# *Aedes albopictus* is rapidly invading its climatic niche in France: wider implications for biting nuisance and arbovirus control in Western Europe

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# Abstract

The Asian tiger mosquito, *Aedes albopictus,* is a competent vector of arboviruses, such as dengue. After its introduction into southern Europe, this invasive species has been rapidly spreading as well as causing autochthonous cases of arboviral diseases. Both *Ae. albopictus* presence and potential to transmit arboviruses are facilitated at warm temperatures, hence global warming is expected to favor their presence in temperate regions.

We use a climate- and environmental-driven mechanistic modelling framework to investigate the impact of recent climate change on *Ae. albopictus* range expansion and its potential to transmit dengue in Western Europe. We simulate climatic suitability, adult density and dengue transmission risk which we compare with a large ensemble of entomological and epidemiological observations over the past 20 years. Most importantly, we analyze a novel dataset about colonization of French municipalities to estimate the spread rate of *Ae. albopictus* and compare it with model predictions. Lastly, we analyze the sensitivity of entomological and epidemiological risk to changes in temperature, rainfall and human density.

Distribution of both simulated mosquito populations and dengue transmission risk satisfactorily match entomological and dengue observations for  Western Europe (AUC = 0.90 and 0.75 respectively). While lowlands in southern Europe were already climatically suitable for hosting *Ae. albopictus* around 2010,  Western France, together with large populated cities such as London, Zagreb and Vienna, have only become suitable recently. Importantly, the accelerating colonization of *Ae. albopictus* in France may be approaching its theoretical climatic suitability limit, which could determine future expansion. Area at risk of local dengue transmission has recently expanded from the Mediterranean coasts over northern Spain and  Western France. The sensitivity analysis suggests that climate change may expose medium-sized cities to the highest epidemiological risk; this finding is consistent with recently reported dengue outbreaks in Italy and southern France.

## Introduction

The Asian tiger mosquito *Aedes albopictus* is native to southeast Asian forests (Porretta et al., 2012). Over the past century, it has undergone genetic, phenotypic and behavioral adaptations, which favored its establishment outside its native range (Sherpa et al., 2019), in human habitats and temperate climates. Such features include laying eggs in man-made water reservoirs (Paupy et al., 2009), and producing diapausing eggs, which can withstand prolonged dry (Kraemer et al., 2019) or cold seasons (Batz et al., 2020; Medley et al., 2019). Facilitated by the worldwide shipping of used tires, *Ae. albopictus* successfully colonized areas in all continents except Antarctica (Kraemer et al., 2019; Sherpa et al., 2019).

Together with *Aedes aegypti*, *Ae. albopictus* is a competent vector of arboviral diseases, such as dengue, Zika and chikungunya (Bohers et al., 2024). Their ability to spread and adapt, coupled with their biting nuisance and competence to transmit arboviruses, makes the *Aedes* spp. the costliest among the invasive species (Diagne et al., 2021), with social and economic impacts expected to rise in the future (Roiz et al., 2024). Consequently, investigating and anticipating the spread of *Ae. albopictus* is a compelling issue in global public health (Ryan et al., 2019).

*Aedes albopictus* was likely introduced into Europe multiple times since its first establishment in Albania in 1979 (Sherpa et al., 2019). Following these introductions, *Ae. albopictus* spread like wildfire over south Western Europe. Such spread was associated with sporadic long-distance dispersal events, most likely facilitated by vehicular traffic (Roche et al., 2015; Roques & Bonnefon, 2016). The establishment of *Ae. albopictus* populations, combined with viral introductions by infected travelers returning from endemic countries, resulted in sporadic – but increasing in frequency and intensity – autochthonous cases of arboviruses (Angelini et al., 2007; Manica et al., 2017; Roiz et al., 2015). Worryingly, local cases of dengue have recently increased in Italy and France; 83 autochthonous cases of dengue were reported over southern French departments in 2024 (Santé Publique France, 2024), strongly associated with imported cases from the French Antilles, while 213 local dengue cases were reported in Italy in 2024, with about 199 cases occurring in Fano, in the Marches region (Sacco et al., 2024).

The life cycle of *Ae. albopictus* depends on a number of environmental factors (Cruz et al., 2024; Gizaw et al., 2024). The presence of water reservoirs, either rain-fed or man-fed, is indispensable for hosting the aquatic stages (Paupy et al., 2009). In temperate climates, the photoperiod regulates the

production of diapausing eggs and their hatching (Lacour et al., 2015). As an ectotherm species, the

life cycle of *Ae. albopictus* is primarily regulated by temperature (Mordecai et al., 2019). Being a

species typical of tropical areas, climatic suitability for *Ae. albopictus* is favored at warm

temperatures (Mordecai et al., 2017, 2019; Souza & Weaver, 2024). In addition, warm temperatures

facilitate the transmission of arboviruses, since they reduce the extrinsic incubation period (EIP) in

*Aedes* mosquitoes and accelerate their gonotrophic cycle (Mordecai et al., 2017).

Given the broad interest in anticipating public health risks induced by climate change (Souza &

Weaver, 2024), there is a growing body of literature focusing on simulating the impact of a warming

climate on *Ae. albopictus* (Barman et al., 2024; Caminade et al., 2012; Couper et al., 2021; Garrido

Zornoza et al., 2024; Kraemer et al., 2019; Lamy et al., 2023; Metelmann et al., 2019) and its potential

to transmit arboviruses at continental and regional scales (Blagrove et al., 2020; Caminade et al.,

2017; Colón-González et al., 2021; Mordecai et al., 2017; Ryan et al., 2019; Zardini et al., 2024).

