How do I bite thee? Let me count the ways: Exploring the Implications of Individual

Biting Habits of Aedes aegypti for Dengue Transmission

Rebecca C. Christofferson[1,2*‡], Helen J. Wearing[3,4*], Erik A. Turner[1], Christine S. Walsh[1], Henrik

Salje[5,6], Cécile Tran Kiem[6,7], Simon Cauchemez[6]

[1]Department of Pathobiological Sciences, School of Veterinary Medicine, Louisiana State University, Skip Bertman
Dr., Baton Rouge, LA, 70803, USA
[2]Center for Computation and Technology, Louisiana State University, Skip Bertman Dr., Baton Rouge, LA, 70803,
USA
[3]Department of Biology, University of New Mexico, Albuquerque, NM, United States
[4]Department of Mathematics and Statistics, University of New Mexico, Albuquerque, NM, United States
[5] Department of Genetics, University of Cambridge, Cambridge, UK
[6]Mathematical Modelling of Infectious Diseases Unit, Institut Pasteur, CNRS, UMR, 2000, Paris, France
[7]Collège Doctoral, Sorbonne Université, Paris, France
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17　*co-primary authors

18　‡Indicates corresponding author.
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20　Key words: Arbovirus, Aedes aegypti, biting, dengue, transmission, bite, temperature
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Abstract

   In models of mosquito-borne transmission, the mosquito biting rate is an influential parameter, and understanding the heterogeneity of the process of biting is important, as biting is usually assumed to be relatively homogeneous across individuals, with time-between-bites described by an exponentially distributed process. However, these assumptions have not been addressed through laboratory experimentation. We experimentally investigated the daily biting habits of *Ae. aegypti* at three temperatures (24°C, 28°C, and 32°C) and determined that there was individual heterogeneity in biting habits (number of bites, timing of bites, etc.). We further explored the consequences of biting heterogeneity using an individual-based model designed to examine whether a particular biting profile determines whether a mosquito is more or less likely to 1) become exposed given a single index case of dengue (DENV) and 2) transmit to a susceptible human individual. Our experimental results indicate that there is heterogeneity among individuals and among temperature treatments. We further show that this results in altered probabilities of transmission of DENV to and from individual mosquitoes based on biting profiles. While current model representation of biting may work under some conditions, it might not uniformly be the best fit for this process. Our data also confirm that biting is a non-monotonic process with temperatures around 28°C being optimum.

Keywords: Aedes aegypti, arbovirus, biting, dengue, temperature, transmission

Author Summary

Mosquito biting is a necessary and critical part of arbovirus transmission. The mosquito must

bite once to acquire a virus and again to transmit, and these two bites must be separated by

sufficient time for the virus to get to the salivary glands of the mosquito. Thus, both the number

and timing of bites is important. We experimentally investigated how these bite characteristics

might be different among individuals and further explored how temperature affected the overall

heterogeneity of biting in *Aedes aegypti* mosquitoes, which carry many arboviruses like dengue

virus (DENV). We found that the biting profiles – including number and timing thereof – did vary

within temperature groups among individuals and compared outcomes associated with each

individual in an individual based model of household DENV transmission. Our results further

confirmed that temperatures around 28°C are optimal for mosquito biting (and transmission),

that correlations between biting characteristics and transmission were not uniform across

temperature, which represents another layer of heterogeneity, and that – at least at 28°C – the

null assumption of an exponential or an exponential like (geometric) distribution of biting in

mathematical models of transmission is not the best and offer an alternative.

Introduction

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Aedes aegypti are the primary vectors for several arboviruses of public health

importance and are primarily found in tropical regions. *Ae. aegypti* tend to be in urban areas and

often live close to or within human dwellings with typically limited flight ranges. There are usually

abundant opportunities for daily blood meals from human hosts [1-4]. Bloodmeal analyses from

mosquito trapping experiments have identified that the majority of bloodmeals in female *Ae.*

aegypti mosquitoes are taken from humans that live in the same household where the mosquito

was found [5]. Moreover, *Ae. aegypti* are known to take multiple bloodmeals – sometimes from

multiple individuals – within a single gonotrophic cycle [5-7]. These bites provide opportunity for

transmission of *Aedes*-borne viruses such as dengue (DENV), which is primarily moved by

humans rather than mosquitoes among households [8]. The transmission system is therefore

reliant upon contact between susceptible mosquito vectors and infectious and susceptible

humans.

Biting is an influential component of the transmission cycle of *Ae. aegypti*-borne viruses

[9-17]. Mosquitoes must bite once to acquire a virus and again to transmit that virus [18, 19].

