#### ROLE OF VECTOR PHENOTYPIC PLASTICITY IN DISEASE TRANSMISSION AS ILLUSTRATED BY THE SPREAD OF DENGUE VIRUS BY *AEDES ALBOPICTUS*

Dominic Brass, Christina Cobbold, Bethan Purse, David Ewing, Amanda Callaghan, Steven White



UK Centre for Ecology & Hydrology

Understand patterns in Aedes albopictus driven dengue risk



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This requires understanding differences in global vector dynamics



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This requires understanding differences in global vector dynamics

We need a model that makes predictions that generalise between environments



[1] Lencioni, V. et al. Multi-year dynamics of the Aedes albopictus occurrence in two neighbouring cities in the alps. The European Zoological Journal **90**, 101–112 (2023).

[2] Gouagna, L. C. et al. Strategic approach, advances, and challenges in the development and application of the SIT for area-wide control of Aedes albopictus mosquitoes in Reunion island. Insects 11, 1–24 (2020).

# THE IDEA

#### Accounting for densitydependence



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# THE IDEA

#### Accounting for densitydependence

Traits important for disease transmission exhibit delayed density-dependence

Larval conditions **Temperature** Competition Adult traits Wing length



Wing length Fecundity Survival



# THE IDEA

Accounting for densitydependence

Traits important for disease transmission exhibit delayed density-dependence

We consider the effect of developmental plasticity on disease dynamics

#### Definition - Phenotypic plasticity

The ability of a single genotype to produce multiple phenotypes when exposed to different environmental conditions



# MODELLING AEDES Albopictus

Developed a stagephenotypically structured system of delay-differential equations

#### The Systematic Formulation of Population Models for Insects with Dynamically Varying Instar Duration

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FFD



LETTER

Phenotypic plasticity as a cause and consequence of population dynamics

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Developed a stagephenotypically structured system of delay-differential equations

Adult population structured by infection status and winglength



Developed a stagephenotypically structured system of delay-differential equations

Adult population structured by infection status and winglength

Use historic experience of larval competition to determine the wing-length of emerging adults



Parametrise reaction norms linking environmental drivers to trait value using laboratory data

	1.00			- 20		20°C				
	0.80-	1_				n	Mean	SE	min	
۵				Cycle	1	11	6.7	0.8	3 a	
	0.60	<u> </u>		Cycle 2		5	8.4	1.6	5 a	
į	0.00	- 1		Cycle	3	2	13.5	4.5	9 a	
	0.40 -			Cycle 4 Cycle 5		1	10.0		10	
;										
			Cycle 6							
	0.20	Cycle 7								
Т	Egg	Ll	L2	L3	L4		Pupae	_	L1-adult	Const.
(°C)	Mean SE	Mean SE	Mean SE	Mean SE	Mean SE		Mean SE	-	Mean SE	Sex ratio
5	$11 \pm 1.3$									
10	$2 \pm 0$									
15	7.4 ± 1.8 a	$5.6 \pm 0.3$ a	$3.3 \pm 0.2$ a	$4.6 \pm 0.2 \mathrm{a}$	13.4 ± 0.8 a		$8.7 \pm 0.6$ a		35.0 ± 0.9 a	47.5%
20	$2.9 \pm 0.4$ b	$3.0 \pm 0.2 \mathrm{b}$	$1.4 \pm 0.2 \mathrm{b}$	$2.1 \pm 0.3 \mathrm{b}$	$4.1 \pm 0.3 \mathrm{b}$		$4.1 \pm 0.2 \mathrm{b}$	•	$14.4 \pm 0.4$ b	43.5%
25	$4.5\pm0.7\mathrm{c}$	$2.1 \pm 0.2$ c	$1.2 \pm 0.2 \mathrm{b}$	$1.2\pm0.1~{ m c}$	$3.3 \pm 0.2 c$		$2.7\pm0.1~{ m cm}$		$10.4\pm0.7\mathrm{c}$	41.0%
30	$6.7 \pm 0.7$ a	$1.4 \pm 0.1 \mathrm{d}$	$1.3 \pm 0.1 \text{ b}$	$1.4\pm0.2\mathrm{c}$	$3.0\pm0.3\mathrm{c}$		$1.9 \pm 0.1  d$		$8.8 \pm 0.6$ d	46.3%
35	$7.1\pm0.8\mathrm{a}$	$1.7\pm0.1~{\rm c}$	$1.2 \pm 0.1 \mathrm{b}$	$2.4\pm0.4\mathrm{b}$	$6.8 \pm 1.1 \mathrm{~d}$		$1.7\pm0.7$		$12.3 \pm 0.7$	66.6%



Parametrise reaction norms linking environmental drivers to trait value using laboratory data