*Aedes albopictus* presence and abundance have been successfully modelled using both statistical

(Kobayashi et al., 2002; Petrić et al., 2021; Ryan et al., 2019; Tran et al., 2020) and mechanistic (Brass

et al., 2024; Da Re et al., 2022; Erguler et al., 2016; Metelmann et al., 2019; Tran et al., 2013)

approaches. Some of these models have been used to investigate historical and future distribution

and abundance of *Ae. albopictus* under different climate emission scenarios. Simulations usually

show a general increase in suitable areas over the temperate fringes of the mosquito's current

distribution. For example, Metelmann et al. (2019) predicted that *Ae. albopictus* could theoretically

spread to England in future, even if the species is not established there yet. Brass et al. (2024)

suggests that recent climatic conditions are already suitable for *Ae. albopictus* in central Europe,

while Zardini et al. (2024) suggest that central Europe may represent the northernmost limit for its

presence.

Dengue transmission risk is usually assessed with the basic reproduction number $R_0$. Some

examples of integrating simulated mosquito density into $R_0$ estimates have already been developed

for dengue (Benkimoun et al., 2021). A multi-model multi-scenario inter-comparison study shows a

northern shift of the dengue transmission belt in temperate Europe for the future (Colón-González et

al., 2021). Another study estimates that, whatever the climate emission scenarios, transmission of

arboviruses will be possible as far north as the Scandinavian peninsula by the 2050s (Ryan et al.,

2019).

While scientific evidence suggests that future climatic conditions will become increasingly suitable
for *Ae. albopictus* and the arboviruses it can transmit, the role of recent global warming in driving the
current range expansion of *Ae. albopictus* is still under debate. As a matter of fact, *Ae. albopictus*
showed remarkable adaptive capabilities, suggesting that its colonization process results from
progressive adaptation of this species to new environments (Batz et al., 2020; Urbanski et al., 2012)
rather than from the enlargement of already suitable areas. Despite the growing body of literature
focusing on the impact of future climate change on the presence and epidemic activity of the Asian
tiger mosquito, only a few studies have analyzed the contribution of recent climate change to its
observed colonization of temperate zones.

In this study, we use a climate-driven mechanistic model (Metelmann et al., 2019) to assess the
impact of recent climate change in determining the presence and the epidemic activity of *Ae.*
*albopictus* in  Western Europe. We first compare simulated mosquito dynamics with observed
presence/absence data for Europe. In addition, we assess the capability of the vector model in
reproducing observed seasonal and interannual variability using aggregated ovitrap data for different
European locations. We employ a novel spatiotemporal dataset of colonized municipalities in
metropolitan France to estimate the relative importance of recent climate change in driving the
observed spread of *Ae. albopictus* at national scale. We then compare $R_0$ dengue model output with
reported autochthonous cases of dengue in Europe. Finally, we conduct sensitivity experiments to
disentangle the relative importance of climatic parameters and human population density on
simulated suitability, vector density and epidemic risk. Based on our findings, we then formulate
recommendations for vector control operators and public health officers in France and Europe.

## Material and methods

### Entomological presence-absence data

118 Adult mosquito occurrence data was derived from the Global Biodiversity Information Facility (GBIF)
119 repository up to 2024 (GBIF.org, 2024). GBIF collects species occurrence data from heterogeneous
120 sources and does not provide absences. Therefore we used the lower resolution European Centre
121 for Disease Control (ECDC) data, available at NUTS3 administrative level, for defining absence of
122 mosquitoes (ECDC, 2024a).

In metropolitan France, we employed the national surveillance data provided by *"Direction générale de la santé"*, Ministry of Health. This dataset includes a comprehensive list of municipalities and the corresponding year of colonization from 2004 to 2024. According to the national surveillance criteria updated in the Decree of the 23$^{rd}$ of July 2019, the Asian tiger mosquito is considered as established in a municipality if one of the following criteria is met: (i) eggs are collected in ovitrap during three consecutive inspections; (ii) larvae or adults were observed in traps in a 150m radius around a positive ovitrap; (iii) larvae, adults or eggs are found in traps placed at a mutual distance larger than 500m (Ministère des Solidarités et de la Santé, 2019).

## Entomological longitudinal data

We based our analysis on ovitrap data from three different sources:

- The VectAbundance repository (Da Re et al., 2024), which is a standardized dataset of ovitrap records from various locations in Italy, France, Switzerland and Albania covering the period from 2010 to 2022. These records have been homogenized by Da Re et al. via temporal downscaling on a weekly basis and spatially aggregated on a 9 x 9 km grid using the median value of all ovitraps included in such grid box for each observation date.

- The Altopictus (a French private vector control operator) ovitrap collection in Rennes (Bretagne region), Bayonne and Saint-Médard-en-Jalles (Nouvelle Aquitaine), Toulouse, Pérols and Murviel-les-Montpellier (Occitanie), based on a varying number of georeferenced traps (between 15 and 44) that were inspected every 1-2 weeks between 2023 and 2024.

- The EID Méditerranée (*"Entente Interdépartementale pour la Démoustication du Littoral Méditerranéen"*, a public vector control operator operating along the French Mediterranean littoral) ovitrap data collection in Provence-Alpes-Côte d'Azur region: in Nice, based on a varying number of georeferenced traps (between 30 and 50) inspected every 2-4 weeks between 2008 and 2023, and in Cagnes-sur-Mer, based on ~15 georeferenced ovitraps inspected every week between 2011 and 2012 (Lacour et al. 2015).

In order to homogenize ovitrap data available from French areas with the VectAbundance data, records have been aggregated using the median value for each date.

## Epidemiological data

We retrieved epidemiological data about local cases of dengue in  Western Europe from ECDC (ECDC, 2024b). Data are available from 2010 to 2022 at NUTS3 administrative level.