The timing of these two events is also important, as these two bites must be separated by a

sufficient period such that the virus can disseminate through the mosquito and establish an

infection in the salivary glands [18]. After this period of time, called the extrinsic incubation

period (EIP), the virus is transmissible via the next bite from that mosquito to a susceptible

human [20]. Vector competence – the ability of a mosquito to become infected with and

ultimately transmit the virus – and EIP are both temperature dependent where higher

temperatures in general shorten the EIP and temperature affects transmission in a non-linear

fashion [11, 20, 21]. Similarly, mosquito life traits are temperature dependent, and in

combination with viral-vector kinetics define the transmission potential of the virus-vector pairing

[11, 18, 22-24].

Ae. aegypti are known to stay relatively near or within domiciles, and transmission of

arboviruses is believed to be moved among households by people, rather than by inter-

household mosquito-based transmission [25]. That means that when a household outbreak

occurs, it likely began with an index case within the household rather than the introduction of an

infectious mosquito, and mosquitoes already within the household become exposed from that

human infectious case and then spread to other members of the household [25]. This is

supported by evidence that the more time spent at home increased the likelihood of being bitten

by *Ae. aegypti* in Iquitos, Peru [26]. Heterogeneity in the biting habits of *Aedes aegypti*

mosquitoes has been observed [5, 6, 27] but modeling frameworks do not always explicitly

account for the potential for individual-level *Ae. aegypti* heterogeneity in biting.

Biting is commonly represented in model structures such as compartmental or SEIR

models as a constant mosquito biting rate, which describes a time-between-bites process that is

exponentially distributed. This governs the movement of mosquitoes from susceptible to

exposed as well as the movement of humans from susceptible to exposed classes [28-30]. In

two individual-based models (IBM), biting was represented as both an activity (0/1) and

associated with altered survival (due to risk of death during biting) [31, 32]. Still other efforts

have addressed heterogeneity in *Ae. aegypti* biting of humans, highlighting the importance of

such for understanding transmission dynamics of arboviruses [15, 27, 33-35].

Through a combination of experimental laboratory work and a computational framework,

we investigated the hypothesis that heterogeneity in individual biting habits of *Ae. aegypti* was

quantifiable and had the potential to affect small-scale DENV transmission at the individual

level.

Methods:

Determination of Individual Biting Habits:

Aedes aegypti (Rockefeller colony) were vacuum hatched and placed in a rearing pan

with deionized water and fish food as in [18, 36]. The aquatic stages were held at a constant

temperature of 28°C in environmental chambers with a photoperiod of 16:8 light:dark hours [18].

On the day after emergence, adults were cold anesthetized, and females were placed in 4-

ounce, white disposable paper cartons with screen fastened around the top to provide

containment. Cartons were placed in 24°C, 28°C, or 32°C environmental chambers with 16

individual females per temperature group. Females that died within the 2 days immediately

following transfer to the 4 oz carton were censored, as death was likely due to handling (n=3).

Female *Ae. aegypti* will take a blood meal in the absence of water to satisfy thirst which may

artificially inflate biting [37, 38], so mosquitoes were maintained on 10% sucrose solution at all

times (i.e., not starved prior to blood-meal offering). Mosquitoes had constant access to

oviposition paper which was kept damp throughout.

Beginning on day 2 post-emergence, mosquitoes were offered a 20-minute blood meal

consisting of bovine blood in Alsever's anticoagulant via Hemotek feeding device, daily at

sunrise for 23 days (ending on day 25 post-emergence) [39]. *Ae. aegypti* were previously

determined to bite repeatedly with highest frequency at 24-hour intervals [40]. The Hemotek

arms were threaded through a port in the environmental chamber so that mosquitoes were

never removed from the chamber, and thus temperature remained consistent. Cartons were

blown on to introduce CO_2 cues and then the discs were placed directly on top of the screen at

the time of feeding. Blood feeding was recorded at each blood meal offering when the presence

of fresh (bright red) blood in the abdomen was observed. Only one person observed each feed

to eliminate that as a source of variation. Deaths were documented as they occurred. Two

biological replicates at each temperature were performed (n=10-16). There were no significant

differences in the distributions of measures of interest (total # bites, time to first bite, time

between first and second bites, and time between first and last bites) between replicates across

all three temperatures (Kolmogorov-Smirnov test), so replicates were collapsed into three

temperature groups (Supplemental Information Figures **S1-4**).

IBM Model Structure and Parameterization:

To investigate the role of mosquito biting heterogeneity on small-scale, household

transmission, an individual-based model was developed that simulated household transmission

patterns using the heterogeneity in biting behavior observed in our experiments. A schematic of

the model is given in Supplemental Figure **S5**. Briefly, a household of two human individuals

was assumed, one of whom is initially infectious (the index case), with a single susceptible

mosquito which is allowed to bite according to her experimentally-determined bite profile. The

mosquito may bite each person with equal probability, regardless of infection status. If the

mosquito bites the index case, she is exposed until completion of the EIP, at which time she is

infectious. The additional human is initially susceptible until bitten by an infectious mosquito.

