

G OPEN ACCESS

Citation: Walker M, Chandrasegaran K, Vinauger C, Robert MA, Childs LM (2021) Modeling the effects of *Aedes aegypti's* larval environment on adult body mass at emergence. PLoS Comput Biol 17(11): e1009102. https://doi.org/10.1371/journal. pcbi.1009102

Editor: Alex Perkins, University of Notre Dame, UNITED STATES

Received: May 10, 2021

Accepted: October 21, 2021

Published: November 22, 2021

Copyright: © 2021 Walker et al. This is an open access article distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

Data Availability Statement: All experimental data is available at https://data.mendeley.com/datasets/ 2ddbwns4gb/1 All code used for running the model is available on a GitHub repository at https:// github.com/melody289/Modeling-Ae-aegyptilarvae-mass-discrete.

Funding: CV was supported by the USDA, National Institute of Food and Agriculture, Hatch project #1017860. CV, KC, LC, and MW received support from the Virginia Polytechnic Institute and State University (Virginia Tech) via a Center for RESEARCH ARTICLE

Modeling the effects of *Aedes aegypti*'s larval environment on adult body mass at emergence

Melody Walker^{1,2}, Karthikeyan Chandrasegaran³, Clément Vinauger^{3,4}, Michael A. Robert⁵, Lauren M. Childs^{1,4}*

 Department of Mathematics, Virginia Tech, Blacksburg, Virginia, United States of America, 2 Current address: Laboratory for Systems Medicine, University of Florida Health, Gainesville, Florida, United States of America, 3 Department of Biochemistry, Virginia Tech, Blacksburg, Virginia, United States of America, 4 Center for Emerging Zoonotic and Arthropod-Borne Pathogens, Virginia Tech, Blacksburg, Virginia, United States of America, 5 Department of Mathematics and Applied Mathematics, Virginia Commonwealth University, Richmond, Virginia, United States of America

* Ichilds@vt.edu

Abstract

Mosquitoes vector harmful pathogens that infect millions of people every year, and developing approaches to effectively control mosquitoes is a topic of great interest. However, the success of many control measures is highly dependent upon ecological, physiological, and life history traits of mosquito species. The behavior of mosquitoes and their potential to vector pathogens can also be impacted by these traits. One trait of interest is mosquito body mass, which depends upon many factors associated with the environment in which juvenile mosquitoes develop. Our experiments examined the impact of larval density on the body mass of Aedes aegypti mosquitoes, which are important vectors of dengue, Zika, yellow fever, and other pathogens. To investigate the interactions between the larval environment and mosquito body mass, we built a discrete time mathematical model that incorporates body mass, larval density, and food availability and fit the model to our experimental data. We considered three categories of model complexity informed by data, and selected the best model within each category using Akaike's Information Criterion. We found that the larval environment is an important determinant of the body mass of mosquitoes upon emergence. Furthermore, we found that larval density has greater impact on body mass of adults at emergence than on development time, and that inclusion of density dependence in the survival of female aquatic stages in models is important. We discuss the implications of our results for the control of Aedes mosquitoes and on their potential to spread disease.

Author summary

In this work we examined how the environment in which juvenile mosquitoes develop affects their adult body size as measured by adult body mass. Adult size has potential impacts on mosquito behavior and the ability of mosquitoes to transmit disease. We used a combination of experimental work and mathematical modeling to determine important Emerging, Zoonotic, and Arthropod-borne Pathogens grant. LC and MW were supported by NSF Grant #1853495. The funders had no role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

Competing interests: The authors have declared that no competing interests exist.

factors affecting adult mosquito body size. In our model, we incorporated potentially interacting aspects of the mosquito life cycle and traits that affect mosquito growth as juveniles. These aspects include body mass, density of the population, and level of available resource. We compared different models to determine the one that best describes the data. As mass at emergence is linked to the success of adult mosquitoes to produce offspring and to their ability transmit pathogens, we discuss how important influences on development and survival of young mosquitoes affect mosquito control and disease spread.

1 Introduction

Mosquito-borne diseases pose a significant global health threat, impacting over 300 million people each year [1]. In this work, we consider populations of *Aedes aegypti* mosquitoes, vectors of the viruses that cause dengue, Zika, chikungunya, and yellow fever, among other diseases [2]. These mosquitoes are known to lay eggs in small, natural or man-made containers such as tires, buckets, and tree holes [3, 4]. Given the inherent constraints on resources of container habitats, larger larval populations can experience greater competition for space, food, and other resources than smaller populations [5, 6]. As a result, this often leads to morphological changes in emerging adult mosquitoes such as variation in adult body size [7]. Because body size is important to adult mosquito life history characteristics and a mosquito's potential as a vector of pathogens, it is important to develop a better understanding of the contributions of larval competition and density dependence in larval populations to adult body size. Herein, we utilize data collected from experiments aimed at characterizing relationships between larval environment and adult body size to develop a discrete time mathematical model that accounts for effects of density and resource availability on larval mass and, ultimately, on adult mosquito body mass. This mathematical formulation can be used to assess variability in mass as a function of larval environment.

While all mosquito species follow a similar life cycle, details of particular stages can be species specific. For example, mosquitoes such as *Aedes aegypti* and *Aedes albopictus*, upon acquiring sufficient reserves of blood, lay eggs on damp surfaces, in small holes, or on the inner walls of man-made containers (e.g. tires or rain barrels) [8]. A single egg batch typically ranges from approximately 50 to 100 eggs [9, 10]. Once the eggs are sufficiently hydrated, they hatch into larvae. The larvae progress through four stages, called instars, becoming larger in each stage, and ultimately molting into a pupa. During the pupal stage, the mosquitoes do not feed [11], and after a short time, adult mosquitoes emerge from the pupae. Development time from hatching of an egg to emergence as an adult depends on several factors including food availability, population density, and ambient temperature [12, 13]. This development time is therefore highly variable and can range from as short as 7 days to more than 90 days [13, 14].

The conditions experienced during each juvenile stage differentially affect outcomes later in the mosquito's life, such as age and mass at emergence. Of particular interest, mosquito body size is correlated with several other adult traits such as fecundity and longevity [9, 15–19]. Studies show a positive correlation with body size and fecundity, such as Briegel's work that found that female *Aedes aegypti* with larger body sizes had two to three times as many mature eggs as smaller mosquitoes [17]. Large males, too, have been associated with greater fecundity in females who mate with them versus those who mate with small males [20, 21]. In regards to survival, larger mosquitoes were shown to live longer on average [19, 22–24]. In a study on competition between *Aedes aegypti* and *Aedes albopictus*, body size was found to be a significant indicator of survival regardless of the competition levels [22].

Body size also affects the behavior of adult mosquitoes. Larger mosquitoes were shown to be more successful at blood feeding [25] and more persistent in acquiring a sufficient blood meal from a particular host [26]. In contrast, small mosquitoes are less persistent on a particular host and often do not obtain a complete blood meal in a single bite. Thus, smaller mosquitoes frequently bite more times and across more individual hosts [9, 26]. Additionally, large mosquitoes take larger blood meals [27]. These relationships between body size and feeding behavior can impact the level and spread of mosquito-borne disease. For example, Juliano et al. showed important relationships between body size and vector competence in Aedes aegypti, the primary vector of dengue [28]. In particular, field collected mosquitoes were more likely to be infected with dengue if they were larger in size [28]. However, other studies have found negative associations between adult size and viral metrics. Bara et al. observed that smaller mosquitoes showed greater dissemination rates of dengue compared to larger mosquitoes [29], and Alto et al. found increased infection and dissemination of DENV-2 infection under high competition between Ae. albopictus and Ae. aegypti, which led to lower survival, longer development time, and smaller adult females [30]. While these studies demonstrated relationships between body size and adult traits, other work has found no such relationship, e.g. [31, 32].

Despite the rich modeling literature on mosquito population dynamics, intricacies of the early stages of mosquito development are often ignored, and only a few modeling studies focus on mosquito body size in Aedes [33-35]. In 1979, Gilpin and McClelland developed one of the first models to consider how the mass of Aedes aegypti larvae changes due to food resources [34]. They formulated equations that dictated the amount of food that is converted into mosquito mass and assumed that larvae must reach a minimum weight and persist for a minimum time before they pupate. This framework is the foundation of several models that include resource-dependent juvenile population dynamics and impacts on the body size of Aedes mosquitoes [36, 37]. Padmanabha et. al. considered that reserves, i.e. stored energy that can be used for needs (excluding weight that is structural), were more important than weight itself [35]. They compared models with reserves-dependent growth to those with mass-dependent growth and showed reserves to be more important to modeling growth. In a more recent study, Romeo Aznar et al. compared the Gilpin-McClelland model with a compartment model [33], where they considered that development time of each biological stage was gamma distributed and the variance of the gamma distribution was food-dependent for some stages. They found that the mean development time changed with food availability. Although this model was adapted to study impacts of the sterile insect technique on a wild population [38], a comprehensive model that explains variability in adult sizes based on characteristics of larval environments remained to be developed.

In the present study, we collected experimental data to inform structure and parameter values of multiple possible models, and we compared fit across models to determine which model is the most appropriate. For any parameter not fit to data, we considered variations and sensitivity of the model to that parameter. Finally, we discuss the importance of our results for mosquito life history characteristics, pathogen transmission, and efforts to control mosquito populations.

2 Materials and methods

2.1 Data collection

Data was collected from a study of the mosquito species *Aedes aegypti*. We conducted two types of experiments. In the first, we tracked development through each aquatic stage under low and high larval density conditions. In the second, we tracked total length of development time and mass upon emergence under low and high larval density conditions.

We studied *Aedes aegypti* mosquitoes due to their medical relevance [39]. We used the Rockefeller strain (MR-734, MR4, ATCC, Manassas, VA, USA) due to the extensive literature on their behavior, which makes comparisons with other studies easier [40]. Larvae were reared in densities of 26 and 78 per 300 mL of nutrient medium to simulate 'low' and 'high' intraspecific competition, respectively. We refer to these settings as low and high density conditions or treatments throughout. Larval densities were chosen to vary per capita nutrition available in the two treatments. The chosen values are comparable with larval densities in natural habitats [41], and previous studies involving similar larval densities identified direct and indirect effects of intraspecific larval competition on larval and adult traits of mosquitoes, including body size [42]. Across the literature, wet and dry body mass, wing length and pupae mass are used as proxies for mosquito size, and the correlation between these morphometric measures are well described [43, 44]. Here, we opted for measurements of adult body mass given the high degree of precision provided by weighing mosquitoes on a micro-balance.

For all experiments, the stock larval nutrient medium was prepared at 3.3 mg/mL using standard fish food (Hikari Tropical First Bites, Petco, San Diego, CA, USA), incubated at 26°C for 24 hours, and used to prepare 12.5% stock dilution. Eggs were synchronously hatched in deionized (DI) water using a vacuum chamber and transferred to 300 mL of nutrient medium stock dilutions. Freshly hatched mosquito larvae in 12.5% dilution of the nutrient broth were housed in an incubator at 26±0.5°C and 70±10% relative humidity with a 14:10 hour day-night cycle. These rearing conditions represent ideal temperatures and relative humidity for mid-summer development, as shown in [45]. Mosquito larvae were monitored every 6 hours until pupation. Upon pupation, they were transferred to individually labeled vials containing water and monitored until emergence using a locomotor activity monitor (Trikinetics, LAM25). The activity monitor contains three infrared beams bisecting each vial just above the water level, repeatedly recognized by opposing infrared detectors. The LAM25 software recorded infrared beam breaks at repeated 60 second intervals to detect and record emerging mosquitoes. Modifying the locomotor activity monitor setup, which otherwise is generally used for tracking mosquito activity, allowed us to accurately quantify the development time of individual mosquitoes.