## Meteorological and environmental data

The dynamic mosquito model is driven by daily meteorological inputs (Metelmann et al., 2019). We retrieved daily precipitation (mm), daily mean, maximum and minimum temperatures (°C) from two main sources: E-OBS gridded dataset (at 0.1° x 0.1°) and MétéoFrance weather stations. E-OBS data were obtained for the period from 2001 to 2023 (Cornes et al., 2018) and its grid was intersected with the human population density raster at 30" based on the 2015 GPWv4 data for 2015 (Doxsey-Whitfield et al., 2015) over Western Europe (36° N-60°N, 15°W-19°E) to run our *in silico* experiments. Since VectAbundance data are arranged on a larger grid, we used E-OBS at a 0.25° resolution. When cells in this grid covered more than one longitudinal record, we retained only the longest record. For simulation in Nice and Montpellier areas, we retrieved weather data from MétéoFrance weather stations at the Nice and Montpellier airports, respectively.

We derived the photoperiod (in hours) using the *getSunlightTimes* function from the R package *suncalc*, based on the latitude, longitude and date for each input.

## Model description and experiments

The dynamic vector model divides mosquito population into five stages (eggs $E$, diapausing eggs $E_d$, larvae and pupae as juveniles $J$, unfed immature female adults $I$, female adults $A$) and expresses the dynamics of their density (mosquitos/ha) via ordinary differential equations (Figure 1). These equations specify the transition rates from one life stage to the next, the mortality and the fertility rates rely on environmental (photoperiod and human density) and meteorological (rainfall and temperature) variables.

$$\begin{cases} \dot{E} = \ \beta(1-\omega)A - (h\delta_E + \ \mu_E)E \\ \dot{J} = \ h(\delta_E E + \ \sigma\gamma E_d) - (\delta_J + \ \mu_J + \frac{J}{K})J \\ \dot{I} = \ \frac{1}{2}\delta_J J - (\delta_I + \mu_A)I \\ \dot{A} = \ \delta_I I - \mu_A A \\ \dot{E}_d = \ \beta\omega A - h\sigma E_d \end{cases}$$

Development rates of larvae and nymphs $\delta_J$ and of immature adults $\delta_I$ (but not for eggs $\delta_E$) depend on the instantaneous temperature, as well as the mortality rate of eggs $\mu_E$ and of juveniles $\mu_J$ and the fertility rate $\beta$. Adult mortality rate $\mu_A$ instead depends on daily temperature. Production of diapausing eggs is regulated by the parameter $\omega$, which depends on the daily photoperiod, while

hatching of diapausing eggs depends on both photoperiod and mean weekly temperature. Eggs

hatching $h$ and the density dependent mortality rate of larvae and nymphs, regulated by the carrying

capacity $K$, depend on the water availability of both natural and anthropic origin, and are

consequently relying on daily precipitation and fixed human density (For further details about the

vector model see Metelmann et al. 2019 and Table SI1).

We run the model under two settings. In the first case, we focus on the estimation of spatial climatic

suitability: we imposed an initial diapausing egg density = 1 ha$^{-1}$, which we reset after each simulated

year. In the second set of experiments, we focus on population dynamics and associated

parameters (adult density, egg laying or $R_0$), hence diapausing eggs are not reset on a yearly basis.

We approximated the initial state by running a preliminary 10-years spin-up simulation (with initial

diapausing egg density = $10^4$ ha$^{-1}$).

## Spatial suitability

We estimated if a particular location for a given period was suitable for *Ae. albopictus* by using the

$E_0$ index (Metelmann et al., 2019). $E_0$ is defined as the expected number of diapausing eggs at the end

of a year after having initialized a population with just one diapausing egg at start. This value is

averaged (using the geometric mean) over multiple years. If $E_0 > 1$ for a particular site, the growing

rate is positive and we assume the site to be suitable for the establishment of *Ae. albopictus*

population. Conversely, $E_0 < 1$ indicates an unsuitable site.

In order to validate the model, we compared mean $E_0$ values in 2018-2023 with observed presence

(GBIF.org, 2024) and absence (ECDC, 2024a) data for Europe. Observed data was re-gridded at the

same spatial resolution than the EOBS driving climate dataset (0.1° x 0.1°). We then estimated the

spatial skill of the model using the Area Under the curve (AUC) of the Relative Operating

Characteristics (ROC) score (Petrić et al., 2021).

We compared the observed distribution of colonized French municipalities by the "*Direction*

*générale de la santé"* (interpolated at the same spatial resolution of the simulations) with the model

predictions year by year. Simulated establishment of *Ae. albopictus* population was estimated using

the aforementioned criterion ($E_0 > 1$), considering a geometric average over the previous 6 years. We

classified the goodness of the classification using the Cohen's K score (Metelmann et al., 2021). We

also calculated the spread rate of *Ae. albopictus* (km/y) in observations and simulations. Spread rate

was estimated in each grid cell using the thin plate spline regression of the year of colonization and

trend surface analysis (Kraemer et al., 2019; Tisseuil et al., 2016).

## Entomological and epidemiological indices

210  We simulated the abundance of *Ae. albopictus* at each life stage, the number of laid eggs (per day)

211  and the basic reproduction number $R_0$ in Western Europe over the period 2006-23.

212  The number $LE_{it}$ of eggs laid per hectare per day in cell $i$ at time $t$ was estimated using the following:

213
$$LE_{it} = \tilde{\beta}_{it} A_{it}$$

Where $A_{it}$ is the aforementioned simulated density of adult female mosquitoes and $\tilde{\beta}_{it}$ is the

approximated fertility rate dependent on mean daily temperature. In order to compare our

simulations with observed ovitrap data, we cumulated $LE_{it}$ with a moving window of one week

(which is the standard trapping period in the VectAbundance dataset), and normalized both

simulated and observed laid eggs with respect to maximum value of each trajectory within the time

periods of the ovitrap data record. For each time series, we estimated Pearson correlation

coefficients and their statistical significance using a Student t-test.

