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Geographic variation in vulnerability to warming temperatures in an intertidal barnacle species.

Short title: Geographic variation in vulnerability to warming

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Abstract

For most species, we lack a detailed understanding of how vulnerability to warming temperatures under climate change varies across a its geographic distribution. Vulnerability arises when there is a gap between local climate and local physiology. Intertidal species are unique because they face two distinct thermal environments, and it is unclear which is the bigger driver of thermal physiology and vulnerability. Here we compare the thermal environments and physiology of three populations of the intertidal barnacle *Balanus glandula*, spanning 1460 km of its geographic range. We measured energy consumption in the laboratory across a 5-hour emersion and subsequent 6-hour immersion at 7 different emersion temperatures. We compared these results to one year of emersion and immersion temperature data from each location. Our results suggest that the temperatures experienced during emersion are a bigger driver of each population's thermal physiology than those experienced during immersion. We also estimated vulnerability to future warming in two ways: by calculating the total energy consumption over a year and by calculating the number of days that temperature exceeded each population's thermal peak. These produced conflicting results. The central population spent the most days over its thermal peak, but the northernmost population had the greatest total costs over a year. This difference may be explained in part by a strong latitudinal gradient in primary productivity that selects for higher energy demand in higher latitude populations. Thus, accurate predictions of B. glandula's response to warming temperatures will require knowledge of both future temperature and food availability.

Introduction

Identifying the species, and populations within species, most vulnerable to warming

temperatures under climate change is a pressing goal of ecology [1,2]. A population's

vulnerability is determined by the magnitude of expected warming relative to its current thermal

physiology and to its capacity to alter its physiology through genetic adaptation or phenotypic

plasticity. Populations from different locations within a species' geographic range don't

experience the same temperatures [3,4], unlikely to experience the same amounts of warming

[5], and likely exhibit different degrees of local adaptation and thermal plasticity [6,7]. This is

particularly true for ectotherms, whose body temperatures reflect a complex interaction between

the individual organism and its environment [8–10]. While several theories predict how

vulnerability to warming should vary across a species range [2,11], our understanding is

constrained by a limited number of empirical studies [2]. Without information for a wide range

of species and habitats, it will be difficult to accurately predict and ameliorate vulnerabilities to

climate change.

Hayford et al. [10] distinguish two general approaches to assessing a population's

vulnerability to warming under climate change. Discrete or threshold approaches [e.g., 2,11,12]

focus on identifying the frequency of exposure to a threshold temperature above which

organisms experience thermal stress and/or death. Populations that spend more time over such a

threshold, either currently or under predicted future temperatures, are considered more

vulnerable to warming. Threshold temperatures are usually identified from a thermal

performance curve ("TPC", Fig 1) of a trait related to fitness. Commonly used thresholds include

the critical thermal maximum and lethal limit [13], but the peak of the thermal performance

curve ("T_{peak}", also known as the thermal optimum) may be more relevant physiologically

[12,14] and easier to accurately estimate [15]. Continuous approaches [12,16] calculate the full

energetic cost of exposure to a continuous time series of temperatures for either current or future

climates at that location. Populations that show higher total costs under those current or future

temperatures are predicted to be more vulnerable to warming. Costs are usually estimated from a

TPC curve fitted for oxygen consumption [16,17], with the assumption that costs continue to

increase above T_{peak} (Fig. 1) [18]. Costs can also be incorporated into more detailed energetic

models that predict growth rates and/or energy balance [19,20], although this requires knowledge

of food availability and intake rates, both of which also may change with warming temperatures

[21]. Of the two, the threshold approach is more commonly used because it requires less detailed

information, but it has also been criticized for that lack of detail [12]. It is not known if the two

approaches produce similar predictions.

Figure 1. Stereotypical patterns of thermal performance and energetic cost for a

poikilotherm. For most traits (solid line), performance increases up to a peak beyond which it

decreases. Further warming can lead to reversible (CT_{max}) and irreversible (LT_{max}) loss of

function. Energetic costs of exposure to temperature (dashed line) are usually estimated from a

TPC curve for metabolic rate or respiration, with the assumption that the full cost of exposure

continues to increase exponentially above T_{peak}, even as organism function declines. Figure

modified from Hayford et al. [10].

