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Carry-over effects of urban larval environment on the transmission potential of a mosquito-borne pathogen

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Abstract

Background: Mosquitoes are strongly influenced by environmental temperatures, both directly and indirectly via carry-over effects, a phenomenon by which adult phenotypes are shaped indirectly by the environmental conditions experienced in previous life stages. In landscapes with spatially varying microclimates, such as a city, the effects of environmental temperature can therefore lead to spatial patterns in disease dynamics. To explore the contribution of carry-over effects on the transmission of dengue virus-2 (DENV-2), we conducted a semi-field experiment comparing the demographic and transmission rates of *Aedes albopictus* reared on different urban land classes in the summer and fall season. We parameterized a model of vectorial capacity using field- and literature-derived measurements to estimate the bias introduced into predictions of vectorial capacity not accounting for carry-over effects.

Results: The larval environment of different land classes and seasons significantly impacted mosquito life history traits. Larval development and survival rates were higher in the summer than the fall, with no difference across land class. The effect of land class on adult body size differed across season, with suburban mosquitoes having the smallest wing length in the summer and the largest wing length in the fall, when compared to other land classes. Infection and dissemination rates were higher in the fall and on suburban and rural land classes compared to urban. Infectiousness did not differ across land class or season. We estimate that not accounting for carry-over effects can underestimate disease transmission potential in suburban and urban sites in the summer by up to 25%.

Conclusions: Our findings demonstrate the potential of the larval environment to differentially impact stages of DENV-2 infection in *Ae. albopictus* mosquitoes via carry-over effects. Failure to account for carry-over effects of the larval environment in mechanistic models can lead to biased estimates of disease transmission potential at fine-scales in urban environments.

Keywords: *Aedes albopictus*; dengue; carry-over effects; urban microclimate

Background

Climate plays an important role in the transmission of mosquito-borne pathogens, determining the geographic range of disease vectors and shaping transmission dynamics [1, 2]. Heterogeneity in environmental conditions can directly shape

¹individual-level variation in traits relevant to mosquito population dynamics [3] and¹
²pathogen transmission [4]. In addition to these direct effects, mosquito phenotypes²
³can be shaped indirectly by the environmental conditions experienced in previous³
⁴life history stages, a phenomenon known as carry-over effects [5]. Carry-over effects⁴
⁵have been documented in a wide-range of species with complex life cycles, such⁵
⁶as amphibians [6], migratory birds [7], and damselflies [8]. Similarly, the mosquito⁶
⁷life cycle is characterized by ontogenetic niche shifts, with a larval aquatic stage⁷
⁸and an adult terrestrial stage. Following these studies, we reason that the thermal⁸
⁹environment a mosquito experiences during its larval stage is likely to have lasting⁹
¹⁰impacts on adult traits, and, ultimately, on transmission potential. ¹⁰

¹¹ Although it has been previously demonstrated that larval environmental temper-¹¹
¹²ature can alter individual mosquito traits important for transmission [9, 10], the net¹²
¹³effect of temperature-mediated carry-over effects on overall transmission potential¹³
¹⁴is ambiguous. Current models of mosquito-borne disease typically only incorporate¹⁴
¹⁵direct effects of temperature, despite evidence that carry-over effects can have large¹⁵
¹⁶impacts on adult phenotypes [11, 12, 13]. Additionally, laboratory studies designed¹⁶
¹⁷to estimate temperature-mediated carry-over effects are often conducted across a¹⁷
¹⁸wider range of temperatures than mosquitoes typically experience in the field [14].¹⁸
¹⁹The studies are not easily “scaled-up” to explain transmission across a landscape¹⁹
²⁰when incorporated into temperature-dependent models of mosquito-borne disease²⁰
²¹[15]. Urban landscapes, in particular, are composed of a variety of microclimates,²¹
²²which can differentially impact mosquito life-history traits leading to heterogeneity²²
²³in vector population dynamics across the landscape [16]. However, it is unknown if²³
²⁴variation in microclimate across an urban area also has implications for carry-over²⁴
²⁵effects of the larval environment on adult phenotypes. ²⁵

²⁶ We hypothesize that relevant environmental variation across an urban landscape²⁶
²⁷during the larval stage will have lasting impacts on adult traits that are important²⁷
²⁸for mosquito population dynamics and pathogen transmission. Further, we predict²⁸
²⁹that failure to account for carry-over effects will result in a biased estimate of vecto-²⁹
³⁰rial capacity, the rate at which future infections arise from one infectious mosquito.³⁰
³¹To estimate the effects of the larval environment in a spatially heterogeneous, ur-³¹
³²ban environment, we conducted a semi-field experiment exploring population and³²
³³dengue virus 2 (DENV-2) transmission relevant life-history traits from *Aedes al-*³³

¹*bopictus* mosquitoes reared in three urban land classes across the summer and fall.¹
²We used a mixture of field-derived and temperature-dependent parameters to con-²
³struct a model of vectorial capacity. Our modeled vectorial capacity was then com-³
⁴pared to a calculation using the experimental grand mean for parameters affected⁴
⁵by carry-over effects in order to estimate the bias introduced by not including these⁵
⁶indirect effects.⁶

