

1 **Dispersal of engineered male *Aedes*** 2 ***aegypti* mosquitoes.** 3

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28 **Abstract**

29 **Background**

30 *Aedes aegypti*, the principal vector of dengue fever, have been genetically engineered for use in a
31 sterile insect control programme. To improve our understanding of the dispersal ecology of
32 mosquitoes and to inform appropriate release strategies of 'genetically sterile' male *Aedes aegypti*
33 detailed knowledge of the dispersal ability of the released insects is needed.

34 **Methodology/Principal findings**

35 The dispersal ability of released 'genetically sterile' male *Aedes aegypti* at a field site in Brazil has
36 been estimated. Dispersal kernels embedded within a generalized linear model framework were
37 used to analyse data collected from three large scale mark release recapture studies. The
38 methodology has been applied to previously published dispersal data to compare the dispersal
39 ability of 'genetically sterile' male *Aedes aegypti* in contrasting environments. We parameterised
40 dispersal kernels and estimated the mean distance travelled for insects in Brazil: 52.8m (95% CI:
41 49.9m, 56.8m) and Malaysia: 58.0m (95% CI: 51.1m, 71.0m).

42 **Conclusions/Significance**

43 Our results provide specific, detailed estimates of the dispersal characteristics of released
44 'genetically sterile' male *Aedes aegypti* in the field. The comparative analysis indicates that despite
45 differing environments and recapture rates, key features of the insects' dispersal kernels are
46 conserved across the two studies. The results can be used to inform both risk assessments and
47 release programmes using 'genetically sterile' male *Aedes aegypti*.

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56 **Key words**

57 *Aedes aegypti*, dispersal, mean distance travelled, dispersal kernel, Brazil, transgenic, mark release
58 recapture, SIT

59 **Author summary**

60 Vector control using releases of sterile insects is a well-known approach. 'Genetically sterile' male
61 *Aedes aegypti* have been developed and released in a modern realisation of the sterile insect
62 technique. Released engineered males seek out and mate with wild females, with the resultant
63 offspring dying before they reach maturity. Control of a wild vector population can therefore be
64 achieved by maintaining sustained releases of sterile males whilst ensuring sufficient distribution
65 and coverage of released males across the target area. In order to efficiently plan releases of these
66 individuals detailed knowledge of how they disperse in the field is required. We present an analysis
67 of the dispersal of these engineered male *Aedes aegypti* using data from field experiments in Brazil.
68 Our results provide detailed information on how the mosquitoes disperse over their potential flight
69 range.

70 **Introduction**

71 Dengue, an arbovirus, has seen recent re-emergence and spread on a global scale [1] and is now
72 responsible for an estimated 390 million infections annually [2]. The vector of dengue is the *Aedes*
73 mosquito, with *Ae. aegypti* and *Ae. albopictus* responsible for the majority of disease transmission
74 [3]. The release of 'genetically sterile' male *Aedes* mosquitoes has been demonstrated to be a
75 valuable additional tool by which the vector can be controlled [4]. Understanding the ability of the
76 released 'genetically sterile' insects to disperse, and their behaviour whilst doing so, is an important
77 step in designing robust, efficient and effective releases. Attaining adequate coverage of released
78 sterile insects across a given area is a major operational challenge of a sterile insect control effort
79 [4]. Knowledge of the distribution of dispersal distances of released insects will improve our ability
80 to target releases, obtain required coverage densities, confidently predict the potential spatial range
81 of a release and is a key element for the assessment of risk.

82 Independently conceived by Petersen in 1896 and Lincoln in 1930 [5], mark-recapture, capture-
83 recapture or mark-release-recapture studies (hereafter referred to as MRR) have since become a key
84 set of ecological methods. MRR allows inference to be drawn about a number of important
85 ecological factors including the estimation of population size and quantification of dispersal and
86 survival. The methods have been used across a diverse range of species, from whales [6] to fruit flies
87 [7], and have seen extensive use in mosquito ecological studies.

88 Analysis of the location of recaptured marked insects with respect to the release point allows
89 inference to be made about the dispersal of the released insects. A number of MRR studies have
90 been performed with the aim of assessing the dispersal ability of both lab [8–11] and 'genetically
91 sterile' [12] strains of male *Ae. aegypti*. However, these studies often document only the mean
92 distance travelled (MDT) [8–12] or range [11,12] of dispersal of the released insects. Common
93 measures of range are the flight range 50% and flight range 90% (FR50 and FR90 respectively) which
94 are estimates of the distance within which 50% or 90% of all insects are expected to disperse
95 [12,13]. The MDT and flight range are intuitive summary measures but do not characterise dispersal
96 well when the distribution of dispersal distances is positively skewed with a long tail.

97 For greater insight, a better understanding of the distribution of dispersal distances can be obtained
98 by incorporating dispersal kernel theory, popular in studies of population spread [14] and seed
99 dispersal [15], into a generalised linear model (GLM) framework [16,17]. Dispersal kernels represent
100 the distribution of dispersal distances over the whole flight range. They can take a wide range of
101 forms with the flexibility to represent dispersal for a diverse range of species [18].

102 This study attempts improve the characterisation of the dispersal ability of 'genetically sterile' male
103 *Ae. aegypti* mosquitoes using data from large-scale MRR experiments carried out at an urban field
104 site in Brazil. The analysis facilitates the quantification of dispersal through the parameterisation of a
105 dispersal kernel for the released insects. Many summary measures of interest relating to dispersal
106 may be drawn from such a kernel. To enable a comparison of both the biological outcomes and
107 methodology employed, the analytical methods are also used to re-analyse published data on the
108 dispersal of 'genetically sterile' male *Ae. aegypti* at an uninhabited forested site in Pahang, Malaysia
109 [12]. The dispersal ability of the 'genetically sterile' insects was previously analysed in the Malaysian
110 study using methods detailed in Morris et al. (1991) [13] and evaluated the MDT to be 52.4m (95%
111 CI: 41.6m, 61.4m) [12]. The aims of the re-assessment of dispersal ability are: i) to assess the
112 robustness of the estimate of dispersal from Brazil data to habitat and locational heterogeneities, ii)
113 to explore potential differences in dispersal behaviour between sites and iii) to assess the

114 applicability of the dispersal kernel method in comparison with more common approaches to
115 estimating and quantifying dispersal.

116 **Methods**

117 **Ethics statement**

118 Before establishment of the 'genetically sterile' male *Ae aegypti* line in the mass rearing laboratory
119 and subsequent open releases, regulatory approvals were obtained from the appropriate Brazilian
120 national regulatory body: the Brazilian National Biosafety Technical Commission (CTNBio). Releases
121 were preceded by community engagement with consent and support obtained from regional (Bahia
122 health secretary) and local community leaders (Town Mayor, health secretary and vector control
123 authorities). Prior to sampling, informed consent was received from the landowners.