We recorded larval development times and proportion surviving through each molting event (four larval instars and pupa) and the metamorphosis into adults by monitoring the number of individuals in each stage across five replicates in low density and five replicates in high density. Each day, all larvae were identified for stage and counted. On the day when an individual emerged as an adult, it was sexed using morphological features such as the structure of the antennae [8].

To determine the body mass, we conducted a separate experiment on 13 replicates in low density and 8 replicates in high density. These different replicate numbers allowed us to obtain comparable final numbers of adults females despite differences in larval densities and ensured ample sample sizes while accounting for mortality, sex ratio, and controlling for "block effects" [42]. The mass of adult females, but not males, was recorded using a high precision analytical balance (\pm 0.0002 mg, Mettler Toledo, USA), resulting in measurements for a total of 133 and 198 adult females from low and high density conditions, respectively. Each adult female was weighed on emergence, and her total development time was recorded and measured in hours; however, as our model uses a time step of one day, we divided the development time by twenty-four and rounded up to the next day. Note that in this experiment, development time of each larval stage was not recorded.

We determined survival using all of the above replicates plus two additional replicates at low density. The two additional replicates were conducted similarly to the experiments in which we studied body mass, except mass was not measured (i.e. only the total number of male and female that emerged was recorded). Thus, the data to determine survival values consisted of 20 replicates in the low density treatment and 13 replicates in the high density treatment, resulting in a total of 520 and 1014 first instar larvae initially, of which 412 and 656 survived to emergence in low and high density conditions, respectively. For calculations of survival by sex, we assumed equal sex ratios initially, such that each replicate began with 50% males and 50% females. As individual replicates may vary from this assumption, we do observe calculated survival fractions above one (Fig 1C).



Fig 1. Experimental data on development time and survival. (A) Development time, in days, of mosquitoes across all experiments, categorized by (left) sex and (right) density treatment. (B) Development time by mass group and density treatment. Box plots of each mass group emerging under (left) low density and (right) high density treatments. This includes mass data, but excludes stage data. In (A)-(B) red lines represent the median of the data; the blue box indicates the upper and lower limits of the interquartile range (IQR); whiskers indicate 2^* IQR; and red plus signs represent outliers. (C) The proportion of mosquitoes which survive by density treatment and sex for each replicate of the low (left) and high (middle) density treatments. The mean and median survival proportion for all experiments by density (right). In (C), blue triangles represent females and red triangles males. The experiments are sorted by female survival proportion, where we assume 50% males and 50% females initially. As individual replicates may differ in their initial sex percentages, some calculated survival proportions are greater than one. (D) Female mosquito emergence by body mass group: < 1.5 mg (small) in blue, 1.5–2.5 mg (medium) in green and > 2.5 mg (large) in red. The top row is the low density treatment and the bottom row is high density.

https://doi.org/10.1371/journal.pcbi.1009102.g001

2.1.1 Key features of data. We aim for our model to reproduce key features observed in our experimental data. In particular, the model should differentiate the development time of individuals based on sex and body size in different larval environments. In Fig 1A, we present data on development time aggregated by sex and density treatment. On average, males emerge slightly before females, and individuals in the high density treatment exhibited greater variance in development time. Using a Welch *t*-test, we found a significant difference in development time for males and females (df = 192.92, *p*-value = $8.252e^{-12}$ for low density and df = 300.53, *p*-value < $2.2e^{-16}$ for high density). When we considered experiments by sex and density treatment (Fig 1C), we noticed that density impacts female survival: the average female survival in the high density treatment is approximately two-thirds of the low density treatment. Males showed a similar trend but to a lesser extent. With a *t*-test comparing between density treatments for each sex, we found that the difference in low and high density is statistically significant for females but not for males (df = 30.131, *p*-value = 0.001364 for females and df = 28.35, *p*-value = 0.1196 for males).

In our analyses, we considered three general mass groups: small (less than 1.5 mg), medium (between 1.5 and 2.5 mg), and large (greater than 2.5 mg). These three groups were chosen by dividing the data from mass at emergence, which ranged from 0.69 to 3.35 mg, into approximately equal thirds. There was a noticeable difference in the proportion of each mass group emerging based on the environmental conditions (Table 1). In both density conditions, food was only provided at the beginning of the experiment. In the low density treatment, this food was sufficient for all larvae over the time span required for development. Thus, the majority of mosquitoes that emerged were of the large mass group and no mosquitoes emerged in the small mass group. In the high density treatment, the amount of food was likely insufficient for successful development of all mosquitoes. As a result, none of the mosquitoes emerged in the large mass group and the majority were in the small mass group.

In Fig 1B, we show data on development time by mass group for each of the density treatments. There was a significant difference in the development time of the small mass group compared to the medium group in the low density treatment (p = 0.012, n = 158). No differences between mass groups were seen in the high density treatment. The data did not meet the normality assumptions necessary for ANOVA; thus, we performed a Kruskal-Wallis and Dunn test to determine if, and which groups, were different. Although the small mass group was significantly different, the Kruskal-Wallis test effect size was small (0.046), so we did not consider a variation based on this difference.

2.2 Model

We developed a discrete time model of an *Aedes* mosquito population which incorporates each biological stage of the mosquito's life cycle: four larval instars, pupa, and adult. In our model, we included effects of larval density and resources on development time and growth (as measured by mass). To that end, we divided each of the stages further by mass. Our determination of the mass range for each of the larval and pupal stages is as described in section

Table 1. Mosquito emergence by body mass in low and high density conditions. The number (percentage) of mosquitoes in each body mass group: <1.5 mg (small), 1.5–2.5 mg (medium), and >2.5 mg (large) divided by environmental conditions: low density and high density. A histogram of masses by density treatment is found in Fig 1D.

			-
Environment	Small	Medium	Large
Low	0 (0%)	38 (28%)	95 (72%)
High	150 (76%)	48 (24%)	0 (0%)

https://doi.org/10.1371/journal.pcbi.1009102.t001

2.1.1. The time step for our model is one day. The system of equations and a detailed description of the model are given in <u>S1 File</u>.

We assume all individuals begin in the first instar larvae to be consistent with our data which does not include eggs. Individuals develop through successive larval stages. The proportion of juvenile mosquitoes that develops each day is governed by the function f(N). We assume that development of juvenile mosquitoes is density-dependent [46, 47], and let the density in the model equal the current larval population (N). See details on f(N) in section 2.2.2. At each transition, a proportion of individuals die. See details on mortality $\mu(N)$ in section 2.2.3. During the larval stages, mosquitoes may remain in their current mass group or transition to a higher mass group. We assume that mosquitoes changing mass groups can only transition to the next highest mass group in a single time step (i.e., a small mosquito cannot become a large mosquito in a single day). We note that once mosquitoes enter the large mass group, they cannot grow larger in mass. During the pupal stage individuals do not eat, so they remain in their current mass group. A diagram depicting our model is shown in Fig 2A. At time step t, the growth functions, G_i , determining transition to a higher mass group depend on resource (r) and larval density (N). The proportion of individuals that grow from small to medium mass is given by $G_1(r/N)$, and from medium to large mass by $G_2(r/N)$. See details on $G_i(r/N)$ in section 2.2.4. We do not track adult populations, but only record the emergence of adults by sex and mass group through time.

2.2.1 Resources. We incorporate the amount of available resources (i.e. food) to larvae in our functions for growth and development. We assume growth is a function of food per larvae, $\frac{r}{N}$, where *r* is the resource and *N* is the current total larvae. In the model, we assume that there is a fixed maximum amount of food necessary for each larvae and any excess food does not help or hinder larval growth or development, although recent evidence suggests that continuous exposure to high food levels is associated with lower survival and smaller adults (as compared to continuous exposure to optimal food levels) [48]. In our model, the amount of food (*r*) that is sufficient for a single larva occurs when *r* = 1. If there are *N* larvae, *r* = *N* is the necessary amount of food for all larvae. However, we assume that when there are lower levels of



(A) Schematic of model

Fig 2. Model formulation and variations. (A) Schematic of our mathematical model describing stage and mass. As an individual moves horizontally (age axis), it advances to a later biological stage from larval stages 1 to 4 to pupae to adult. As an individual moves vertically (mass axis), it grows larger in mass. (B) Flow chart of each variation considered. The first variation (top row) is in regards to the number of compartments. We include 5, 6, or 7 total compartments, in C1, C2, and C3 respectively. The second variation is in mortality: single constant death (D1), two constant deaths for each of male and female (D2), and a density-dependent death for females and constant death for males (D3). The third variation is the inclusion of a growth function, E. The darker boxes are the version of each variation chosen as best based on the AIC. See section 2.4 for description of the fitting.

https://doi.org/10.1371/journal.pcbi.1009102.g002

food, for example, as the result of decreasing resources, larvae experience slower growth. This is consistent with [48], where development time lengthened with lower food availability.

We model resources, such that they are used up over the course of the day dependent upon the number of larvae at fixed rate *q*. Thus the available resources on day *t* are given by

$$r(t) = r(t-1)e^{-qN(t-1)},$$
(1)

where N(t) is the total larvae at time t, and q is the per capita rate of resource usage. The value for q, fixed to 0.01, was chosen such that the resources available in the low density treatment would be sufficient for all larvae while the resources available in the high density treatment would run out before all larvae pupate. In section 3.4.1, we discuss our choice for q in more detail and include a sensitivity analysis.

2.2.2 Development time. In previous work, we showed that development time is affected by density [47], which was also noted by others [46]. To model density-dependent development time, we incorporate the function f(N), which is the proportion of individuals that develop to the next life stage. For f(N), we use the Maynard-Smith-Slatkin density function, which is the best fit formulation from [47], and is given by

$$f(N) = \frac{k}{1 + (aN)^b}$$

where *k* is the maximum development rate, *N* is the total current larvae, and *a* and *b* scale the importance of density. We use the parameter choice a = 0.0043 and b = 1.61 found in previous work [47]. Here, we employ different f(N) for males and females and fit two different values for the maximum development proportion, *k*: one for males, denoted k_m , and the other for females, denoted k_f .

2.2.3 Death proportion. A proportion of individuals from each compartment die each day. In an initial pass, the death proportion for each day is taken directly from the experimental data. In this case, we compare each replicate separately and use the proportion of individuals that die each day from that specific replicate. We use the daily death proportions by replicate to fit the development time. After we fit the daily maximum development proportion (k_m , k_f) see section 2.4.1), we combine data on mortality from all replicates and fit a single constant for the death proportion.

