Title: Autumn larval cold tolerance does not predict the northern range limit of a widespread

butterfly species

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Author's Contributions

PT and HMK: conceived the ideas, designed the methodology and wrote the paper. PT
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All authors contributed critically to the drafts and gave final approval for publication.

Data availability

Should the manuscript be accepted, the data supporting the results will be archived in an
appropriate public repository (Dryad or Figshare) and the data DOI will be included at the end of
the article.

Abstract

Climate change is driving range shifts, and a lack of cold tolerance is hypothesized to constrain insect range expansion at poleward latitudes. However, few, if any, studies have tested this hypothesis during autumn when organisms are subjected to sporadic low temperature exposure but may not have become cold tolerant yet. In this study, we integrated organismal thermal tolerance measures into species distribution models for larvae of the Giant Swallowtail butterfly, *Papilio cresphontes,* living at the northern edge of its actively expanding range. Cold hardiness of field-collected larvae was determined using three common metrics of cold-induced physiological thresholds: the supercooling point (SCP), critical thermal minimum (CT_{min}), and survival following cold exposure. *P. cresphontes* larvae in autumn have a CT_{min} of 2.14°C, and were determined to be tolerant of chilling. These larvae have a SCP of -6.6°C and can survive prolonged exposure to -2°C. They generally die, however, at temperatures below their SCP (-8°C), suggesting they are chill tolerant or modestly freeze avoidant. Using this information, we examined the importance of low temperatures at a broad scale, by comparing species distribution models of *P. cresphontes* based only on environmental data derived from other sources to models that also included the cold tolerance parameters generated experimentally. Our modelling revealed that growing degree-days and precipitation best predicted the distribution of *P. cresphontes*, while the cold tolerance variables did not explain much variation in habitat suitability. As such, the modelling results were consistent with our experimental results: low temperatures in autumn are unlikely to limit the distribution of *P. cresphontes*. Further investigation into the ecological relevance of the physiological thresholds determined here will help determine how climate limits the distribution of *P. cresphontes*. Understanding the factors that limit species distributions is key to predicting how climate change will drive species range shifts.

Keywords: mechanistic species distribution model, survival assays, Maxent, global warming, cold distribution limits, insect, low temperature limit

Introduction

Over the past few decades, range shifts due to climate and land use changes have been

reported in both plants and animals. Many are experiencing a poleward and/or upward shift in

distribution, pushing their northern and upper range limit to higher latitudes and altitudes (Chen

et al., 2011; VanDerWal et al., 2013; but see Kerr et al. 2015). However, there has been

substantial variation in the degree and direction of range shift across taxonomic groups (Chen et

al., 2011; Lenoir & Svenning 2015). For example, some insect taxa (i.e., dragonflies, spiders)

have shifted northwards to a much greater extent than mammals and birds, whereas other

groups have expanded southwards. This variation in range shifts remains difficult to explain,

and thus necessitates a better understanding of the factors influencing species' distributions

(Buckley & Kingsolver, 2012; Sunday et al. 2012; Sgrò et al., 2016).

A climate change-driven range shift in insects is often attributed to the relaxation of a

harsher poleward climate since temperature acts as a physiological limitation and as a

phenological cue (Logan & Bentz, 1999; Root et al., 2003; Elkinton et al., 2008). For those

species with northern range edges associated with temperature clines, climatic warming has

frequently resulted in the colonization of suitable areas in more northern latitudes (Parmesan,

1996; Parmesan & Yohe, 2003). For example, the northern range expansion of the deer fly

(*Lipoptena cervi*) in Finland is due to warmer temperatures during the summer (Härkönen et al.,

2010). However, for many other species, it remains unclear how climate change has led to

range shifts (Chen et al., 2011; Sunday et al., 2012).

Low temperatures throughout the year can constrain the ability of insects to persist at

northern range limits as they can influence any life cycle stage (Ungerer et al., 1999).

Overwinter survival has been shown to be a key factor limiting the ranges of insects (Crozier,

2003). For example, the northern range edges of the Southern pine beetle (*Dendroctonus*

frontalis) and Sachem skipper butterfly (*Atalopedes campestris*) are limited by temperatures

below -16°C (Lesk et al., 2017) and the -4°C January isotherm (Crozier, 2003), respectively.

Overwintering insects can adopt one of multiple physiological strategies for surviving winter

(Williams et al., 2015); some insects can survive internal ice formation and are termed freeze

tolerant, some physiologically supress the temperature at which their body fluids spontaneously

freeze (the supercooling point; SCP) and are termed freeze avoiding, while others remain

susceptible to chilling (termed chill susceptible), and seek relatively warm microhabitats to

overwinter (Sinclair, 1999; Denlinger & Lee, 2010; Overgaard & MacMillan, 2017).

While winter represents a considerable challenge (Robinet & Roques, 2010; Williams et

al., 2015), insects are also likely to be particularly vulnerable to low temperatures during

autumn. Autumn is when most insects are acquiring cold tolerance through physiological

adjustments (acclimatization) and have therefore not reached their peak cold hardiness (i.e., the

capacity to tolerate intensity and duration of cold exposure; Tauber et al., 1986). For example,

monarch butterflies, *Danaus plexippus,* that were not yet acclimated had a mortality rate of 60%

when exposed to low temperatures in September and October compared to 20% for acclimated

ones (Larsen & Lee, 1994). There is also a much higher frequency of cold temperature events in

autumn than in the summer (Danks, 1978). Yet, no study, to our knowledge, has considered low

temperatures during autumn as a limiting factor on the geographic distributions of insects.