Intertidal habitats present additional challenges to, and opportunities for, understanding

biogeographic variation in thermal physiology. Intertidal organisms alternate between terrestrial

and marine conditions with the daily oscillation of the tides, each with a distinct thermal

environment. Because terrestrial climates are more variable and exhibit greater thermal extremes

than marine climates, we would expect the two environments to select for very different

physiologies. It is generally unknown which of the two environments is the bigger driver of an

intertidal organism's thermal physiology. Many intertidal species are active mainly when

immersed and minimize activity and metabolism during low tide [22,23]. Thus natural selection

might favor a thermal physiology that is closely matched to the aquatic thermal environment

[24,25]. But intertidal species experience their greatest thermal extremes during emersion and

must survive and maintain some metabolic function during those times. Therefore natural

selection could also favor physiologies that enhance tolerance and/or performance during

emersion. Indeed high shore intertidal species frequently exhibit greater thermal tolerances than

their lower shore congeners, despite experiencing the same aquatic thermal environments [26].

This latter pattern is predicted by the Climate Extremes and Climate Variability Hypotheses

(CEH/CVH), which argue that species, or populations within species, experiencing greater

thermal variability should have broader thermal tolerances [27].

Approximately 60% of the world's coastlines experience diurnal or mixed semi-diurnal

tides [28]. In these tidal regimes, aerial exposure is concentrated in a single low or lower-low

tide, the timing of which may vary with season, latitude, and coastal geomorphology [29]. The

timing of this low tide can decouple emersion temperatures from latitudinal gradients in

terrestrial climate, by filtering the time of day that organisms are exposed to terrestrial climate

[3,30]. For example, in the northeastern Pacific, intertidal sites at higher latitudes in Oregon and

Washington are much more likely to be exposed to air in middle of the day during the summer

than lower latitude locations in California, but the opposite is true in winter (Fig 2B). This

difference in the timing of the lower low tide creates thermal hotspots – high latitude locations

that experience greater daily maximum temperatures than lower latitude locations, despite have a

cooler terrestrial climate overall [3,30]. This decoupling of emersion and immersion thermal

regimes creates a unique opportunity to ask which environment is more important in driving

thermal physiology and potentially test predictions of the CEH/CVH, which have not explicitly

been tested in intertidal species [27]. However, past studies of thermal physiology of NE Pacific

intertidal species provide conflicting results. While strong physiological differences have been

identified between populations north and south of a known biogeographic break at Pt.

Conception, CA (Fig 2A) [e.g., 31]; intraspecific comparisons at higher latitudes have variously

shown associations with low tide temperatures [32], water temperatures [33], or no

differentiation at all [34].

Figure 2. Location of the three populations used in this study (A) and representative

summer and winter spring tides for the sites (B). The double-ended arrow in (A) indicates the

geographic distribution of *B. glandula*, which extends from Mexico to Alaska [35]. The flat-

headed arrow indicates the region of genetic sampling in Sotka et al. [36] and Wares and Skozen

[37], with the genetic cline indicted by the dashed portion. "PC" indicates the thermal break at

Point Conception, California. The predicted tides in (B) for June 15, 2022 ("Summer") and

December 23, 2002 ("Winter") were calculated using the xtide program

(http://tbone.biol.sc.edu/tide/index.html)

Here we extend an earlier study of the thermal physiology of the barnacle *Balanus*

glandula from single a high latitude northeastern Pacific population [38] to two additional sites

spanning 1460 km (Fig 2A) to address two specific research questions: 1) Is terrestrial or marine

climate the stronger driver of thermal physiology? and 2) which population is most vulnerable to

warming temperatures under climate change? *Balanus glandula* is a common mid to high shore

barnacle along the Pacific Coast of North America from Mexico to Alaska [35]. It has also been

introduced to Asia, Africa, and South America [39,40]. For each study population, we measured

respiration over a 5-hour emersion and a subsequent 6-hour immersion at 7 different emersion

temperatures. We compare these results to intertidal and water temperature data collected for one

year at each study site.

Based on previous studies [3,30,41], we expected the aquatic thermal environments of

these three sites to correlate with latitude, but the emersion thermal regimes to be uncorrelated.

Specifically, we expected the southernmost site, Newport Bay (hereafter referred to as SC) to

have the warmest immersion and emersion conditions, because it occurs south of Point

Conception, CA. We expected the northernmost site, Friday Harbor (hereafter referred to as FH)

to have the coldest immersion thermal regime, but an intermediate emersion thermal regime

because it had previously been identified as a high latitude hotspot [3]. These three sites also

span a known genetic cline between two mitochondrial genotypes in *Balanus glandula* [37,42].

The central site, Bodega Bay (hereafter referred to as BB), occurs in the middle of this cline and

the other two sites are beyond its endpoints (Fig 2A). Limited evidence suggests the genetic

cline may reflect adaptation to different thermal regimes [37,42].

To address our first question, we estimated the thermal performance curve (Fig 1) for

each population and tested for differences in the peaks of the curves among populations.