⁸Methods⁸

⁹We conducted our semi-field experiment across an urban gradient in Athens, GA in⁹
¹⁰the summer and fall of 2016. To explore the effects of microclimate variation across¹⁰
¹¹an urban landscape, we used an impervious surface map (National Land Cover¹¹
¹²Database 2011 [17]) to select three replicate sites ($30m \times 30m$) each of low (0-5%),¹²
¹³intermediate (6-40%), and high (41-100%) impervious surface. Percent impervious¹³
¹⁴surface is an accurate predictor of land surface temperature, particularly for urban¹⁴
¹⁵landscapes [18], and allowed us to ensure our sites exhibited the full range of urban¹⁵
¹⁶microclimates. To select our sites, we calculated the percent impervious surface¹⁶
¹⁷of each $30m \times 30m$ pixel using a moving focal window of $210m \times 210m$, as the¹⁷
¹⁸surrounding impervious surface can affect the microclimate in the pixel of interest.¹⁸
¹⁹We then classified each pixel based on the mean impervious surface within its focal¹⁹
²⁰window, with 0 - 5 % representing low, 6 - 40 % representing intermediate, and²⁰
²¹41 - 100% representing high. Because impervious surface is an effective classifier²¹
²²of urban land classes [19], we identified the sites as rural, suburban, and urban²²
²³with low, intermediate, and high impervious surface scores, respectively. Final site²³
²⁴selection was constrained by access and permissions, however, the final distribution²⁴
²⁵of sites was chosen to ensure all sites were at least 2 miles from others of the same²⁵
²⁶land class, and were interspersed across the study area (Fig. 1).²⁶

²⁷Within each site, we evenly distributed four plastic trays (Sterilite, 13.625" x²⁷
²⁸8.25" x 4.875"), each containing 100 first instar *Ae. albopictus* larvae and 1L of leaf²⁸
²⁹infusion. Leaf infusion was prepared as described in Murdock et al. [16]. Briefly, 80²⁹
³⁰g live oak (*Quercus virginiana*) leaves and 3 g of 1:1 yeast:albumin mixture were³⁰
³¹infused in deionized water. Trays were screened with a fine mesh, placed in a wire³¹
³²cage to deter wildlife, covered with clear plastic vinyl to keep rainwater from enter-³²
³³ing, and placed in full shade. We added deionized water to trays after two weeks to³³

maintain a total water volume at 1L. We placed data loggers (Monarch Instruments:
 Radio Frequency Identification (RFID) Temperature Track-It Logger) in vegetation
 next to each tray, approximately 3 feet above the ground. Data loggers recorded in-
 stantaneous temperature and relative humidity at ten minute intervals throughout
 the study period. Data loggers were also placed in the trays to measure the lar-
 val, aquatic temperature, however three and 17 loggers (of 36) failed due to water
 damage in the summer and fall, respectively. Of loggers that did not fail during the
 experiment, water temperatures were highly correlated with ambient temperatures
 $(\rho = 0.929)$; thus, only ambient temperatures are used as an approximation of lar-
 val environmental temperature. Sites were visited daily to collect emerging adults
 until all larvae had emerged or died (Summer Replicate: Aug. 1 to Sept. 3, 2016,
 Fall Replicate: Sept. 26 to Nov. 8, 2016). We quantified the total number of adults
 emerging per day, and recorded the sex and wing length of each emerged adult.
 Adult females were collected to use in vector competence assays.

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Dengue virus *in vitro* culturing and mosquito infections

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DENV-2 stock was obtained from the World Reference Center for Emerging Viruses
 and Arboviruses at the University of Texas Medical Branch (PRS 225 488, origi-
 nally isolated from human serum in Thailand in 1974 [20]). We propagated virus by
 inoculating Vero (African green monkey kidney epithelial) cells with a low MOI in-
 fection. Virus-containing supernatant was harvested when the cells exhibited more
 than 80% cytopathic effect. Supernatant was cleared of cell debris by centrifugation
 (1000xg, 1 min), aliquoted into cryo-vials, and stored at -80 °C. We quantified vi-
 ral titers of virus stock using TCID₅₀ assays, calculated by the Spearman-Kärber
 method [21, 22]. When mixed 1:1 with the red blood cell mixture, the final concen-
 tration of virus in the blood meal was 3.540×10^6 TCID₅₀/mL.

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Adult mosquitoes were collected as they emerged from trays, aggregated by site,
 and stored in reach-in incubators at $27^\circ\text{C} \pm 0.5^\circ\text{C}$, $80\% \pm 5\%$ relative humidity,
 and a 12:12 hour light:dark cycle. To ensure infected mosquitoes were of a simi-
 lar age, mosquitoes were pooled into cohorts of 4-6 days old in the summer and
 4-9 days old in the fall (due to slower and more asynchronous emergence rates).
 Mosquitoes were allowed to mate and fed *ad libitum* with a 10% sucrose solution.
 Forty-eight hours prior to infection, the sucrose was replaced with deionized wa-

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ter, which was then removed 12-14 hours before infection to encourage feeding.¹
 Infectious blood meals were administered to mosquitoes through a water-jacketed²
 membrane feeder and consisted of 47% human red blood cells washed in DMEM³
 (vol/vol), 1% sucrose(weight/vol), 20% FBS (vol/vol), 5 mM ATP, and 33% DMEM⁴
 medium combined with 1 mL of virus stock [23]. Blood-fed female mosquitoes were⁵
 then maintained as described above for the duration of the experiment.⁶

For a mosquito to become infectious, arboviruses must pass through multiple¹⁰
 tissues that impose significant barriers to infection, namely the midgut and salivary¹¹
 glands [24]. Therefore, we assessed mosquitoes for infection, dissemination, and¹²
 infectiousness through salivation assays and tissue dissections 21 days post infection¹³
 [25]. First, mosquitoes were cold anesthetized and immobilized by removing their¹⁴
 legs and wings. Wings were mounted on a glass slide to measure wing length from the¹⁵
 distal end of the alula to the apex of the wing via a dissecting scope and micrometer.¹⁶
 The proboscis of each female was then inserted into a sterile pipette tip containing¹⁷
 10-20 μ L of FBS (with 3mM ATP and red food coloring) and allowed to salivate on¹⁸
 a plate kept at 27 °C for 15 minutes, after which the salivation media was expelled¹⁹
 into 500 μ L of DMEM and stored at -80 °C. After salivation, we removed the head²⁰
 of each individual and stored the body and head separately at -80 °C.²¹