We computed the basic reproduction number, $R_0$, for dengue in Europe for each cell $i$ and time $t$

using the following formulation (Blagrove et al., 2020; Caminade et al., 2017; Zardini et al., 2024):

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$$R_{0,it} = (a_{it}\,\varphi_i)^2 \frac{A_{it}}{H_{it}} \frac{v_{it}}{v_{it} + \mu_{A,it}} \frac{bB}{\mu_{A,it}r}$$

Where $(a_{it}\,\varphi)$ is the biting rate $a_{it}$ (day$^{-1}$) adjusted for vector preference $\varphi_i$, $\frac{A_{it}}{H_{it}}$ is the vector to host

ratio, $v_{it}$ is the reciprocal of the EIP for dengue (day$^{-1}$), $r$ is the human recovery rate for dengue (day$^{-}$

$^{1)}$, $\mu_{A,it}$ is the aforementioned adult vector mortality rate, $b$ and $B$ are the probability of transmission

of the disease from an infected vector to a susceptible host and *vice versa*. Parameters $a_{it}, v_{it}, \mu_{A,it}$,

depend on temperature, $\varphi_i$ depends on human density (per km$^2$) and $A_{it}$ depends on time as it is an

output of the dynamical model (Table SI1). We considered the number of days for which $R_0 > 1$ (the

length of the dengue transmission risk season, LTS) as a proxy for the epidemic risk. We then

compared observed local cases of dengue in 2010-2022 in Europe (ECDC, 2024b), interpolated at

the same spatial resolution of the simulations, with the simulated duration of the epidemic risk. We

then estimated the skill of our prediction using the AUC ROC.

*Figure 1 - Model outline. In the grey box, the model as depicted by Metelmann et al. (2019): adults lay eggs which will either hatch or go into diapause depending on the photoperiod. Diapausing eggs will exit diapause because of favorable conditions in terms of temperature and photoperiod. Hatching is due to the presence of water, from rainfall or human irrigation. Juvenile survival depends on the presence of water as well. Remaining development, mortality and egg-laying rates depend on temperature. Immature adults become adults after the first (human) blood meal.*

## Sensitivity analysis using the exposure state

To explore the sensitivity of our simulations to rainfall, temperature and population density, we performed a sensitivity analysis by assessing the variation of the model outputs (i.e. suitability, density of adults and LTS) by perturbing historical meteorological inputs for specific locations. First, we selected 6 sites (from the E-OBS 0.25° grid) along two gradients of host density and types of climate. The first 3 sites have a similar climate and different population densities: Paris city center (11,000 hab/km$^2$), Paris inner suburbs (2,000 hab/km$^2$), and Paris outer suburbs (500 hab/km$^2$). The other three selected locations have similar population density but different types of climates according to the Köppen classification (Beck et al., 2023): Montpellier, France, Mediterranean climate "Csa"; Bilbao, Spain, oceanic climate "Cfb"; Augsburg, Germany, humid continental climate "Dfb". For each location we calculated an exposure space (Culley et al., 2016) by perturbing recent historical climate conditions (2019-2023) for temperatures $T_h$ (using and additive disturbance, from $T_1 = T_h - 2$ to $T_2 = T_h + 8$ °C) and precipitation $R_h$ (using a multiplicative disturbance from $R_1 = 0$ to $R_2 = 2R_h$). We then showed simulated suitability, density of adults and duration of the epidemic risk in the 2D exposure space (X-axis average MJJASO temperature in °C, Y-axis total MJJASO rainfall in mm) for the aforementioned locations (more details in Figure SI1).

 *Table 1 - Glossary*

| Term | Definition |
|---|---|
| Epidemic risk | Epidemic risk is defined as the interaction between a hazard (the presence of an infected virus), the exposure (the contact between an infected vector and a human) and the vulnerability (the probability of developing a disease after a contact). Here we measure it as the length of the transmission risk season (LTS), *i.e.* the number of days for which the basic reproduction number $R_0 > 1$. |
| EIP | The extrinsic incubation period is the time needed for a virus to develop in a mosquito until it reaches the salivary glands, making the mosquito infectious after an infected blood meal. |
| Climatic suitability | Capacity of a site to host a stable mosquito population based on its climate, here defined by $E_0 > 1$ for a given period. |
| Autochthonous case | One person infected in Europe, in opposition to an imported case (infected abroad). For dengue, Zika or chikungunya, a person that is found positive to any of these arboviruses without having visited regions where these arboviruses are endemic. |
| Ovitrap | A trap designed to mimic a development site for collecting eggs laid by a local mosquito population to estimate their oviposition activity. |
| (Immature) development site | A small, natural or man-made, water reservoir suitable for the aquatic life cycle of the mosquito (from eggs hatching to larvae, pupae and the emergence of immature adults). |
| ROC AUC | The Area Under the Curve of the Relative Operating Characteristics is a metric that quantifies the agreement between a presence-absence dataset and a modelled continuous variable (such as $E_0$). |
| Cohen's K | It is a metric that quantifies the accuracy of a prediction of a presence-absence model against a presence-absence dataset which makes full use of the confusion matrix (legend in Figure 3). |
| MJJASO | From May to October (seasonal activity of *Ae. albopictus* in most European sites). |
| Thin plate spline regression | An interpolation method of bivariate data. |
| Trend surface analysis | A method to estimate the spatial gradient over an interpolated surface. |
| Exposure space | The space generated by combining perturbed historical meteorological trajectories (details in Figure SI1). |
| NUTS3 | Third administrative level of the Nomenclature of Territorial Units for Statistics of the European Union. |

# Results

## Current climatic suitability for *Ae. albopictus* and risk of autochthonous dengue transmission in Western Europe.

Suitable areas for the establishment of the Asian tiger mosquito, as estimated by $E_0$, correspond to the lowland and coastal areas in Italy, the Iberian Peninsula, central and southern France, and the

Adriatic Balkans (Figure 2a). Climate is also suitable over northern European cities such as Paris,

London, Vienna, Strasburg and Frankfurt. Observed occurrence of *Ae. albopictus* (GBIF.org, 2024)

roughly follows a similar pattern, with several false positives simulated over the inner southern

Iberian Peninsula; such mismatch is also consistent with ECDC absence data (ECDC, 2024a).