Data from the field suggests that, while mosquitoes bite multiple times in a single gonotrophic

cycle, each biting event was from a single source [7]. The simulation stopped once a secondary

case was achieved or at the end of 25 days, whichever came first. We also assumed that

contact with an infected human resulted in perfect transmission to the mosquito and, likewise,

that a bite from an infectious mosquito always resulted in transmission to the susceptible human

in the main model. However, we also explored the sensitivity of the model to variation in the

probability of transmission from the infectious index case to the mosquito (β_v) as well as the

probability of transmission from the exposed mosquito to susceptible humans (β_h).

Model parameters are given in **Table 1**. But briefly, the DENV infectious period in

humans was assumed to 8 days [41]. We varied the mosquito biting behavior in separate

simulations to match the different mosquitoes in our experiment. The model was realized for

1000 simulations per single mosquito, with its biting parameterized according to its distinct

profile. The mosquito bit with a probability of 0 or 1 determined by the experimental data. That

is, if a particular mosquito from the data bites on days 2 and 15 from the experimental data, she
will only bite on days 2 and 15 of the simulation. The temperature-dependent EIP for DENV
virus was defined as 11.5, 7.9, and 6.4 days for 24°C, 28°C, and 32°C, respectively [24].
164

165 Table 1: Parameter values for the models were either generated in the Christofferson laboratory or taken from the
166 literature.

Temp	Parameter	Value (rounded up to nearest day)
24	EIP [20]	12 days
28		8 days
32		7 days
All	Infectious Period of Human [41]	8 days

All individuals (humans and mosquitoes) are initialized at day 1 of the simulation, which
was run for 25 days, and the total number of transmission events to the mosquito and
secondary transmission events to the susceptible human among all simulations was recorded.
This was done to control for variability in timing and to enable direct comparison across
mosquito individuals and their biting profiles. The probability of secondary transmission was
calculated as the number of simulations where the susceptible human was infected divided by
the total number of simulations.
The output from these models was used to determine which aspects of the bite profiles
were most correlated with transmission using the "rcorr" function in R. Specifically, correlation
matrices were produced to determine the association between the total number of bites a
mosquito performed, time to first bite, the total number of times the mosquito became infected
out of 1000 simulations, and the total number of times out of 1000 simulations that the
susceptible human became infected.
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182 Local Sensitivity Analysis:
183 A sensitivity analysis was conducted for the human infectious period (+/-2 days) and the EIP (+/-
184 2 days), as well as β_v and β_h (0.25-0.75) [36, 42], for each temperature, compared to the
185 parameter values presented in Table 1. The same metrics were calculated and observed for
186 changes from the base scenario.
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189 Analytic Computation of Mosquito Transmission:

190   We can analytically derive mathematical expressions for the probability that the

191   mosquito is successfully exposed (p_{ME}) and the probability that transmission to the second

192   human occurs (p_{HE}):

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$$p_{ME} = 1 - [1 - 0.5\beta_v]^{B_{Inf}} \qquad \text{Eq. 1}$$

194   Where B_{Inf} is the total number of bites that a mosquito takes during the human infectious

195   period, i.e. the total number of bites taken during the first 8 days. This expression can be viewed

196   as 1 – the probability that every bite taken during the human infectious period did not result in

197   an exposure.

199
$$p_{HE} = \sum_{i=1}^{B_{Inf}} 0.5\beta_v (1 - 0.5\beta_v)^{(i-1)} (1 - [1 - 0.5\beta_h]^{B_{t_i+EIP}}) \qquad \text{Eq. 2}$$

The probability p_{HE} sums over the number of bites taken during the human infectious

period. The first terms represent the probability that bite i results in a successful mosquito

exposure ($0.5\beta_v$) multiplied by the probability that previous bites have not yet resulted in

exposure, $(1 - 0.5\beta_v)^{(i-1)}$. B_{t_i+EIP} is the total number of bites that a mosquito takes *after* t_i+EIP

days, where t_i represents the day that bite i occurred.

Further, we can calculate the total possible number of transmissible bites by counting

pairs of bites separated by the EIP, regardless of where they occur. This gives a measure of

transmission potential for each mosquito under relaxed assumptions about synchrony with the

index case and is a vectorial capacity-like metric. These computational measures were

completed for each mosquito biting profile.

<u>Exploring common mathematical representation of biting:</u>

Commonly used compartmental models assume that the time-between-bites for a given

population is distributed exponentially with a single rate parameter, λ. We tested this

assumption by fitting the empirical data to a null model of exponentially distributed time-

between-bites. We estimated the λ_{TEMP} parameter of the exponential distribution from the

empirical data using the maximum likelihood estimate (sample mean time between bites)$^{-1}$.