To determine variation in the proportion of mosquitoes that become infected²⁴
 (bodies positive for virus), disseminated (heads positive for virus), and infectious²⁵
 (saliva positive for virus), we used cytopathic effect (CPE) assays to test for the²⁶
 presence of virus in each collected tissue [22]. Individual bodies and heads were²⁷
 homogenized in 500 μ L of DMEM and centrifuged at 2,500 rcf for 5 minutes. 200²⁸
 μ L of homogenate was added to Vero cells in a solution of DMEM (1% pen-strep,²⁹
 5% FBS by volume) in a 24-well plate and kept at 37 °C and 5 % CO_2 . Salivation³⁰
 media was thawed and plated on Vero cells as above. After 5 days, Vero cells were³¹
 assessed for presence of DENV-2 via CPE assays. Samples were identified as positive³²
 for virus if CPE was present in the well.³³

¹Intrinsic growth rates (r') and vectorial capacity (VC)

²We calculated the per capita population growth rate per tray following Livdahl and²

³Sugihara [26] Eq. 1:

$$r' = \frac{\ln(\frac{1}{N_0} \sum_x A_x f(\bar{w}_x))}{D + \frac{\sum_x x A_x f(\bar{w}_x)}{\sum_x A_x f(\bar{w}_x)}} \quad (1)$$

⁷ Where N_0 is the initial number of female mosquitoes (assumed to be 50% of the⁷
⁸larvae, 50), A_x is the number of mosquitoes emerging on day x , D is the time to⁸
⁹reproduction following emergence (assumed to be 14 days [27]), and $f(\bar{w}_x)$ is fecun-⁹
¹⁰dity as a function of mean wing size on day x (\bar{w}_x ; Equation 2). This relationship¹⁰
¹¹is assumed to be linear and calculated via Lounibos et al. [28]:¹¹

$$f(\bar{w}_x) = -121.240 + (78.02 \times \bar{w}_x) \quad (2)$$

¹⁵ While it is possible to reason how changes in each parameter will result in carry-¹⁵
¹⁶over effects that individually affect disease transmission, determining the overall net¹⁶
¹⁷effect and magnitude of the change is less straight forward. Therefore, we calculated¹⁷
¹⁸the vectorial capacity (VC; Equation 3) for each site and season using a modified¹⁸
¹⁹temperature-dependent dengue calculation defined in Mordecai et al. [29] to create a¹⁹
²⁰quantitative estimate of the influence of carry-over effects on disease transmission.²⁰
²¹Using the experimental mean for field-derived parameters affected by carry-over²¹
²²effects (fecundity and vector competence), we calculated an additional site-level²²
²³VC to serve as an estimate of this value when *not* accounting for site-specific carry-²³
²⁴over effects.²⁴

$$VC(T) = \frac{a(T)^2 b(T) c(T) e^{-\mu(T)/EIR(T)} EFD(T) p_{EA}(T) MDR(T)}{\mu(T)^2} \quad (3)$$

²⁷ Here, mosquito traits are a function of temperature, T , as described in Table 1.²⁷
²⁸Site-level VC was calculated using a combination of traits empirically measured²⁸
²⁹in this study and traits estimated from thermal response models as described in²⁹
³⁰Mordecai et al. [29]. The bite rate ($a(T)$), adult mosquito mortality rate ($\mu(T)$), and³⁰
³¹extrinsic incubation rate ($EIR(T)$), were calculated for mosquitoes at a constant³¹
³²27 °C using temperature dependent functions from Mordecai et al. [29], to match³²
³³the adult environment used in the experiment. Vector competence ($b(T)c(T)$) was³³

¹calculated as the proportion of infectious mosquitoes per site as determined by¹
²our DENV-2 infection assays. The number of eggs produced per female per day²
³($EFD(T)$) was calculated by estimating fecundity from average female wing length³
⁴following Eq. 2, and then dividing this by the expected lifespan of mosquitoes ($1/\mu$).⁴
⁵The egg-to-adult survival probability ($p_{EA}(T)$) was defined as empirically measured⁵
⁶egg-to-adult survival probability (the average proportion of adult female mosquitoes⁶
⁷emerging per site). The mosquito immature development rate ($MDR(T)$) was cal-⁷
⁸culated as the inverse of the mean time to emergence for female mosquitoes per site,⁸
⁹resulting in a daily rate of development. To estimate bias introduced by not includ-⁹
¹⁰ing carry-over effects, we compared our site-level calculated VC to one calculated¹⁰
¹¹using the experimental grand mean for site-level EFD and bc . All other parameters¹¹
¹²were the same across the two models. 12

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¹⁴Statistical Analysis 14

¹⁵We used generalized linear mixed models (GZLMs) to explore if microclimate (i.e. ¹⁵
¹⁶mean, minimum, maximum, and daily ranges of temperature and relative humidity),¹⁶
¹⁷egg-to-adult survival (the proportion of adult females emerging per tray), larval¹⁷
¹⁸development rate (1/days to emergence), female body size, per capita growth rate,¹⁸
¹⁹and metrics of vector competence (i.e. infection, dissemination, and infectiousness)¹⁹
²⁰differed across land class and season. In all models, fixed effects included land class,²⁰
²¹season, and their interaction, with site as a random effect. The effect of body size on²¹
²²infection dynamics was also explored at the level of the individual mosquito, fitting²²
²³a binomial GZLM including wing size as a fixed effect and site as a random effect.²³
²⁴Vectorial capacity was calculated at the site-level, and so a two-way ANOVA was²⁴
²⁵used to estimate the effect of land class, season, and their interaction, on vectorial²⁵
²⁶capacity. 26