Because we expected the FH site to have warmer low tide and cooler aquatic conditions than the

BB site, we predicted the FH population would have a warmer thermal peak (T_{peak}) than BB if

emersion temperatures influence physiology but the opposite pattern if water temperature is more

important. We also examined temporal patterns of oxygen demand over the duration of

individual emersion and immersion trials to identify additional signs of increased energy demand

or physiological stress that could signify a thermal threshold. To address ours second question,

we calculated a combined TPC across emersion and immersion phases. We used this curve, in

combination with local environmental data, to calculate vulnerability in two ways: the number

of days in which temperatures exceeded T_{peak} and the total annual cost of low tide exposure for

each population at each site.

Methods

Study sites

All three study sites (Fig 2) experience mixed semi-diurnal tidal cycles with two unequal

low tides. At all three sites animals were collected inside bays or fjords that were protected from

most wave action. At the northern Friday Harbor ("FH") site, we collected barnacles by

deploying 15cm x 15cm PVC settlement plates intertidally off the Friday Harbor Labs dock

(48.5457°N, 123.0127W). The dock also contained a NOAA monitoring station [43], which

records water temperatures every 6 minutes. Intertidal temperature dataloggers were placed on a

west-facing rock wall, approximately 100 m from the collection site. Additional details for this

site are described in Ober et al. [38]. At the central site in Bodega Bay CA ("BB", 38.3037N,

123.0532°W), barnacles were collected both growing on mussel shells within the bay and on

PVC settlement plates attached to a rope hung between pieces of iron rebar [44]. Temperature

dataloggers were placed on the northern face of the southern jetty, approximately 500-1000m in

between the two collecting sites. Water temperature data for BB came from the CeNOOS

Automated Shore Station at Bodega Marine Laboratory [45], approximately 2km away, which

records every 6 minutes. At the southern field site (southern California "SC"), barnacles were

collected on oyster shells growing on an artificial jetty on the eastern side of the mouth of

Alamitos Bay (33.7484N, 118.1155W). Temperature data loggers were placed on a southeast-

facing artificial sea wall in Newport Bay (33.6022°N,117.8824°W), approximately 27 km from

the collecting site. Water temperature data came from the SCCOS Automated Shore station at

Newport Pier [46], approximately 1.7 km from the sea wall.

Intertidal temperature measurement

Intertidal temperature dataloggers were established at each site as part of separate study

of barnacle growth rates in the field (Gilman et al. *in prep*). For this study we used 1 year of data

from loggers located in the approximate vertical center of the barnacle zone. The dataloggers

were affixed directly to the rock with epoxy and recorded temperature every 15-20 minutes.

This method directly mimics barnacle temperatures [41,47], as a large proportion of a barnacle's

surface area is in direct contact with the rock. At FH and BB, we used three replicate Tidbit

loggers (v1 or v2 Onset Computer Corp., Bourne, MA, USA). At SC, two replicate DS-1921G

ibutton dataloggers (Analog Devices Inc., Wilmington MA, USA) were used. Loggers were

downloaded and replaced every 6 months at BB and FH and every 2-4 weeks at SC.

We calculated daily physiological maxima (DPM) for the intertidal and water temperature

datasets as the 97.9 percentile of the daily data, after first selecting the maximum temperature

among any replicate dataloggers within each site and timepoint. This percentile reflects a

temperature exposure of at least 30 minutes, corresponding to the minimum time threshold for an

organism to experience physiological consequences [5,48].

Animal collection and maintenance

As described in Ober et al. [38], collected barnacles were isolated, tagged, and their maximum opercular diameter measured. Barnacles were maintained in the laboratory in a tidally cycling tank with a daily light cycle (Table 1). They were fed brine shrimp (*Artemia spp.*, Pentair Aquatic Eco-Systems, Apopka, FL 32703) two to three times per week [49]. Laboratory tidal regimes (Table 1) were based on local tidal cycles for the warmest part of the year at each site (approximately July-October for SC and BB and April – August for FH), at a height equivalent to the middle of the barnacle zone. The acclimation water temperatures were calculated by averaging at least one year of water temperature data from the nearest water monitoring station. Animals were acclimated to the conditions for at least a week before each respiration trial and re-acclimated for at least one week between trials. They were starved for 48 hours prior to each trial.

Table 1. Laboratory tidal cycles and abiotic conditions used for each population.