Because the experimental data were measured discretely (daily), we discretized the continuous

~EXP(λ_{TEMP}) to an analogous geometric distribution. The resulting probability distribution was

then tested against the time-between-bites of the experimental data using the Chi-square test

for goodness-of-fit (*chisq.test* function with simulated p-values) to determine whether we could

reject the null hypothesis that the experimental data could be derived from the discretized

~EXP(λ_{TEMP}) distribution. Further, we tested the proportion of mosquitoes at each simulated

scenario that bit at least once and at least twice using the *prop.test* function in R. The empirical

and theoretically-derived distributions of time-to-first-bite, time-between-first-and-last bites, and

total number of bites were compared with the Chi-square goodness-of-fit as above.

For the time-between-bites data where the Chi-square test for goodness-of-fit rejected

the null hypothesis, we assessed whether a negative binomial distribution described the data

better. To do this we fit both geometric and negative binomial distributions by maximum

likelihood estimation and calculated Akaike Information Criterion (AIC) values for model

comparison.

Results:

Experimentally measured heterogeneity in biting among individual mosquitoes

The experimental data showed that biting is heterogenous among mosquitoes at the

individual level across temperature conditions. At 32°C, mosquitoes were less likely to bite

overall within the timeline of the study (25 days post-emergence) with only 46.2% biting at least

once (12/26), and 34.6% biting twice or more (9/26). At 28°C, all (26/26) mosquitoes bit at least

once while 84.6% (22/26) bit twice or more. At 24°C, 77.8% (21/27) of the mosquitoes bit at

least once and 59.3% (16/27) bit at least twice. The average time between bites was shortest,

the average total number of bites higher, and the average daily number of bites highest at 28°C.

This provide further evidence that temperatures like 28°C are optimum for transmission [11].

Summary metrics for each temperature are given in Table 2. Further, the distributions of total

number of bites, time to first bite, time between first and second bites, and the time between first

and last bites are displayed in Figure 1. We explore the time between bites in detail below.

Table 2: Summary metrics of biting profiles at each temperature.

Temperature	Total # that bit 1+	Total # bit 2+	Empirical average of total bites (range)	Empirical average Time between Bites	Empirical average daily # bites (range)
24	21/27 (77.78%)	16/27 (59.3%)	2 (0-5)	5.99	0.083 (0-0.185)
28	26/26 (100%)	22/26 (84.6%)	3.92 (0-12)	3.41	0.163 (0.038-0.346)
32	12/26 (46.15%)	9/26 (34.6%)	0.96 (0-3)	5.62	0.040 (0-0.154)

Figure 1: Summary of Ae. aegypti *biting characteristics at three temperatures: Total number of bites, time to first bite, time between the first and second bites, and time between first and last bites.*

We measured the contribution of individual bite profiles to the likelihood that 1) a

mosquito becomes exposed and infectious and 2) a secondary (human) infection occurs given

the introduction of a single infectious index case. Not all mosquitoes that bit became exposed

and/or transmitted (**Figure 2**). That is, they either did not bite soon enough to become exposed

by the viremic index case and/or there were no bites after the EIP resulting in subsequent

transmission to susceptible household members. At 28°C, nine mosquitoes that bit at least once

never became exposed; at 24°C, 7 mosquitoes that bit did not become exposed; and 3

mosquitoes that bit at least once at 32°C did not become exposed (**Figure 2A.1-C.1**). On the

other hand, all the exposed mosquitoes who bit at least twice (6/6) subsequently transmitted

with non-zero probability at 32°C. At 24°C, 80% of exposed mosquitoes that bit at least twice

(8/10) transmitted to susceptible individuals; and at 28°C 93.75% (15/16) exposed mosquitoes

that bite twice transmitted to susceptible household members with a non-zero probability as one

mosquito's only subsequent bite was not outside the EIP (**Figure 2A.2-C.2**).

To determine whether these differences in patterns were related to the static parameters

of EIP and human infectious period, we tested the sensitivity of the model to these quantities.

We found that the system was relatively insensitive to small changes in either parameter. The

system was not overly sensitive to changes in EIP, which may be an artifact of initializing the

model in sync with the index case. Overall, changes in probability were modest (Supplemental

Figure S7). At 24°C there was no change in the overall outcome of transmission for any

individual mosquito. At 28°C, only two mosquitoes (#23 and #24) had any change in the overall

binary outcomes of supporting any transmission or not. With an increase in EIP, #23 did not

subsequently transmit after exposure while a decrease in EIP allowed #24 to transmit to the

susceptible human. At 32°C, a lengthening of the EIP by 2 days resulted in #3 and #7 having a

zero probability of transmitting to the susceptible household member (**Supplemental Figure**

S7).