²⁷To confirm the relationship between the categorical variables of land use and²⁷
²⁸season and temperature, we fit additional models containing mean temperature as²⁸
²⁹a covariate to the residuals of the original GZLMs including season and land use²⁹
³⁰as fixed effects. This test explored if there was additional variation in the response³⁰
³¹variable due to temperature that was not explained by land class and season. To³¹
³²explore if the effect of temperature differed across season, we fit individual GZLMs³²
³³to the above response variables including mean temperature as a covariate. For egg-³³

to-adult survival, larval development, body size, and the per capita growth rate,¹
 mean temperature was calculated over the season at the tray level, and site was²
 included as a random effect. Because mosquitoes were pooled by site for infection³
 assays, temperature was aggregated to the site level and no random effects were⁴
 included for analyses of infection metrics and *VC*.⁵

All analyses were conducted with respect to the female subset of the population,⁶
 as they are the subpopulation responsible for disease transmission. In the case of⁷
 data logger failure ($N = 3$), imputed means from the site were used to replace⁸
 microclimate data. In the case of trays failing due to wildlife tampering (two urban⁹
 and one suburban in the fall replicate), collected mosquitoes were used for infection¹⁰
 assays, but trays were excluded from demographic analyses. For all mixed-models,¹¹
 significance was assessed through Wald Chi-square tests ($\alpha = 0.05$) and examination¹²
 of 95% confidence intervals. Pearson residuals and Q-Q plots were visually inspected¹³
 for normality. All mixed models were fit using the `lme4` package in *R*.¹⁴

Results¹⁵

Effects of land class and season on microclimate¹⁶

We found that microclimate profiles differed significantly across both season and¹⁷
 land class (Fig. 2, Table 2). In general, temperatures were warmer in the summer and¹⁸
 on urban sites, replicating what was found in a prior study in this system [16]. We¹⁹
 did observe a significant interaction between season and land use on the mean daily²⁰
 minimum temperature and diurnal temperature range, with no effect of land use on²¹
 these response variables in the summer. Urban sites in the fall were characterized²²
 by significantly higher daily average minimum temperature and smaller diurnal²³
 temperature range relative to rural sites (Table 2). Mean relative humidity was²⁴
 higher in the summer than the fall (mean \pm SE, summer: $27.576\% \pm 0.199\%$, fall:²⁵
 $19.450\% \pm 0.194\%$). In the summer, minimum and mean relative humidity was²⁶
 significantly lower on urban sites compared to rural and suburban sites (Table 2).²⁷
 A similar trend was seen in the fall, with urban sites having lower mean relative²⁸
 humidity compared to other land classes, but no difference in minimum relative²⁹
 humidity (Table 2).³⁰

¹Direct and carry-over effects of land class and season on population growth ¹

²Of the 3,600 first-instar larvae placed in each season, a total of 2595 and 1128 ²
³mosquitoes emerged in the summer and fall, respectively. The total egg-to-adult ³
⁴survival per tray was significantly higher in summer than fall (Table 3, mean \pm ⁴
⁵SE, summer: 0.670 ± 0.158 , fall: 0.297 ± 0.160), but did not differ across land class ⁵
⁶(Fig. 3A, Table 3). The mean rate of larval development per tray was significantly ⁶
⁷different between summer and fall (Fig. 3B, Table 3), with daily development rates ⁷
⁸of 0.074 ± 0.002 SE and 0.0387 ± 0.002 SE, respectively. There were no significant ⁸
⁹differences in larval survival or development rates across land class. We did not ⁹
¹⁰observe a significant carry-over effect of land class or season on mosquito wing size, ¹⁰
¹¹however there was a significant interaction between the two (Table 3). We found a ¹¹
¹²significant difference in wing size across season for mosquitoes on rural sites only, ¹²
¹³with larger bodied mosquitoes in the summer (mean \pm se: 2.451 ± 0.054), than ¹³
¹⁴the fall (2.300 ± 0.052). While urban mosquitoes tended to be larger in the fall, ¹⁴
¹⁵and suburban mosquitoes tended to be larger in the summer, these effects were not ¹⁵
¹⁶significant. ¹⁶

¹⁷After incorporating the number of adult females emerging per day, the day of ¹⁷
¹⁸emergence, and their body size into the per capita growth rate equation (Eq. 1), we ¹⁸
¹⁹found that the estimated per capita growth rate was higher in the summer season ¹⁹
²⁰than the fall season (Fig. 3C, Table 3, mean \pm SE, summer: 0.135 ± 0.005 , fall: ²⁰
²¹ 0.068 ± 0.006), with no difference across land class. The effect of temperature within a ²¹
²²season was only significant for egg-to-adult survival, and differed in direction across ²²
²³season (mean $\beta \pm$ SE, summer: -0.328 ± 0.148 , fall: 0.368 ± 0.135 , Table S1). This ²³
²⁴mirrors a trend for the effect of land class on egg-to-adult survival to differ across ²⁴
²⁵season (Table 3). When controlling for land class and season, temperature explained ²⁵
²⁶no additional variation for any response variable (Table S2). ²⁶

²⁸Carry-over effects of land class and season on vector competence ²⁸

²⁹A total of 319 female mosquitoes were assessed for infection status, 20 per site in ²⁹
³⁰the summer and varying numbers per site in the fall due to lower emergence rates ³⁰
³¹(sample sizes reported in Table 4). Carry-over effects of the larval environment ³¹
³²on infection status were limited to infection and dissemination rates. We found ³²
³³that land class and season did significantly impact the probability of a mosquito ³³