	Friday Harbor, WA	Bodega Bay, CA	Southern California
Low Tide 1		7:30am – 12:30 pm	
High Tide 1		12:30pm – 7:30pm	
Low Tide 2	7:30pm – 10:30pm	7:30pm-3:30am	7:30pm – 1:30am
High Tide 2	10:30pm – 7:30am	3:30am-7:30am	1:30am – 7:30 am
Lights on	03:30am – 7:30pm	04:45am – 07:00pm	
Water Temperature	10°C	12°C	18°C
Air Temperature		18-20 °C	

Respiration experiment

Full methods for the respiration protocol are reported in Ober et al. [38]. Briefly, about
30 minutes before the end of the "High Tide 2" phase of the laboratory tidal cycle (Table 1),
barnacles were placed into individual aerial respiration chambers filled with filtered seawater.
These were submerged in a temperature-controlled water bath at the acclimation water
temperature. After 30 minutes, the water was siphoned out of the chambers and the water bath
was heated or cooled until the chambers reached the experimental temperature. They were
maintained at the experimental temperature for the remainder of the 5-hr low tide period.
Oxygen concentrations in the chambers were recorded every 10-seconds with a fluorometric
oxygen system with non-invasive sensor spots (Fibox 4 with Pst3-NAU spots, PreSens Precision
Sensing, Regensburg, Germany). An empty chamber was included as a control. Afterwards, the
barnacles were transferred to individual aquatic respiration chambers and submerged in a second
water bath at the population's acclimation water temperature. Aquatic respiration was monitored
for 6 hours as intertidal species can show elevated oxygen demand after emersion ["oxygen
debt", 50].

While Ober et al. [38] used a fixed ramping rate of 10°C hr-1 for the FH population, this
study used a fixed ramping period of 2.5 hrs. To facilitate comparisons among sites, we only
used emersion data from the last 2.5 hours of each emersion trial, when all trials were at the
experimental temperature. Additionally, in the present study we used 8 blocks of 4 barnacles
from SC and BB, with each block used at two different temperatures, for a total of 8 barnacles
per temperatures. The FH experiment used 9 barnacles per temperature, in blocks of 3.

Calculation of respiration rates

Slopes of change in oxygen concentration were calculated every five minutes for the last 2.5 hours of the emersion phase and the full six hours of the immersion phase. Slopes were only calculated for intervals with at least 10 data points. We divided the emersion observations by the control value before calculating the slope, in order to correct for the effect of pressure changes on the oxygen readings within the vials during the ramping phase. For aquatic data, we subtracted the slope of the control chamber from each experimental chamber to control for backgrounds rates of oxygen consumption within the seawater. If a control slope was missing for a five-minute aquatic interval, we substituted the average control slope for that trial. We used the five-minute slopes to calculate mean respiration rates for every 30 minutes of the last 2.5 hours of the emersion phase and for every hour of the immersion phase. For this we used 10% trimmed means to systematically exclude outliers. All calculations were completed using the Presens Oxygen Calculator (v 2.2.6) and R (v 4.2.2).

Statistical analysis

All statistical analyses were done in R 4.2.2, using the packages lme4 v. 3.1.160 [51], nlme v. 1.1.31 [52], car 3.1-1[53], and emmeans 1.8.3 [54]. Models were tested for departures from normality and homoscedasticity as described in [55]. In all models that included size as covariate, we used (Operculum Length)$^{1.8533}$, based on a previously reported allometric relationship [49]. The fixed effects were tested with type II likelihood ratio χ^2 tests, using the Anova function of car. We adjusted for multiple comparisons among sites using the Tukey method. We used the deltaMethod function of the car package to compare nonlinear parameter combinations.

To compare the intertidal and water DPMs, we used analysis of variance with

autocorrelated errors in the gls function of nlme. We also compared the 75%ile of the intertidal

temperatures using the rq function of the quantreg package v. 5.94 [56].

To compare thermal physiology thresholds among populations, we fit quadratic thermal

performance curves in each medium and compared the location of T_{peak}. We first averaged

across the hourly or half-hourly time points to calculate an average respiration rate per barnacle.

We then estimated the thermal performance curves for each population in each medium as a

parabola of the form: $\mu mol\ O_2 = \beta_{1,P} + \beta_{2,P}T + \beta_{3,P}T^2 + \beta_4 Op^{1.8533}$. Where P=source population,

T=temperature, and Op=operculum length. From this we calculated T_{peak} for each population as

$\frac{-\beta_2}{2\beta_3}$-. We selected this simple quadratic curve as curves with fewer parameters are

recommended when there are a small number of study temperatures [15]. We included barnacle

identity as a random factor to control for repeated measures on the same barnacle.   Due to

significant error heteroskedasticity, both models included extra error terms to allow for

heterogeneous variances among temperature treatments (immersion) or temperatures and source

populations (emersion).