The length of infectious period had more of an impact, likely due to the synchrony of

initialization of the model. Nevertheless, at 24°C, a 2-day shorter infectious period changed the

overall yes/no outcomes of mosquitoes #10 and 23 to a zero probability of mosquito exposure

and subsequent transmission to a susceptible human. (**Supplemental Figure S8**). On the other

hand, increasing the infectious period by 1 or 2 days resulted in 4 additional mosquitoes (#8, #3,

#22, and #25) having successful mosquito exposure. Only two (#13 and #25) also had

subsequent transmission to a susceptible human with a 2-day increase in the infectious period.

At 28°C, three mosquitoes (#11, #12, and #26) went from no transmission to successful

mosquito exposure with +2 days on the infectious period and two (#11 and #12) also resulted in

subsequent transmission to a susceptible human. A shortening of the infectious period resulted

in #8, #17 and #25 having no simulations successfully expose the mosquito. Finally, at 32°C, a

shorter infectious period resulted in #9 not successfully becoming exposed or subsequently

transmitting, while a 2-day longer infectious period resulted in #26 having a non-zero (and

relatively high) probability of exposure (~90%), though this did not result in subsequent

mosquito-to-human transmission (**Supplemental Figure S8**). Finally, we varied both the

probability of transmission from the index case to the mosquito (Bv) and the probability of

transmission from exposed mosquitoes to susceptible humans (Bh) to determine how this

parameter altered the overall outcomes. The change in probabilities was largely proportional

and did not result in a change in overall yes/no outcomes for any individuals across all three

temperatures (**Supplemental Figure S9-10**).

Figure 2: The proportion of mosquito-specific simulations (y-axes) that resulted in successful mosquito exposure from the index case (A-C.1) and subsequent transmission to the susceptible human (A-C.2) at each temperature: 24°C (A), 28°C (B), and 32°C (C). Color gradients indicate total number of bites per individual that transmitted.

Correlation of biting profile with simulated transmission potential

The correlation between mosquito exposure and subsequent transmission was uniformly

positive, and moderate to high (0.6 at 24°C, 0.90 at 28°C, and 0.73 at 32°C). The total number

of bites was highly and significantly correlated with mosquito exposure and subsequent

transmission across all temperatures (**Table 3**). Time-to-first bite (TTFB) was also significantly

correlated with mosquito exposure across all temperatures but was not significantly correlated

at 32°C for transmission to a susceptible human (**Table 3, Supplemental Figure S6**). These

relationships were inverse, indicating that a shorter time-to-first bite correlates with higher

probability of transmission. While this is likely at least partially due to the model structure,

previous work has also suggested that biting sooner is an important component of transmission

[18].

The time between first and second bites was not significantly correlated with either

transmission event at either 24°C or 28°C but was positively correlated with mosquito exposure

at 32°C (ρ=0.60). However, the time between first and last bites was significantly, positively

correlated with transmission to susceptible household members from exposed mosquitoes

across all temperatures. However, only at 28°C and 32°C was this significantly correlated with

the initial mosquito exposure itself (**Table 3, Supplemental Figure S6**).

Table 3: The significantly correlated variables associated with successful exposure/ transmission demonstrate
that factors associated with successful transmission are not uniform across all temperatures.
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Event	Temp	Variable	Rho
Mosquito exposure (IC -> Ms)	24°C	Time to 1st bite	-0.86
		Total # Bites	0.38
	28°C	Time to 1st bite	-0.84
		Total # Bites	0.66
		Time b/t 1st and last bites	0.72
	32°C	Time to 1st bite	-0.94
		Total # Bites	0.66
		Time b/t 1st and last bites	0.83
		Time b/t 1st and 2nd bites	0.70
Mosquito to Susceptible Human Transmission (Me->Hs)	24°C	Time to 1st bite	-0.61
		Total # Bites	0.59
		Time b/t 1st and last bites	0.72
	28°C	Time to 1st bite	-0.75
		Total # Bites	0.84
		Time b/t 1st and last bites	0.86
	32°C	Total # Bites	0.74
		Time b/t 1st and last bites	0.86

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We determined which bites most contributed to transmission out of all successful

transmission simulations for both IC->Ms and Me->Hs. To show the hypothetical role of each

bite within an individual biting profile, we calculated the proportion of transmission events

attributable to a particular bite out of all simulations that resulted in a transmission event (not

all bites) (**Figure 3**). In most cases, the first bite did account for the majority of that mosquito's

transmission when exposure was limited to a single index case as in our model, especially at

32°C where 100% of mosquito exposure occurred upon the first bite. At 28°C and 24°C, there

was more variability in the bite at which a mosquito was exposed, though the first bite still

accounted for more than other bites.