¹becoming infected and disseminating dengue infection (Table 3). Both metrics were¹
²higher in the fall compared to the summer replicate, with urban sites having the²
³lowest infection and dissemination rates across both seasons (Fig. 4A, B). While³
⁴there was a trend for a higher proportion of mosquitoes becoming infectious in the⁴
⁵summer (Fig. 4C), this was not significant ($\chi^2 = 3.63$, $p = 0.057$). The probability⁵
⁶of becoming infectious did not differ across land class, nor season (Fig. 4C, Table 3),⁶
⁷despite the higher probability of mosquito infection and dissemination in the fall,⁷
⁸and on suburban and rural sites. Similarly, there was no effect of temperature on⁸
⁹any infection metric within a season (Table S1), and temperature did not explain⁹
¹⁰any additional variation after controlling for land class and seasons (Table S2). This¹⁰
¹¹suggests that the ability of virus to escape the midgut and invade the salivary glands¹¹
¹²differs in adults reared in the summer vs. the fall and across land class, with a higher¹²
¹³proportion of dengue infected mosquitoes becoming infectious in the summer and¹³
¹⁴on urban sites (Table 4, $\chi^2 = 13.65$, $p < 0.001$). We also found the probability of¹⁴
¹⁵infection to decline with increasing body size ($\chi^2 = 4.776$, $p = 0.0289$), although¹⁵
¹⁶there was no evidence for a relationship between body size and the probability of¹⁶
¹⁷dissemination or infectiousness.¹⁷

¹⁹Integrating direct and carry-over effects into estimates of transmission potential¹⁹

²⁰We found VC to be higher in the summer (mean: $5.847 \pm 0.0768SE$) than the fall²⁰
²¹($0.252 \pm 1.087SE$) (Fig. 5, Table 3). In the summer season, there was a trend for²¹
²² VC to increase with increasing urbanization (Fig. 5). This trend was not signifi-²²
²³cant, however, given the small sample size ($n=9$) and the disproportional impact²³
²⁴of having no infectious mosquitoes at one site, resulting in a value of $VC = 0$ for²⁴
²⁵one sample. There was no effect of temperature on VC within a season (Table S1),²⁵
²⁶and temperature did not explain any additional variation after controlling for land²⁶
²⁷class and season. When comparing VC calculations using field-based or grand mean²⁷
²⁸estimates of EFD and bc , we found that the effect of land class and season were²⁸
²⁹not significantly different (land class: $\chi^2 = 0.381$, $p = 0.826$, season: $\chi^2 = 1.408$,²⁹
³⁰ $p = 0.235$), suggesting that the omission of carry-over effects in calculations did not³⁰
³¹lead to biased estimates of relative VC in different seasons or land classes. However,³¹
³²the use of the grand mean did lead to an underestimate of VC on some suburban³²
³³and urban sites in the summer, with a two-fold decrease in predicted VC (Fig. 5,³³

¹Supp. Fig. 1). The calculated *VC* for rural sites in the summer and across all land¹
²classes in the fall more closely resembled the grand mean calculated *VC*.²

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⁵**Discussion**⁵

⁶Mathematical models of mosquito-borne disease rarely include mosquito larval⁶
⁷stages [15], and of those that do, few include the influence of carry-over effects⁷
⁸on important mosquito life-history traits (but see [30]). This is likely because there⁸
⁹are relatively few empirical studies parameterizing carry-over effects in mosquito-⁹
¹⁰pathogen systems [2], and most are laboratory studies conducted across a wider¹⁰
¹¹range of temperatures than those seen in the field. Here, we demonstrate that¹¹
¹²fine-scale differences in larval microclimate across land class and season generate¹²
¹³carry-over effects on adult fecundity and vector competence. When integrated into a¹³
¹⁴model of vectorial capacity, we find that vectorial capacity differs across season, but¹⁴
¹⁵not land class. Further, failure to account for site-specific carry-over effects across¹⁵
¹⁶urban land classes results in biased estimates of vectorial capacity, underestimating¹⁶
¹⁷potential disease transmission in urban areas.¹⁷

¹⁸ The subtle heterogeneity in microclimate we observed across season resulted in¹⁸
¹⁹significantly different predicted population growth rates through its effects on de-¹⁹
²⁰mographic traits. Daily mean temperatures (25.43 °C) across all sites in the summer²⁰
²¹were closer to the predicted thermal optimum of *Ae. albopictus* (24-25 °C) [29] than²¹
²²in the fall (17.69 °C), leading to higher egg-to-adult survival rates. We also observed²²
²³more rapid larval development rates in the summer relative to the fall. This is likely²³
²⁴due to the strong positive relationship observed between development rates and²⁴
²⁵mean larval temperature, as the metabolic rate of mosquitoes will increase with²⁵
²⁶warming temperatures [3]. Temperature explained no additional variation in any²⁶
²⁷response variable after accounting for land class and season, suggesting that our²⁷
²⁸coarser characterizations of land class and season contain the temperature varia-²⁸
²⁹tion necessary to predict changes in demographic and infection rates. Additionally,²⁹
³⁰we only found an effect of temperature within a season for egg-to-adult survival (Ta-³⁰
³¹ble S1). While we did not find a significant influence on many traits, our trends do³¹
³²agree with a previous study in this system that found lower egg-to-adult survival on³²
³³urban sites [16]. The variation in mean temperature across land class in our study³³