124 **Study site**

125 The field site is located in Itaberaba, a suburb of the city of Juazeiro, Bahia, Brazil (Latitude: -9° 26'
126 59", Longitude: -40° 28' 53") (Fig 1A). The site is located in a semi-arid part of Brazil and consists
127 mainly of low-socioeconomic status residential housing. The majority of houses at the study site
128 were single-story brick and concrete buildings with unscreened windows. The habitat across the
129 sampled region was a homogenous urban environment.

130 **Mark-Release-Recapture**

131 The 'genetically sterile' line used was OX513A and was reared according to methodologies given in
132 Carvalho *et al.* (2014) [19].

133 A total of 19,164 'genetically sterile' male *Ae. aegypti* formed three releases. Individuals from each
134 release were marked with the same coloured fluorescent powder (www.dayglo.com). Release 1 (red
135 release) and release 2 (blue release) were performed on 21 February 2011 and consisted of 5,349
136 and 5,465 individuals released from points 1 and 2 (Fig 1B) respectively. Release 3 (yellow release)
137 was performed on 25 February 2011 with 8,350 individuals released from point 1 (Fig 1B).

138 Aspiration sampling was used to recapture marked adults. Sampling was conducted using locally
139 made battery powered hand-held aspirators. After obtaining consent from the respective property
140 owner, each building was sampled for a set period of 15 minutes. Sampling locations were
141 distributed across the study site (Fig 1B) and sampling was conducted for up to nine days post
142 release (S1 Data). Sampling locations were chosen by randomly selecting one household from each
143 of the 47 (60m by 60m) grid squares each day (with the exception of day 5 for the red and blue
144 release and day 1 for the yellow release where, for logistical reasons, the number of households
145 sampled was lower). Each mosquito collected was assessed to determine the i) origin ('genetically
146 sterile' or wild as indicated by the presence/absence of fluorescent powder respectively), ii) sex and
147 iii) genus (*Aedes* or non-*Aedes*). Weather variables (daily maximum temperature and maximum
148 humidity) were recorded from a local weather station (situated approximately 10.2km north-west of
149 Itaberaba).

150 A secondary analysis of the MRR data from a previously published study in Malaysia [12] was
151 undertaken to obtain a comparative second estimate of the 'genetically sterile' insect's dispersal
152 ability and associated density kernel. In this study a total of 6,045 marked 'genetically sterile' males
153 were released at a forested site in Pahang, Malaysia. For a detailed description of the study site and
154 MRR methods please see Lacroix *et al.* (2012) [12].

155 **Model framework**

156 All multivariable analyses were performed within a GLM framework. In instances where the number
 157 of recaptures is small relative to the total releases, the Poisson regression model may be used as an
 158 appropriate approximation [20].

159 We assume that the count response variable (recaptures) is Poisson distributed with mean μ and
 160 variance μ

$$161 \quad Y_i \sim \text{Pois}(\mu_i). \quad (1)$$

162 The response must be ≥ 0 . Therefore a log link function is used to link the mean to the explanatory
 163 variables

$$164 \quad \ln(\mu_i) = x_i \beta, \quad (2)$$

165 where $x_i \beta$ is a linear predictor

$$166 \quad x_i \beta = \beta_0 + \beta_1 x_{i1} + \dots + \beta_p x_{ip}, \quad (3)$$

167 where β denotes the unknown parameters to be estimated and x_i , the explanatory variables.

168 Parameter estimates were obtained by maximising the log-likelihood (ℓ) of the data:

$$169 \quad \begin{aligned} \ell \quad Y \quad & \sum_{i=1}^n (y_i \ln \mu_i - \mu_i - \ln(y_i!)) \\ & = \sum_{i=1}^n (y_i x_i \beta - \exp(x_i \beta) - \ln y_i!). \end{aligned} \quad (4)$$

170 In the situation where the response variable is overdispersed (variance \gg mean) a Poisson GLM,
 171 where the variance is assumed to equal the mean, would be misspecified. In this instance, the
 172 negative binomial GLM, detailed below, may be used [21–23].

173 We assume that the count response variable follows a negative binomial distribution. A Poisson
 174 model is used for the count, conditional on the mean value, Z_i , which is assumed to have a gamma
 175 distribution, with mean, μ_i , and constant scale parameter, θ

$$176 \quad Y_i \sim \text{Poisson}(Z_i), \quad Z_i \sim \text{gamma}(\mu_i, \theta). \quad (5)$$

177 Therefore the expected value of Y and the variance of Y are as follows

$$178 \quad E(Y_i) = \mu_i, \quad \text{Var}(Y_i) = \mu_i + \frac{\mu_i^2}{\theta} \quad (6)$$

179 The mean response, μ_i , may be linked to a linear combination of explanatory variables using the log
 180 link function (Equation (2)) and linear predictor (Equation (3)). Parameters were estimated by
 181 maximising the log-likelihood (ℓ) of the model:

182
$$\ell = \sum_{i=1}^{i=n} \theta(\ln(\theta) - \ln(\theta + \mu_i)) + \ln(\Gamma(\theta + y_i)) - \ln(y_i! \Gamma(\theta)) - y_i(\ln(\mu_i) - \ln(\theta + \mu_i)), \quad (7)$$

183 where the gamma function, Γ , is

184
$$\Gamma(n) = (n-1)! \quad (8)$$

185 **Dispersal kernels**

186 Considerable inconsistencies abound regarding different interpretations of the term ‘dispersal
 187 kernel’ [18,24,25]. Two kernel definitions, often used interchangeably, are i) the probability density
 188 function (pdf) of the dispersal distance of each disperser. Referred to as the distance kernel [18] or
 189 the distance pdf [25] and ii) the density of probability of a given bearing and dispersal distance from
 190 the source. Referred to as the location kernel [18] or the density pdf [25].

191 We adopt the terminology of Cousens *et al.* [25], henceforth referring to kernel type 1 as the
 192 distance pdf and type 2 as the density pdf. Both kernel types are true pdfs, integrating to 1 (the
 193 density pdf is integrated over the whole 2d space). Both kernel types are closely related. The
 194 distance pdf can be derived by multiplying the density pdf by $2\pi d$ where d is the distance from the
 195 source [18] (assuming radial symmetry). Examples of these kernel types are shown in figure 2.