Regardless of season, cold exposure can cause insects to cross physiological thresholds

leading to sublethal effects. At a species-specific low temperature, most insects lose the ability

for coordinated movement (a critical thermal minimum; CT_{min}) after which they enter a state of

complete neuromuscular silence termed chill coma (Gibert & Huey, 2001; MacMillan & Sinclair,

2011; Oyen & Dillon, 2018). Many insects can recover from this state with no immediate

evidence of injury; but prolonged or severe cold exposure can cause behavioural defects, or

slow or halt development (Asahina, 1970; Rojas & Leopold, 1996; Overgaard & MacMillan,

2017). Chill susceptible insects suffer from a loss of homeostasis at low temperatures well

above their SCP, while those more tolerant to chilling survive such exposures (Overgaard &

MacMillan, 2017). While freeze tolerant insects can survive freezing of the extracellular fluid at

their SCP, they can suffer from ice-related injuries below this temperature. By contrast, freeze

avoidant insects cannot survive at temperatures below their SCP. Thus, the relationship

between the SCP and injury/mortality is critical to determining the cold tolerance of these insects

(Sinclair et al., 2015).

Cold tolerance traits, like the CT_{min}, SCP, or survival following a cold stress frequently

correlate strongly with insect distribution (Bozinovic et al. 2011; Gouveia et al. 2014; Andersen

et al. 2015). For example, the CT_{min} and CT_{max} of *Drosophila* species have been used to

accurately predict their current distributions(Overgaard et al., 2014). Similarly, the SCPs and

potential for mortality at more mild temperatures were used to understand and predict the

potential invasion limits of pine beetles in North America (Ungerer et al., 1999; Bleiker et al.,

2019). However, given the limited number of taxa in which these relationships have been

explored, we have little to no ability to generalize predictions about how climate influences

species' distributions via low temperatures to other taxa (Ouimette 2018).

Species distribution modelling is a commonly used approach to evaluate recent range

shifts and forecast future shifts due to climate change. For example, they have been used to

predict invasive species' distributions (Jiménez-Valverde et al., 2011), define species' ranges

and predict impacts of climate change on available habitat (Araujo et al. 2005). Using

correlations between georeferenced occurrence records and a set of environmental variables

with geospatial data, these models predict a species' suitable habitat (Elith et al., 2011). While

the approach is useful for many applications, these models often violate key assumptions, such

as predicting habitat suitability in novel conditions and omitting key biotic variables known to

influence geographic distributions such as species interactions (Tingley et al. 2014; Briscoe et

al. 2019).

One way that has been demonstrated to improve the accuracy of these models is to

incorporate physiological variables or stress tolerance thresholds (e.g., heat tolerance,

metabolic needs, thermal limits; (Kearney & Porter, 2009; Overgaard et al. 2014). By using

variables derived directly from these physiological traits, the underlying processes explaining

the species' distribution are thought to be better incorporated into the model (Kearney & Porter,

2009). For example, Elith et al. (2010) built a distribution model for the cane toad (*Bufo marinus*)

based on the impact of topography on the above-ground activity of adults and the temperature

requirement of larval growth and survival. Models produced in such a way are called

mechanistic or process-based models. They calculate the likelihood of presence by eliminating

regions where the conditions are detrimental to species persistence (Buckley et al., 2010). In

some contexts, these models have been shown to be more accurate than (Peterson, 2011), or

complement weaknesses of (Martinez 2015), correlative models when modelling the

fundamental niche and can strengthen predictions about future distributions (Buckley et al.,

2011; Martinez 2015; Evans et al. 2016; Kotta et al. 2019). The main limitation with this

approach is that experimentally derived variables can be time-consuming to develop and thus

not feasible to do for large assemblages of species or over broad ranges (Peterson 2011; Evans

et al. 2016). It can also be difficult to find traits that are predictive and functional in the species'

environment (Kearney & Porter, 2009).

Here, we use a widespread butterfly, the Giant Swallowtail, *Papilio cresphontes*, as a case

study to test the hypothesis that low temperatures during autumn are the limiting factor of the

northern edges of insect species' distributions. Specifically, the two main objectives are to: i) find

relevant cold tolerance thresholds of *P. cresphontes* collected in autumn; and ii) determine the

relative importance of these thresholds on the distribution of *P. cresphontes* at the landscape

level. Simultaneously, we aim to determine the cold tolerance strategy of larvae at the northern

range limit. Determining the species' cold tolerance strategy improves our knowledge about the

importance of low temperatures on *P. cresphontes* survival and helps identify which low

temperature thresholds are likely to be most relevant at the species' northern range. In this

study, we focus on the pre-overwinter life stage (i.e., larval stage), which is in contrast to most

other studies which have determined the cold tolerance strategy of the life stage that

overwinters (Radchuk et al., 2013). To determine the importance of low temperatures at a broad

scale, we model the geographic distribution of *P. cresphontes* with the cold tolerance

parameters we generated experimentally.