¹was very small ($< 1^{\circ}\text{C}$), and we expect these relationships would be magnified in ¹
²mega-cities that can have urban heat island effects of up to 6°C [31]. ²
³ Surprisingly, we found no main effect of land class or season on female mosquito ³
⁴body size, despite the difference in temperatures across season. Following allomet- ⁴
⁵ric temperature-size relationships of ectotherms, warmer larval temperatures should ⁵
⁶lead to smaller bodied mosquitoes [32]. However, contrary to predictions generated ⁶
⁷from the allometric temperature-size relationship, we observed mosquitoes on rural ⁷
⁸sites to be larger in the summer despite the fact that all land classes were cooler in ⁸
⁹the fall relative to the summer. Our results contrast with many laboratory studies ⁹
¹⁰that have found a negative relationship between rearing temperature and mosquito ¹⁰
¹¹body size (*Ae. albopictus* [33], *Culex tarsalis* [34], *Anopheles gambiae* [35]). How- ¹¹
¹²ever, these studies all used a constant temperature treatment, while mosquitoes in ¹²
¹³our field-based study experienced fluctuating temperatures. Among studies using ¹³
¹⁴fluctuating temperatures, there is mixed evidence for a relationship between rear- ¹⁴
¹⁵ing temperature and mosquito body size [16, 36]. Larger temperature fluctuations ¹⁵
¹⁶at the more extreme temperatures (cool and warm) can lead to counterintuitive ¹⁶
¹⁷effects of temperature on organismal traits if these temperatures approach or cross ¹⁷
¹⁸the thermal maximum or minimum (at which trait performance is zero) and in- ¹⁸
¹⁹duce thermal stress [37, 38]. Rural sites in the fall did experience a larger average ¹⁹
²⁰diurnal range of temperatures than in the summer, suggesting this differential ef- ²⁰
²¹fect of temperature fluctuations at thermal extremes could be acting on body size. ²¹
²²Our findings demonstrate that, while the use of fluctuating temperatures in studies ²²
²³of mosquito life-history traits is relatively new, these fluctuations can have signifi- ²³
²⁴cant impacts on mosquito ecology and should be integrated in lab-based studies of ²⁴
²⁵mosquito vectors to more closely approximate field conditions. ²⁵
²⁶Our results agree with laboratory studies in other arboviral systems (chikungunya ²⁶
²⁷[39], yellow fever [39], and Rift Valley fever [40]) that found cool larval environmental ²⁷
²⁸temperatures to enhance arbovirus infection relative to warmer larval environments. ²⁸
²⁹Studies in the *Ae. albopictus*-dengue virus system have also found that low larval ²⁹
³⁰temperatures enhance mosquito susceptibility to viral infection, although this is ³⁰
³¹dependent on larval nutrition [10] and the stage of the infection (i.e. mid-gut vs. ³¹
³²dissemination vs. saliva) [9]. While we found infection and dissemination to decrease ³²
³³with increasing temperatures across season and land class, there was no effect on ³³

¹viral presence in the saliva, suggesting carry over effects due to microclimate vari-¹
²ation may alter the overall efficiency of dengue infection. Thus, even though a²
³smaller proportion of mosquitoes reared on urban sites and in the summer became³
⁴infected and disseminated infection, these mosquitoes were more likely to become⁴
⁵infectious, resulting in no net difference in overall vector competence across land⁵
⁶class and season. Larval environmental temperature may differentially impact later⁶
⁷stages of viral infection (i.e. salivary gland penetration) compared to earlier stages⁷
⁸(i.e. midgut escape) through effects on mosquito physiology and immunity, as well⁸
⁹as on important tissue barriers to infection [4, 39, 41, 42].⁹

¹¹Current models of vector-borne disease focus primarily on direct effects of en-¹¹
¹²vironmental variables on mosquito densities and disease transmission and rarely¹²
¹³include the effects of the larval stage, either directly or via carry-over effects [15].¹³
¹⁴While we found carry-over effects due to seasonal and urban environments to have¹⁴
¹⁵a significant impact on virus infection and dissemination, we found no net effects on¹⁵
¹⁶saliva positivity for the virus. Therefore, when incorporating parameters into cal-¹⁶
¹⁷culations of vectorial capacity, we did not find a significant difference in predicted¹⁷
¹⁸vectorial capacity due to land class. However, we did find VC to be higher in the¹⁸
¹⁹summer relative to the fall, driven by differences in demographic rates such as larval¹⁹
²⁰survival and development rates rather than differences in adult vector competence.²⁰
²¹Unfortunately, given the logistical limitations imposed by a field experiment set-²¹
²²ting, we were unable to measure additional life-history traits important for disease²²
²³transmission in conjunction with vector competence. Lab studies have found that²³
²⁴factors such as adult longevity [43], biting rate [44], and pathogen extrinsic incuba-²⁴
²⁵tion period [45, 46] are also be impacted by carry-over effects. For example, warmer²⁵
²⁶larval temperatures correspond with decreased adult longevity in mosquitoes [43],²⁶
²⁷and including this relationship could mediate the seasonal differences in VC found²⁷
²⁸in our study, with decreased adult longevity in the summer corresponding to de-²⁸
²⁹creased VC . Less is known about traits specific to transmission such as biting rate²⁹
³⁰and EIP, which have only been investigated in response to larval diet and compe-³⁰
³¹tition [44, 45, 46]. Carry-over effects of the larval environment can act on multiple³¹
³²adult phenotypes, often in conflicting ways, and the net effect of this on disease³²
³³transmission has yet to be fully explored.³³

Carry-over effects are not simply limited to microclimate, and can result due to variation in larval nutrition [44], intra- and inter-specific densities [47], and predation [30] in mosquito systems. Further, abiotic and biotic factors will likely interact to influence carry over effects [10, 48], and this interaction could be scale-dependent [49]. For example, biotic processes are predicted to be more important at local geographic scales, while abiotic processes dominate at regional geographic scales in species distribution models [50]. Future exploration of the scale-dependent contribution of different environmental factors and their interactive influence on both direct and carry-over effects is needed to improve models predicting the distribution of mosquito vector species, mosquito population dynamics, and disease transmission.