Despite this regional discrepancy, the suitability for establishment reproduces well the current

distribution of *Ae. albopictus* in Western Europe (ROC AUC = 0.90).

Areas potentially at risk of dengue transmission correspond to a subset of areas suitable for *Ae.*

*albopictus* (Figure 2b). The highest LTS values are simulated over coastal areas along the

Mediterranean and Adriatic Sea where $R_0 > 1$ for more than 3 months on average (e.g., LTS in

Barcelona is 18.1 weeks, Naples 15.8 weeks, Dubrovnik 14.8 weeks). Over the Atlantic coasts of

France and Spain, this period is shorter and ranges between 1 and 3 weeks per year (e.g. Bordeaux

2.9 weeks). Local transmission of dengue might also be possible in large urban cities such as Zagreb

(LTS = 3.0 weeks), Vienna (LTS = 1.1 weeks) and Paris (LTS = 1.0 week).

Observed autochthonous cases of dengue were mostly reported in southern France, Italy, Croatia

and over the eastern coasts of Spain. The number of reported cases was smaller over Croatia (n=10),

southern Spain (n=6) and Paris where only once local case was reported in 2023, and larger in the

Italian cities of Lodi (n=41), Rome (n=38). Notably, we did not consider the 2024 epidemic in Fano

(n=199; Sacco et al., 2024) since it occurred outside the study period. The performance of LTS to

predict autochthonous cases is moderate (ROC AUC = 0.75).

*Figure 2 - Suitability of Western Europe for the establishment of* Ae. albopictus *and epidemic risk of dengue compared to observations. (a) Geometric average of $E_0$ in 2018-2023, GBIF observed occurrence data (yellow dots) and absence data (transparency layer) from ECDC. (b) LTS, i.e. the annual mean duration of the dengue transmission season in 2010-2022, (defined as the period for which $R_0 > 1$), and autochthonous dengue cases as reported by ECDC at the NUTS3 level. Missing values due to unavailability of weather data are depicted by the gray shading.*

## Temporal analysis of the colonization of France by *Ae. albopictus*

The colonization of France by the Asian tiger mosquito started from the southeastern regions,
namely in Alpes-Maritimes in 2004 (Roche et al., 2015). The model predicts a wider area to be already
suitable for *Ae. albopictus* in 2006 (false positive), around the Mediterranean coasts, the southern
Atlantic coasts and cities like Paris and Lyon (Figure 3a). This spatial mismatch between newly
colonized areas and the simulated environmental niche leads to a poor model performance, with a
Cohen's K of 0.02 (Figure 3e). During the following years, both suitable and colonized areas have
gradually expanded (Figure 3b-c-d). The observed colonization front accelerated roughly from 10 to
20 km/y (median value) in 10 years, while climatically suitable areas expanded at a slower rate, less
than 5 km/y, with an acceleration simulated over the last 5 years (Figure 3f). Therefore, the model
skill increases with time towards higher values of Cohen's K (0.5 in 2014), with a maximum of 0.64
reached in 2018. We also show an overestimation of suitable areas (e.g. false positives) over the
central- Western part of the country and an underestimation (e.g. false negatives) over mountain
fringe areas in Corse, the Alps, the Pyrenees and the Massif Central, but also in Alsace (Figure 3d).

Our findings suggest that (i) a larger geographical domain was already suitable for the establishment of *Ae. albopictus* when he first colonized Nice and parts of Corsica in 2006 and (ii) that *Ae. albopictus* has gradually invaded its theoretical climatic niche over the 2006-2023 period.

*Figure 3 - Comparison between the observed colonization of* Ae. albopictus *in metropolitan France (a-d) and simulated suitability niche ($E_0 > 1$) between 2005 and 2024, (e) model accuracy (Cohen's K) and (f) spread rate (km/year) estimated from both observations and predictions.*

## Estimation of egg laying activity

We simulated egg laying activity for 74 sites in France, Switzerland and the Italian peninsula. We show positive linear correlations between simulations and ovitrap observations for 70 sites (see Figure SI2). 66 out of 74 correlation coefficient values were statistically significant at the 99% confidence interval (p-value < 0.01 based on a Student T-test, Figure 4a). The seasonal cycle of simulated egg abundance was well captured by the model for some locations of the VectAbundance dataset (Figure 4b-e). The lowest correlations are shown over the northeastern part of Italy.