We then consider a density-dependent death function for females. In the experiments, we observed that females had greater survival in low density compared to high density (Fig 1C). While males showed a similar trend as females, the difference observed between low and high density treatments was not statistically significant. Thus, we do not consider density-dependent mortality for males. To incorporate density dependence into our death function for females, we use a Hill function, given by

$$\mu(N) = \frac{\mu_f(N^n + h)}{N^n + f^n},$$

where n = 3 is the Hill exponent, h and f are constants such that h < f and $\frac{\mu_f h}{f^3}$ is the minimum proportion of individuals that die and μ_f is the maximum proportion. The inclusion of the h in the numerator, a departure from a traditional Hill function, allows the lower bound of the function to be greater than zero. If h equals zero, this returns the traditional Hill function. We fix h = 100, so that the lower bound is $\frac{100\mu_f}{f^3}$. See section 3.4.4 for more discussion and a sensitivity analysis of f.

2.2.4 Growth. The growth functions determine the proportion of mosquitoes that move from one mass group to the next largest mass group. We fit to the functional form for G_1 and G_2 which are Holling type III forms, a standard functional form in ecology, where growth is maximized at intermediate values [49, 50]. These are sigmoidal functions that rise from near zero, when r/N is small, to a maximum of g_i , when r/N is large. The steepness of the curve is governed by the magnitude of n_i and the point of inflection of the curve occurs at c_i , each of which are defined below. At each time step, the function G_1 determines the proportion of individuals that grow from small to medium mass

$$G_1\left(\frac{r}{N}\right) = \frac{g_1\left(\frac{r}{N}\right)^{n_1}}{\left(\frac{r}{N}\right)^{n_1} + c_1^{n_1},}$$

and G_2 the proportion that grow from medium to large mass

$$G_2\left(\frac{r}{N}\right) = \frac{g_2\left(\frac{r}{N}\right)^{n_2}}{\left(\frac{r}{N}\right)^{n_2} + c_2^{n_2}},$$

where n_1 and n_2 are Hill coefficients; c_1 and c_2 are the value where half maximal growth occurs; and g_1 and g_2 are the maximum growth proportions. G_1 and G_2 inherently depend on time as rand N change with time. Note that for implementations of the model that do not involve mass (sections 2.4.1–2.4.2), we assume the growth functions are zero. In such cases, nothing depends on mass so there is no need to track the mass group of mosquitoes in a particular life stage.

The proportion which grows from one mass group into a higher mass group is reduced when $\frac{r}{N} < 1$. Furthermore, in the absence of food (r = 0), the proportion which grows is defined to be zero. If the amount of food is well above that necessary for the current number of larvae, i.e. $\frac{r}{N} > 1$, the proportion which grows does not exceed $g_i \le 1$. As r = 1 is sufficient food for one larva, we expect the growth function to reach the maximum approximately when $\frac{r}{N} = 1$. Thus, we assume that half maximal growth occurs when $\frac{r}{N} = 0.5$, i.e. $c_1 = c_2 = 0.5$. See section 3.4.3 for more discussion and a sensitivity analysis on c_1 and c_2 .

The parameters g_1 , g_2 , n_1 and n_2 are determined to be consistent with the resulting mass difference observed under low and high density treatments. We assume that growth to the large mass group is more harshly affected by lack of resources than growth to the medium mass group. Furthermore, we consider different growth functions by sex, such that maximal growth proportion differs for males and females. Specifically, we assume that a lower proportion of males grow, and let maximum male growth be vg_1 and vg_2 where $0 < v \le 1$.

2.3 Model variations

In our base model, we assume a single constant daily death proportion for both males and females (μ), but have different daily maximum development proportions for males (k_m) and females (k_f). As we are not initially incorporating data on mass, we set maximum growth proportions for both males and females to zero, $g_i = 0$.

In our analysis, we consider three sets of variations to the base model: variations in the number of compartments (denoted by C); variations in death proportions (denoted by D); and inclusion of the growth function (denoted by E). A diagram of the variations is shown in Fig 2B.

2.3.1 Compartment variations: C1-C3. First, we altered the total number of compartments, which resulted in three different variations of the model. To begin, we allow for each aquatic stage to encompass only a single compartment in the version denoted C1. We then add a second compartment to the pupae stage, so that there are two compartments for each sex and body size of pupae in the version denoted C2. Finally, we split apart both pupae and L4 into two compartments, instead of a single compartment for each, in the version denoted C3. The addition of these compartments forces the minimum time in L4 and pupae to be longer as well as the development time to be longer, but is not meant to suggest additional biological stages. The choice of including the compartments in L4 and pupae is motivated by observations on the time to emergence in our experiments (Fig 1C). It is important to note that the additional compartments only extend the development time but do not impact growth as no growth can happen during transition between the first and second sub-compartments within L4 or during the pupal stages.

Finally, as detailed in section 2.4.1, we compare the results obtained from the three variations and choose the variation that best fits larval timing and adult emergence. Once we determine the optimal number of compartments for the model structure, we use this as the starting point to consider variations in how we model death.

2.3.2 Death variations: D1-D3. Next, we consider different ways to incorporate death by studying three variations with different assumptions on mortality. The death proportion is initially a single constant value for both males and females in the version denoted D1. The second version, denoted D2, uses a different constant proportion for each sex. Finally, we consider a density-dependent death function for females, but not males, in the version denoted D3. We do not include a density-dependent death function for males as the data did not support differences for the males by density treatment. We compare all three versions of the incorporation of mortality as described in section 2.4.2. We use the best fitting model to consider inclusion of the growth functions.

2.3.3 Growth function included: E. After determining the number of compartments and the form for the death function, we fit the two growth functions, G_1 and G_2 . At each time step, these functions determine what proportion grows from small to medium mass and from medium to large mass, respectively. Details of the growth functions are found in section 2.2.4. We denote this as variation E.

2.4 Fitting parameters

We describe our fitting for each variation of the model. Throughout, we use \mathbb{M} to refer to model output and \mathbb{D} for experimental data. We list all parameters with their description, the standard value, and variations considered in <u>Table 2</u>. The values of fitted parameters for each variation are found in Table A in <u>S1 File</u>.

2.4.1 Estimating development time: Variations C1-C3. In our compartment variations, we ultimately fitted three parameters $(k_m, k_f, \mu(N) = \mu)$ using a two step process. We began by simultaneously fitting the maximum development proportion for males (k_m) and females (k_f) . In order to separate effects of density on death and on development time, we calculated the daily death proportion directly from each replicate by dividing the number of individuals that died on the previous day by the total number of larvae and pupae present on the previous day.

We fit the maximum development proportion for males and females by minimizing a summed squared error of the difference in total larvae time and the total emergence of adults from the five replicates in high density and five replicates in low density where we monitored the development time at each stage.

Table 2. Model parameters. For each parameter, we include its representative symbol, a description, the standard value used, and the range of values considered. For parameters with a value given, we used that value during the fitting process, but performed univariate sensitivity. For parameters marked fitted, the range listed under variation is the constraint when fitting using MatLab function fmincon.

Symbol	Description		Variation
9	Per capita usage of resources		0.0005-0.05
Ь	Exponent on density in development time		0.001, 0.01
а	Coefficient on density in development time		1, 2
k_m	Maximum male daily development proportion		0-1
k_f	Maximum female daily development proportion		0-1
μ_m	Maximum male daily death proportion	Fitted	0-0.5
μ_f	Maximum female daily death proportion	Fitted	0-0.5
f	Female density-dependent death parameter	Fitted	1-100
h	Constant in density-dependent death	100	1-2000
n	Exponent in female density-dependent death	3	1-10
g_1	Maximum daily growth proportion from small to medium	Fitted	0-1
<i>g</i> ₂	Maximum daily growth proportion from medium to large	Fitted	0-1
n_1	Exponent of growth function from small to medium	Fitted	0.5-12
<i>n</i> ₂	Exponent of growth function from medium to large		0.5-12
c_1	Per capita resource for half maximal growth for small to medium		0.3, 0.5, 0.69
<i>c</i> ₂	Per capita resource for half maximal growth for medium to large		0.3, 0.5, 0.69
ν	Male proportional growth	0.52	0.1-1

https://doi.org/10.1371/journal.pcbi.1009102.t002

Let the $\hat{L}(t)$ be the vector of the sum of each aquatic stage (across mass groups) at each time t given by

$$\hat{L}(t) = [L1(t), L2(t), L3(t), L4(t), P(t)]$$

Let the total sum of the aquatic stages from a given model simulation be $\mathbb{M}_{(\hat{L},j)}$, where $j \in \{L, H\}$ represents low and high model density treatments. The first subscript of $\mathbb{M}_{(\hat{L},j)}$ indicates element-wise sum over $\hat{L}(t)$.

Let the total number of emerged adult females from a given model simulation be $\mathbb{M}_{(F,j)}$ for the *j*th treatment. The total number of emerged male mosquitoes are represented similarly by $\mathbb{M}_{(M,j)}$. Thus, model output for emerged adults is given by

$$\mathbb{M}_{(F,j)} = \sum_{t} (F_s(t) + F_m(t) + F_l(t)),$$

 $\mathbb{M}_{(M,j)} = \sum_{t} (M_s(t) + M_m(t) + M_l(t)).$

where $M_i(t)$ and $F_i(t)$ for $i \in \{s, m, l\}$ are the number of emerging males and females, respectively, of a given mass group at time t. While \hat{L} is a vector, $\mathbb{M}_{(F,j)}$ and $\mathbb{M}_{(M,j)}$ are each scalars. The data from a particular replicate j is represented similarly but with \mathbb{D} rather than \mathbb{M} .

We fit the development proportions by finding the square of the difference between model output and data, weighting each term, and then summing across all replicates. Specifically, our error formula is given by

$$\mathcal{E}_{C} = \sum_{j=1}^{10} \left(\left(\frac{\mathbb{M}_{(\hat{L},j)} - \mathbb{D}_{(\hat{L},j)}}{\mathbb{D}_{(\hat{L},j)}} \right)^{2} + \left(-\frac{\mathbb{M}_{(F,j)} - \mathbb{D}_{(F,j)}}{4} \right)^{2} + \left(\frac{\mathbb{M}_{(M,j)} - \mathbb{D}_{(M,j)}}{4} \right)^{2} \right)$$

where the division and the square occur element-wise in the first term. We chose the weights in the error formula so that the data on the emergence of adults (second and third term) has more weight than data on time spent in the biological stages (first term). We placed more weight on the adult results as sex is not separable in the data until the adult stage.

To find the minimum error \mathcal{E}_C , we used fmincon in MatLab allowing both k_m and k_f to be constrained between 0 and 1. We choose a nine by nine grid (values between 0.1 and 0.9 incremented by 0.1 for each of k_m and k_f) of initial starting points. Once we determined optimal k_m and k_f in our first step, we used these values in estimating a single daily death proportion, $\mu(N) = \mu^*$, for all replicates. In this case, the only difference between model output from high (*H*) and low (*L*) density treatments is the initial number of mosquitoes, i.e. our initial condition. We found a constant $\mu(N) = \mu^*$ that minimizes the death error, \mathcal{E}_D , found in section 2.4.2. We used equal weighting for low and high density with 5 replicates for each condition.

Overall, we fit three parameters: k_m , k_f , and μ^* . In the first step, we fit k_m and k_f simultaneously. Then in the second step, we used these values when we estimate μ^* .

