#### Climate, human population density, and land cover link the distributions of two globally important dengue vectors from local to continental scales

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#### Abstract

The distributions of mosquito vectors are expected to shift with rising temperatures due to climate change. But other global change patterns, like land cover change and human population growth, are simultaneously occurring. How will these changes interact to shift the future distributions of these vectors? Here, we analyzed how climate, land cover, and human population density regulate and predict habitat for two mosquito species, Aedes aegypti and Ae. albopictus, the primary vectors of dengue, Zika and chikungunya. We asked the following questions: How do environmental response curves derived from vector occurrence data compare to lab-derived responses? Based on these environmental response curves, which environmental drivers best predict the spatial distribution of each vector? Are environmental responses derived from large-scale (continental) occurrence data consistent at fine spatial scales? To answer these questions, we analyzed 6,317 Ae. aegypti occurrence records, 3,629 Ae. albopictus records, 10 satellite-derived environmental covariates, and two independent field surveys cover 134 sites. We found close agreement in the range of lab and environmental temperature responses, though the mean of observed temperatures was higher in the environment (31.0 °C for Ae. aegypti, 29.1 °C for Ae. albopictus) than lab predictions of the thermal optimum for transmission (29.1 °C for Ae. aegypti, 26.4 °C for Ae. albopictus). Using presence-only species distribution modeling approaches, we found that human population density was the best predictor for each vector's spatial distribution (explaining 68.4% of model performance for Ae. aegypti, 48.7% for Ae. albopictus). These patterns were consistent in the field for presence/absence Ae. aegypti data (0.71 AUC, 0.80 recall), but failed to predict Ae. albopictus distributions in the sites we surveyed (0.53 AUC, 0.20 recall). In this session, we will explore these results and discuss the potential to predict and monitor Aedes habitat using satellite data.

## Introduction

The distributions of mosquito vectors are expected to shift with rising temperatures due to climate change. But other global change patterns, like land cover change and human population growth, are simultaneously occurring. How will these changes interact to shift the future distributions of these vectors? Here, we analyzed how climate, land cover, and human population density regulate and predict habitat for two mosquito species, Aedes aegypti and Ae. albopictus, the primary vectors of dengue, Zika and chikungunya.

#### **Research questions**

1. How do environmental response curves derived from vector occurrence data compare to lab-derived responses? 2. Based on these environmental response curves, which drivers best predict the spatial distribution of each vector? 3. Are environmental responses derived from large-scale (continental) occurrence data consistent at fine spatial scales?

## Methods

Using species distribution models, we analyzed 6,317 Ae. aegypti occurrence records, 3,629 *Ae. albopictus* records, 10 satellite-derived environmental covariates, and two independent field surveys over 134 sites. Occurrence records were aggregated from two global datasets. All covariates were measured at or resampled to 1 km<sup>2</sup> grid cells. We derived pixel-wise mean, variance, and skewness statistics from all MODIS temperature (°C) and TRMM precipitation measurements (mm mo<sup>-1</sup>) over 2002-2017, capturing 16 years of spatial and temporal climatic variation. Land cover covariates included tree cover (%) and the mean and variance of leaf area index (LAI; m<sup>2</sup> m<sup>-2</sup>) from 2002-2017. We included just one human population covariate, population density (people ha<sup>-1</sup>), which was natural log-transformed. We used the target group selection method to generate a sampling bias map. We assumed 0 that neither the vectors nor the vector sampling efforts were uniformly distributed. Suitability was therefore calculated relative to a statistically derived null distribution of locations where people live to account for these biases



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Temperature ʹ°C)

## Precipitation (mm/month)

Land cover  $(LAI: m^2/m^2)$ (cover: %)

Population density (people/ha)

![](_page_1_Figure_14.jpeg)

![](_page_1_Picture_15.jpeg)

C) Mexico City

![](_page_1_Figure_17.jpeg)

D) Port-Au-Prince

![](_page_1_Picture_19.jpeg)

E) São Paulo

![](_page_1_Figure_21.jpeg)

0 200 km

![](_page_1_Figure_23.jpeg)

#### Results

*Ae. albopictus* tended to occur in areas with slightly lower mean temperatures (mean = 29.1 °C) than *Ae. aegypti* (mean = 31.0 °C), and these mean observed values are slightly higher than lab-derived predictions of dengue transmission optima (26.4 °C for *Ae. albopictus*, 29.1 °C for *Ae. aegypti*). *Ae. albopictus* records were not significantly different from the background mean temperature (p = 0.738), but were significantly different from *Ae. aegypti* distributions (p < 0.001). Both vectors were observed in areas of higher population density than would be predicted by randomly sampling human populations, and in areas with low temperature variance, in areas skewed towards higher extreme temperatures, and with high variance in leaf area index (all distributions tested for stochastic equality using the two-sided, nonparametric Brunner-Munzel test, all with p < 0.001).

For Ae. aegypti, we found that human population density alone best predicted habitat suitability (AUC  $\mu$  = 0.853,  $\sigma$  = 0.003), followed by temperature (AUC  $\mu$  = 0.765,  $\sigma$  = 0.002), land cover (AUC  $\mu$  = 0.750,  $\sigma$  = 0.002), and precipitation (AUC  $\mu$  = 0.741,  $\sigma$  = 0.010). We found similar trends for *Ae. albopictus*, where population density best predicted habitat suitability (AUC  $\mu$  = 0.853,  $\sigma$  = 0.010), followed closely by temperature (AUC  $\mu$  = 0.842,  $\sigma$  = 0.002), land cover (AUC  $\mu$  = 0.783,  $\sigma$  = 0.007), and precipitation (AUC  $\mu$  = 0.743,  $\sigma$  = 0.005).

Population density was the most important covariate for each vector based on both percent contribution (68.4% for *Ae. aegypti*, 48.7% for *Ae.* albopictus) and permutation importance (72.5% for Ae. aegypti, 53.1% for Ae. albopictus). The next three most important covariates (based on percent contribution) were tree cover (15.5%), mean temperature (5.5%) and precipitation variance (4.0%) for *Ae. aegypti*, and temperature variance (20.3%), tree cover (17.0%) and precipitation variance (7.0%) for Ae. albopictus. These final models performed better than any of the independent model runs reported above (cross-validation AUC  $\mu$  = 0.897,  $\sigma$ = 0.002 for Ae. aegypti,  $\mu$  = 0.917,  $\sigma$  = 0.002 for Ae. albopictus). The joint spatial predictions cluster around population centers, with differences between vector distributions driven by vector-specific responses to the underlying covariate data.

	Aedes aegypti			Aedes albopictus	
	Percent contribution	Permutation importance	Variable	Percent contribution	Permutation importance
	68.4	72.5	Population density	48.7	53.1
	15.5	0.6	Temperature - variance	20.3	20.2
	5.5	6.1	Land cover - tree cover	17	1.8
	4	9.3	Precipitation - variance	7	11.6
	2.9	2.1	Land cover - LAI mean	2.1	1.7
e	2	3.6	Precipitation - mean	1.4	6
	0.7	2.3	Land cover - LAI variance	e 1.3	2.4
	0.7	2.9	Temperature - mean	1.2	1.1
	0.2	0.6	Temperature - skew	0.7	0.9
	0.1	0	Precipitation - skew	0.3	1.1