Density and temperature dependent variable time delays



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Density and temperature dependent variable time delays

Relationship between average larval temperature and average food per larvae per day and adult wing length



Parametrise reaction norms linking environmental drivers to trait value using laboratory data

Density and temperature dependent variable time delays

Relationship between average larval temperature and average food per larvae per day and adult wing length



Stage-phenotypically structured delay-differential equations

Input environmental variables

Output population & disease dynamics

<u>No backfitting</u>



# MODEL VALIDATION & RESULTS

Rimini, Italy



#### Rimini, Italy

Temperate climate



Rimini, Italy

Temperate climate

Oviposition activity monitored in 2008





#### ENVIRONMENTAL CUES

Rimini, Italy

Temperate climate

Oviposition activity monitored in 2008



Carrieri, M., Angelini, P., Venturelli, C., Maccagnani, B. & Bellini, R. Aedes albopictus (Diptera: Culicidae) Population size survey in the 2007 Chikungunya outbreak area in Italy. II: Estimating epidemic thresholds. Journal of Medical Entomology (2012).

## TRENTO, ITALY

Oviposition activity monitored over 10 years





Lencioni, V. et al. Multi-year dynamics of the Aedes albopictus occurrence in two neighbouring cities in the alps. The European Zoological Journal 90, 101–112 (2023).

### BOLOGNA, ITALY

Oviposition activity monitored over 10 years (taken from VectAbundance)





Lake Charles, Louisiana



Lake Charles, Louisiana Subtropical climate



Lake Charles, Louisiana Subtropical climate

Adults trapped



Willis, F. S. & Nasci, R. S. Aedes albopictus (Diptera: Culicidae) population density and structure in southwest Louisiana. Journal of Medical Entomology **31**, 594–599 (1994).

Lake Charles, Louisiana Subtropical climate Adults trapped Average wing-length of adults measured



Willis, F. S. & Nasci, R. S. Aedes albopictus (Diptera: Culicidae) population density and structure in southwest Louisiana. Journal of Medical Entomology 31, 594–599 (1994).

### GOUGNA ET AL. (2020)

Saint Paul, Reunion Tropical climate Larvae sampled



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Gouagna, L. C. *et al.* Strategic approach, advances, and challenges in the development and application of the SIT for area-wide control of *Aedes albopictus* mosquitoes in Reunion island. *Insects* **11**, 1-24 (2020).



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# SEIR MODEL

Validate disease dynamics by comparing predictions to historic dengue outbreaks



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Validate disease dynamics by comparing predictions to historic dengue outbreaks

We select plausible introduction scenarios for dengue cases based on case reports



# **SEIR MODEL**

Validate disease dynamics by comparing predictions to historic dengue outbreaks We select plausible introduction scenarios for dengue cases based on case reports

These are often uncertain





# RESULTS

The wing-length distribution of infected and uninfected mosquitoes are different



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Large mosquitoes drive increase in dengue case numbers before peak infection



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Phenotypic plasticity alters disease dynamics



The wing-length distribution of infected and uninfected mosquitoes are different

Large mosquitoes drive increase in dengue case numbers before peak infection

Phenotypic plasticity alters disease dynamics

Trait structure alters disease dynamics



![](_page_46_Figure_2.jpeg)

![](_page_47_Figure_2.jpeg)

![](_page_48_Figure_2.jpeg)

![](_page_49_Figure_2.jpeg)

# CONCLUSIONS

We have produced a globally validated model of mosquito and disease dynamics

![](_page_50_Picture_2.jpeg)

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This can be used to produce accurate predictions of relative disease risk

![](_page_51_Picture_3.jpeg)

# CONCLUSIONS

We have produced a globally validated model of mosquito and disease dynamics

This can be used to produce accurate predictions of relative disease risk

Mosquito trait variation in response to developmental environmental experience alters disease dynamics

![](_page_52_Picture_4.jpeg)

TICKSOLVE Stage-Structured Delay-Differential Equations for *ixodes ricinus* 

![](_page_53_Picture_1.jpeg)

![](_page_53_Picture_2.jpeg)

#### TICKSOLVE STAGE-STRUCTURED DELAY-DIFFERENTIAL EQUATIONS FOR *IXODES RICINUS*

![](_page_54_Picture_1.jpeg)

![](_page_54_Figure_2.jpeg)

### THANKS FOR LISTENING!

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![](_page_55_Picture_2.jpeg)