2.4.2 Estimating daily death proportions: Variation D1-D3. For our death variations (D1-D3), we fitted parameters related to death proportion (μ^* , μ_β , μ_m , f) and calculated the total number of males and females that emerged. The data consists of 13 replicates in high density and 20 replicates in low density. For these replicates, we used the total number of males and females that emerged as adults.

Let the total number of emerged adult females from the model in low density be $\mathbb{M}_{(F,L)}$ and in high density as $\mathbb{M}_{(F,H)}$. Similarly, the total number of emerged adult males in low and high density is $\mathbb{M}_{(M,H)}$ and $\mathbb{M}_{(M,H)}$, respectively. Thus, our model output is given by

$$\begin{split} \mathbb{M}_{(F,i)} &= \sum_{t} (F_{s}(t) + F_{m}(t) + F_{l}(t)), \\ \mathbb{M}_{(M,i)} &= \sum_{t} (M_{s}(t) + M_{m}(t) + M_{l}(t)), \end{split}$$

where $i \in \{L, H\}$. Similarly, let the total number of females from the *k*th replicate in low density be given as $\mathbb{D}_{(F,L_k)}$ and from *j*th replicate in high density as $\mathbb{D}_{(F,H_i)}$.

We determined the sum of the squared difference between data and model output of the total number of emerged males and total number of emerged females in each density treatment, as given by

$$\begin{split} \mathcal{E}_{D} &= \sum_{k}^{20} \left((\mathbb{M}_{(F,L)} - \mathbb{D}_{(F,L_{k})})^{2} + (\mathbb{M}_{(M,L)} - \mathbb{D}_{(M,L_{k})})^{2} \right) \\ &+ \frac{1}{2} \sum_{j}^{13} \left((\mathbb{M}_{(F,H)} - \mathbb{D}_{(F,H_{j})})^{2} + (\mathbb{M}_{(M,H)} - \mathbb{D}_{(M,H_{j})})^{2} \right). \end{split}$$

Note that we scaled the sum for the high density treatments by one-half as we aimed for approximately equal weighting for data from both high and low density treatments. Recall there are a total of 20 replicates in low density, each assumed to start with 13 individuals of each sex, and 13 replicates in high density, each assumed to start with 39 individuals of each sex. Thus, there are approximately twice as many total larvae across all high density replicates, so we scale the sum by one-half.

In order to minimize our error \mathcal{E}_D , we used MatLab function fmincon on a range of initial values for the parameters. The precise parameters that we fitted was dependent on the death proportion variation considered. In variation D1, we fitted a single death constant ($\mu(N) = \mu^*$).

In D2, we fitted two death constants, one for males $(\mu(N) = \mu_m)$ and one for females $(\mu(N) = \mu_f)$. Finally, in D3, we fitted two parameters (μ_f, f) for a density-dependent death function for the females $\left(\mu(N) = \mu_f \frac{N^3 + h}{N^3 + f^3}\right)$ and a single constant death $(\mu(N) = \mu_m)$ for the males. See section 2.2.3 for more details on the density-dependent death function. In our fitting, the values $\mu_f \mu_m$, and μ^* are constrained between 0.001 and 0.5, and *f* is constrained between 1 and 100.

2.4.3 Estimating growth function parameters: Variation E. In variation E, we investigated mass-dependent growth. As previously noted, we split individuals by body mass into three groups: less then 1.5 mg (small), between 1.5 and 2.5 mg (medium), and greater than 2.5 mg (large). To determine the growth function, we fitted the parameters n_1 , g_1 , n_2 , and g_2 as described in section 2.2.4 and fixed the resource usage rate to q = 0.01 as described in section 2.2.1. The latter resulted in complete resource usages in the high density treatment prior to the emergence of all individuals as adults, but allowed resources to remain in the low density treatment even after all individuals emerged.

We have less data on mass size for males. Thus, we first found the average development time for females and males separately. Then, we determined v = 0.52 to be the relative proportion that males grow compared to females, by dividing the male average with the female average. See section 3.4.2 for more discussion on v and a sensitivity analysis.

The data is comprised of 13 replicates in low density and 8 replicates in high density. For each replicate, we used the total number of emerged adult females in each mass group. Recall that $F_s(t)$ ($F_m(t)$, $F_l(t)$) are the small (medium, large) females that emerge as adults at time t. Let the total number of small females that emerged in the kth replicate be given as $\mathbb{D}_{(F_s,k)}$ (similarly, $\mathbb{D}_{(F_m,k)}$ and $\mathbb{D}_{(F_l,k)}$ for medium and large, respectively). The model output for the total number of small females emerged is given by $\mathbb{M}_{(F_s,j)}$ (similarly $\mathbb{M}_{(F_m,j)}$ and $\mathbb{M}_{(F_l,j)}$ for medium and large, respectively) where j = L for low density and j = H for high density. In order to fit the parameters for the growth functions, $G_1(r/N)$ and $G_2(r/N)$, we minimized the squared difference between the proportion of each mass group (small, medium, and large) in the data compared to the model run. The error function is given by

$$\mathcal{E}_{\scriptscriptstyle E} = \sum_{k=1}^{21} \left(\frac{\mathbb{M}_{\scriptscriptstyle (F_s j)}}{\mathbb{M}_{\scriptscriptstyle (F,j)}} - \frac{\mathbb{D}_{\scriptscriptstyle (F_s k)}}{\mathbb{D}_{\scriptscriptstyle (F,k)}} \right)^2 + \sum_{k=1}^{21} \left(\frac{\mathbb{M}_{\scriptscriptstyle (F_m j)}}{\mathbb{D}_{\scriptscriptstyle (F,j)}} - \frac{\mathbb{D}_{\scriptscriptstyle (F_m k)}}{\mathbb{D}_{\scriptscriptstyle (F,k)}} \right)^2 + \sum_{k=1}^{21} \left(\frac{\mathbb{M}_{\scriptscriptstyle (F_l j)}}{\mathbb{D}_{\scriptscriptstyle (F,j)}} - \frac{\mathbb{D}_{\scriptscriptstyle (F_l k)}}{\mathbb{D}_{\scriptscriptstyle (F,k)}} \right)^2.$$

Note that the value for *j* in the subscript of \mathbb{M} is determined by the specific data replicate considered: if it is low density, then *j* = *L*, and if it is high density, then *j* = *H*.

In order to find the minimum error \mathcal{E}_E , we used the function fmincon in MatLab with over 150 initial value choices. We constrain g_1 and g_2 between 0 and 1, and n_1 and n_2 between 0.5 and 12.

2.5 Model comparison

We have three different sets of variations of the model we fitted: C1-C3, D1-D3, E (Fig 2B). We compared each version within its variation group (row in Fig 2B). We used the Akaike information criterion (AIC), a common metric to compare between models, to choose the best-fitting model. The AIC is a relative measure of the maximum likelihood of a model that deducts for the complexity of the model by reducing by the number of parameters needed. We assumed errors are normally distributed and used different versions of least squares error from the data (ϵ_i) described above. Our model structure meets the criteria for AIC: the same outcome variables are measured, the same data is used to compare, and the sample is sufficiently large [51]. As the sample sizes in our data were relatively small, we used a second order

biased correction, AIC_c. The AIC_c we employed is given by

$$\operatorname{AIC}_{c} = n \log\left(\frac{\epsilon_{i}}{n}\right) + 2K + \frac{2K(K+1)}{n-K-1},$$

where *K* is the number of parameters in the model, ϵ_i is the error as described above, and *n* is the sample size of the data [52].

For the first variations (C1-C3), we varied the number of compartments, but the number of parameters being fitted did not change. Since there was no complexity difference in the number of parameters, the AIC_c simply compared the errors without any difference due to parameters. After determining the maximum male and female development proportions, these parameters are fixed for the next set of variations.

Next, we determined the best representation of death (D1-D3). The number of parameters in these versions varies from one in the simplest case to three in the most complex case. Using the AIC_c we determined which version is best and used that version when we estimated parameters in the growth function.

Finally, in variation E, we fitted four parameters in the growth functions. We only considered a single mathematical formulation of the growth functions as mass-dependent growth of larvae is a poorly understood process. Furthermore, our growth functions do not mechanistically describe growth but only represent the proportion that emerge as a larger mass, using a standard Holling type III formulation.

3 Results

3.1 Development time fit: Variation C1-C3

We observed that the predicted adult emergence in the model generally occurred at similar times to that in the data. The model, however, spreads emergence across multiple days compared to the data, which tends to sharply peak across one or two days. This results in the model output showing lower peaks on any given day. In the absence of extra compartments in L4 and pupae, the model results show the timing of emergence of both males and females in either density treatment begins two to three days earlier in the model than observed in the data (Fig 3A, dashed dotted green line). Adding in a second pupal compartment reduced this difference to one to two days (Fig 3A, dashed maroon line). Finally, adding in a second compartment to the L4 stage in addition to the second compartment for the pupal stage further reduced this difference (Fig 3A, solid blue line). Emergence in high density started on the same day in the model output and in the data, and emergence in the model output under low density started one day earlier. In all versions of the model, the model predicted emergence spread across a wider range of days. In particular, the model suggests a longer tail of later times of emergence, i.e. longer development time, than in the data Fig 3A).

As we employ a constant death rate (rather than time varying) in the model, we expect slightly more males than females as they develop faster. Even with a shorter development time for males, we found that the model predicts that the number of males and females emerging are nearly identical when assuming identical death rates. Experimentally, in the high density treatment 17–35 (mean 28.2) males and 16–25 (mean 20.2) females emerged. In contrast, in the model version approximately 26 males and 25 females emerged. In the low density treatment, 8–13 (mean 10) males and 7–13 (mean 10.4) females emerged in the experiments. While in the model, approximately 9 females and 9 males emerged. The model results are approximate as fractions of individuals can be represented so we round to the closest whole number.

Using the AIC_c value to compare the model output with the data on the development time, we found that all versions of the model are remarkably similar. We found AIC_c values of 22.9,



(A) Timing of emergence by sex and density treatment under different compartment versions, C1-C3.



(B) Timing of each developmental stage with version C3.

Fig 3. Fitting results for compartment versions, C1-C3. (A) Data and model output in the high density (top row) and low density (bottom row) treatments. The results are split by sex: males (left column) and females (right column). Each panel shows the number of adults that emerged on a given day. The gray lines each represent an individual experimental replicate. For the model, the only difference between low and high density treatments is that the initial values are different. The best fit for: C1, the base model with only one compartment per biological stage (dashed dotted green line); C2, including a second compartment for pupae (dashed maroon line); and C3, including second compartments for both pupae and L4 (solid blue line). (B) Model output in version C3, where both L4 and pupae have two compartments, in high density (top row) or low density (bottom row) treatments. From left to right, the panels show: combined L1 and L2 stages; L3 stage; L4 stage; pupae; emerging males; and emerging females. Each gray line is an experimental replicate.

https://doi.org/10.1371/journal.pcbi.1009102.g003

22.3, and 21.3 for the base model (C1), including a second compartment for pupae (C2), and including a second compartment for both L4 and pupae (C3), respectively. We found the model with the two extra compartments to be best among the three as it had the lowest AICc value. When we examined our results by individual stage separately, rather than by total development time, we saw that the model matches the data well at each stage (Fig 3B). It is clear, however, that the model does not capture all the features of each stage for individual data replicates, such as the persistence of individuals in L4 in high density. This is unsurprising as the data is more varied in the later stages, and the model fits the average of the data, not each replicate individually.