P. cresphontes has undergone a rapid expansion over the past decade and now occurs

as far north as Ottawa, Ontario, Canada. Finkbeiner (2011) hypothesized that the range

expansion into New York state from 2000 to 2010 was due to the disappearance of frost (as

defined meteorologically) in September. In spite of this, they still showed frost resistance in

larvae and that individuals could survive exposure to -3.9°C. However, these observations were

based on field observations of only a few specimens and the cold tolerance strategy was not

determined. Therefore, a more in-depth study of the impact of low temperature on

P. cresphontes is warranted.

To establish the larval cold tolerance strategy, we measured three common cold-induced

physiological thresholds: SCP, CT_{min}, and survival following cold exposure. To gain a better

estimate of the cold tolerance strategy, thresholds (SCP and low temperature survival) were

measured for two generations and multiple sites across a latitudinal gradient at the northern

range limit. Latitude has the potential to affect cold tolerance due to its correlation with climate

(e.g., photoperiod, temperature; Sømme, 1982; Tanaka, 1996) but see Yoshio & Ishii, 2001).

Therefore, latitude could affect when and to what extent organisms are cold hardy. Combined,

these experiments were used to identify a potentially relevant low temperature threshold of

larvae collected at the northern range edge. We examined the importance of low temperatures

at a broad scale by comparing species distribution models of *P. cresphontes* based only on

environmental data derived from other sources to models that also included the cold tolerance

parameters generated experimentally.

Materials and methods

1. Study system

The Giant Swallowtail (*Papilio cresphontes*) butterfly is a member of the Papilionidae

family. The species' range extends from Costa Rica throughout the North American continent.

The bulk of the population is concentrated in the Eastern United States. It ranges as far north as

the Ottawa region, Ontario, Canada. (Figure 1)

P. cresphontes larvae undergo five instars before entering the chrysalis stage (Bullock,

1991). In Ontario, there are two generations with flights occurring from May to July and again

from late July to late September (Layberry, Hall & Lafontaine, 1998). The larval stage lasts from

3-4 weeks and pupae formed in the summer emerge after 10-12 days, but those that develop in

autumn will remain as pupae until spring. The overwintering pupae undergoes winter diapause,

a state of lowered metabolism, with reduced respiration, and no feeding or growth occurs (Scott,

1997). Once eclosed, adults live for 6-14 days (Layberry, Hall & Lafontaine, 1998).

P. cresphontes is known to use plants of the Rutaceae family as their primary food

source in the larval stage. In Ontario, those plants consist mainly of Northern Prickly Ash

(*Zanthoxylum americanum*) and Hoptree (*Ptelea trifoliata*). The adults are generalists and will

gather nectar from most flowering plants, for example, goldenrod (*Solidago*), swamp milkweed

(*Asclepias incarnata*) and azalea (*Rhododendron azaleastrum*)(McAuslane, 2004).

2. Cold tolerance experiments

i) Experimental overview

To determine the impact of low temperatures on *P. cresphontes* larval survival and

developmental success, three experiments were conducted (Figure 2; Figure S1). The first

experiment was done to determine the cold tolerance strategy of the larvae by measuring the

SCP in relation to cold survival. Additionally, since the results from this initial experiment were

not enough to clearly define the cold tolerance strategy, further low-temperature survival assays

(i.e. the ambient exposure of larvae to low temperatures for an extended period of time) were

conducted to better determine the impacts of prolonged low temperature exposure on survival

and developmental success. A third experiment was conducted to measure the CT_{min}.

The SCP and low-temperature survival assays were measured for both generations and

multiple sites (n=4) across a latitudinal gradient at the northern range limit (Figure 1, 2).

(ii) Field sampling

To gain a better understanding of the cold tolerance strategy, larval sampling occurred

during the two *P. cresphontes* generations: July and August of 2018. A total of 117 larvae were

collected from four sites around Ottawa, Ontario, Canada (Mud Lake, Shirley's Bay, Brockville

and Queens University Biological Station (QUBS)) spanning a 100 km latitudinal gradient at the

northern range limit of *P. cresphontes* (Figure 1). These four sites were accessible and had a

high abundance of larvae.

Until the start of the experiments, captured individuals in July were provided with a fresh

supply of *Z. americanum* and kept in an LTCP-19 Biochamber at the University of Ottawa,

Canada (45.423325, -75.683177) on a 21°C/25°C 15h: 9h L:D cycle with light intensity peaking

at noon. These conditions were chosen to match the average environmental conditions in July

for the Ottawa region. To replicate the conditions required for larvae to prepare to overwinter,

the chamber parameters were modified weekly for the August generation to match the

conditions from August to October (meteomedia.ca; Figure S1). Over the time spent in the

chamber, peak daily temperatures reached a maximum 25 °C and a minimum of 15°C overnight

at the beginning and by the end, daytime temperatures only reached 10°C and a minimum of

6°C overnight (Figure S1). Likewise, photoperiod fell from 14.5h to 11h. To validate our

treatments, we recorded the realized climate regime in the environmental chamber using a

temperature logger (HOBO Pendant, Onset Computer Corporation, Bourne, MA).

(iii) Experimental details

Experiment 1: Supercooling point (SCP) test

To identify the cold tolerance strategy of *P. cresphontes* larvae, we quantified the SCP

and survival following a freezing event in the first generation (i.e. the July generation; n=27 from

QUBS; Figure 2; Figure S1). The SCP test was done following the recommended methodology

of Sinclair (2015) and details can be found in Appendix S1.