We also looked for temperature thresholds by comparing the temporal pattern of oxygen

consumption in each medium. For example, a decreasing emersion respiration rate over time

could suggest physiological impairment, while the same pattern during immersion would be

consistent with an oxygen debt [50]. For this analysis, we fit separate mixed-model ANCOVAs

for each medium. These models treated temperature (T) as a categorical variable and also

included source population (P) and time point (H) as categorical factors and Operculum Length

(Op) as a covariate. They were of the form: $\mu mol\ O_2 = \beta_1 + \beta_2 P + \beta_3 H + \beta_4 T + \beta_5 PH + \beta_6 PT +$

$\beta_7 HT + \beta_8 PHT + \beta_9 Op^{1.8533}$. Barnacle identity was included as a random factor to control for

repeated measurements on the same barnacle. Due to significant error heteroskedasticity, both

models included an extra error term to allow heterogeneous variances among trial dates. One

data point for an FH barnacle at 38°C at the 30-minute point was excluded from the emersion

data set, as it was 5-fold larger than any other data point. Polynomial contrasts were used to

identify respiration patterns over time within each combination of source population and

experimental temperature.

Vulnerability to warming

To estimate each population's vulnerability to warming temperatures, we first calculated

the cost of a 30-emersion at each temperature for each phase of the trial, using the estimated

marginal means (EMM) from the analyses above. For the cost during emersion, we multiplied

the 5-minute EMM values from the emersion TPC by 6 to scale up to 30 minutes. For the cost

during immersion, we subtracted off a basal rate of oxygen consumption from the immersion

TPC EMMs, using the EMM for hour 6 of each trial, and then multiplied by 72 to get the full

oxygen debt for the 6-hr immersion trial. We divided this value by 5 to adjust it from the 2.5-

hour emersion exposure of the trial to a 30-emersion exposure. Finally, we summed the

emersion and immersion values to get the total energetic cost of a 30-minute emersion at each

temperature.

We used these values to calculate vulnerability in two ways: as the number of days spent

above T_{peak} and as the total costs of emersion over the year. In the first case, we fit a new TPC

parabola for each population from the total cost values and calculated T_{peak} as described above.

In the second case, because costs increase with temperature even though TPC curves decline at

high temperatures (Fig 1), we used the regression approach of Ober et al. [38] to calculate an

exponentially increasing total cost function based on the metabolic theory of ecology equation

[21,57], excluding temperatures above the peak of the total cost curve (or below the minimum

for SC).

Results

Field temperatures

There was clear pattern of declining water temperatures with increasing latitude, but no

such pattern for intertidal temperatures (Fig 3). Both intertidal ($F_{2,1093} = 21.8$, p<0.0001) and

water ($F_{2,1093} = 38.02$, p<0.0001) DPMs showed significant differences among sites, driven

mainly by the southernmost SC site. Compared to BB and FH, the mean intertidal DPM at the

SC site warmer 7-10°C warmer and the mean water DPM were more than 6°C warmer (p <

0.01). The northernmost FH site had 3°C cooler water DPMs and 1.5°C warmer intertidal DPMs

than the BB site, but these differences were not significant (p >0.05). However, a comparison of

the 75th percentile of the intertidal DPMs, using quantile regression, revealed statistically

significant differences among all pairs of populations (p<0.004 for each comparison), with the

northernmost (FH) site being 4.5°C cooler than SC and 1.9°C warmer than BB.

Figure 3. Box plot of daily 97.9%ile of intertidal data logger and water temperatures at

each site over one year. The top and bottom of the box indicate the 25th and 75th percentiles.

The notched line is the median. The notch height estimates 95% CIs around the median.

Thermal performance curves

When respiration rates were pooled over timepoints, both the immersion and emersion

analyses revealed a statistically significant quadratic term, indicating that parabolas fit the

respiration data better than linear regression (Table 2A). There were also significant interactions

between population and both the linear and quadratic temperature terms, indicating differences in

the shape or position of the parabolas among the three populations (Table 2A, Fig 4). The fitted

parabolas (Table 2B) for emersion for the two most northern populations were concave, with

T_{peak} estimated at 27.5°C (23.8-31.2 95%CI) for FH and 24.9°C (19.2-30.5 95%CI) for BB (Fig

4a and 4c). In contrast, the parabola for SC (Table 2B) was convex (Fig 4e), with a minimum

respiration rate at 20.0°C (14.4-25.6 95%CI). The shallow, convex shape of the SC curve

suggests that we did not measure temperatures high enough to capture T_{peak} for this population.