Fig 4. Total survival proportion by sex and density treatment for experimental data and model versions D1-D3. Survival proportion of males (red) and females (blue). The model (D1-D3) under low density (left) and high density (right) treatments. On the far right of each panel, the values for median and mean of all 13 high and 20 low density replicates, with dashed lines indicating one quartile above and below the median and one standard deviation around the mean.

https://doi.org/10.1371/journal.pcbi.1009102.g004

3.2 Death proportion fit: Variation D1-D3

We found that the model, with the best fit for each variation, reproduced the median value of the data well (Fig 4). For the death proportion versions, we obtained AIC_c of 59.52, 61.1, and 59.51 for D1, D2, and D3, respectively. Thus, version D3, with density-dependent death for females, is the best fit model.

In version D1, we fit a single constant for both male and female mortality across high and low density treatments. While we fit daily mortality, we discuss results in terms of overall survival, where survival can be determined approximately from mortality as: total survival proportion = $(1 - \text{mortality proportion})^{\text{development time}}$. Experimentally, we observed that while females in the low density treatment and males in both density treatments had a survival proportion near 0.75, the females in the high density treatment fared much worse. As all four situations are indistinguishable in D1, the single constant fit resulted in a survival proportion of 0.70. This is lower than the observed survival of 0.75 in three of the cases as it accounts for the lower survival exhibited by females in the high density treatment.

In D2, two parameters are included for mortality: one for males and one for females. Separating by sexes did not produce a better fit considering the added complexity of an extra parameter. In contrast, the inclusion of density-dependent death for females, as in version D3, allowed for different mortality by density condition for females. With this model version, we found a mortality proportion close to the median of the experimental data, and this version had the lowest AIC_c value. The total number of males that emerged in version D3 of the model is approximately 10 and 29 in low and high density, respectively, and the median of the number of total males in the data is 10 and 29 (low and high density, respectively). For females, the model outcomes for D3 were approximately 10 and 23 total females, and the median of the data was 10 and 22 (low and high density, respectively).

3.3 Growth function fit: Variation E

In the experiments, several low density replicates were almost entirely composed of large individuals (Fig 5A, left). In fact, nearly half of emerging adults in the low density replicates were entirely from the large mass group. In contrast, there were several replicates in high density treatment with only small individuals (Fig 5A, right). When we estimated parameters in our growth functions, the model fit to the mean of the experimental data, which is skewed by



(B) Proportion of emerging females by mass through time

Fig 5. Fitting results for inclusion of the growth function, variation E. (A) The proportion of emerged females in a particular mass group is represented on the x-axis, and the mass group of these females on the y-axis. The green diamond is the best fit model. The median of the data is represented with a purple triangle with a dashed line for one quartile above and below. The blue triangle represents the mean of the data, and the dashed line is one standard deviation around the mean. The left panel is low density and right is high density. (B) Proportion of females that emerge from each mass group through time: small (left), medium (middle), large (right). The top row is high density and the bottom is low density. The model output is in solid blue. Individual replicates of the data are in gray.

https://doi.org/10.1371/journal.pcbi.1009102.g005

replicates with all individuals in a single mass group. This is particularly true for the large mass group in the low density treatment and the small mass group in the high density treatment.

The model with our determined growth functions fit the mean proportion of females that emerged in both low and high density closely (Fig 5A). First, we calculated *v*, the relative proportion that males grow compared to females, directly from the data for a value of v = 0.52. We then use this value to fit the growth function to the female data alone.

Although we do not fit development time based on mass, we generally saw that individuals emerge in the model around the same time as observed in the data (Fig 5B). Comparing the model to data, the large individuals in low density emerged slightly earlier. Furthermore, the modeled development time of the medium individuals in high density was much flatter than the individual data replicates. In all cases the model output showed lower peaks and emergence across a longer time period than found in any the graphs of individual data replicates. However, the model output was close to the average trend in the data.

3.4 Fixed parameter variation

Some of the parameters in our study were set as fixed values. We now focus on each parameter that was fixed to a particular value, and discuss its effect on the model fits when varied. The following fixed parameters arise in different variations: the resource usage rate q, the relative male growth v, the half maximal constants (c_1, c_2) in the growth functions, the minimum mortality h in the density-dependent death function, and the exponent in the density-dependent female death proportion n.

3.4.1 Resource decay, *q*. We repeated the fits for variation E using different resource usage rates, *q*. In this study, we aimed to choose a *q* that allowed all resources to be consumed in the high density treatment, but for some resources to remain in the low density treatment. We considered seven different *q* values ranging from 0.0005 to 0.05. Fig 6A left, shows how resources decay through time in our model in the low density treatment. For all values of *q* that are 0.01 or smaller, more than 20% of the original resources remained after 15 days in the low density treatment. Fig 6A right, shows resource levels through time in the high density treatment. For $q \ge 0.01$, nearly all resources were used by day 15. Most values of *q* such that 0.005 $\le q \le 0.2$ would sufficiently fit our desired condition: complete loss of resources in high density, but not in low density. Our choice of q = 0.01 for this study falls within this range.

The timing of female emergence of each mass group is similar for all intermediate q values. For the smallest two q values, q = 0.0005 and q = 0.001, there were noticeable differences in timing of female emergence in the high density treatment (Fig 6C). The small mass group was slower to emerge and the medium mass group was slightly faster compared to other q values as well as to the average trend of the data (Fig 6C, black dashed line). For q = 0.05, the function did not obtain similar proportions to that seen in the data (Fig 6B). This occurred to a lesser extent for = q = 0.025 and q = 0.02. Overall, for choices of q that are small enough but not too small, the proportion of females emerging through time by each mass group was close to the average of the data.

3.4.2 Relative male growth, v. The value v is the proportion of growth of males relative to that of females. Our model focuses on female body size and time of emergence, and v does not affect the development time or growth of females. The growth of females does implicitly depend on the number of males through total larval density, but does not change regardless of the mass group of each male. In order to confirm this, we varied v between 0.1 and 1, and compared female mass and total population size. The results are identical, in all aspects, for all values of v in this range, apart from the proportion of males in each mass group.

3.4.3 Per capita resource for half maximal growth, c_1 and c_2 . For the majority of this study, we fixed the constants of the location of half maximal growth, c_1 and c_2 , to equal 0.5 in both growth functions (see section 2.2.4). As we want values of the growth function to approach a maximum of one as $\frac{r}{N}$ gets near one, we choose the constants of half maximal growth to occur when $\frac{r}{N} = 0.5$. We now consider nine pairs for c_1 and c_2 with combinations of c_i in {0.3, 0.5, 0.69 $\approx \log(2)$ }. From the combinations examined, $c_1 = 0.5$ resulted in the proportions of emerging females of small, medium, and large mass that fit the average of the data regardless of whether $c_2 = 0.3$, 0.5 or 0.69 (S1A Fig). The choice of $c_1 = 0.69$ appropriately determined the proportion of emerging females in the high density treatment, but $c_1 = 0.69$ resulted in more small mosquitoes in the low density treatment than observed in the data. We saw no difference in timing based on the choice of c_1 and c_2 (S1B Fig).

3.4.4 Minimum mortality in female density-dependent death, *h*. We fixed the constant *h* in the numerator of the female density-dependent function to be h = 100 (see section 2.2.3). A positive choice for *h* ensures that the death proportion is not zero at low population size. However, the final survival proportion is quite insensitive to the value for minimum mortality,



(C) Proportion females emerging through time

Fig 6. Proportion of females that emerge as resource usage rate, *q***, varies.** (A) Simulated available resources throughout the time of the experiment for different values of *q*. (B) The proportion of emerging females in each mass group as *q* varies. Blue, red, and gold represent small, medium, and large mass groups, respectively. The black dashed lines indicate the divisions at which different mass groups were expected based on means of proportions of the mass groups from the data. In particular, the lower dashed line separates small and medium mosquitoes, and the upper dashed line, the red bar would be entirely between the two dashed lines, and the gold bar would be above the higher dashed line. (C) The proportion of females emerging over time by mass group: small (left), medium (middle), and large (right). The top row is the high density treatment, and the bottom row is the low density treatment. The solid color lines are model output with different *q* values. The black dashed line represents the mean of the data.

https://doi.org/10.1371/journal.pcbi.1009102.g006

h. This is because the denominator contains a large number, e.g. $f^3 \approx 24^3$. Alternative choices for *h* (up to *h* = 1000) produced very similar curves (S2 Fig). Once *h* = 2000, there were observable differences, but mostly when population levels were very low. As population levels were only low late in experiments, different values of *h* did not appreciably change the final survival proportion.

Variation	Original							
	<i>a</i> = 0.0043	<i>a</i> = 0.001	<i>a</i> = 0.01	<i>a</i> = 0.0043	<i>a</i> = 0.0043			
	<i>b</i> = 1.61	<i>b</i> = 1.61	<i>b</i> = 1.61	<i>b</i> = 1	<i>b</i> = 2			
C1	22.93	22.90	23.26	22.94	22.93			
C2	22.36	22.42	22.97	22.41	22.36			
C3	21.32	21.09	23.25	21.76	21.18			

Table 3. AIC_c values for density-dependent functions with different parameters. The original parameter choice was a = 0.0043 and b = 1.61. We then consider a = 0.001 and a = 0.01, each with b = 1.61, as well as b = 1 and b = 2, each with a = 0.0043.

https://doi.org/10.1371/journal.pcbi.1009102.t003

3.4.5 Exponent in density-dependent death, *n*. In the density-dependent death function for females (see section 2.2.3), we fixed the Hill exponent to three, n = 3. Similar survival proportions occurred for any Hill exponent greater than one (S3 Fig). However, an exponent of one deviates considerably from the sex-specific survival proportion. While higher values of the Hill exponent produced nearly identical survival proportions, the choice of a Hill exponent of three was in the region of values we considered where survival proportions were not changing with changes in the exponent. Note that for each Hill exponent chosen, while the survival proportion did not differ, other parameter estimates did change.

3.4.6 Density-dependent parameters *a*, *b*. In Walker et. al. [47], both parameters *a* and *b* in the density-dependent function were fitted to a different data set with *Aedes aegypti*. We used the same values in this study for the model selection for the standard choices (a = 0.0043, b = 1.61), but vary both *a* and *b* univariately here. We considered a = 0.001 and 0.01 with b = 1.61 and then b = 1 and 2 with a = 0.0043. For each parameter combination, we fitted the best k_m and k_f for each variation C1-C3 as describe previously in section 2.4.1. Table 3 gives the AICc values for each combination of *a* and *b*.

The lowest overall AIC_c was found with the parameter set of a = 0.001 and b = 1.61 (Table 3), although it is similar to to the AIC_c of the original parameter set. The three best choices (default values a = 0.0043 with b = 1.61, a = 0.001 with b = 1.61, and b = 2 with a = 0.0043) all produced very similar results (S4A Fig, compare solid blue line with other lines).