To get a more accurate estimate of SCP, another SCP test was done on the second

generation (i.e. the August generation; n=29; Figure 2, Table S1; Figure S1). The individuals

from the August generation were subjected to the same cooling protocol as described above but

were kept submerged until all of the larvae had frozen. This was done to obtain the full

distribution of SCP values.

Experiment 2: Low-temperature survival assays

To identify the consequences of chronic low temperature likely to be experienced by the

larvae before pupation, three low-temperature survival assays were conducted (Figure 2; Figure

S1). The duration and temperature of the assays were chosen based on the SCP determined in

the first experiment and on meteorological data in the Ottawa area representing the

developmental period before pupation. In the Ottawa region, this period corresponds to the

month of October. We considered temperatures from 2012 to 2017, which correspond with the

timing of the recent range expansion of *P. cresphontes* into the Ottawa area (Ontario Butterfly

Atlas: www.ontarioinsects.org/atlas_online.htm) and used the 'OTTAWA CDA RCS' weather

station (45.38, -75.71). For this timeframe, the mean and lowest temperatures recorded during

frost events (as determined meteorologically: the span of time in which the temperature is below

0°C for more than one hour) were -1.33°C and -6.3°C, respectively, and the average frost time

was 6.24 hours. Consequently, and in conjunction with the measured SCP of -5.6°C determined

in the first experiment, trials were run at -2°C, -6°C, -8°C for 7 hours each.

To determine whether larvae were chill susceptible, the first assay was done at -2°C

(Figure 2). Mortality at this temperature (i.e., above the SCP; see results) would indicate this

species is chill susceptible whereas survival would confirm the ability of larvae to tolerate

exposure to low temperatures above their SCP, thus making them chill tolerant or freeze

avoidant. The second test was done at -6°C to distinguish between freeze avoidant and freeze

tolerant strategies (Figure 2). Considerable mortality near the mean SCP would mean larvae are

freeze avoidant whereas high survival would mean they are more likely freeze tolerant. Finally,

to find a potentially ecologically relevant low temperature threshold causing high mortality (i.e. a

potential lower lethal limit), we also tested larval survival at the lowest recorded temperature for

October at this weather station from 2012-2017 (i.e. -8°C). Exposure at this temperature tests

the maximum larval cold tolerance and their resistance to freezing since it is well below the

mean SCP obtained in the previous experiment (i.e., -5.77 °C; Figure 2).

The low-temperature assays were done in an environmental growth chamber

(Biochambers model LTCB-19) for both larval generations and across multiple sites (Figure 2;

Table S1). The -2°C test was not repeated in August because i) there was high survival with the

July individuals (see results); ii) a low number of specimens were collected in the field, and iii)

this temperature was unlikely to be a limiting factor for larval survival and thus was not a priority

to test. Larvae were weighed and measured before being placed in individual containers inside

the chamber. The chamber temperature was brought down from 21°C to the predetermined test

temperature at a rate of 0.1°C per minute. Once the desired minimum temperature was

reached, the temperature remained constant for seven hours before being increased back to

21°C at the same rate. Larvae were then left to recuperate. They were allowed to feed on

freshly harvested *Z. americanum* leaves during the trial and were checked daily for mortality.

Death was determined when larvae did not react to physical stimuli: larvae were poked with a

stick, sprayed with water and then shaken. All larvae that survived to pupation were kept until

emergence.

Experiment 3: Critical thermal minimum (CT_{min})

To identify the temperature at which larvae lose coordination, a CT_{min} test was done on

twenty larvae were caught in July from the QUBS site, following the method of Andersen et al

(2014)(see Appendix S1 for details). The CT_{min} test was only done on July individuals since it

required a higher number of specimens than the other tests and our sample size was limited.

iv) Statistical analysis

To compare survival rates of larvae of the July generation that froze to those that did not

freeze during the SCP experiment, a χ2 test was used. The rate above and below the SCP for

the August generation could not be compared since all larvae were exposed to low

temperatures until SCP was reached (i.e. no survival rate was calculated above SCP). To test

whether mean SCP differed across sites for the August generation, an ANOVA was run. Since

there was no significant difference in mean SCP across sites (df=3, F-value=0.94 p = 0.44), site-

level data were pooled. The assumptions of normality and homogeneity were evaluated using

the Shapiro-Wilk test and the Brown-Forsythe for homogeneity of variance test, respectively.

To compare the larval survival rate and the rates of pupation and adult emergence

between generations and sites for the low-temperature assays, a χ2 test was used. There was

no impact of site on survival rate or developmental success in any assay (Table S2), so site-

level data were pooled. All analyses were conducted in R (version 3.4.1).

2. Distribution modelling

i) Overall approach

To determine the limiting factors of the current geographic distribution of *P. cresphontes*,

we modelled its range at two different spatial extents: northern range and the entire distribution

(Figure 1); and using two different modelling approaches. We wanted to determine whether the

relative importance of variables differs between the northern range and the entire distribution.

We defined the northern range as the area encompassing the upper 50th percentile of

occurrences based on both latitude and longitude in the northeastern part of the range (Figure

1). The 50th percentile represented a clear threshold in the latitudinal distribution of occurrences.