*Figure 4 – Comparison between observed (orange dots) and simulated (blue line) standardized laid eggs rate per day (%). (a) Localization of ovitraps and correlation coefficient values (r) with model predictions. Comparison between simulations and observations for four different locations: (b) Locarno (CH, r: 0.58), (c) San Marino (SM, r: 0.58), (d) Palermo (IT, r: 0.37), (e) Nice (FR, r: 0.66). The other trajectories are reported in Figure SI2. Ovitrap data was mainly derived from VectAbundace (Da Re et al., 2024). Since the normalization is done with respect to the maximum value within the record period, simulation values over 100% are admissible outside the record period (such as in Locarno).*

## Effect of recent climate change conditions on simulated suitability, adult density and dengue epidemic risk

For estimating the impact of recent climate change conditions, we compared two periods: 2006-
2014 ("2010s") and 2015-2023 ("2020s"). By the 2010s, a large part of southern Europe was already
suitable ($E_0 > 1$) for the establishment of the Asian tiger mosquito, with the exception of mountainous
areas (such as the Pyrenees between Spain and France or the Apennines in Italy; Figure 5a). During
the 2020s, an increase in $E_0$ is simulated over Western Europe (Figure 5b). Importantly, 2020s
climatic conditions are simulated to become suitable for the establishment of *Ae. albopictus* over
the Western lowlands of France, over the central plateau of Spain, and in large urban cities such as
London, Vienna and Frankfurt. The suburbs of Paris and Zagreb are also simulated to become
increasingly suitable during the 2020s (Figure 5c).

Changes in simulated adult density (per ha) follow a similar spatial pattern (Figure 5d). There is an

increasing trend in simulated adult density over the southern part of Western Europe and Paris

during the 2020s (Figure 5e). Interestingly, a slight decrease in adult density is simulated over large

areas over the southernmost part of Europe, particularly over southern Spain (Sevilla) and southern

Italy (Naples; Figure 5f). Highest simulated densities of mosquitoes are reached in Barcelona (23,000

adults/ha), Valencia (19,000 adults/ha), Milan and Rome (15,000 adults/ha).

Most of Western Europe is still unsuitable for local transmission of dengue for the 2020s (Figure 5h).

However, the LTS is higher than 2 months on average over most of Italy, south-eastern France, Spain

and Portugal. Largest values (> 3 months) are simulated in coastal cities bordering the

Mediterranean and Adriatic Sea. Simulated LTS for dengue show large values in Italy, Spain and

Portugal (maximum values up to 37 weeks in Sicily and Sardinia, 29 in southern Portugal) and coastal

areas in Croatia (up to 18 weeks) during the 2010s (Figure 5g). New regions at risk (LTS >0) emerge in

2020s over most of central- Western Europe, while it decreases over the warmest regions of the

southern Iberian Peninsula and Italy (consistently with the simulated decrease of adult density).

Importantly, the south Western Nouvelle Aquitaine region in France, parts of central Spain, inner

Croatia and Paris become suitable for autochthonous transmission of dengue during the 2020s

(Figure 5i).

*Figure 5 – Effect of recent climatic conditions on the suitability of* Ae. albopictus *(upper row), simulated adult density during boreal summer (middle row), and on dengue transmission risk (bottom row) over Western Europe. (a) Suitable areas, measured using $E_0$ in 2006-2014, (b) in 2015-2023 and (c) qualitative variation in suitability ($E_0 > 1$). (d) Average simulated adult abundance (per ha) between May and October for the period 2006-14, (b) 2015-23, (c) and simulated changes between 2015-23 and 2006-14. (g) Simulated LTS for dengue (number of days in which $R_0 > 1$) for 2006-14, (h) 2015-23 (i) and simulated changes between both periods.*

Model sensitivity to rainfall, temperature and human population density

By using the sensitivity analysis approach, we assessed the influence of temperature, precipitation

variations and host population density gradients on *Ae. albopictus* dynamics and dengue

transmission risk in Europe. Regarding host density, we considered the three Parisian sites along a

decreasing host population gradient with an oceanic climate "Cfb" of the Köppen classification

(Beck et al., 2023). We mapped the recent climatic evolution of these three areas of Paris with varying

population densities in the exposure space. We observe a 1.3 °C increase in averaged MJJASO

temperature with no clear changes in annual rainfall between 2004 and 2023 in Paris (gray lines on

Figure 6a-i). Both simulated $E_0$ and adult density are larger in the densely populated area of Paris

(direct impact of carrying capacity that depends on rainfall and human host density in the model).

Paris city center has been suitable for the whole study period (Figure 6a), while suburbs became

suitable only during recent years with recent increases in temperature (Figure 6d, g). Concerning the

dependency of adult density to rainfall, contour lines tend to be rounded – rather than straight

vertical – while decreasing in population density (Figure 6b, e, h). This finding suggests that in densely

populated areas rain is less limiting, due to permanent man-made development sites, whereas in

sparsely populated areas, rain becomes a limiting factor. The effect of host density on LTS is more

complex, since a higher host abundance also reduces the vector to host ratio in simulated $R_0$. For

such a reason, the suburbs of Paris are simulated to experience a higher risk of dengue transmission

at higher temperatures (Figure 6i) compared to the city center (Figure 6c).

For climatic variation, we compared three different types of climate on simulated $E_0$, adult density

and LTS while keeping similar population density values (~800 hab/km$^2$). The Mediterranean climate

in Montpellier is highly conducive for the establishment of *Ae. albopictus* population. However, a

gradual decrease in all simulated indicators is shown when MJJASO temperatures exceed about

27°C (Figure 6j, k, l). For Bilbao, a southern European oceanic city, climatic conditions became

suitable for disease transmission in 2023 (red line on Figure 6o). Recent humid continental climate

conditions, as experienced in Augsburg, southern Germany, are still far from being suitable for the

vector and potential risk of dengue transmission (Figure 6r).