Among all combinations of *a* and *b*, the best model overall based on the AIC_c remains when there are two compartments for both L4 and pupae (variation C3). In fact, variation C3 was the best choice in all parameter combinations considered, except when a = 0.01 with b = 1.61. In this case, the best model was C2, but with a higher AIC_c than with other combinations of *a* and *b*. Visually, this parameter choice poorly fit the data and would not be the optimal choice (S4B Fig).

4 Discussion

In this work, we developed a discrete time mathematical model parameterized with laboratory data that demonstrated how density in the larval environment affects variability in mosquito body size. Our model separates masses into three groups—small, medium, and large—and tracks mosquito growth through aquatic stages. Using our model, we determined the distribution of mass and sex at adult emergence under different larval density treatments, and we illustrated the interactions between larval environmental conditions and adult body mass, which could have important implications for mosquito population and mosquito-borne pathogen control. This work is an important contribution towards understanding how body mass affects mosquito development and how conditions in the early developmental stages may have longer-term consequences.

Larvae require more than a single day in each developmental stage. The trigger to transition between larval stages, and from pupae to adults, is under the control of the insect's hormonal and physiological machinery, a function of development time, and not directly linked to the absolute size of the individual [53, 54]. When we used only a single compartment for each aquatic stage, more than a quarter of the individuals in the model emerged earlier than the timing observed in the data. Furthermore, the model showed emergence that is distributed over many days with much lower numbers per day than seen in the data. Comparing our model variations, the AICc selected the model with a second compartment in both L4 and pupae. This model variation extends the minimum time of emergence by two days. Importantly, while forcing the minimum to be larger by two days resulted in model output with time of emergence closer to that of the data, it also produced variance that was more similar to the data. The additional time spent in L4 was suggested previously by Levi et al. [55]. In their study, they found that in nutrient rich environments larvae initially grew quickly, but then spent a longer time in L4. In less rich nutrient environments, individuals developed more slowly throughout all stages. In both situations, they noticed similar overall times until emergence. In studies using similar temperatures to our own, pupae had an average development time close to two days [56-58]. While in our model we set the minimum development time of pupae to two days, the overall development time is often longer than the sum of the minimum time per stage because individuals may remain in any stage for a longer period.

We found that including density-dependent death is important to accurately represent the development of female larvae. Our data showed that the survival of female mosquitoes is diminished in the high density treatment. The difference in survival, compared to the low density treatment, was significant enough that the AICc selected variation D3 even though it required two additional parameters to incorporate this density dependence. In contrast, the difference between male and female survival in our experiment was negligible compared to the difference in female survival in the two density treatments. This is emphasized further by the lower AICc score for the model with a single constant parameter representing both male and female survival compared to the model with two parameters for sex-dependent survival. Density plays a key role in females' survival, and survival to adulthood is an important factor in mosquito population and disease dynamics. In particular, because female mosquitoes transmit pathogens, any significant alterations of female survival could propagate through population dynamic processes to cause profound impacts on population magnitude and potentially pathogen transmission. This is consistent with previous work in Ae. aegypti where density-dependent effects were more pronounced on a range of female traits including development time, body mass, and longevity [42]. Furthermore, similar sex-specific density-dependent effects are observed across taxa, e.g. in ectoparasites [59], fishes [60], lizards [61] and large herbivores [62].

Our model fitting demonstrates that growth to the largest mass group quickly becomes restricted in resource-limited environments. We determined the exponent of the function determining growth from medium to large mass groups (G_2) to be $n_2 = 9$. Given that the exponent determines the steepness of the growth curve, the proportion growing from medium to large mosquitoes rapidly diminishes to nearly zero as resources decay. This indicates that only in environments with ample resources will individuals grow into the largest mass groups. It should be noted that we fixed rather than fit several parameters in part due to the lack of variation in feeding regimes in our data. To further explore the consequences of this choice, in section 3.4 we considered the sensitivity of our model to these parameters. We showed that our choices of parameters either give the best fits overall or the results were not sensitive to the parameter. While examining various feeding regimes will be important future work, our model and choices of parameters described our data well.

In the model, development time was less important in determining adult size at emergence compared to the larval density environment. Our experiments show significant differences in the emergence of different mass groups between the high and low density treatments. In particular, no large mass female individuals emerged in high density and no small mass individuals emerged in low density. The only significant difference observed for development time as a function of mass occurred from small to medium mass groups in the high density treatment, but the effect size was small. In contrast to previous work, our model focuses on the distributions of mosquito mass with different larval environments. Romeo Aznar et. al. [33] modeled saturating monotonic growth in size along with food-independent and food-dependent transition between stages, but did not explicitly distinguish between maturation and weight gain. They found that for the same environmental conditions variation in body size at emergence is incidental and based on relative time of emergence within the cohort. However, they observed increased dispersion in body weight and time of emergence in suboptimal feeding conditions. In a modeling investigation, Gilpin and McClelland considered a uniform distribution for size, and found the range of the spread in size increased linearly over time [34]. This quickly led to very large and very small mosquitoes with equal probability, and there was no relationship between size and larval environment. In other work, Padmanabha et. al. aimed to predict time of pupation, but not variation in mass at emergence [35]. They did consider variation in time to emergence as well as survival by temperature, but did not track changes in mass.

Our model is limited in a few key ways: we focus on a strain of a single mosquito species, Aedes aegypti; we assume a constant temperature setting; and we assume the absence of any additional resources over the course of the experiment. While there are other species of the genus Aedes that typically have similar behavior, even within the Aedes genus there are differences that could change the results. Here, we employ the Rockefeller strain of Aedes aegypti which has been laboratory-reared for several generations with an *ad libitum* diet. Despite this, the strain is not resilient to effects of food stress resulting from poor nutrition or competition for resources [32]. Furthermore, the distribution of adult body mass of mosquitoes reared in low and high larval densities in this study is representative of body mass of Ae. aegypti in other laboratory and field studies [63, 64]. Regardless, extrapolations to other species would require experiments for species-specific parameterization. Temperature has a key role in development time and mortality [56, 57], and inclusion of temperature variation would improve the model's utility. Adding in effects of temperature would significantly increase the complexity of the model; however, it will be important for future iterations of the model to consider variation in temperature. The level of resources available alters the potential for growth of individuals. In field conditions, resources may decrease for several reasons including competition with other species or environmental changes (e.g. flushing due to rain) as well as increase through the influx of detritus (e.g. falling leaves) in the water source. An open system would likely have more complex resource dynamics with many different levels and timing of resource changes. We leave an examination of more complex resource settings as future work. In addition, features found to be important in other models have been omitted because we focus on results at emergence rather than specific results at each stage. For example, the inclusion of resource dependent mortality at young stages and resource dependent delay of L4 would more accurately describe behavior of individual stages. Additionally, we use mass as a measurement in our model, while Padmanabha et. al. [35] found in their model that reserves, rather than raw weight, more accurately describe when individual mosquitoes pupate. While reserves are an important indicator of success as an adult, using mass as a proxy still performs well and is easier to measure.

The work described herein is an important contribution towards understanding how environmental conditions during juvenile growth affect mosquito mass and development, and thus control of mosquito populations and the diseases whose causative agents they transmit. Historically, control of most mosquito-borne diseases focuses on decreasing mosquito population size through methods such as insecticide applications [65, 66]. In the past two decades, novel strategies aimed at reducing wild populations or replacing them with those that cannot transmit pathogens have gained traction and are being tested—and implemented across the globe [67]. These strategies include releases of Wolbachia-infected mosquitoes that induce females sterility as well as releases of genetically modified mosquitoes that pass on lethal genes, decrease sterility, inhibit flight, or change males to females [68–76]. While these traditional and novel interventions directly alter the adult mosquito population, it is essential to consider the potential effects on juvenile stages as well because many important morphological processes occur in the early developmental stages. These processes can significantly influence mosquito life history characteristics and the role of mosquitoes in pathogen transmission. Although we did not model how mass affects the adult population dynamics or disease spread, our model can be adapted to consider the importance of larval environmental heterogeneity on mosquito and disease mitigation strategies. For example, in methods inducing late-acting lethality in larval populations, competition is high in early larval stages, which could result in longer development time and smaller adults emerging. Longer development times lead to delayed impacts of density dependence, and smaller female mosquitoes may bite more frequently across numerous hosts, leading to wider spread of pathogens [73, 77, 78]. If a control method targets adults only, then adult mosquito population drops, which leads to fewer eggs laid, which in turn leads to lower aquatic density and thus higher survival of females and larger females in the next generation who may be more efficient in obtaining blood meals [7]. These two examples could be explored in depth with an extension to our model, and different strategies that target different or multiple life stages could be evaluated against one another in the presence of mass- and density- mediated effects on larval population dynamics.

Supporting information

S1 File. Detailed model description, equations and Table A of fitted model parameters. A mathematical description of the full system, along with exemplary equations are provided. In Table A, for each fitted parameter, we list the fitted value under the variation considered. A dash indicates that parameter was not fitted under that variation. (PDF)

S1 Fig. Proportion of females that emerge as half maximal growth varies. (A) The proportion of emerging females in each mass group as c_1 and c_2 vary for low density (left) and high density (right) treatments. Blue, red, and gold represent small, medium, and large mass groups, respectively. The black dashed lines indicate the divisions at which different mass groups were expected based on means of proportions of the mass groups from the data. In particular, the lower dashed line separates small and medium mosquitoes, and the upper dashed line separates medium and large mosquitoes. For close fits to the data, the blue bar would be below the lower dashed line, the red bar would be entirely between the two dashed lines, and the gold bar would be above the higher dashed line. (B) The proportion of females emerging over time by mass group: small (left), medium (middle), and large (right). The top row is the high density treatment, and the bottom row is the low density treatment. The solid color lines are model output with different c_1 and c_2 values. The black dashed line represents the mean of the data.

(TIFF)

S2 Fig. Density-dependent death proportion as the minimum death varies. Density-dependent death function $f(N) = \mu_f \frac{N^3 + h}{N^3 + f^3}$ with f = 23.5 and $\mu_f = 0.0661$. The minimum constant h varies from 1 to 2000. See section 2.2.3 for details on the functional form. (TIFF)

S3 Fig. Density-dependent death proportion as the Hill exponent changes. Model results employing the density-dependent death function $f(N) = \mu_f \frac{N^n + h}{N^n + f^n}$ with f = 23.5, $\mu_f = 0.0661$, and h = 100. The density-dependent death exponent, n, varies along the x-axis from 1 to 10. The total larvae at a given time, N, changes in the course of the model simulations. The dashed lines represent the median values from the data and the diamonds the model results for the survival proportion of females in low density (blue), females in high density (yellow), males in low density (red), and males in high density (purple). The survival proportion for males and females in low density is indistinguishable in the data. See section 2.2.3 for details on the functional form of f(N).

(TIFF)

S4 Fig. Varying parameters *a*, *b* in the density-dependent function. (A) The solid blue line shows the choice of our model with our original parameters a = 0.0043 and b = 1.61. The two other parameter choices with similar AIC_c values are shown, when a = 0.001 and b = 1.61 (solid green line) and when a = 0.0043 and b = 2 (solid dark maroon line). (B) This shows all three variations C1 (dashed dotted green line), C2 (dashed dotted maroon line), and C3 (solid blue line) with the parameters set at a = 0.01 and b = 1.61. (TIFF)

Author Contributions

- **Conceptualization:** Melody Walker, Karthikeyan Chandrasegaran, Clément Vinauger, Michael A. Robert, Lauren M. Childs.
- Data curation: Melody Walker, Karthikeyan Chandrasegaran.
- Formal analysis: Melody Walker, Karthikeyan Chandrasegaran, Michael A. Robert, Lauren M. Childs.