196 The density pdfs, assuming radial symmetry, used in this analysis are defined by the following
 197 functions

198
$$\text{Negative exponential kernel} = \frac{1}{2\pi a^2} e^{\left(-\frac{d}{a}\right)} \quad a > 0, \quad (9)$$

199
$$\text{Exponential power kernel} = \frac{b}{2\pi a^2 \Gamma\left(\frac{2}{b}\right)} e^{\left(-\frac{d^b}{a^b}\right)} \quad a, b > 0, \quad (10)$$

200 where d is the distance (metres), a and b are kernel parameters and Γ the gamma function
 201 (Equation (8)) [18]. The negative exponential kernel is characterised by the exponential power kernel
 202 when $b = 1$. The associated MDT functions are

203
$$\text{Negative exponential MDT} = 2a \quad (11)$$

204
$$\text{Exponential power MDT} = a \left(\frac{\Gamma\left(\frac{3}{b}\right)}{\Gamma\left(\frac{2}{b}\right)} \right). \quad (12)$$

205 Estimates of FR₅₀ and FR₉₀ are made by assessing the cumulative distribution of the distance pdf at
 206 the 50% and 90% levels.

207 **Variables**

208 The outcome variable was the number (count) of marked ‘genetically sterile’ male *Ae. aegypti*
 209 recaptured. Potential explanatory variables included in the Brazil analysis were: i) a spatial measure
 210 which could either be the measured distance (m) between release and recapture or the density as
 211 calculated by a parameterisation of a given density pdf, ii) the number of days post release, the

212 effect of which is assumed to be linear, iii) the number of wild *Aedes* species collected iv) the
213 number of non-*Aedes* wild mosquitoes collected, v) the maximum temperature (°C) on the day of
214 collection, vi) the maximum humidity (%) on the day of collection and vii) the directional quadrant,
215 North, South, East or West (relative to release point) that the collection was made in.

216 Due to the relatively low recapture number, data from all three MRR experiments were combined
217 for analysis.

218 For the Malaysia analysis the outcome variable was the number (count) of marked ‘genetically
219 sterile’ male *Ae. aegypti* recaptured. Potential explanatory variables included in the analysis were: i)
220 a spatial measure which could either be the measured distance (m) between release and recapture
221 or the density as calculated by a parameterisation of a given density pdf, ii) the number of days post
222 release, the effect of which is assumed to be linear iii) the number of wild *Aedes* species (specifically:
223 *aegypti*, *albopictus* and *togoi*) collected, iv) the number of wild *Culex* collected and v) a categorical
224 variable indicating if the recapture location was uphill or downhill from the release site.

225 Three models were evaluated to compare different transformations of the distance explanatory
226 variable. Model 1 incorporated all explanatory variable including distance, Model 2 all explanatory
227 variables including distance density (negative exponential kernel) and model 3 all explanatory
228 variable including distance density (exponential power kernel).

229 For model 1 the full model was fitted using maximum likelihood techniques, utilising the GLM and
230 Negative binomial GLM function of the statistical software package R [26] with the MASS package
231 [21]. All explanatory variables were included in the initial model as well as an interaction term
232 between distance and day post release. Model selection by minimising the Akaike information
233 criterion (AIC), a penalised likelihood score, was then performed using the MASS package [21]. The
234 AIC is calculated as

$$235 \quad \quad \quad AIC = 2k - 2\ln(\mathcal{L}) \quad \quad \quad (13)$$

236 where k is the number of parameters and \mathcal{L} the maximised likelihood value.

237 For models 2 and 3 fitting was performed using the following process. First, the distance density was
238 estimated using the assigned kernel. The GLM was then fitted using the same process as for model 1,
239 as a function of the transformed distance explanatory variable. These steps were then optimised
240 over the kernel parameter space allowing identification of the optimal combination of explanatory
241 variables and kernel parameters as indicated by the AIC. The best overall model was judged to be
242 the one with the minimum AIC value.

243 For comparison, the estimated survival of released insects in the Brazil study predicted using the
244 GLM was also calculated using a non-linear regression approach [27], that was also used in the
245 original analysis of the Malaysia data [12].

246 **Kernel confidence intervals**

247 Following model estimation, 95% confidence intervals were calculated for the maximum likelihood
248 kernel parameter estimates using the profile likelihood method. The maximised log-likelihood with
249 respect to β , α and b , i.e. that corresponding to the maximum likelihood estimates (MLEs) of β ,
250 α and b , is defined as $\ell(a, b, Y)$.

251 First kernel parameter a was increased or decreased in small increments whilst β was held at the

252 MLE ($\hat{\beta}$) and kernel parameter b was optimized conditional on $\hat{\beta}$ and the assumed value of α
 253 (a_0), giving the log-likelihood:

$$254 \quad \ell(a, \tilde{b} | Y), \quad (14)$$

255 where $\tilde{b} = \text{MLE } b | a_0, \hat{\beta}$.

256 Secondly, kernel parameter b was increased or decreased in small increments whilst β was held at
 257 the MLE ($\hat{\beta}$) and kernel parameter α was optimized conditional on $\hat{\beta}$ and the assumed value of
 258 $b(b_0)$, giving the log-likelihood:

$$259 \quad \ell(a, b | Y), \quad (15)$$

260 where $\tilde{a} = \text{MLE } a | b_0, \hat{\beta}$.

261 After each change the log-likelihood of the model was recalculated and a corresponding test statistic
 262 assessed. For example, evaluating for a , the G^2 statistic was calculated

$$263 \quad G^2 = 2 \left(\ell(a, b | Y) - \ell(\tilde{a}, \tilde{b} | Y) \right). \quad (16)$$

264 The G^2 statistic was compared to the χ^2 distribution (with 1 degree of freedom) for the (1- α)
 265 percentile. Thus for 95% confidence intervals the critical G^2 value is 3.84. The log-likelihood surface
 266 was calculated with respect to kernel parameters of the optimal model for exploration and
 267 visualisation of the parameter space for both the Brazil and Malaysia analyses.

268 Results

269 Primary Analysis - Brazil

270 Recaptures for the three MRR experiments are summarised in Table 1. The locations of recaptures
 271 are shown in figure 3. The mean count of recaptured marked males (the response), per sample, per
 272 day was 0.077 (variance = 0.73). Over the recapture period the maximum daily temperature ranged
 273 between 25.4°C-34.6°C and the maximum relative humidity between 66%-92%.

274 A summary of model performance using the untransformed- and transformed-distance explanatory
 275 variable is shown in Table 2. Combining all available data (from the red, yellow and blue releases)
 276 the best fitting model (lowest AIC) incorporated the exponential power kernel. The maximum
 277 likelihood exponential power density pdf has an associated MDT of 52.8m (95% CI: 49.9m, 56.8m),
 278 FR_{50} of 52.4m (95% CI: 50.6m, 54.7m) and FR_{90} of 83.0m (95% CI: 74.8m, 93.9m). The MLE kernel,
 279 log-likelihood surface and examples of kernels drawn from 95% CI parameter values are shown in
 280 figure 4.