To assess if the physiological-based cold tolerance metrics were significant predictors of *P.*

cresphontes' distribution, the modelling was done using two approaches: correlative and

mechanistic. The correlative approach was based on environmental data derived from other

sources (e.g., remote sensing, weather stations) whereas the mechanistic approach additionally

included parameters derived from the cold tolerance experiments.

For both approaches, we tested the importance of the cold tolerance strategy

parameters by comparing i) model fit with and without them and ii) the proportion of suitability

explained among a number of climate-related factors hypothesized to constrain the range of *P.*

cresphontes at the landscape scale.

ii) Data

Models were built using all available occurrence records for North America from e-

butterfly (www.e-butterfly.org, accessed July 2018, 2255 records), the Global Biodiversity

Information Facility (GBIF; https://www.gbif.org, accessed September 2018, 3420 records) and

Moths and Butterflies of North America (www.butterfliesandmoth.org, accessed October 2018,

1206 records) from 1980 to 2018. This timeframe was chosen to match the timeframe of the

available environmental variables. Records that were obviously inaccurate (e.g., in the middle of

the ocean) or had no geographical coordinates were removed. To reduce the likelihood of an

occurrence being based on the same individual being observed on multiple occasions, duplicate

records across data sources were also removed. To reduce sampling bias, we used a low-end

estimate of dispersal ability (1 km) and randomly selected a single occurrence across data

sources per 1 km^2. In total, 3510 and 1457 occurrences were used for the North American and

the northern range extent, respectively. Occurrences were then mapped onto a North America

Lambert Conical Conic projection.

Sixteen environmental variables were initially considered for the model building based on

previous work done on butterflies (Appendix S1; Table S3). The variables included represented

both the direct and indirect effects (i.e. via a species' interaction) climate can have on a species'

range (Appendix S1). Variables were created for both range extents and then screened for

collinearity (Appendix S1; Table S4). All climatic data were downloaded at a 1 km resolution and

restricted to 1980-2010. All data was downloaded at a North American extent, with the

exception of Normalized Difference Vegetation Index (NDVI), which was global in extent.

For the mechanistic models, we created variables for both range extents based on the

results from the cold tolerance experiments and minimum daily temperature data from Daymet

(Table S3). Three variables were built for each extent by summing the number of days below

the following thresholds: 2.14°C (i.e., CT_{min}), -6.6°C (i.e. the SCP from the August generation)

and -8°C (i.e. potential lower lethal limit) for each year of the study period 1980-2010 (Table S3).

The SCP from the August generation was chosen since it is the one most likely to match when

larvae are more likely to experience low temperatures.

iii) Modelling

Models were built using Maxent (version 3.4.1, Philipps et al. 2006) with the BIOMOD2

package (Damien Georges) in R (version 3.41.1). Maxent is a commonly used species

distribution modelling technique for presence-only data (Merow et al., 2013). Maxent uses a

maximum-entropy approach to model species distributions. It relates the species' presences

and pseudo-absences to the environmental variables to predict where the habitat is suitable

across the mapped geographical space. Maxent is one of the more common species distribution

modelling software (Elith et al., 2011) and it performs very well compared to other modelling

techniques when dealing with presence-only data (Hernandez et al., 2006). Details on the

modelling are presented in Appendix S1.

iv) Statistical analysis

To determine the role of low temperature in limiting the distribution of *P. cresphontes*,

model accuracy was compared across the two extents and approaches using t-tests.

Comparisons were based on AUC, Kappa and TSS.

Results

1. Cold tolerance experiments

Experiment 1: Supercooling point (SCP) test

The results from the July and August SCP experiments provide support for a chill

susceptible strategy in *P. cresphontes*. The mean SCP for July was -5.8°C (0.2SE; n=14). There

was no significant difference in survival for July larvae that froze compared those that did not

freeze during the SCP experiment (above: 77% (10/13) vs. below: 43% (6/14 larvae); χ2= 1.2, df

= 1, p-value = 0.2).

The mean SCP for all August larvae was -6.6°C (0.2SE; n=28) and for the upper 50% of

larvae (directly comparable to the July experiment), the mean SCP was -5.9°C (0.2SE; n=14).

The August larvae had a 3.5% (1/29) survival rate following freezing (i.e., exposure to

temperatures below the mean SCP). The low survival below SCP and apparent lack of change

in the SCP suggests that larvae are tolerant of moderate chilling, but intolerant of freezing, and

are not undergoing a change in SCP under simulated autumn conditions.

Experiment 2: Low-temperature survival assays

i) Survival

Our low-temperature survival assays further demonstrate that larvae are chill tolerant

or modestly freeze avoidant, as they are unable to survive exposure to temperatures below their

SCP. In the -2°C test, July larvae had a survival rate of 93% after 24 hours (Table 1). Therefore,

single ecologically-relevant exposures to temperatures above their SCP do not have a large

impact on larval survival. Moreover, while the survival rate of larvae in July was more affected

by exposure to -6°C than -2°C (Table 1), it was still rather high (70%). Therefore, larvae are able

to survive a 7 h cold exposure near their SCP and thus may be chill tolerant or freeze avoidant.