Figure 3: Experimental bite data from mosquitoes shows heterogeneity among individuals. Bolded cells represent days on which a mosquito bit. **Left Column:** During the first 8 days of simulation, gradations of red represent the proportion of IC->Ms events established by that bite out of total bites resulting in IC-Ms transmission. **Right column:** For days 9-25, gradations of red represent the proportion of exposed mosquito to susceptible human events established by that bite out of total bites resulting in Me->Hs transmission.

336   <u>Analytic computation of mosquito transmission:</u>

337   The probability that the mosquito is successfully exposed (p_{ME}) and the probability that

transmission to the second human occurs (p_{HE}) can be analytically derived as above (Eq. 1-2).

With perfect transmission ($\beta_v = 1$), the p_{ME} is entirely determined by the number of bites that

an individual takes during the human infectious period. And once the number of bites taken

during the human infectious period is accounted for, p_{HE} is largely determined by how many

bites occur EIP days after a successful mosquito exposure. The outcomes of the simulations

(Figure 2) are in agreement with these calculations (**Supplemental Figure S11**). In parallel, we

counted the pairs of bites that occurred at least one EIP apart, which relaxes the assumption of

asynchrony. This is because it quantifies, for a given profile, how many pairs out of the total

bite-pair combinations could theoretically support transmission. This measure is an

approximation of the vectorial capacity equation, which is an extension of the Ross-Macdonald

framework [15, 43]. Figure 4 demonstrates that at 28°C more mosquitoes have a greater

likelihood of transmission due to the timing of pairs of bites based on the temperature-

dependent EIP. At 24°C and 32°C, especially, there are fewer overall opportunities to transmit

given the timing of pairs of bites, while at 28°C there is much more variability and overall

opportunity (**Figure 4**). This demonstrates that even when the assumption of synchrony

between the introduction of the infectious index case and mosquito bite timeline, the potential

for consequences on transmission directly related to bite heterogeneity exist.

Figure 4: Left: At each temperature, the number of pairs of bites for each mosquito (x-axis) that are separated by at minimum the temperature dependent EIP. Right: Two representative mosquitoes from 28°C with high frequency of biting demonstrating the number of bites occurring after the temperature dependent EIP (histogram) corresponding to each bite (black and white bars below).

Null model assumptions about biting do not represent the observed process

We investigated whether the distribution of time-between-bites is consistent with the

commonly assumed distribution: ~EXP(λ_{Temp}) after calculating λ_{Temp} as the reciprocal of the

sample mean of time between bites (0.167 for 24°C, 0.293 for 28°C, and 0.178 for 32°C). We

compared the experimental and theoretical times-between-bites and found that there was a

significant difference at 28°C, but not at 24°C or 32°C (**Figure 5, row 1**). We then simulated

1000 hypothetical bite profiles at each temperature using λ_{Temp} and compared the proportion of

mosquitoes that bit at least once and the proportion that bit at least twice between the

experimental and simulated mosquito populations. There was no difference between the

empirical data and theoretical population at 28°C in the proportion that bit at least once while

both the 24°C and 32°C empirical data had significantly lower proportions (**Table S1).** When we

compared the proportion that bit at least twice, the empirical data had significantly lower

proportions compared to the simulated population at all temperatures (**Table S2**).

We then compared the distributions of several key metrics found to be correlated with

transmission (Table 3). At all temperatures, the distribution of the total number of bites was

different between the empirical and simulated populations (**Figure 5, row 2**). The time to first

bite was not significantly different for 32°C, again due to overall lack of biting. This metric was

significantly different for 24°C and 28°C (**Figure 5, row 3).** Finally, at only 28°C was the time

between first and last bites significantly different, though again the distributions at 24°C and

32°C are visually different and this is again likely due to an overall relative dearth of biting at

these temperatures. For 28 degrees, we also assessed whether an alternative discrete

distribution with additional parameter, the negative binomial, fit the time-between-bites data

better. The negative binomial distribution has a much lower AIC score (difference of -16.79)

than the geometric distribution (analogous to a discretized exponential distribution), which

provides strong support that this 2-parameter distribution describes the data better than the

single parameter geometric distribution.

Table 4: Comparison of distributions shows that negative binomial is a better fit to the 28°C bite data (time between
bites) compared to the geometric.

Figure 5: Distributions of metrics describing biting from the experimental data (white bars) to data simulated under the null assumptions of ~EXP(λ_{Temp}) (grey bars). Top Row: Time between bites. Second Row: the total number of bites. Third Row: the time to first bite. Bottom Row: Time between first and last bites. Distribution differences were tested usng the Chi-square goodness of fit and NS indicates not significant.