281 For all three models the distance or distance density and the number of days post release were
 282 strongly associated with recapture number. There was no evidence of an interaction between
 283 distance (untransformed or transformed) and the number of days post release in any of the models
 284 considered. Assuming no emigration, the estimated mortality rate of released insects of 0.62 (95%
 285 profile likelihood CI: 0.84, 0.42) would equate to a mean average lifespan of $0.62^{-1} = 1.61$ days (95%

286 CI: 1.19 days, 2.38 days). Estimates of the mean average lifespan calculated using the non-linear
287 regression approach [27] were 1.00 days (95% bootstrapped CI: 0.63 days, 1.48 days).

288 Other significant explanatory variables were the number of non-*Aedes* mosquitoes recorded from
289 the sample and the maximum humidity. There was evidence of a lack of radial symmetry in dispersal
290 from the release point as the quadrant explanatory variable was also associated with recapture
291 number. A summary of the parameter estimates from the optimal model, using the exponential
292 power transformation of distance as an explanatory variable is shown in Table 4.

293 **Secondary analysis - Malaysia**

294 The MRR performed in Malaysia was associated with consistently higher recaptures than the MRR
295 experiments in Brazil. Of 6,045 released 'genetically sterile' males 3,034 (50.2%) were recaptured
296 over the 15-day course of the experiment. The count of recaptured, marked males (the response)
297 was very overdispersed (mean=2.6 per sample per day, variance=523), and therefore a negative
298 binomial GLM was fitted. A summary of the model performance using the untransformed and
299 transformed distance explanatory variable is shown in Table 6.

300 Again, the optimal model, as determined by AIC, used the exponential power density pdf, although
301 the negative exponential density pdf produced only marginally inferior fit (AIC = 668.0 and 668.4 for
302 the exponential power and negative exponential kernels respectively). The MLE exponential power
303 pdf estimates a MDT for the 'genetically sterile' release of 58.0m (95% CI: 51.1m, 71.0m), FR_{50} of
304 51.8m (95% CI: 47.9m, 58.7m) and FR_{90} of 105.7m (95% CI: 86.5m, 141.1m). The MLE kernel, log-
305 likelihood surface and examples of kernels drawn from 95% CI parameter values are shown in figure
306 5.

307 The coefficient summaries from the negative binomial model using the exponential power
308 transformed distance explanatory variable are shown in Table 8. This second analysis again indicates
309 that distance is an important significant predictor of the expected count of recaptures. The number
310 of days post release was also significantly associated with recapture number. Assuming no
311 emigration, the estimated mortality rate of released insects of 0.46 (95% profile likelihood CI: 1.03,
312 0.13) would equate to a mean average lifespan of $0.46^{-1}=2.17$ days (95% CI: 0.97 days, 8.85 days).
313 Unlike the analysis for Brazil there was evidence of an association between the distance and the
314 number of days post release explanatory variables.

315 For a direct comparison the distance pdf and density with respect to distance for the optimal kernels
316 estimated from the Itaberaba and Malaysia MRR experiments have been overlaid (Fig 6).

317 **Discussion**

318 An in-depth analysis of the dispersal ability of released 'genetically sterile' male *Ae. aegypti*
319 mosquitoes in the field has been conducted. The primary analysis, of MRR data from Brazil, indicates
320 distance from the release point to be an important predictor of the expected number of recaptures.
321 The relationship between the recapture number and the distance from the release point is highly
322 non-linear. The regression model performed optimally when an exponential power dispersal kernel
323 was used to transform distances. The analysis methodology was also used to re-analyse MRR from
324 Malaysia, where again an exponential power kernel provided best model fit.

325 The optimal Brazil GLM performed well, explaining around half of the variation observed in the data.
326 Transformed distance and the number of days post release were the most influential, highly
327 significant predictors of recapture number. The decline in numbers temporally after release is
328 considered to be predominantly due to the effect of mortality. The effect of emigration from the

329 study area can be important [28] and would further reduce numbers but was assumed to be small
330 due to the size of the study area and low recaptures at the periphery.

331 Distance was highly significantly correlated with recapture number in all models considered. The
332 exponential power dispersal kernel provided the optimum model fit. This kernel parametric form is
333 slightly more flexible than the negative exponential. The kernel produced showed a high and
334 consistent level of dispersal from 0-33m from the release site. After this point the density falls fairly
335 steeply, reaching very low levels shortly after 100m ($FR_{90} = 83.0m$), indicating that coverage
336 decreases quickly at increasing distances more than 33m from a release point. The MDT estimated
337 using the best-fit kernel parameters was 52.8m (95% CI: 49.9m, 56.8m). This is consistent with a
338 number of published field studies of male *Ae. aegypti* dispersal which estimate mean distance
339 travelled to range from 10m to 100m (Table 10). There are however, a limited number of studies of
340 male *Ae. aegypti* dispersal as focus has been on the biting females. It is however, important to note
341 that for skewed distributions of dispersal distances the MDT as a measure of central tendency
342 should be interpreted with some caution.

343 The number of non-*Aedes* mosquitoes was significantly positively correlated with the number of
344 recaptured 'genetically sterile' mosquitoes. This explanatory variable is thought to be a proxy for the
345 house-attractiveness or accessibility of a house to mosquitoes; large numbers of other mosquito
346 species may indicate that the household is situated in a favourable location or particularly amenable
347 or attractive to mosquitoes. Clustering of *Aedes* mosquitoes at the household level is a commonly
348 observed phenomenon [29–31]. An alternative explanation could also lie in differences in the
349 abilities of operators to sample mosquitoes.

350 The humidity and directional quadrants were significantly associated with the number of recaptured
351 'genetically sterile' mosquitoes but had small effect sizes. It would be expected that any directional
352 differences are attributable to site- and time-specific heterogeneities in terrain, habitat type, wind
353 direction or other external factors [32–34]. Humidity was positively associated with recapture
354 number, which could be due to an increased tendency for *Ae. aegypti* to seek shelter with increasing
355 humidity [35]. These explanatory variables must be interpreted with some caution as there is the
356 potential for selection by AIC to overfit models [36,37]. The covariates may therefore be included in
357 the final models despite their relatively small influence on model fit.

358 One limitation of this study was the low number of recaptures in the Brazil dataset. For this reason
359 there was little power to analyse individual releases separately, necessitating the analysis of the
360 combined datasets and the assumption that influences not accounted for would be similar across all
361 three releases. The proportion of individuals recaptured may have been improved by a more
362 targeted, or higher intensity, sampling effort or increased survival of the released individuals.
363 Alternatively, the absolute number of recaptures could have been increased with larger release
364 numbers. We have assumed throughout that aspiration will not remove all fluorescent dust from the
365 marked individuals [38]. The dataset could also have been further improved with household-specific
366 monitoring of climatic factors to give greater resolution to observations on the relationship between
367 meteorological variables and recapture numbers. The standard errors of the GLM coefficients (Table
368 4, Table 8) were estimated conditional upon the MLEs for kernel parameters a and b whilst the
369 confidence intervals for the kernel parameters were obtained conditional upon the MLE for the GLM
370 coefficients. Thus the reported standard errors are likely to be smaller than if we had been able to
371 compute the unconditional standard errors and confidence intervals for all of the parameters.
372 Analysis of the residuals indicated little residual spatial autocorrelation with perhaps the exception
373 of some under-estimation of recapture numbers at further distances (>150 m), potentially indicating

374 the influence of long-distance dispersers [39], although, due to the small number of recaptures, this
375 is difficult to verify.