August larval survival was significantly higher than that of the July larvae following

exposure to -6°C (100%, 17/17; χ^2-squared = 5.1, df = 1, p-value = 0.024; Table 1). This finding

provides further evidence in support of a modestly freeze avoidant strategy since the mean SCP

for the August generation (-6.6°C) was lower than the test temperature (-6°C) and survival was

still quite high.

In comparison, in the extreme low-temperature test (-8°C), larvae had marginal

survival in both generations (July: 12.5% (1/8); August: 10% (1/10)) and there was no

improvement in survival across generations (χ^2 = 3.03e-31, df = 1, p-value = 0.99; Table 1). This

low survival further suggests larvae may use a freeze avoidant strategy since this temperature is

below the mean SCP measured in the August generation.

ii) Developmental success

Overall, low temperature exposure impacted pupation and adult eclosion. The

percentage of larvae that successfully pupated and eclosed generally decreased as the

temperatures got lower (Table 1). The percentage of larvae that successfully pupated at -8°C

was low (July: 12.5%; August: 10%). Therefore, although larval survival can be high after

exposure to low temperatures, chilling also had effect on developmental success, which is

indicative of sub-lethal chilling injury occurring, even in the absence of freezing.

Experiment 3: CT_{min}

The average CT_{min} of larvae was 2.14°C (0.26SE; n=20).

2. Distribution modelling

i) Model performance

Model predictive ability varied depending on the metric of accuracy used. All models

were considered good based on TSS (i.e. between 0.4-0.75) and AUC (>0.8), whereas based

on Kappa, the northern range models were weak (<0.4) and the full range models were fair

(>0.5; Figure 3).

Statistically, the mechanistic models were significantly more accurate than the

correlative models in some but not all cases (Figure 3; Table S5). However, in the select cases

where there was a significant difference between modeling approaches, the effect size was

small, implying the difference is unlikely to be biologically significant (e.g., AUC difference for

northern range comparison: 0.011; Figure 3, Table S5). Therefore, it is unlikely the inclusion of

cold tolerance metrics significantly improves the prediction of habitat suitability of *P.*

cresphontes.

438   The accuracy of the full range models was significantly higher than the accuracy of the

439   northern range models for all evaluation metrics and approaches (Figure 3; Table S6). Although

440   the effect size is also not large, the significant difference in accuracy was consistent across all

441   comparisons (Figure 3; Table S6). Therefore, the habitat suitability of *P. cresphontes* is

442   predicted more accurately across its full range than its northern range.

444   *ii) Variable importance*

445   Growing degree-days was the most important factor explaining the habitat suitability of

P. cresphontes across all model extents and approaches, explaining between 27-35% of the

variation in habitat suitability (Table 3). Precipitation was the second most important factor

across both extents and approaches. Across the full and northern range extents, GDD and

precipitation combined explained the majority of the variation in habitat suitability (~50%; Table

3). The physiologically derived variables included in the final model (i.e., CT_{min} and the potential

lower lethal limit) only explained a small part of variation in habitat suitability (~7.5%) and ranked

5th and 6th in model contribution (Table 3).

DISCUSSION

455   Although insects are likely to be particularly vulnerable to low temperatures during

autumn, no study, to our knowledge, has considered low temperatures during autumn as a

limiting factor on the geographic distributions of insects. Here, we test this hypothesis at the

northern edge of the distribution of the widespread butterfly, the Giant Swallowtail,

P. cresphontes. Our study contributes three main findings that lead to a rejection of this

hypothesis. First, given the survival of larvae to prolonged exposures to temperatures close to

but not below their SCP, they are chill tolerant or modestly freeze avoidant at their northern

range limit at this time of year. Their low survival below the SCP provides support that this

species is unable to handle freezing before pupation and that they may use a freeze avoidant

strategy. However, further testing for seasonal plasticity in the SCP is required to confirm a

freeze avoidant strategy as we had limited power and were unable to directly compare the

change in SCP between the two generations. An active depression of the SCP over the season

would indicate that larvae are accumulating cryoprotectants and that the larval cold tolerance

could be increasing, thus supporting a freeze avoidant strategy.

Our results demonstrate that ecologically relevant exposures to temperatures above the

SCP are not lethal, but below the SCP are; thus providing a clear picture of the lower end of

thermal tolerance for this species. This means that for *P. cresphontes* larvae, a single overnight

exposure to temperatures near -8°C will impede their survival and could represent a potential

lower lethal limit. Consequently, areas with early autumn temperatures that reach -8°C should

be inhospitable for *P. cresphontes*. However, in the Ottawa region, which is at the northern

range limit, there are only 0-3 frost episodes on average from September to October. Therefore,

low temperatures (i.e. <-7°C) during early autumn are unlikely to be experienced by the larvae

there or anywhere in its current range. This suggests that a single exposure to low temperatures

in early autumn is unlikely to limit the current northern range of *P. cresphontes* through cold-

induced mortality. However, there may be sublethal effects of chilling (e.g., effects on

developmental success), which could still impact overall insect fitness.