A similar pattern was seen in the immersion curves (Table 2B). The two most northern

populations were concave, with T_{peak} estimated at 23.6°C (22.3-25.0 95%CI) for FH and 19.4 °C

(13.2-25.5 95%CI) for BB (Fig 4b and 4d). But the southern site (SC) again showed a convex

pattern (Fig 4f), with a minimum respiration rate at 20.9°C (4.9 – 36.9 %CI).   Multiple

comparisons tests revealed a significant difference in the location between the emersion peaks

for FH and SC (p=0.028), but no other differences were significant (p>0.05).

Table 2. Model results for average oxygen consumption, treating temperature as a

continuous variable. Analysis of deviance statistics for each medium (A) and parameter

estimates (+se) for the line of best fit from the model for each population (B).

A)		Emersion			Immersion		
	Term	χ^2	df	p	χ^2	df	p
	Population	32.5011	2	<0.0001	3.2237	2	0.1995
	Temperature	2.5823	1	0.1081	13.0248	1	0.0003
	Temperature^2	0.8269	1	0.3632	16.6997	1	< 0.0001

Size	5.1147	1	0.0237	12.8529	1	0.0003
Pop x Temp	26.224	2	<0.0001	13.5512	2	0.0011
Pop x Temp^2	32.5011	2	<0.0001	16.1523	2	0.0003

B)	**Bodega Bay**	**Friday Harbor**	**Southern California**
Emersion			
Intercept	-0.03280 (0.02094)	-0.09348 (0.0235)	0.00059 (0.01494)
Temperature	0.00122 (0.00056)	0.00353 (0.00185)	0.00935 (0.00201)
Temperature^2	-0.00142 (0.00101)	-0.00007 (0.00004)	-0.00017 (0.00005)
Operculum Length	0.00004 (0.00002)	0.00004 (0.00002)	0.00004 (0.00002)
Immersion			
Intercept	-0.01169 (0.01698)	-0.05631 (0.01585)	-0.00053 (0.01652)
Temperature	0.0022 (0.00138)	0.00546 (0.00117)	-0.00071 (0.00132)
Temperature^2	-0.00006 (0.00003)	-0.00012 (0.00002)	0.00002 (0.00003)
Operculum Length	0.00131 (0.00038)	0.00131 (0.00038)	0.00131 (0.00038)

346 **Figure 4. Oxygen consumption by temperature for each population during emersion (A-C)**

347 **and Immersion (D-F)**. Parabolas represent the predicted respiration rates from the mixed

348 model, calculated at the mean overall size of the experiment (5.05 mm operculum length).

349 Dashed lines estimate T_{peak} values. The individual points represent mean respiration rates for

350 individual barnacles, with the darkness of the fill indicating size.

Temporal patterns of respiration by medium

Emersion

The emersion respiration rates ranged from 0.0018 to 0.057 μmol/5min, depending on the source population, temperature, and time point (Fig 5). All terms in the mixed model ANCOVA for emersion were statistically significant, including the three-way interaction between source population, time, and treatment temperature (Table 3). There was also a significant main effect of population, with the highest overall respiration rates at FH at most temperatures. Most combinations of source population and temperature showed little variation across timepoints, with two exceptions. Within the southernmost (SC) population, all but the two lowest and one highest temperature showed a significantly increasing linear trend in respiration rates over the 2.5 hr. period. Above 20 °C, both BB and FH showed linearly decreasing trends over time.

Figure 5. Average emersion oxygen consumption by temperature and time point for each population. Gray lines represent data for individual barnacles, darkness reflects size. Estimated marginal means (±95CI), for a barnacle of 5.05 mm operculum length, are shown in red. Letters indicate significant polynomial contrasts (p<0.05, L=Linear, Q=Quadratic)

Table 3. Analysis of deviance for respiration by population and timepoint during emersion and immersion.

	Emersion			Immersion		
Term	χ^2	df	p	χ^2	df	p
Population	36.6454	2	<0.0001	4.8558	2	0.0882
Time	7.7978	4	0.0993	76.7112	5	<0.0001

Temperature	204.9101	6	<0.0001	288.4884	6	<0.0001
Size	4.5776	1	0.0324	8.8019	1	0.0030
Pop x Time	71.1303	8	<0.0001	10.7962	10	0.3736
Pop x Temp	237.3474	12	<0.0001	349.9856	12	<0.0001
Temp x Time	235.9039	24	<0.0001	97.2488	30	0.0000
Pop x Time x Temp	294.976	48	<0.0001	128.1517	60	0.0000