UK Centre for Ecology & Hydrology

# THE MODEL

 $\boldsymbol{E}_{\boldsymbol{\gamma}}\left(t\right)-Active~eggs$ 

 $\boldsymbol{E}_{D}\left(t\right)-Diapausing~eggs$ 

 $E_{Q}(t) - Quiescent eggs$ 

L(t) – Larvae

 $A_{j}(t)$  – Adults in environmental class j

R – Recruitment terms

M – Maturation terms

- P Survival
- $\delta-\text{Mortality rate}$

 $\tau$  – Stage duration

$$\frac{dE_{\gamma}(t)}{dt} = R_{E_{\gamma}}(t) - M_{E_{\gamma}}(t) - \delta_{E_{\gamma}}(t)E_{\gamma}(t),$$

$$\frac{dE_{D}(t)}{dt} = R_{E_{D}}(t) - M_{E_{D}}(t) - \delta_{E_{D}}(t)E_{D}(t),$$

$$\frac{dE_Q(t)}{dt} = R_{E_Q}(t) - M_{E_Q}(t) - \delta_{E_Q}(t)E_Q(t),$$

$$\frac{dL(t)}{dt} = R_L(t) - M_L(t) - \delta_L(t)L(t),$$

$$\frac{dA_{j}(t)}{dt} = R_{A_{j}}(t) - M_{A_{j}}(t) - \delta_{A_{j}}(t)A_{j}(t), \text{ for } j \in 1, ..., m$$

$$\frac{dI_j(t)}{dt} = M_{A_j}(t) - \delta_{I_j}(t)I_j(t), \text{ for } j \in 1, \dots, m.$$

# **RATE OF DEVELOPMENT**

$$\frac{d\tau_{E_{\gamma}}(t)}{dt}=1-\frac{g_{E_{\gamma}}(t)}{g_{E_{\gamma}}(t-\tau_{E_{\gamma}}(t))},$$

$$\frac{d\tau_L(t)}{dt} = 1 - \frac{g_L(t)}{g_L(t - \tau_L(t))},$$

$$\frac{d\tau_P(t)}{dt} = 1 - \frac{g_P(t)}{g_P(t - \tau_P(t))},$$

$$\frac{d\tau_{\rm EIP}(t)}{dt} = 1 - \frac{g_{\rm EIP}(t)}{g_{\rm EIP}(t - \tau_{\rm EIP}(t))}.$$

### **SURVIVAL EQUATIONS**

$$\frac{dS_{E_{\gamma}}(t)}{dt} = S_{E_{\gamma}}(t) \left( \frac{g_{E_{\gamma}}(t)\delta_{E_{\gamma}}(t-\tau_{E_{\gamma}}(t))}{g_{E_{\gamma}}(t-\tau_{E_{\gamma}}(t))} - \delta_{E_{\gamma}}(t) \right),$$

$$\frac{dS_L(t)}{dt} = S_L(t) \left( \frac{g_L(t)\delta_L(t-\tau_L(t))}{g_L(t-\tau_L(t))} - \delta_L(t) \right),$$

$$\frac{dS_P(t)}{dt} = S_P(t) \left( \frac{g_P(t)\delta_P(t-\tau_P(t))}{g_P(t-\tau_P(t))} - \delta_P(t) \right), \tag{63}$$

$$\frac{dS_{\text{EIP}_j}(t)}{dt} = S_{\text{EIP}_j}(t) \left( \frac{g_{\text{EIP}}(t)\delta_{A_j}(t - \tau_{\text{EIP}}(t))}{g_{\text{EIP}}(t - \tau_{\text{EIP}}(t))} - \delta_{A_j}(t) \right), \text{ for } j \in 1, ..., m.$$

# **TRANSITION FUNCTIONS**

$$\bar{\alpha}(t) = \frac{\int_{t-\tau_P(t)-\tau_L(t-\tau_P(t))}^{t-\tau_P(t)} \frac{F(s)}{L(s)} ds}{\tau_L(t-\tau_P(t))}.$$

-

$$w_j(T_{avg}(t),\bar{\alpha}(t)) = \begin{cases} 1, & \text{if } g(w(T_{avg}(t),\bar{\alpha})(t)) = w_j \\ 0, & \text{otherwise} \end{cases}$$

# RT

$$\begin{split} R_{t-\tau_{\text{EIP}}(t)} &= \left[ \sum_{j=1}^{m} \left( \int_{t}^{t+\tau_{\text{REC}}} \frac{\frac{g_{\text{EIP}}(s)}{g_{\text{EIP}}(s-\tau_{\text{EIP}}(s))} b(s-\tau_{\text{EIP}}(s)) h_{v}(s-\tau_{\text{EIP}}(s)) 2\kappa A_{j}(s-\tau_{\text{EIP}}(s)) S_{\text{EIP}_{j}}(s)}{H_{T}} \right] \\ &\times \left( \int_{s}^{s+1/\delta_{A_{j}}(s)} \frac{b(u)v_{h}(u)H_{s}(u)}{H_{T}} du du ds \right) \right]^{\frac{1}{2}}. \end{split}$$