Funding acquisition: Clément Vinauger, Lauren M. Childs.

Investigation: Melody Walker, Karthikeyan Chandrasegaran.

Methodology: Melody Walker, Karthikeyan Chandrasegaran, Clément Vinauger, Michael A. Robert, Lauren M. Childs.

Project administration: Clément Vinauger, Michael A. Robert, Lauren M. Childs.

Resources: Clément Vinauger, Michael A. Robert, Lauren M. Childs.

Software: Melody Walker, Lauren M. Childs.

Supervision: Clément Vinauger, Michael A. Robert, Lauren M. Childs.

Validation: Melody Walker, Karthikeyan Chandrasegaran, Clément Vinauger, Michael A. Robert, Lauren M. Childs.

Visualization: Melody Walker, Karthikeyan Chandrasegaran, Lauren M. Childs.

Writing - original draft: Melody Walker.

Writing – review & editing: Melody Walker, Karthikeyan Chandrasegaran, Clément Vinauger, Michael A. Robert, Lauren M. Childs.

References

- Franklinos LH, Jones KE, Redding DW, Abubakar I. The effect of global change on mosquito-borne disease. Lancet Infect Dis. 2019; 19(9):e302–e312. <u>https://doi.org/10.1016/S1473-3099(19)30161-6</u> PMID: 31227327
- Souza-Neto JA, Powell JR, Bonizzoni M. Aedes aegypti vector competence studies: A review. Infect Genet Evol. 2019; 67:191–209. https://doi.org/10.1016/j.meegid.2018.11.009 PMID: 30465912
- Wong J, Stoddard ST, Astete H, Morrison AC, Scott TW. Oviposition site selection by the dengue vector Aedes aegypti and its implications for dengue control. PLOS Negl Trop Dis. 2011; 5(4):e1015. <u>https://</u> doi.org/10.1371/journal.pntd.0001015 PMID: 21532736
- Yee DA, Kneitel JM, Juliano SA. Environmental correlates of abundances of mosquito species and stages in discarded vehicle tires. J Med Entomol. 2010; 47(1):53–62. <u>https://doi.org/10.1603/033.047.</u> 0107 PMID: 20180308
- Juliano SA. Coexistence, exclusion, or neutrality? A meta-analysis of competition between Aedes albopictus and resident mosquitoes. Isr J Ecol Evol. 2010; 56(3-4):325–351. https://doi.org/10.1560/IJEE. 55.3-4.325 PMID: 23482823
- Parker AT, Gardner AM, Perez M, Allan BF, Muturi EJ. Container size alters the outcome of interspecific competition between *Aedes aegypti* (Diptera: Culicidae) and Aedes albopictus. J Med Entomol. 2019; 56(3):708–715. https://doi.org/10.1093/jme/tjy215 PMID: 30566608
- 7. Steinwascher K. Competition among *Aedes aegypti* larvae. PLOS One. 2018; 13(11):e0202455. https://doi.org/10.1371/journal.pone.0202455 PMID: 30439951
- 8. Christophers SR. *Aedes aegypti*: the yellow fever mosquito: its life history, bionomics and structure. London: The Syndics of the Cambridge University Press, Bentley House; 1960.
- Farjana T, Tuno N. Multiple blood feeding and host-seeking behavior in Aedes aegypti and Aedes albopictus (Diptera: Culicidae). J Med Entomol. 2013; 50(4):838–846. https://doi.org/10.1603/ME12146 PMID: 23926783
- Noden BH, O'NEAL PA, Fader JE, Juliano SA. Impact of inter-and intra-specific competition among larvae on larval, adult, and life-table traits of *Aedes aegypti* and *Aedes albopictus* females. Ecol Entomol. 2016; 41(2):192–200. https://doi.org/10.1111/een.12290 PMID: 27141149
- Nelson MJ, et al. Aedes aegypti: Biology and ecology; 1986. Available from: https://iris.paho.org/ handle/10665.2/28514.
- Delatte H, Gimonneau G, Triboire A, Fontenille D. Influence of temperature on immature development, survival, longevity, fecundity, and gonotrophic cycles of *Aedes albopictus*, vector of chikungunya and dengue in the Indian Ocean. J Med Entomol. 2009; 46(1):33–41. <u>https://doi.org/10.1603/033.046.0105</u> PMID: 19198515
- Souza RS, Virginio F, Riback TIS, Suesdek L, Barufi JB, Genta FA. Microorganism-based larval diets affect mosquito development, size and nutritional reserves in the yellow fever mosquito *Aedes aegypti* (Diptera: Culicidae). Front physiol. 2019; 10:152. https://doi.org/10.3389/fphys.2019.00152 PMID: 31024326
- Briegel H, Timmermann SE. Aedes albopictus (Diptera: Culicidae): physiological aspects of development and reproduction. J Med Entomol. 2001; 38(4):566–571. https://doi.org/10.1603/0022-2585-38.4. 566 PMID: 11476337
- Armbruster P, Hutchinson RA. Pupal mass and wing length as indicators of fecundity in Aedes albopictus and Aedes geniculatus (Diptera: Culicidae). J Med Entomol. 2002; 39(4):699–704. https://doi.org/ 10.1603/0022-2585-39.4.699 PMID: 12144308
- Blackmore MS, Lord CC. The relationship between size and fecundity in Aedes albopictus. J Vector Ecol. 2000; 25(2):212–217. PMID: 11217219
- Briegel H. Metabolic relationship between female body size, reserves, and fecundity of Aedes aegypti. J Insect Physiol. 1990; 36(3):165–172. https://doi.org/10.1016/0022-1910(90)90118-Y
- Ower GD, Juliano SA. Effects of larval density on a natural population of *Culex restuans* (Diptera: Culicidae): No evidence of compensatory mortality. Ecol Entomol. 2019; 44(2):197–205. https://doi.org/10. 1111/een.12689 PMID: 31178624
- Steinwascher K. Relationship between pupal mass and adult survivorship and fecundity for Aedes aegypti. Environ Entomol. 1982; 11(1):150–153. https://doi.org/10.1093/ee/11.1.150

- De Jesus CE, Reiskind MH. The importance of male body size on sperm uptake and usage, and female fecundity in *Aedes aegypti* and *Aedes albopictus*. Parasit Vectors. 2016; 9(1):1–7. https://doi.org/10. 1186/s13071-016-1734-8 PMID: 27519588
- Ponlawat A, Harrington LC. Factors associated with male mating success of the dengue vector mosquito, *Aedes aegypti*. Am J Trop Med Hyg. 2009; 80(3):395–400. https://doi.org/10.4269/ajtmh.2009. 80.395 PMID: 19270288
- Alto BW, Bettinardi DJ, Ortiz S. Interspecific larval competition differentially impacts adult survival in dengue vectors. J Med Entomol. 2015; 52(2):163–170. <u>https://doi.org/10.1093/jme/tju062</u> PMID: 26336301
- EH JG, Walker K, Ernst K, Riehle M, Davidowitz G. Size as a Proxy for Survival in Aedes aegypti (Diptera: Culicidae) Mosquitoes. J Med Entomol. 2020; 57(4):1228–1238. https://doi.org/10.1093/jme/ tjaa055
- 24. Reiskind M, Lounibos L. Effects of intraspecific larval competition on adult longevity in the mosquitoes *Aedes aegypti* and *Aedes albopictus*. Med Vet Entomol. 2009; 23(1):62–68. https://doi.org/10.1111/j. 1365-2915.2008.00782.x PMID: 19239615
- Nasci RS. Relationship between adult mosquito (Diptera: Culicidae) body size and parity in field populations. Environ Entomol. 1986; 15(4):874–876. https://doi.org/10.1093/ee/15.4.874
- Roitberg BD, Keiser S, Hoffmeister T. State-dependent attacks in a mosquito. Physiol Entomol. 2010; 35(1):46–51. https://doi.org/10.1111/j.1365-3032.2009.00709.x
- Rocha-Santos C, Dutra ACVPL, Fróes Santos R, Cupolillo CD, de Melo Rodovalho C, Bellinato DF, et al. Effect of Larval Food Availability on Adult *Aedes Aegypti* (Diptera: Culicidae) fitness and susceptibility to Zika infection. J Med Entomol. 2020; 58(2):535–547. https://doi.org/10.1093/jme/tjaa249
- Juliano SA, Ribeiro GS, Maciel-de Freitas R, Castro MG, Codeço C, Lourenço-de Oliveira R, et al. She's a femme fatale: low-density larval development produces good disease vectors. Mem I Oswaldo Cruz. 2014; 109(8):1070–1077. https://doi.org/10.1590/0074-02760140455 PMID: 25591112
- Bara J, Rapti Z, Cáceres CE, Muturi EJ. Effect of larval competition on extrinsic incubation period and vectorial capacity of *Aedes albopictus* for dengue virus. PLOS One. 2015; 10(5):e0126703. https://doi. org/10.1371/journal.pone.0126703 PMID: 25951173
- Alto BW, Lounibos LP, Mores CN, Reiskind MH. Larval competition alters susceptibility of adult Aedes mosquitoes to dengue infection. Proceedings of the Royal Society B: Biological Sciences. 2008; 275 (1633):463–471. https://doi.org/10.1098/rspb.2007.1497 PMID: 18077250
- Chandrasegaran K, Juliano SA. How do trait-mediated non-lethal effects of predation affect populationlevel performance of mosquitoes? Frontiers in Ecology and Evolution. 2019; 7:25. <u>https://doi.org/10.3389/fevo.2019.00025</u> PMID: 31218216
- Alto BW, Lounibos LP, Higgs S, Juliano SA. Larval competition differentially affects arbovirus infection in *Aedes* mosquitoes. Ecology. 2005; 86(12):3279–3288. <u>https://doi.org/10.1890/05-0209</u> PMID: 19096729
- Romeo Aznar V, De Majo MS, Fischer S, Francisco D, Natiello MA, Solari HG. A model for the development of *Aedes* (Stegomyia) *aegypti* as a function of the available food. J Theor Biol. 2015; 365:311–324. https://doi.org/10.1016/j.jtbi.2014.10.016 PMID: 25451964
- Gilpin ME, McClelland GAH. Systems analysis of the yellow fever mosquito Aedes aegypti. Fortschr Zool. 1979; 25(2-3):355–388. PMID: 535875
- Padmanabha H, Correa F, Legros M, Nijhout HF, Lord C, Lounibos LP. An eco-physiological model of the impact of temperature on *Aedes aegypti* life history traits. J Insect Physiol. 2012; 58(12):1597– 1608. https://doi.org/10.1016/j.jinsphys.2012.09.015 PMID: 23068992
- Focks DA, Haile D, Daniels E, Mount GA. Dynamic life table model for Aedes albopictus (Diptera: Culicidae): analysis of the literature and model development. J Med Entomol. 1993; 30(6):1003–1017.
- Magori K, Legros M, Puente ME, Focks DA, Scott TW, Lloyd AL, et al. Skeeter Buster: a stochastic, spatially explicit modeling tool for studying *Aedes aegypti* population replacement and population suppression strategies. PLOS Negl Trop Dis. 2009; 3(9):e508. <u>https://doi.org/10.1371/journal.pntd.</u> 0000508 PMID: 19721700
- Natiello MA, Solari HG. Modelling population dynamics based on experimental trials with genetically modified (RIDL) mosquitoes. Ecol Modell. 2020; 424:108986. <u>https://doi.org/10.1016/j.ecolmodel.2020</u>. 108986
- Matthews BJ, McBride CS, DeGennaro M, Despo O, Vosshall LB. The neurotranscriptome of the Aedes aegypti mosquito. BMC Genomics. 2016; 17(1):1–20. <u>https://doi.org/10.1186/s12864-015-2239-</u> 0 PMID: 26738925