Immersion

Average immersion respiration rates ranged from 0.0071 to 0.0431 µmol/5min,

depending on the source population, temperature, and time point (Fig 6). The mixed model

analysis revealed a significant overall effect of emersion temperature on immersion respiration,

as well as significant differences among populations in the form of interactions between

population and temperature and between population, temperature, and hour (Table 3). There

were three distinct temporal patterns among the temperatures (Fig 6). At the coolest

temperatures we frequently detected a significant quadratic trend, where immersion respiration

rates tended to be lowest in the first hour post emersion, peak in the second or third hour, and

then decline. This pattern occurred at temperatures between 10-20°C at the central BB site,

between 10-25°C at the northern FH site and at temperatures up to 30°C at the southern SC site,

although the quadratic trend was not always statistically significant (Fig 6).   Above these

temperatures, the pattern shifted to a linear negative trend, indicating higher respirations rates in

the first 1-2 hours post emersion. This occurred at 25 and 35°C for BB, 30 and 35°C for FH and

35°C for SC. The transition between the first two patterns for FH and BB corresponds with the

estimated emersion T_{peak} values for those populations. Finally, at 35 and 38°C for BB and at

38°C for the other two sites, respiration was initially depressed and then increased over time.

Figure 6. Aquatic oxygen consumption by temperature and hour for each population.
Estimated marginal means (±95CI), for a barnacle of 5.05 mm operculum length, are shown in
red. Gray lines are data for individual barnacles, with the darkness of the line indicating size.
Letters indicate significant polynomial contrasts (p<0.05, L=Linear, Q=Quadratic). Colored bars
indicate qualitative assignment of respiration rates pattern to non-stressful (Blue), increased
energy demand (Orange), and physiological damage (Yellow).

Total cost of emersion and vulnerability to warming

The estimated total cost of a 30-minute emersion ranged from -0.07μmol for BB
barnacles at 38°C to 0.66μmol for FH barnacles at 25°C (Fig 7A). On average, immersion
respiration accounted for 18% of the total cost of exposure, but this ranged from 0 to nearly 50%,
depending on the temperature and population. As before, the fitted parabolas for the total cost for
the two most northern populations were concave (Table 4), with T_{peak} estimated at 25.1°C
(25.12-25.12%CI) for FH and 21.3 (21.3-21.3%CI) for BB (Fig 7); while the parabola for SC
was convex, with a minimum cost at 20.5°C (20.5-20.5 95%CI, Table 4). The MTE-based cost
curves showed increasing costs with latitude (Fig 7A, Table 4).

When we compared the number of days above T_{peak} for each site, substituting 38°C for
the SC T_{peak} (Fig 7C), we found BB to be most vulnerable, with 149 days over T_{peak}, followed by
FH (93 days) and SC (4 days). The total annual costs (Fig 7B) followed the pattern of the MTE
costs curves (Fig 7A), with FH showing the highest energy costs and SC the lowest.

411 **Figure 7. Total cost of a 30-minute emersion exposure by temperature (A), total annual**

412 **cost for each population (B), and number of days over T_{Peak} (C).**

414 **Table 4. Parameter values (±SE) for the total costs equations fitted for each population.**

	Friday Harbor	Bodega Bay	Southern California
Parabola TPC			
Intercept	-1.04 (1.63E-15)	-0.285 (1.63E-15)	0.257 (1.63E-15)
Temperature	0.135 (1.49E-16)	0.0529 (1.49E-16)	-0.0187 (1.49E-16)
Temperature^2	-0.00268 (3.04E-18)	-0.00124 (3.04E-18)	0.000457 (3.04E-18)
Exponential TPC			
-E	48.43 (11.32)	22.53 (17.74)	17.77 (8.99)
ln(R$_o$)	-14434.01 (3287.7)	-6961.4 (5111.17)	-6042.32 (2721.12)

Discussion

Identifying the species and populations most vulnerable to climate change is a critical

priority for ecologists. Studies of intraspecific biogeographic variation in thermal physiology,

particularly of species that do not experience strong latitudinal clines in thermal conditions, are

needed to test and further develop models of geographic-scale thermal physiology. Overall, our

results suggest that, consistent with the Climate Extremes and Climate Variability hypotheses,

the extreme temperatures experienced during emersion are driving the thermal physiology of

these three populations of *Balanus glandula*. We found that the peaks (T_{peak}) of the respiration

curves in all three populations were well above the range of aquatic temperatures at each site and

better matched to differences in the emersion thermal regimes among sites. These conclusions

are tempered by the small thermal differences between our two northernmost sites which limited

the statistical significance of some comparisons. We also found different patterns of

vulnerability to warming temperatures, depending on the metric of vulnerability used. The most

vulnerable site according to the cost approach was the northernmost site, while it was the central

site according to the threshold approach. This discrepancy arises from the large differences we

observed in energy demand among our three populations, which may reflect local adaptation to

variation in nearshore ocean productivity and food supply.