376 The optimal GLM associated with the Malaysia data also explained approximately 50% of the
377 variation observed in the recapture data. For the optimal model only two covariates, number of days
378 post release and distance, plus their interaction term were included. The optimal model again used
379 the exponential power dispersal kernel, with a corresponding MDT estimate of 58m (95% CI: 51.1m,
380 71.0m) that corroborated the MDT estimate of 52.4m (95% CI: 41.6m, 61.4m) from the previously
381 published analysis of these data [12]. The FR_{50} estimate of 51.8m (95% CI: 47.9m, 58.7m) was
382 substantially different from the previously published estimates of 16.2m (95% CI: 10.5m, 22.5m), a
383 product of the different underlying models for dispersal with respect to distance used in each
384 analysis. This deviation further highlights the potential benefits of more accurately characterising
385 dispersal behaviour. The number of days post release covariate, as expected, was significantly
386 negatively associated with recapture number. The associated average lifespan of 2.17 days (95% CI:
387 0.97 days, 8.85 days) was in good agreement with the previously published estimate of 2 days (95%
388 CI: 1.8 days, 2.2 days). The coefficient indicated a smaller effect size than seen in the Brazil MRR
389 data, implying improved survival, less emigration or a combination of the two for individuals in the
390 Malaysia releases. The recapture rate in the Malaysia MRR experiment was very high, approximately
391 50% of all individuals released were recaptured. This may bias the results and could violate the
392 underlying assumption that the negative binomial distribution approximates proportions when
393 recapture numbers are small relative to the release size.

394 The lack of significant interaction between the number of days post release and distance in the Brazil
395 data provides evidence for a single main dispersal event on release (the probability of travelling a
396 given distance is not influenced by the number of days post-release). In the Malaysia analysis the
397 significant interaction term indicates a more continual dispersal process over time, possibly due to
398 the lack of favourable (urban/peri-urban) habitat across the whole range of the study site at this
399 location. However the influence of the interaction term on predicted recaptures is very small; the
400 majority of released individuals have died (or emigrated) before the interaction term becomes
401 influential. For both locations the majority of recaptures are predicted spatially and temporally close
402 to the release location and date respectively. There is published evidence to support either the
403 occurrence of a single dispersal event [40–42] or a more continuous dispersal process upon release
404 [8,9,11].

405 For experiments carried out in different habitats, on different continents, the estimated dispersal
406 kernels were very similar (Fig 6). The Brazil dispersal kernel is slightly fatter tailed (larger b
407 parameter) but in general there is evidence for a degree of consistency in the dispersal ability of
408 ‘genetically sterile’ male *Ae. aegypti* across a range of environments. Consistent dispersal may
409 facilitate more generalised release procedures for sterile insect releases across a range of release
410 locations and scenarios.

411 Accurately measuring and assessing the dispersal of released ‘genetically sterile’ male *Ae. aegypti* in
412 the field is a vital component necessary to optimise vector control using these genetically sterile
413 individuals. A successful control program using ‘genetically sterile’ male *Ae. aegypti* would maximise
414 ‘genetically sterile’ insect density over the target area. Knowledge of the released insects’ ability to
415 disperse is vital in predicting their density with respect to specific release points or routes. An ability
416 to predict the coverage of dispersed individuals will facilitate the design and implementation of
417 more efficient control and monitoring programs in the future.

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427

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520 **Supporting information legends**

521 **S1 Data. Brazil MRR data.** Georeferenced recapture number with respect to release, time and meteorological
522 variables.

523 **Figure legends**

524 **Figure 1. Brazil study site sampling locations.** A) The study site, Itaberaba, a suburb of the city of Juazeiro,
525 Bahia State, Brazil. B) MRR release points (numbered squares) and sampling locations (green circles) within
526 households distributed across the sampling grid at the study site.

527 **Figure 2. Dispersal kernels.** Examples of different kernel interpretations for the negative exponential (A, B and
528 C) and exponential power (D, E and F) kernels. The distance pdf is shown in panels A and D. The density with
529 respect to distance is shown in panels B and E and the density pdf is illustrated in panels C and F (after Cousens
530 *et al.* [25]). Kernels in A, D, C and F integrate to unity (in 1 dimension for the distance pdfs and 2 dimensions
531 for the density pdfs).

532 **Figure 3. Recaptures for three MRR experiments at the field site in Brazil.** Numbered squares represent the
533 two release points for MRR experiments. Coloured circles indicate the location and size of recaptures for three
534 separate MRR releases (insects marked with red, yellow and blue fluorescent powder).

535 **Figure 4. Dispersal kernel summary for the Brazil analysis.** A) Maximum likelihood estimate of density with
536 respect to distance for Brazil data. B) Maximum likelihood distance pdf. C) The log-likelihood surface with
537 respect to kernel parameters a and b , coloured points highlight the MLE (black, log-likelihood = -128.1) and
538 examples of extreme 95% CI (green, light blue, dark blue and mauve) kernel parameter combinations. The
539 dotted line demarks the 95% confidence interval contour. Solid black contour lines are at intervals of 10 log-
540 likelihood. Examples of D) distance densities and E) distance pdfs from the 95% confidence interval range
541 corresponding to the coloured points shown in panel C.

542 **Figure 5. Dispersal kernel summary for the Malaysian analysis.** A) Maximum likelihood estimate density with
543 respect to distance for Malaysian data. B) Maximum likelihood distance pdf. C) The log-likelihood surface with
544 respect to kernel parameters a and b , coloured points highlight the MLE (black, log-likelihood = -328) and
545 examples of extreme 95% CI (green, light blue, dark blue and mauve) kernel parameter combinations. The
546 dotted line demarks the 95% confidence interval contour. Solid black contour lines are at intervals of 10 log-
547 likelihood. Examples of D) distance densities and E) distance pdf from the 95% confidence interval range
548 corresponding to the coloured points shown in panel C.

549 **Figure 6. Dispersal kernel comparison.** A comparison of the A) distance pdf and B) density with respect to
550 distance for estimates using MRR data from Brazil (solid blue line) and Malaysia (dashed pink line). The
551 comparison highlights the similarity in estimated kernels for experiments conducted on different continents, in
552 different habitats.