*Figure 6 - Exposure spaces generated by perturbing historical (2019-2023) weather trajectories of temperature (°C) and precipitation (mm) in six cities of Western Europe. For readability, the x-axis shows the average temperature from May to October (MJJASO), while the y-axis the cumulative precipitation. For each location, we plot the climatic suitability as measured by $E_0$ (top panel with a black line denoting threshold $E_0 = 1$), simulated density of adult females between May and October (intermediate panel), the length of the dengue transmission season (in days, bottom panel with a red line depicting the threshold $R_0 = 1$). In Montpellier, $E_0$ and LTS are greater than 1 in the whole exposure space. In each figure, we show the climatic trajectory of each site over the last 20 years.*

# Discussion

## Climate change is key to understand the recent spread of the Asian tiger mosquito in Western Europe

The ongoing invasion of *Ae. albopictus* is increasingly raising concerns about the future of tropical diseases in Europe. 2024 represents a record year in terms of dengue cases at global scale (about 14,000,000, mostly reported in Latin America, ECDC, 2024c), as well as autochthonous cases of

arboviruses in metropolitan France (84; Santé Publique France, 2024) and the largest dengue
outbreak ever reported in Western Europe (199 in Fano, Italy; Sacco et al., 2024).

In this context, our study aimed to (i) validate simulations with respect to an ensemble of
entomological and epidemiological observations and (ii) test the hypothesis that recent climate
change affected the establishment of *Ae. albopictus* population and its potential to transmit dengue
in Western Europe. We confirmed the skill of the $E_0$ index (AUC = 0.90) in reproducing spatio-
temporal dynamics of *Ae. albopictus* in Western Europe. The main discrepancy between
observations and model predictions occurs over the southern Iberian Peninsula, where high
suitability is simulated (as in other studies, such as Zardini et al., 2024), though *Ae. albopictus* has
not been reported there yet. Such mismatch could be considered further by using rainfall thresholds
to calculate aridity conditions in our modeling framework. Former studies based on threshold-based
models show that at least 400-500 mm of total annual rainfall is needed for the survival of the Asian
tiger mosquito (Kobayashi et al., 2002; Petrić et al., 2021). However, *Ae. albopictus* can lay its eggs
in man-made water reservoirs, which will play an important role regardless of rainfall amounts
(Paupy et al., 2009; Tran et al., 2013). Alternatively, the absence of mosquitoes in the region may be
explained by the slow process of arrival, adaptation and colonization of the region (Barman et al.,
2024). As a matter of fact, *Ae. albopictus* was first detected in Portugal in 2017 (Osório et al., 2018),
surprisingly late compared to other southern European countries. Such delay might be related to
passive transportation factors and geographical barriers that might have prevented an earlier
introduction of this species in Portugal (Zé-Zé et al., 2024).

Understanding dynamics is key to interpret the observed colonization of *Ae. albopictus* in France, as
well as estimating colonization in the upcoming years. Importantly, the skill-score we used to
estimate the agreement between our climatic suitability model and observations was high in 2018
(Cohen's K = 0.64) but close to zero in 2006 (Cohen's K = 0.02). We propose that the systematic
overestimation of model predictions around 2010 unravels the gradual colonization dynamics of an
already climatically suitable area. Therefore, the improved model performance in recent years
(Cohen's K ~ 0.5, Figure 3e) reflects a progressive overlap between the simulated and the observed
niches, whose conceptual distinction is crucial in forecasting climate effects on ectotherms
(Braschler et al., 2020). In other words, *Ae. albopictus* has rapidly invaded its theoretical
environmental niche since 2010, and its dispersal was very-likely facilitated by passive motor vehicle
transportation (Roche et al., 2015). This suggests that future colonization of *Ae. albopictus* may be

driven by the climate-induced enlargement of the suitable area. However, one cannot overlook
instances of colonization outside these areas (e.g. in Alsace, Figure 3d), where either local
populations may be adapting to colder climates or environmental drivers of suitability are poorly
captured by the model, and which need further investigation. The accelerating colonization of *Ae.*
*albopictus* in France, with a spread rate from 10 to 40 km/year, is qualitatively consistent with
findings by Kraemer et al. (2019), who obtained larger values (100 to 200 km/year) based on high
resolution mosquito occurrence data (Kraemer et al., 2015) for Europe.

## Recent climate change has different impacts on *Ae. albopictus* suitability, its abundance and its potential to transmit autochthonous cases of dengue

Our simulations highlight the expansion of areas suitable for the establishment of *Ae. albopictus* in
Europe, consistently with predictions by other studies (Colón-González et al., 2021). These areas,
located between 40° and 52°N, are either located over large continuous regions (such as the Spanish
plateau and the French  Western Atlantic region) or in isolated urban centers (London, Vienna and
Zagreb). In cities, the urban heat island effect combined with high population density could provide
suitable conditions for *Ae. albopictus*, compensating for inadequate conditions of the surroundings.
The model predicts a higher density of adults compared to other published estimates (Brass et al.,
2024 and Zardini et al., 2024) with a more heterogeneous distribution, e.g., hotspots simulated in
cities (compared to estimates by Brass et al., 2024). The predicted geographical expansion of adult
presence for the 2020s are similar to estimates by Zardini et al. (2024) and more conservative to
other studies (Brass et al., 2024; Petrić et al., 2021) which simulate adult presence in central and
northern Europe. Simulated adult density seems to increase in 2020 rather than spreading to new
areas, notably between 40° and 49°N. Conversely, in a warm Mediterranean region (between 37° and
43°N), corresponding to the southern Iberian Peninsula, southern and central Italy, we estimate a
localized decreases in simulated adult density. Such finding is consistent with our sensitivity
analysis that reveals a decrease in simulated adult at higher temperatures. Such decrease may be
related to summer temperatures exceeding optimal values for the survival of adult mosquitoes
(around 26-27 °C in the model; Metelmann et al., 2019) as well as decreasing precipitation with a
detrimental effect on oviposition. A recent modeling study is consistent with our findings:
temperature extremes can have a beneficial effect on mosquito abundance in temperate climates
(northern Italy) and a detrimental impact in warmer locations (southern Italy; Garrido Zornoza et al.,
2024). Estimating the effect of climate change on the risk of dengue transmission is more
challenging, since one needs to consider the viral recombination inside the ectotherm vector as well