Discussion

388   The impact of using individual-level data rather than population averages (such as for

389 compartmental models) remains to be elucidated, but it is important to note that the simple act

of biting twice does not translate to a mosquito becoming part of a transmission cycle because

the timing of bites is important [18]. *Ae. aegypti* biting in the field was estimated as 0.63-0.76

bloodmeals per day (a population estimate) within a single gonotrophic period, while our data

suggests this may be lower for some temperatures (Table 2) [7]. Further, different descriptors of

the timing of bites were significantly correlated with transmission for each temperature,

indicating that a more macro-level heterogeneity across temperature exists as well. However, at

this level, it is important to note that biting in the wild and ultimately DENV transmission is

affected by a multitude of environmental factors [15, 33, 44].

When we perturbed the model by varying several parameters, the most sensitive

variable was the infectious period of the index case, which is likely attributable to the synchrony

of initialization. Subsequent exploration of the relative timing of pairs of bites (Figure 5)

displayed the number of potential secondary bites quantified, and with this an estimate of an

individual vectorial capacity-like quantity can be determined to quantify transmission potential

[15, 43]. Further, this shows that supports previous findings that earlier bites are more likely to

contribute to transmission [18]. The general lack of sensitivity of the system to changes in EIP is

also due to the timing of pairs of bites in the empirical data relative to the length of the EIP,

while varying reductions in both βv and βh translated to expected reductions in the probability of

transmission. But again, we observed biting in isolation while there is certainly an interplay

between mortality, biting, and other factors [15, 35, 43]. In addition, the IBM herein provides a

framework with which to test the interaction of biting with these other parameters.

While assessing the goodness of fit of the $\sim\mathrm{EXP}(\lambda_{\mathrm{TEMP}})$, we determined that at the

optimum temperature of 28°C, this distribution did not adequately describe the time between

bites. Nor did the temporal measures of mosquito bite profiles simulated from $\sim\mathrm{EXP}(\lambda_{28°C})$

represent the observed empirical distribution of total bites, the distribution of time to first bites, or

time between the first and last bites (all of which were significantly correlated with successful

transmission outcomes). We show that a negative binomial distribution better fits the time

between bites at 28°C, suggesting that models describing optimum conditions should consider

this alternative fit, or a similar continuous distribution such as a gamma distribution. At 24°C and

32°C, while the ~EXP(λ_{TEMP}) could not be ruled out as an appropriate distribution, there are clear

quantifiable differences observable and likely this is due to the low number of overall bites

compared to 28°C.

We recognize caveats of the study. First, laboratory settings represent ideal conditions

for mosquitoes, and partial and full bloodmeals are probably a conservative estimate of

mosquito-human contact, as probing can lead to transmission [36]. However, recently published

data demonstrate that while there are likely differences in biting habits across geographic and

laboratory populations, a degree of heterogeneity was observed in a field-derived colony,

indicating that heterogeneity among individuals is not limited to laboratory colonies [36].

Constant temperatures may not capture the nuances of temperature effects, though have been

successfully used for important modeling studies and are important to isolate thermal effects

[11, 24]. In addition, mortality is differential among temperatures, as demonstrated in several

previous studies with obvious implications for transmission and biting [11, 22, 24]. While our

study was not designed to explicitly measure mortality, we did find that the mortality of our three

temperatures did follow the same pattern as previously published, with 28°C having greater

average survival (>25 days) than either 24° or 32° days (24 and 23.5 days, respectively) [11,

24]. Finally, it is important to emphasize that our data are not intended to represent the entire

range of heterogeneity potential among individuals, but to explore the implications of possible

heterogenous biting profiles on theoretical transmission scenarios.

These data indicate that there is considerable heterogeneity in not only total number of

bites, but the timing of bites as well. Further, these metrics were differentially affected by

temperature. Several studies have demonstrated the non-monotonic nature of mosquito life

traits and/or that the optimal temperature for biting is approximately 28°C [11, 17, 24]. Our data

further confirms a thermal optimum for *Ae. aegypti*-driven arbovirus transmission, as well as

suggests that the commonly used representation of biting at this optimal temperature may not

best represent the heterogeneity and associated arbovirus transmission potential.

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<u>Figure & Table Legends:</u>

Table 1: Parameter values for the models were either generated in the Christofferson laboratory
or taken from the literature.

Table 2: Summary metrics of biting profiles at each temperature.

Table 3: The significantly correlated variables associated with successful exposure/
transmission demonstrate that factors associated with successful transmission are not uniform
across all temperatures.

Table 4: Comparison of distributions shows that negative binomial is a better fit to the 28°C bite
data (time between bites) compared to the geometric.

Figure 1: Summary of Ae. aegypti biting characteristics at three temperatures: Total number of
bites, time to first bite, time between the first and second bites, and time between first and last
bites.

Figure 2: The proportion of mosquito-specific simulations (y-axes) that resulted in successful
mosquito exposure from the index case (A-C.1) and subsequent transmission to the susceptible
human (A-C.2) at each temperature: 24°C (A), 28°C (B), and 32°C (C). Color gradients indicate
total number of bites per individual that transmitted.