553

554 **Tables**

555 **Table 1. Summary data of the three MRR experiments in Brazil.**

Release	Release date	Number released	Release point	Number (%) recaptured
Red	21-Feb-2011	5,349	1	22 (0.4)
Yellow	25-Feb-2011	8,350	1	17 (0.2)
Blue	21-Feb-2011	5,465	2	30 (0.5)
Total	-	19,164	-	69 (0.36)

556

557 **Table 2. Brazil model performance and kernel parameter estimates.**

Distance transformation	AIC	Explained variance (%)	Number of covariates	Kernel parameter estimates (95% CI)	
				<i>a</i>	<i>b</i>
Untransformed	293.5	46.7	5	-	-
Negative exponential	286.5	48.6	6	97.8 (57.8,238.3)	-
Exponential power	276.2	51.4	5	75.3 (66.1,85.0)	3.7 (2.0,7.3)

558 **Table 3 legend.** The estimated model performance (minimum AIC indicated in bold) and kernel parameters for different
 559 transformations of the distance explanatory variable for the Brazil analysis.

560

561 **Table 4. Brazil model coefficient estimates.**

Coefficient	Estimate	Standard error	z-value	p-value	
Intercept	-1.35	1.18	-1.16	0.25	
Transformed distance	83,140	7,015	11.85	<0.0001	
Number of days post release	-0.62	0.11	-5.84	<0.0001	
Wild other spp	0.021	0.0079	2.62	0.0088	
Maximum humidity	-0.035	0.016	-2.22	0.027	
Quadrant*	North	1	-	-	
	South	-1.88	0.80	-2.36	0.018
	East	0.76	0.37	2.07	0.039
	West	1.40	0.35	3.97	0.00071

562 *Overall significance level $p < 0.0001$ ($\chi^2 = 46.50$, 3df).

563 **Table 5 legend.** GLM coefficient estimates and associated standard errors, z-value and p-values from the optimal model for
 564 the Brazil analysis. Distance was transformed using the exponential power kernel.

565

566 **Table 6. Malaysia model performance and kernel parameter estimates.**

Model	AIC	Explained variance (%)	Number of covariates*	Kernel parameter estimates (95% CI)	
				<i>a</i>	<i>b</i>
Untransformed	728.7	7.9	3	-	-
Negative exponential	668.4	54.1	3	31.3 (27.7,34.9)	-
Exponential power	668.0	46.8	3	48.1 (45.3, 52.1)	1.4 (1.3,1.5)

567 *Including the interaction term

568 **Table 7 legend.** The estimated model performance (minimum AIC indicated in bold) and kernel parameters for different
 569 transformations of the distance explanatory variable for the Malaysia analysis.

570

571 **Table 8. Malaysia model coefficient estimates.**

Coefficient	Estimate	Standard error	t-value	p-value
Intercept	-0.013	0.79	-0.016	0.98
Transformed distance	96,300	17,500	5.50	<0.0001
Number of days post release	-0.46	0.22	-2.10	0.036
Interaction (Transformed distance × Days post release)	-24420	6467	-3.78	0.0002

572 **Table 9 legend.** GLM coefficient estimates and associated standard errors, t-values and p-values from the optimal model
 573 for the Malaysia analysis. Distance was transformed using the exponential power kernel.

574

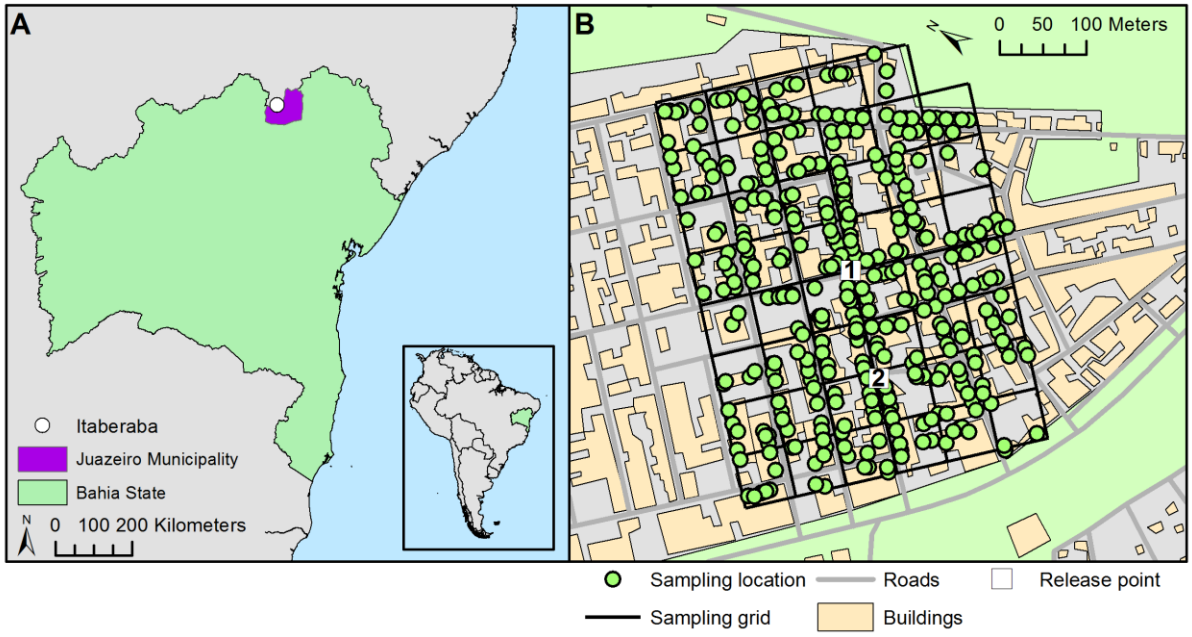
575 **Table 10. Summary of a literature review of male *Ae. aegypti* dispersal estimates.**

MDT/MDT range (m)	Location	Notes	Reference
10-30	Hainan Island, China	Released in the centre of a village	[8]
15-39	Sonepat, India	-	[9]
32	Ilha do Governador, Brazil	Raised on poor diet	[10]
35	Pentland, Australia	-	[11]
35-60	Hainan Island, China	Released at the edge of a village	[8]
42	Ilha do Governador, Brazil	Raised on rich diet	[10]
52	Jalan Tentera, Malaysia	Transgenic	[12]
100	Jalan Tentera, Malaysia	Laboratory strain	[12]

