

## Additional file 2

### Text S1. Additional methods and results for the regression analysis

All models consider the counts of mosquito from each night  $i$  with the response variable,  $\mu_i$ . Explanatory variables include a factor variable for the effect of the region that trapping occurred in ( $region_k$ ) and continuous variables for the effect of wind speed, temperature and relative humidity ( $windspeed_i$ ,  $temperature_i$ ,  $RH_i$ ). We additionally included a site-specific intercept,  $\alpha_{j[i]}$ , with subscript notation to indicate which of the  $j$  sites were sampled in night  $i$ . This results in the following regression model,

$$\ln(\mu_i) = \alpha_{j[i]} + \beta_{1,k}region_k + \beta_2windspeed_i + \beta_3temperature_i + \beta_4RH_i + \varepsilon_i$$

where  $\varepsilon_i$  is the Poisson distributed random error.

Following model selection, we evaluate  $\beta_{1,k}$  to determine how the average nightly counts of mosquitoes in region  $k$  differed from counts in Malelane, and we evaluate  $\beta_2, \beta_3, \beta_4$  to determine how the average counts of mosquitoes vary with wind speed, temperature, and relative humidity, respectively. If the relative values of  $\beta_{1,k}$  are consistent among traps (e.g.  $\beta_{1,1} > \beta_{1,2} > \beta_{1,3}$ ) for models fit to data from different traps, then we conclude that trap choice does not influence spatial comparisons among regions. We were additionally interested in the variance components, as high variances indicate that counts vary across sites. We did not estimate variation due to trap position due to the number of damaged traps. However, regression models fit with the additional random effect to data from a subset of sites with full trap data resulted in similar estimates.

**Table S4.** Model parameters, estimates, standard error (SE) and hypothesis tests for the Poisson regression analyses in Fig. 3.

<b>Model</b>	<b>Estimate</b>	<b>SE</b>	<b>Z value</b>	<b>P</b>
<b>BG data</b> (n = 41; variance across sites = 0.062)				
$\beta_{1,1}$ – Satara vs. Malelane	0.280	0.449	0.623	0.533
$\beta_{1,2}$ – Shingwedzi vs. Malelane	-0.073	0.828	-0.089	0.930
$\beta_{1,3}$ – Punda Maria vs. Malelane	-0.935	1.183	-0.791	0.429
$\beta_3$ – temperature	0.700	0.330	2.117	0.034
<b>CDC data</b> (n = 52; variance across sites = 0.621)				
$\beta_{1,1}$ – Satara vs. Malelane	0.049	0.579	0.085	0.933
$\beta_{1,2}$ – Shingwedzi vs. Malelane	-0.508	0.613	-0.829	0.407
$\beta_{1,3}$ – Punda Maria vs. Malelane	-1.215	0.657	-1.850	0.064
$\beta_2$ – wind speed	-0.098	0.059	-1.653	0.098
$\beta_3$ – temperature	0.498	0.102	4.864	<0.001
<b>Net data</b> (n = 52; variance across sites = 0.326)				
$\beta_{1,1}$ – Satara vs. Malelane	-0.388	0.424	-0.916	0.359
$\beta_{1,2}$ – Shingwedzi vs. Malelane	0.338	0.445	0.759	0.448
$\beta_{1,3}$ – Punda Maria vs. Malelane	-0.522	0.443	-1.179	0.238
$\beta_2$ – wind speed	-0.348	0.063	-5.545	<0.001
$\beta_3$ – temperature	0.344	0.079	4.336	<0.001
$\beta_4$ – relative humidity	-0.271	0.071	-3.823	0.001
<b>Net + CDC data</b> (n = 50, variance across sites = 0.181)				
$\beta_{1,1}$ – Satara vs. Malelane	-0.385	0.330	-0.742	0.458
$\beta_{1,2}$ – Shingwedzi vs. Malelane	-0.245	0.330	-0.724	0.458
$\beta_{1,3}$ – Punda Maria vs. Malelane	-0.651	0.340	-1.916	0.055
$\beta_2$ – wind speed	-0.217	0.044	-4.909	<0.001
$\beta_3$ – temperature	0.298	0.062	4.782	<0.001
<b>All data combined</b> (n = 38, variance across sites = 0.222)				
$\beta_{1,1}$ – Satara vs. Malelane	-0.288	0.382	0.752	0.452
$\beta_{1,2}$ – Shingwedzi vs. Malelane	-0.148	0.382	-0.387	0.699
$\beta_{1,3}$ – Punda Maria vs. Malelane	-0.688	0.404	-1.704	0.088
$\beta_2$ – wind speed	-0.107	0.051	-2.095	0.036
$\beta_3$ – temperature	0.432	0.076	5.702	<0.001