillovers in Global Value Chains.”