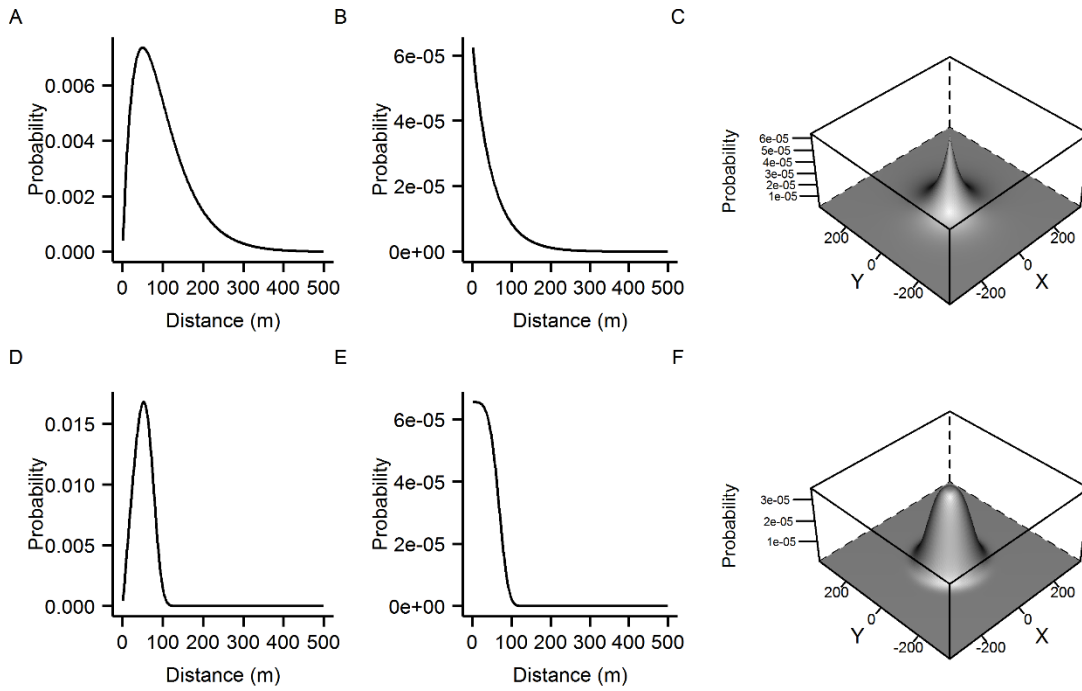
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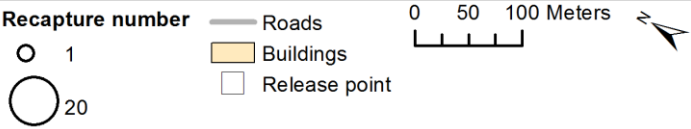
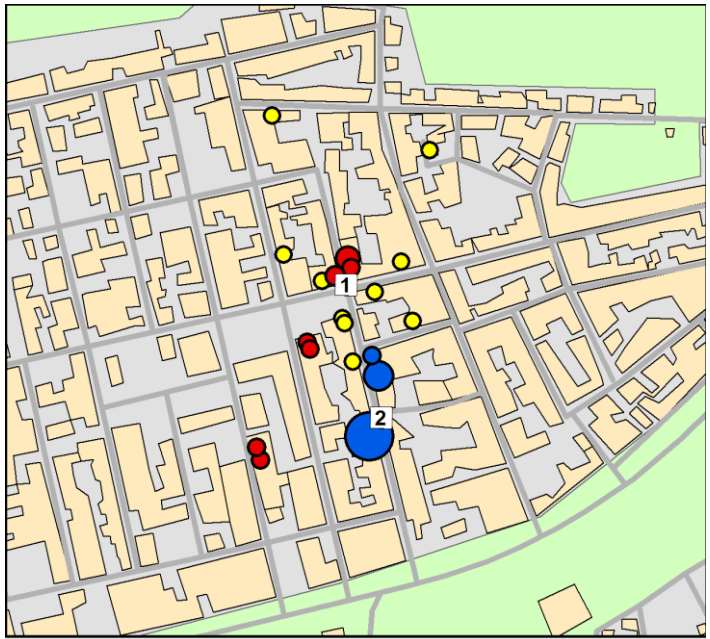
578 **Figures**