Conclusions

We found fine-scale variation in microclimate across season and urban land class to shape *Ae. albopictus* population dynamics and arbovirus transmission potential through direct effects on larval survival and development rates, and indirectly through carry-over effects on vector competence and fecundity. DENV-2 infection and dissemination rates were higher in mosquitoes from rural and suburban land classes than urban ones, and were higher in the fall compared to the summer. However, there was no difference in overall infectiousness. Therefore, the seasonal differences in VC we observed were due to the direct effects of the larval environment on egg-to-adult survival and development rates, rather than carry-over effects. When comparing VC to a calculated VC that did not account for site-specific carry-over effects, we found that not accounting for carry-over effects results in an underestimate of predicted VC in suburban and urban sites in the summer, and an overestimate in the fall.

The interaction between the larval and adult environments, mediated by carry-over effects, could have complex consequences for adult phenotypes relevant to disease transmission for mosquitoes as well as other organisms. Given the devastating impact of disease in other species with complex life histories (e.g. chytridiomycosis in amphibians), carry-over effects in disease transmission are important, though understudied, mechanisms that must be better understood to control disease spread. Incorporating relationships between carry-over effects and organismal life-history traits into statistical and mechanistic models will lead to more accurate predic-

tions on the distributions of species, population dynamics, and the transmission¹
of pathogens and parasites. Mosquito-borne disease incidence is spatially heteroge-²
neous in urban areas [51], and a better understanding of both the larval and adult³
environments, including their interaction, could improve the accuracy of fine-scale⁴
predictions of disease incidence across a city. ⁵

Declarations

Ethics approval and consent to participate

⁸Not applicable.

⁹

Consent for publication

¹⁰Not applicable.

¹¹

Availability of data and material

¹²The datasets and code used in during the current study are available in the figshare repository, LINK.

¹³

Competing interests

¹⁴The authors declare that they have no competing interests.

¹⁵

Funding

¹⁶This work was supported by the University of Georgia (Presidential Fellowship, College of Veterinary Medicine,

¹⁷Department of Infectious Diseases) the National Science Foundation Graduate Research Fellowship, and the

National Science Foundation Research Experiences for Undergraduates (Grant No. 1156707). The funders had no

role in study design, data collection and analysis, decision to publish, or preparation of the manuscript.

¹⁹

Author's contributions

²⁰MVE, JMD, and CCM designed the experiment. MVE, JCS, and NS conducted the field-work portion of the

experiment. MVE and MAB designed and conducted the infection portion of the experiment. MVE, JMD, and CCM

²¹conducted statistical analyses. MVE, JMD, and CCM were involved in original draft preparation and all authors were

²²involved in reviewing and editing. All authors read and approve the final manuscript.

²³

Acknowledgements

²⁴We thank members of the Murdock and Brindley labs for discussion and technical support conducting viral assays.

We thank Diana Diaz, Abigail Lecroy, and Marco Notarangelo for assistance in the field and lab.

²⁵

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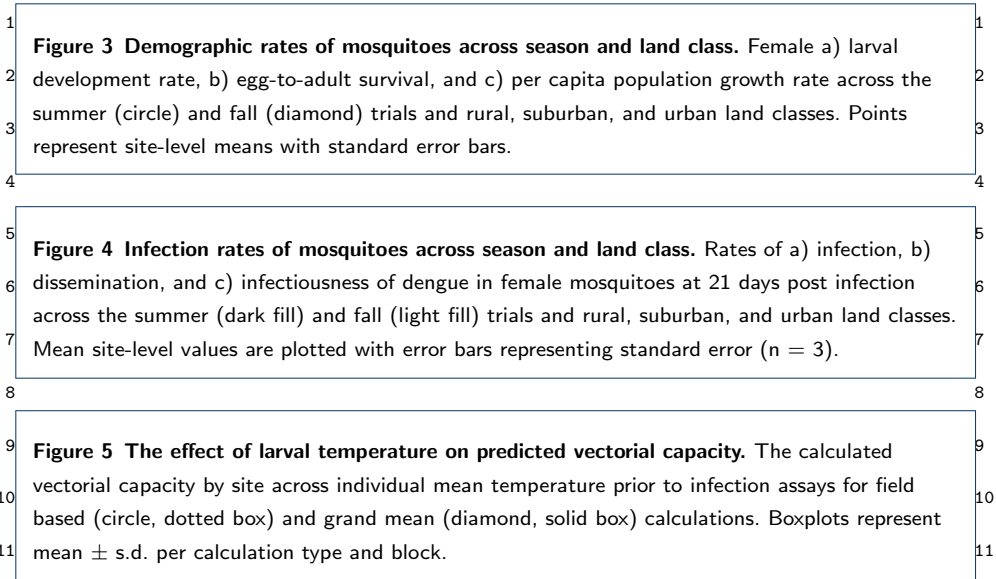
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Figures

Figure 1 Map of study sites in Athens, GA. Inset illustrates location of Athens-Clarke County (black outline) in the state of Georgia. Symbols represent land classes (square: rural, circle: suburban, triangle: urban). Colors represent the amount of impervious surface within the 210m focal area of each pixel, as illustrated on the color bar on the bottom.