Figure 3: Experimental bite data from mosquitoes shows heterogeneity among individuals.
Bolded cells represent days on which a mosquito bit. **Left Column:** During the first 8 days of
simulation, gradations of red represent the proportion of IC->Ms events established by that bite
out of total bites resulting in IC-Ms transmission. **Right column:** For days 9-25, gradations of

red represent the proportion of exposed mosquito to susceptible human events established by
that bite out of total bites resulting in Me->Hs transmission.

Figure 4: Left: At each temperature, the number of pairs of bites for each mosquito (x-axis) that
are separated by at minimum the temperature dependent EIP. Right: Two representative
mosquitoes from 28°C with high frequency of biting demonstrating the number of bites occurring
after the temperature dependent EIP (histogram) corresponding to each bite (black and white
bars below).

Figure 5: Distributions of metrics describing biting from the experimental data (white bars) to
data simulated under the null assumptions of ~EXP(λ_{Temp}) (grey bars). Top Row: Time between
bites. Second Row: the total number of bites. Third Row: the time to first bite. Bottom Row: Time
between first and last bites. Distribution differences were tested usng the Chi-square goodness
of fit and NS indicates not significant.

Supplemental Tables & Figures:

Supplemental Figure S1: There was no significant difference (Kolmogorov-Smirnov test) in the
total number of bites at any of the tested temperatures
600
601 **Supplemental Figure S2:** We also saw no significant difference in the distribution of time to
602 first bite across the three temperatures (Kolmogorov-Smirnov test).
603
604 **Supplemental Figure S3:** Similarly, there was no significant difference between biological
605 replicates in the time between first and second bites across all three temperatures (Kolmogorov-
606 Smirnov test).
607
608 **Supplemental Figure S4:** Finally, there was no significant difference between biological
609 replicates in the time between first and last bites across all three temperatures (Kolmogorov-
610 Smirnov test).
611
Supplemental Figure S5: A schematic of a hypothetical mosquito with an individual bite profile
where a bite occurs when M{bite} = 1. Infectious contact from the index case (P_{H0}) at Time = 2
results in the mosquito becoming exposed (M{status} = 1). A bite from the M{status} = 1
mosquito at Time = 5 did not result in transmission to a susceptible individual (S_H) as the EIP
had not concluded. However, after the EIP, the mosquito status changes to infectious (M{status}
= 2) and bites from the mosquito, such as Time = 12 to S_H results in a transmission event and
changes the status to S_H as "unavailable for infection".
619
620 **Supplemental Figure S6.** Complete correlation matrices for all three temperatures. X indicate ρ
621 is not significantly different from 0.
622
623 **Supplemental Figure S7: Local sensitivity analysis for the extrinsic incubation period of**
624 **DENV in the mosquito vector:** The proportion of simulations (y-axis) where a particular
625 mosquito (x-axis) becomes exposed (P(IC—Ms)) or transmits (P(M->H)) per a range of extrinsic
626 incubation periods. Only mosquitoes with at least one scenario resulting in a non-zero
627 probability of transmission are shown.
628
629 **Supplemental Figures S8: Local sensitivity analysis for infectious period of the index**
630 **case:** The proportion of simulations (y-axis) where a particular mosquito (x-axis) becomes

exposed (P(IC—Ms)) or transmits (P(M->H)) per a range of infectious periods. Only mosquitoes with at least one scenario resulting in a non-zero probability of transmission are shown.

Supplemental Figure S9: Local sensitivity analysis for the probability of transmission given contact between a susceptible mosquito and the infectious index case (β_v): The proportion of simulations (y-axis) where a particular mosquito (x-axis) becomes exposed (P(IC—Ms)) or transmits (P(M->H)) per a range of β_v. Only mosquitoes with at least one scenario resulting in a non-zero probability of transmission are shown

Supplemental Figure S10: Local sensitivity analysis for the probability of transmission given contact between an infectious mosquito and susceptible human (β_h): The proportion of simulations (y-axis) where a particular mosquito (x-axis) transmits (P(M->H)) per a range of β_h. Only mosquitoes with at least one scenario resulting in a non-zero probability of transmission are shown.

Supplemental Figure S11: The outcomes of mosquito exposure and subsequent transmission to the susceptible household member derived directly from Equations 1 & 2 (see Methods).

Supplemental Table S1: Comparison of the proportion of mosquitoes that bit at least once between the experimental data and simulated population. There was a significant difference at 24°C and 32°C, using prop.test function in R.

Supplemental Table S2: Comparison of the proportion of mosquitoes that bit at least twice between the experimental data and simulated population. There was a significant difference at across all temperatures, using prop.test in R.