Second, we found that exposure to normal frost temperatures for this time of year in this

area (i.e., -1°C to -2°C), which are above the SCP, did not affect larval survival. Therefore, it is

also unlikely that frost in early autumn is a limiting factor of *P. cresphontes'* northern range. This

is in concordance with the field observations from Finkbeiner (2011), which showed that

P. cresphontes larvae can survive frost events. Since larvae had no problem reaching pupation

after exposure to -2°C, the most common frost temperature, and the majority of larvae still

successfully pupated (60%) after the low frost temperature test (i.e., -6°C), the cold tolerance of

larvae at the current northern range limit seems sufficient to cope with the climate they

experience. However, further testing in ambient conditions and with a proper control is needed

to determine whether cold exposure on larvae have significant sub-lethal effects on eclosion as

rates of eclosion in this experiment were less than 30% but a control group was not included

due to sample size constraints. It has been demonstrated elsewhere that the success of each

life stage matters in the response of butterfly species to climate change (Radchuk et al. 2012).

494   Third, the species distribution modelling results were consistent with the experimental

results: cold related variables did not explain the distribution of *P. cresphontes* at a broad scale.

Together, these results provide strong support that exposure to temperatures above -6.6°C

during autumn does not limit the northern range of *P. cresphontes*. Instead, growing degree-

days and precipitation are the most important predictors, of those tested here, on the distribution

of *P. cresphontes* at a broad scale. While other factors were not tested here, for example, biotic

interactions (e.g., host plant occurrence), our results suggest that *P. cresphontes* depends on

specific heat accumulation and water availability to complete its life cycle. These factors have

also been identified as important in predicting the range of other butterflies (Luoto et al., 2006;

Eskildsen et al., 2013). Evidence suggests there is a strong relationship between the number of

growing degree-days and the growth rate of larvae, and the foraging activities of adults (Ritland

& Scriber, 1985; Kukal & Dawson, 1989; Schneider & Root, 2002). Precipitation can limit the

range of insects directly due to dehydration or indirectly by imposing limitations on primary

productivity, which in return transposes to higher resource availability.

508   The full range model was more accurate at predicting the distribution of *P. cresphontes*

than the northern range model. Similar results have been found for other species, with model

accuracy and performance increasing with larger study extents (VanDerWal et al., 2009; Connor

et al., 2019). This is likely a result of smaller environmental gradients in the northern range,

which causes the model to have a greater difficulty detecting differences between presences

and absences, and in turn leads to difficulties predicting presences and absences in novel areas

(Smith & Santos, 2019). Prevalence can also decrease with larger study extents (Barbet-Massin

et al., 2012). However, the accuracy of the two models was still different using TSS, a metric

that is not sensitive to prevalence (Table S4; Figure 3). Nevertheless, even if model

performance was lower in the northern range relative to the full range, the model was still valid

and performed adequately across all metrics.

While growing degree-days and precipitation explain a lot of the variation in habitat

suitability for *P. cresphontes*, we may have underestimated the role of cold tolerance at the

northern range edge for three reasons. First, low temperature thresholds, like the CT_{min},

although not lethal, could still impact *P. cresphontes'* survival in natural conditions due to the

enhanced risk of predation or starvation linked with the loss in mobility at temperatures below

2°C. While a possibility, this hypothesis depends on a high likelihood of predation, which is

unknown across *P. cresphontes'* range (Hazel et al., 1998; McAuslane, 2009). Low temperature

may also lead to sublethal effects on developmental success, energy status and behaviour.

Second, it is also possible that more prolonged or repeated exposure to low temperatures could

impact *P. cresphontes* larval survival or have sublethal effects at the northern range edge

because we did not test for the effects of repeated cold injury or chilling. In *Drosophila*, multiple

cold events could cause an accumulation of cold injuries resulting in larval death or in sub-lethal

effects such as decreased feeding, fecundity and dispersal ability (Marshall & Sinclair, 2012).

Third, the SCP estimated here is effectively a theoretical one. Since we could not collect and

test the last emerging larvae of the season due to the difficulty in finding them, they might not

have reached their peak cold tolerance and therefore *in situ,* SCP could be lower (i.e., their cold

hardiness may have been underestimated). On the other hand, the *in situ* SCP may be higher

since environmental factors in natural conditions, such as air moisture or exposure to ice

nucleators, may further affect survival at temperatures above SCP. Nevertheless, because the

low temperature thresholds derived experimentally in this study could be an underestimate, the

mechanistic models may not have been optimized perfectly, thus underestimating the role of

cold tolerance metrics at a broad scale. However, since temperatures below the SCP (i.e.

<-7°C) are very unlikely to occur multiple times before pupation, if at all, low temperatures during

autumn remain unlikely to limit the northern range of *P. cresphontes*.

Given our lack of knowledge about the overwintering ecology and physiology of this

species in its northern range, pupal overwinter survival may be a more important limiting factor

for the species' northern range limit. Unfortunately, it is unknown whether they overwinter in the

leaf litter on the ground or tree branches above the ground. If on the ground, they would likely be

covered in snow (average annual snowfall in Ottawa is 224 cm (Environment and Climate

Change Canada)) and therefore buffered against low temperatures. The temperature 30 cm

below snow cover is usually -4°C (Flin 2008). This behaviour would increase their chances of

surviving the winter. Indeed, West (1996) showed that in Virginia, *P. cresphontes* pupate 3-25

cm off the ground on dead branches. In the northern range, it is likely that even at these heights,

chrysalids would be covered by snow. While ontogenetic variation in thermal tolerance is

present in other species (Terblanche et al. 2006; Marais et al. 2009; MacLean et al., 2016; but

see Ouimette, 2018) and only larvae were tested in this study, *P. cresphontes* larvae purge their

food and liquids before pupation, thus they are likely to be even more cold tolerant at the pupal

stage. Other members of *Papilio* are more cold tolerant than what we determined for the larval

stage of *P. cresphontes. P.canadensis* and *P.glaucus,* are considered to be freeze tolerant at

the pupal stage and have an SCP of -23°C to -27°C (Kukal et al., 1991). *P. xuthus* is freeze

avoidant with an SCP of -25°C (Shimada 1988). Therefore, the pupae of *P. cresphontes* are

likely to have even greater cold hardiness than the larvae and thus be able to survive the winter

at the northern range if they pupate at similar heights as the populations in Virginia.