Figure 2 Temperature and relative humidity across season and land class. The solid line represents the mean temperature and relative humidity across trays in each land class. The dotted lines represent the mean minimum and maximum temperature and relative humidity across trays in each land class.



Tables

Table 1 Parameters used in the VC calculation. Parameters sourced from Mordecai et al. 2017 [29] were mathematically estimated at a constant temperature of 27 °C, the temperature at which our adult mosquitoes were housed. Parameters that included carry-over effects are starred.

Parameter	Definition	Source
$a(T)$	Per-mosquito bite rate	Mordecai et al. 2017
$b(T)c(T)^*$	Vector competence	Current Study
$\mu(T)$	Adult mosquito mortality rate	Mordecai et al. 2017
$EIR(T)$	Extrinsic incubation rate (inverse of extrinsic incubation period)	Mordecai et al. 2017
$EFD(T)^*$	Number of eggs produced per female mosquito per day	Current Study
$p_{EA}(T)$	Egg-to-adult survival probability	Current Study
$MDR(T)$	Larval development rate	Current Study

Additional Files

- Additional file 1 — SuppVCDiff.pdf
- Additional Figure 1. Bias in VC due to not accounting for site-level carry-over effects across land class and season.
- Additional file 2 – SupplementalTables.pdf
- Supplementary Tables S1, S2.

Table 2 Mean microclimate values across season and land class. 95% confidence intervals are listed in parentheses. Superscripts represent differences as measured by pair-wise comparison using Tukey multiple comparison of means, adjusting for significance with the Holm-Bonferroni method.

	Summer			Fall		
	Rural	Suburban	Urban	Rural	Suburban	Urban
Min. Temperature	21.726 (20.926,22.525) ^a	21.997 (21.198,22.797) ^a	22.667 (21.867,23.466) ^a	11.031 (10.231,11.83) ^b	12.231 (11.432,13.031) ^{bc}	13.411 (12.611,14.211) ^c
Mean Temperature	27.577 (27.132,28.021) ^a	27.381 (26.937,27.826) ^a	27.451 (27.007,27.896) ^a	19.45 (19.005,19.894) ^b	19.547 (19.103,19.991) ^b	19.951 (19.507,20.396) ^b
Max. Temperature	31.533 (30.763,32.302) ^a	30.86 (30.091,31.629) ^a	31.399 (30.63,32.168) ^a	27.567 (26.798,28.336) ^b	26.58 (25.811,27.35) ^b	26.846 (26.077,27.615) ^b
Daily Temperature Range	9.807 (8.507,11.107) ^a	8.863 (7.563,10.163) ^a	8.732 (7.432,10.032) ^a	16.536 (15.236,17.836) ^b	14.349 (13.049,15.649) ^{bc}	13.435 (12.135,14.735) ^c
Min. Relative Humidity	73.49 (69.39,77.59) ^{ab}	76.29 (72.19,80.39) ^a	67.403 (63.303,71.503) ^b	47.676 (43.576,51.776) ^c	48.835 (44.735,52.935) ^c	44.143 (40.043,48.243) ^c
Mean Relative Humidity	89.006 (86.232,91.779) ^{ab}	90.382 (87.609,93.155) ^a	84.428 (81.655,87.201) ^b	75.388 (72.614,78.161) ^c	75.567 (72.794,78.34) ^c	69.005 (66.232,71.778) ^d
Max. Relative Humidity	31.533 (30.763,32.302) ^a	30.86 (30.091,31.629) ^a	31.399 (30.63,32.168) ^a	27.567 (26.798,28.336) ^b	26.58 (25.811,27.35) ^b	26.846 (26.077,27.615) ^c
Daily Humidity Range	26.458 (22.065,30.851) ^a	23.69 (19.296,28.083) ^a	30.978 (26.585,35.371) ^a	51.686 (47.292,56.079) ^b	50.094 (45.701,54.487) ^b	47.628 (43.235,52.021) ^b

Table 3 GZLM model results of land class, season and their interaction on demographic and infection rates. Significance was assessed via Wald Chi-square tests ($\alpha = 0.05$).

	Class			Season			Class*Season		
	df	χ^2	p-value	df	χ^2	p-value	df	χ^2	p-value
Egg-to-Adult Survival	2	0.0361	0.982	1	61.129	<0.001	2	5.891	0.0526
Development Rate	2	3.847	0.1461	1	597.51	<0.001	2	3.108	0.2184
Wing Length	2	0.8348	0.6587	1	2.7937	0.0946	2	14.748	<0.001
Per Capita Growth ('r)	2	0.667	0.717	1	219.84	<0.001	2	2.622	0.23
Infection	2	18.168	<0.001	1	12.271	<0.001	2	1.985	0.371
Dissemination	2	14.253	<0.001	1	14.909	<0.001	2	0.941	0.625
Infectiousness	2	1.105	0.575	1	3.63	0.057	2	0.302	0.86
Vectorial Capacity	2	0.161	0.922	1	5.721	0.017	2	0.905	0.636

Table 4 Dengue infection rates. The rates of infection (mosquitoes with dengue positive bodies), dissemination (infected mosquitoes with dengue positive heads) and infectiousness (infected mosquitoes with dengue positive saliva) across season and land class. Raw numbers of positive samples are shown with denominator in parentheses.

Season	Land Class	No. infected (n)	No. disseminated (n)	No. infectious (n)
<i>Summer</i>				
	Rural	22 (56)	19 (60)	6 (60)
	Suburban	32 (57)	26 (57)	10 (57)
	Urban	10 (51)	10 (53)	7 (53)
<i>Fall</i>				
	Rural	32 (50)	30 (50)	3 (47)
	Suburban	28 (43)	25 (41)	3 (43)
	Urban	26 (59)	22 (57)	4 (59)