as other important socio-economic and human behavior factors (Morin et al., 2013). At low
temperatures, the extrinsic incubation period (EIP) is usually too long for the vector to become
infectious during his standard life span (about 30 days). At higher temperatures, there is a
competition between shorter EIPs (favoring disease transmission) and mosquito survival (Blagrove
et al., 2020). The vector to host ratio is also an important factor in $R_0$ formulation. It can lead to
counterintuitive outcomes, since higher host density typically leads to increased carrying capacity
(with a saturating effect for very high human density – a factor neglected by the model) and increased
vector abundance, but also lower vector to host ratio values. We found that epidemic risk in middle-
sized cities is more sensitive to temperature increase due to climate change conditions. This
theoretical result is consistent with recent observations. An outbreak of chikungunya was reported
in Ravenna (about 150k hab.) in 2007 (Angelini et al., 2007) and an outbreak of dengue affected Fano
(about 60k hab.) in 2024 (Sacco et al., 2024). Simulations predict suitability for the transmission of
dengue along the Mediterranean coasts and new areas (mainly in France) become suitable in the
2020s. Overall, simulated LTS increased by 3.6 days for the study area; this is qualitatively consistent
with the projections by Colón-González et al. (2021), who estimated that mountainous European
areas (>1500) would have maximum LTS increase of between 1 and 3 weeks in 2006-2099 compared
to 1951-2006 , depending on altitude and host density.

## Caveats and future directions

The model we employed provides a framework for describing environmentally driven mosquito
population dynamics. It is a parsimonious framework that could be refined by incorporating genetic
and phenotypic plasticity (Brass et al., 2024), delayed biological effects (Metelmann et al., 2021), the
distinction between parous and nulliparous females (Tran et al., 2013), human mobility (Barman et
al., 2024; Da Re et al., 2022), non-linear relationship between the carrying capacity of the juvenile
stage and the human-influenced landscape (Lamy et al., 2023). However, this model takes into
account the latitudinal variation of the critical photoperiod that controls the hatching of
overwintering diapausing eggs. This adaptation feature, which triggers egg hatching in spring when
temperatures are more favorable, was observed in North America already twenty years following the
introduction of *Ae. albopictus* (Urbanski et al., 2012). We argue that evolutionary convergence might
have caused the same adaptation behavior in *Ae. albopictus* populations over the European
continent (Lacour et al., 2015). Yet, other types of adaptation could occur, such as an increased
resistance to warm temperatures once they become the limiting factor (Couper et al., 2021), or that

winter diapause might even become unnecessary for the persistence of *Ae. albopictus* populations
in warm temperate regions. Without further experimental studies about adaptation mechanisms,
future predictions of *Ae. albopictus* mostly rely on historical traits and consequently remains
uncertain. To estimate the length of the dengue transmission season, we used a typical $R_0$
formulation (Blagrove et al., 2020; Caminade et al., 2017; Zardini et al., 2024) that is valid at the
equilibrium for fixed parameters, for completely susceptible population. Since dengue is not
endemic to Europe, we can assume the latter condition to be satisfied. However, $R_0$ computation in
the case of time-varying parameters may lead to an overestimation in disease risk (Djidjou-Demasse
et al., 2020). To overcome this limitation, future modeling work could be based on a SEIR
(Susceptible-Exposed-Infectious-Recovered) model to simulate the number of autochthonous
cases.

A major caveat in this study is the mismatch in the spatiotemporal resolution among different data
sources. In the ECDC dataset of dengue transmission cases, the entire NUTS3 region is marked as
'positive', although the transmission event occurred locally (Figure 3d). The same applies to the
dataset of the colonization in France at the municipality scale, thus resulting in a systematic
overestimation of the colonized areas over mountainous areas (in Corsica, the Alps, the Pyrenees,
the Massif Central; Figure 2). Furthermore, observed entomological data may be biased by delays in
reporting by field operators. Conversely, the E-OBS weather dataset was too coarse when comparing
estimated and observed laid eggs trajectories in high environmental gradients of alpine valleys
(Figure 4). In addition, in order to simulate the future dynamics of *Ae. albopictus* using climate model
outputs for different RCP-SSP scenarios, following studies will need to consider these
discrepancies.

## Recommendations for vector control operators and public health officers

• This work would not have been possible without collecting reliable and up-to-date
observation data on mosquito presence and abundance. This highlights the need to maintain
strong and continuous links between surveillance operators and academic researchers in
terms of optimizing protocols and data sharing. Maintaining and expanding these efforts
would ensure the interception of adaptation mechanisms that would lead to drastic system
changes (e.g. local population adaptation) and the improvement of models.

• The gap between the simulated and the observed niches of *Ae. albopictus* may be
progressively closing. Future entomological surveillance should focus on areas projected to
become suitable, independently of the distance with the presently colonized area. Such
areas include, but are not limited to, the north Western French countryside and large,
populated cities between southern England and Western Germany, such as London and
Frankfurt.

• Countries would benefit from prioritizing surveillance and control measures in areas
predicted to be suitable for *Ae. albopictus* or arboviral transmission. This would help refocus
anticipatory measures and avoid dispersing resources over low-risk areas – as in France, for
example, where the Decree of the 23$^{rd}$ of July 2019 defines the entire national territory as at
risk for vector-borne transmission.

• The sensitivity analysis and observations of autochthonous arbovirus cases (Angelini et al.,
2007; ECDC, 2024b; Roiz et al., 2015; Sacco et al., 2024; Santé Publique France, 2024)
suggest to target prevention and control efforts against arbovirus transmission in mediumly
populated areas that were colonized by *Ae. albopictus*.

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