Conclusions

Our results demonstrate that the cold tolerance of *P. cresphontes* larvae is well matched

to its environment in the northern part of its range and that it is unlikely that low temperatures in

autumn are limiting its range. This suggests that the warming of autumn temperatures over the

past decade is unlikely to have been the factor that led to the recent range expansion of

P.cresphones, as originally hypothesized by Finkbeiner (2011). Instead, the warming of winter

temperatures could have ameliorated overwintering survival, leading to an improvement in

habitat suitability. Further study on determining the key factor(s) limiting the northern range of

P. cresphontes should focus on defining the cold hardiness of the pupal stage, the overwintering

behaviour and the ecological relevance of the physiological thresholds found in this study.

Determining the ultimate factors that limit species' distributions will be critical in accurately

predicting species' range shifts in response to future climate change.

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Tables

Table 1. Survival rates and developmental success for the low-temperature assays. Test
temperature and generation are shown. The percentage of initial larvae that survived the first 24
hours after the experiment, successfully pupated and eclosed as adults is alos shown. All
percentages are based on the number of larva that began the experiment.

Test	Generation	Number of larvae	Larval survival (%)	Pupation (%)	Eclosion (%)
-2°C	July	23	93	78.5	52
-6°C	July	10	70	60	30
	August	17	100	100	0
-8°C	July	8	12.5	12.5	12.5
	August	10	10	10	0

820	Table 2. Contribution of the environmental variables in explaining habitat suitability across different model extents and approaches.
821	Shown is the rank order of variable importance and mean (± S.E) proportion of variance explained. In bold is the variable that
822	explains the most amount of variation for each model type.

Variables	North America				Northern range			
	Correlative		Mechanistic		Correlative		Mechanistic	
	Rank order	Variation explained (SE)	Rank order	Variation explained (SE)	Rank order	Variation explained (SE)	Rank order	Variation explained (SE)
Growing degree-days	**1**	**31.33 (0.19)**	**1**	**27.34 (0.23)**	**1**	**37.17 (0.08)**	**1**	**35.03 (0.06)**
Precipitation	2	24.87 (0.08)	2	21.70 (0.085)	2	23.94 (0.08)	2	22.47 (0.07)
Extreme maximum temperature	3	17.54 (0.06)	3	16.22 (0.06)	6	2.38 (0.02)	7	2.41 (0.02)
Normalized Difference Vegetation Index	4	15.45 (0.10)	4	15.87 (0.11)	4	16.77 (0.11)	4	14.89 (0.11)
Mean temperature of the coldest month	NA	NA	NA	NA	3	16.9 (0.06)	3	15.57 (0.05)
Precipitation as snow	5	10.80 (0.07)	5	10.05 (0.07)	5	2.84 (0.02)	6	3.46 (0.02)
CT_{min}*	NA	NA	NA	NA	NA	NA	5	5.97 (0.03)
Potential lower lethal temperature*	NA	NA	6	8.80 (0.04)	NA	NA	8	0.16 (0.004)

823	* variable derived experimentally

824 **Figures**
825
Figure 1. The geographic distribution and northern range of *Papilio cresphontes*. The first panel
shows the entire range of *P. cresphontes* encompassed by the minimum convex polygon (thin
black line). The second panel shows the position of the northern range, denoted by the solid
blue line, relative to the entire range. The blue circle indicates the locations of the field sites. In
the first two panels, the occurrence points (black) are shown against the predicted habitat
suitability (red). The third panel shows the field sites where larvae were collected at the northern
range limit: the blue dot indicates the Brockville site (44.84952, -75.75226), red indicates the
Queen's University Biological Station site (44.56747, -76.32454), brown indicates the Mud lake
site (5.37192, -75.79451) and yellow indicates the Shirley's Bay site (45.36546, -75.88302).
835
836 **Figure 2.** Overview of cold tolerance experiments conducted in this study. Shown are the
837 supercooling point (SCP), low temperature survival assays, and critical thermal minimum
838 experiments (CT_{min}). Sites included in the experiments are: Brockville, Queen's University
839 Biological Station (QUBS), Mud lake and Shirley's Bay. For number of larvae in each
840 experiment, please refer to Table S1.
841
Figure 3. A comparison of model accuracy for the two model extents (northern and full range)
and approaches (correlative and mechanistic) using all three evaluation metrics (AUC (Area
under the receiver operating characteristic curve), Kappa, TSS (True skills test). Mean accuracy
scores and standard error bars across 100 iterations are shown with the y-axis cropped at 0.2.
Grey and black bars represent the mechanistic and correlative models, respectively. Letters
indicate that mean scores are significantly different from each other ($p < 0.05$).