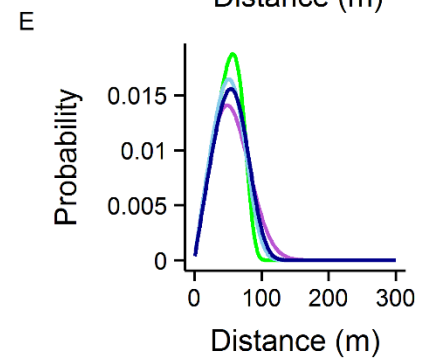
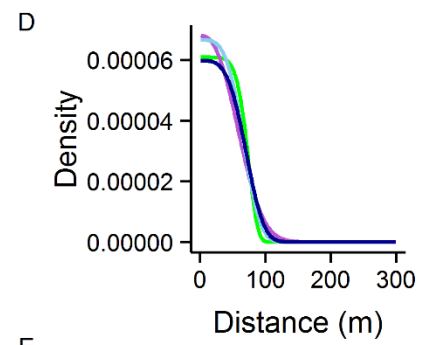
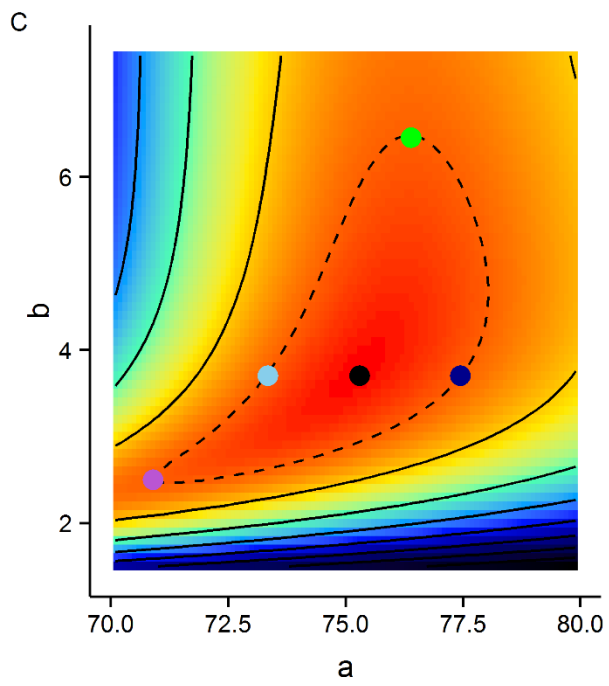
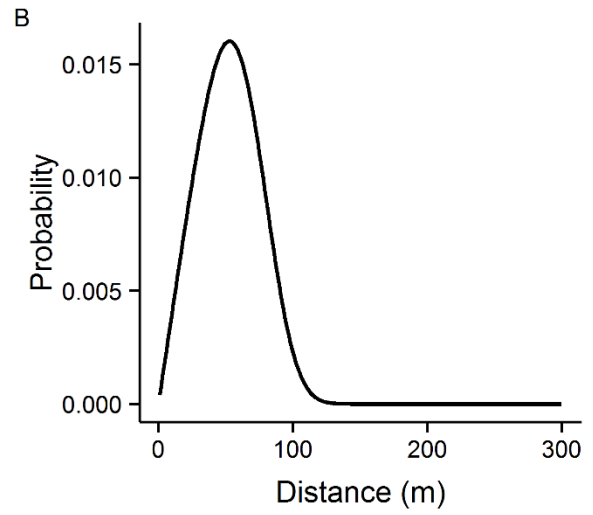
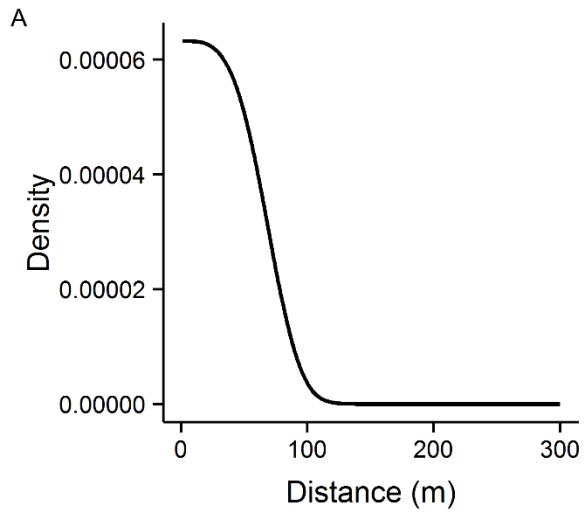
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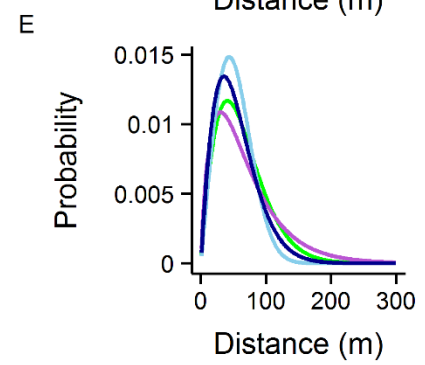
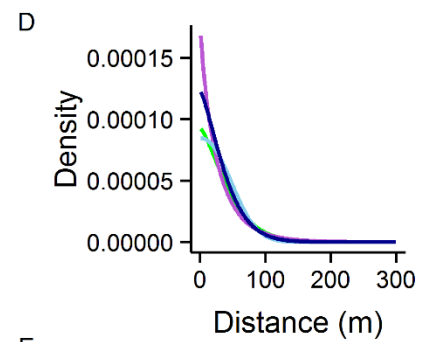
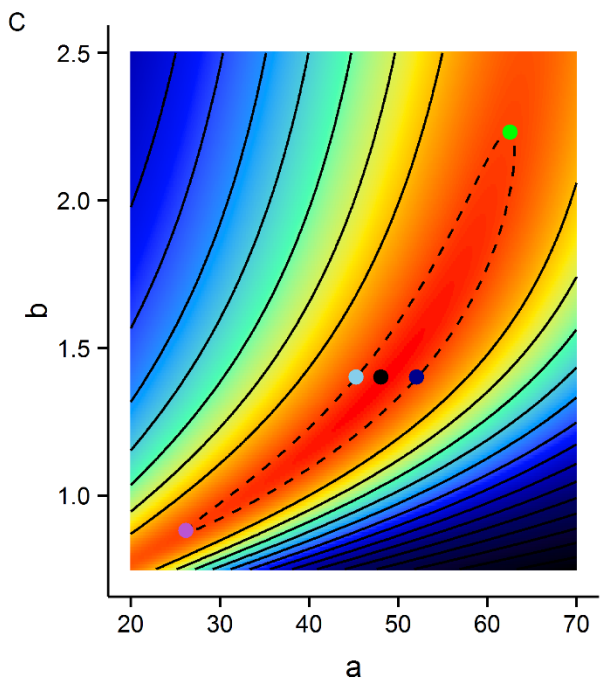
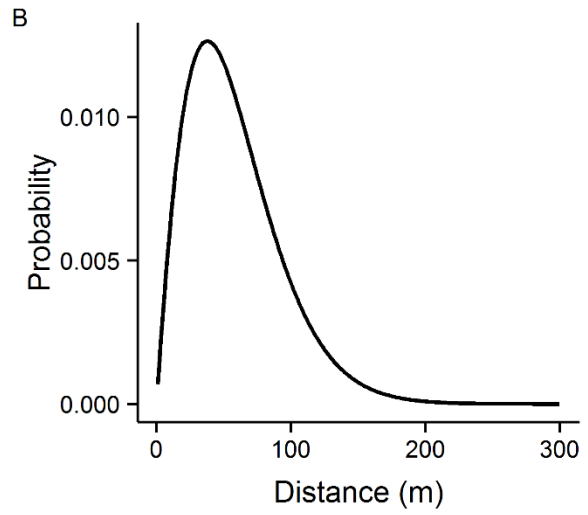
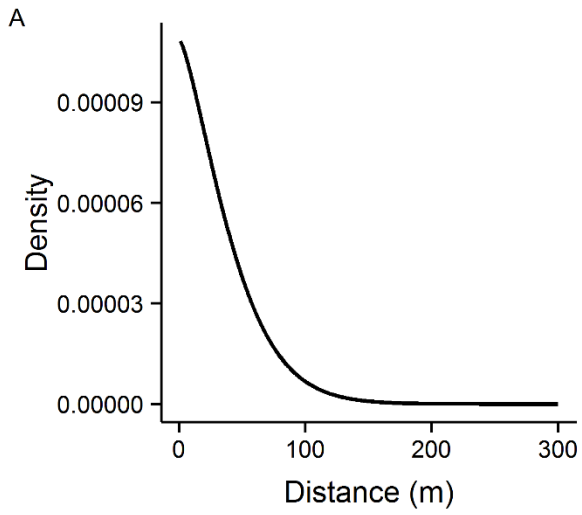
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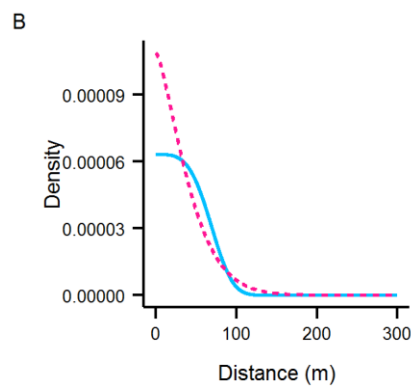
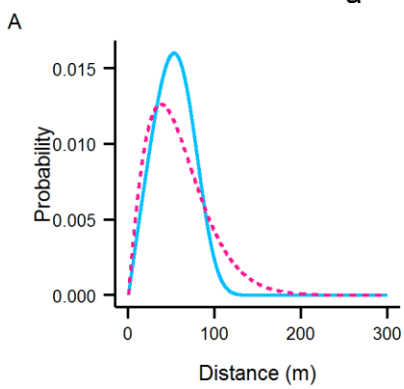
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Study

— Brazil

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