- 40. Tallon AK, Hill SR, Ignell R. Sex and age modulate antennal chemosensory-related genes linked to the onset of host seeking in the yellow-fever mosquito, *Aedes aegypti*. Sci Rep. 2019; 9(1):1–13. https://doi.org/10.1038/s41598-018-36550-6 PMID: 30631085
- Walsh R, Facchinelli L, Ramsey J, Bond J, Gould F. Assessing the impact of density dependence in field populations of *Aedes aegypti*. J Vector Ecol. 2011; 36(2):300–307. https://doi.org/10.1111/j.1948-7134.2011.00170.x PMID: 22129401
- Chandrasegaran K, Kandregula SR, Quader S, Juliano SA. Context-dependent interactive effects of non-lethal predation on larvae impact adult longevity and body composition. PLOS One. 2018; 13(2): e0192104. https://doi.org/10.1371/journal.pone.0192104 PMID: 29401513
- Siegel JP, Novak RJ, Lampman RL, Steinly BA. Statistical appraisal of the weight–wing length relationship of mosquitoes. Journal of Medical Entomology. 1992; 29(4):711–714. https://doi.org/10.1093/ jmedent/29.4.711 PMID: 1353798
- Koenraadt C. Pupal dimensions as predictors of adult size in fitness studies of Aedes aegypti (Diptera: Culicidae). Journal of Medical Entomology. 2014; 45(2):331–336. <u>https://doi.org/10.1093/jmedent/45.2.331</u>
- 45. Carrington LB, Armijos MV, Lambrechts L, Barker CM, Scott TW. Effects of fluctuating daily temperatures at critical thermal extremes on *Aedes aegypti* life-history traits. PLOS One. 2013; 8(3):e58824. https://doi.org/10.1371/journal.pone.0058824 PMID: 23520534
- Hancock PA, White VL, Ritchie SA, Hoffmann AA, Godfray HCJ. Predicting Wolbachia invasion dynamics in *Aedes aegypti* populations using models of density-dependent demographic traits. BMC Biol. 2016; 14(1):96. https://doi.org/10.1186/s12915-016-0319-5 PMID: 27825343
- Walker M, Robert MA, Childs LM. The importance of density dependence in juvenile mosquito development and survival: A model-based investigation. Ecol Modell. 2021; 440:109357. https://doi.org/10. 1016/j.ecolmodel.2020.109357
- Romeo Aznar V, Alem I, De Majo MS, Byttebier B, Solari HG, Fischer S. Effects of scarcity and excess of larval food on life history traits of *Aedes aegypti (Diptera: Culicidae)*. Journal of Vector Ecology. 2018; 43(1):117–124. https://doi.org/10.1111/jvec.12291 PMID: 29757503
- Holling CS. Some characteristics of simple types of predation and parasitism1. The Canadian Entomologist. 1959; 91(7):385–398. https://doi.org/10.4039/Ent91385-7
- 50. Real LA. The kinetics of functional response. The American Naturalist. 1977; 111(978):289–300. https://doi.org/10.1086/283161
- Anderson D, Burnham K. Model selection and multi-model inference. Second NY: Springer-Verlag. 2004; 63(2020):10.
- Burnham KP, Anderson DR. Multimodel inference: understanding AIC and BIC in model selection. Sociol Methods Res. 2004; 33(2):261–304. https://doi.org/10.1177/0049124104268644
- 53. Stern D. Body-size control: how an insect knows it has grown enough. Current Biology. 2003; 13(7): R267–R269. https://doi.org/10.1016/S0960-9822(03)00197-0 PMID: 12676103
- van Schoor T, Kelly ET, Tam N, Attardo GM. Impacts of dietary nutritional composition on larval development and adult body composition in the yellow fever mosquito (Aedes aegypti). Insects. 2020; 11 (8):535. https://doi.org/10.3390/insects11080535
- 55. Levi T, Ben-Dov E, Shahi P, Borovsky D, Zaritsky A. Growth and development of *Aedes aegypti* larvae at limiting food concentrations. Acta Trop. 2014; 133:42–44. https://doi.org/10.1016/j.actatropica.2014. 02.001 PMID: 24524949
- 56. Couret J, Dotson E, Benedict MQ. Temperature, larval diet, and density effects on development rate and survival of *Aedes aegypti* (Diptera: Culicidae). PLOS One. 2014; 9(2):e87468. <u>https://doi.org/10. 1371/journal.pone.0087468 PMID: 24498328</u>
- Rueda L, Patel K, Axtell R, Stinner R. Temperature-dependent development and survival rates of *Culex quinquefasciatus* and *Aedes aegypti* (Diptera: Culicidae). J Med Entomol. 1990; 27(5):892–898. <u>https://doi.org/10.1093/jmedent/27.5.892</u> PMID: 2231624
- Tun-Lin W, Burkot T, Kay B. Effects of temperature and larval diet on development rates and survival of the dengue vector *Aedes aegypti* in north Queensland, Australia. Med Vet Entomol. 2000; 14(1):31–37. https://doi.org/10.1046/j.1365-2915.2000.00207.x PMID: 10759309
- 59. Szentiványi T, Vincze O, Estók P. Density-dependent sex ratio and sex-specific preference for host traits in parasitic bat flies. Parasites & Vectors. 2017; 10(1):1–9. <u>https://doi.org/10.1186/s13071-017-2340-0 PMID: 28851414</u>
- Bartolino V, Ciannelli L, Bacheler NM, Chan KS. Ontogenetic and sex-specific differences in densitydependent habitat selection of a marine fish population. Ecology. 2011; 92(1):189–200. <u>https://doi.org/ 10.1890/09-1129.1 PMID: 21560689</u>

- 61. Dreiss A, Cote J, Richard M, Federici P, Clobert J. Age-and sex-specific response to population density and sex ratio. Behavioral Ecology. 2010; 21(2):356–364. https://doi.org/10.1093/beheco/arp198
- Vander Wal E, Van Beest FM, Brook RK. Density-dependent effects on group size are sex-specific in a gregarious ungulate. PLOS One. 2013; 8(1):e53777. <u>https://doi.org/10.1371/journal.pone.0053777</u> PMID: 23326502
- **63.** Williams CR, Mincham G, Ritchie SA, Viennet E, Harley D. Bionomic response of *Aedes aegypti* to two future climate change scenarios in far north Queensland, Australia: implications for dengue outbreaks. Parasites & Vectors. 2014; 7(1):1–7. https://doi.org/10.1186/1756-3305-7-447 PMID: 25240382
- 64. Arrivillaga J, Barrera R. Food as a limiting factor for *Aedes aegypti* in water-storage containers. Journal of Vector Ecology. 2004; 29:11–20. PMID: <u>15266737</u>
- Rose RI. Pesticides and public health: integrated methods of mosquito management. Emerg Infect Dis. 2001; 7(1):17. PMID: <u>11266290</u>
- World Health Organization, UNICEF. Global vector control response 2017-2030; 2017. Available from: https://apps.who.int/iris/bitstream/handle/10665/259205/9789241512978-eng.pdf.
- 67. Organization WH, et al. Guidance framework for testing genetically modified mosquitoes. World Health Organization; 2021.
- Adelman ZN, Tu Z. Control of mosquito-borne infectious diseases: sex and gene drive. Trends Parasitol. 2016; 32(3):219–229. https://doi.org/10.1016/j.pt.2015.12.003 PMID: 26897660
- Chae K, Valentin C, Dawson C, Jakes E, Myles KM, Adelman ZN. A knockout screen of genes expressed specifically in *Ae. aegypti* pupae reveals a critical role for stretchin in mosquito flight. Insect Biochem Mol Biol. 2021; p. 103565. https://doi.org/10.1016/j.ibmb.2021.103565 PMID: 33716097
- Liu P, Jin B, Li X, Zhao Y, Gu J, Biedler JK, et al. Nix is a male-determining factor in the Asian tiger mosquito Aedes albopictus. Insect Biochem Mol Biol. 2020; 118:103311. <u>https://doi.org/10.1016/j.ibmb.</u> 2019.103311 PMID: 31901476
- Winskill P, Carvalho DO, Capurro ML, Alphey L, Donnelly CA, McKemey AR. Dispersal of engineered male Aedes aegypti mosquitoes. PLOS Negl Trop Dis. 2015; 9(11):e0004156. <u>https://doi.org/10.1371/journal.pntd.0004156</u> PMID: 26554922
- Gilbert JA, Melton L. Verily project releases millions of factory-reared mosquitoes. Nat Biotechnol. 2018; 36:781–782. https://doi.org/10.1038/nbt0918-781a PMID: 30188519
- 73. Phuc HK, Andreasen MH, Burton RS, Vass C, Epton MJ, Pape G, et al. Late-acting dominant lethal genetic systems and mosquito control. BMC Biol. 2007; 5(1):1–11. https://doi.org/10.1186/1741-7007-5-11 PMID: 17374148
- Fu G, Lees RS, Nimmo D, Aw D, Jin L, Gray P, et al. Female-specific flightless phenotype for mosquito control. PNAS. 2010; 107(10):4550–4554. https://doi.org/10.1073/pnas.1000251107 PMID: 20176967
- 75. Heinrich JC, Scott MJ. A repressible female-specific lethal genetic system for making transgenic insect strains suitable for a sterile-release program. PNAS. 2000; 97(15):8229–8232. <u>https://doi.org/10.1073/pnas.140142697 PMID: 10890889</u>
- 76. Smith DL, Perkins TA, Tusting LS, Scott TW, Lindsay SW. Mosquito population regulation and larval source management in heterogeneous environments. PloS One. 2013; 8(8):e71247. <u>https://doi.org/10.1371/journal.pone.0071247</u> PMID: 23951118
- 77. Maciel-de Freitas R, Codeco C, Lourenço-de Oliveira R. Body size-associated survival and dispersal rates of *Aedes aegypti* in Rio de Janeiro. Med Vet Entomol. 2007; 21(3):284–292. https://doi.org/10. 1111/j.1365-2915.2007.00694.x PMID: 17897370
- Russell TL, Lwetoijera DW, Knols BG, Takken W, Killeen GF, Ferguson HM. Linking individual phenotype to density-dependent population growth: the influence of body size on the population dynamics of malaria vectors. Proc R Soc B Biol Sci. 2011; 278(1721):3142–3151. https://doi.org/10.1098/rspb. 2011.0153 PMID: 21389034