Intertidal organisms experience two distinct thermal regimes during emersion and

immersion and it unclear which drives thermal physiology, particularly at warmer temperatures

[24,26,33,34]. Three lines of evidence in this study suggest that the extreme temperatures

experienced during low tide conditions drive the thermal physiology of *B. glandula*. First, the

identified T_{peak} of each population is well above its site's aquatic DPMs. A population's T_{peak}

should exceed its average DPM to minimize the amount of time spent at stressful temperatures

[58]. But, these T_{peak} values exceed even the maximum water temperatures observed at each site.

Second, the ordering of the thermal peaks of the northern FH and central BB sites match the

ordering of emersion DPMs rather than immersion DPMs. The emersion temperature

differences at these two sites were small and sensitive to the temperature metric used (mean vs.

75%ile). However, similarly small temperatures differences can greatly influence intertidal

community composition and population dynamics [59]. We found similar results in a separate

study that measured aquatic respiration rates in barnacles acclimated to different immersion

temperatures (Ober et al. *in prep*): aquatic respiration rates in all three populations peaked at

water temperatures well over local maxima and the northern FH population again had a warmer

T_{peak} than BB.

It also appears that the T_{peak} values for the southernmost (SC) population are greater than

the highest temperature we tested (38°C), placing them well above the T_{peak} values for BB and

FH. This is surprising, given that a 40°C emersion is lethal for FH barnacles [38]. However, it

is consistent with the warmer overall temperatures at this site and with previously identified

genetic differences among populations [37,42]. Unfortunately, because both the aquatic and

intertidal thermal environments at the southern SC site are so much warmer than the two other

sites, comparisons with SC do not inform the question of whether aquatic or terrestrial

environments are the main driver of these physiological differences.

The third line of evidence that emersion conditions drive warm temperature

physiology is in the temporal patterns of oxygen use during the immersion phase of the

experiment. We detected three distinct temporal patterns of respiration during immersion.

Consistent with FH having a more warm-adapted physiology than BB, the transition between

each pattern occurred at a warmer temperature for FH than for BB. At the coolest temperatures,

barnacles exhibited parabolic patterns with initially low respiration rates that increased in the

second or third hour of immersion and then declined. This pattern indicates a non-stressful

emersion temperature. We hypothesize that barnacles initially reduced respiration upon

immersion because they had access to sufficient oxygen during emersion and were not

experiencing high energy demand. Video observations of the FH population [38] indicate that *B.*

glandula are frequently closed in the first few hours of immersion at these cooler temperatures.

Similar patterns of low post-emersion oxygen demand have been reported in other high shore

barnacle species [60]. The concentration of oxygen in air is roughly 20-fold higher than in

seawater, so oxygen may be more accessible during emersion at moderate temperatures then

during immersion [61]. By staying closed barnacles may also reduce to vulnerability predation

during immersion [62].

At warmer temperatures, barnacles began immersion with higher initial respiration rates

that declined gradually over the 6-hr immersion period. These barnacles also often showed a

decline in emersion respiration over time. The high initial immersion activity was also observed

in video data recorded from FH barnacles [38] and suggests that high energy demand during

emersion created an immediate need for respiration upon resubmersion. This is typically

described as an oxygen debt in the literature [50,63]. Finally, at the warmest temperatures,

respiration was depressed and remained low for several hours. This third pattern suggests

physiological impairment.   At these temperatures the barnacles often appeared in a heat coma

during immersion with cirri extended but not moving.

The second goal of this study was to assess each population's vulnerability to warming

temperatures under climate change. Our assessment of vulnerability varied with the metric of

vulnerability. When we calculated the total annual costs of exposure to intertidal DPM

temperatures, the northernmost FH population had the greatest total energy cost among the three

sites. However, when we calculated the number of days above T_{peak}, the central BB site

appeared most vulnerable. We were unable to identify any published studies that compared

these two metrics of vulnerability, so it is unclear how common our result is. But it suggests that

estimates of vulnerability to warming are highly sensitive to the metric used.

One possible explanation for the greater energetic costs at FH may be differences in food

availability at each site. *Balanus glandula*'s exact diet is unknown [64], but they likely consume

a range of phytoplankton and zooplankton, as adults can thrive on either in the laboratory

[49,65,66]. There is a strong latitudinal gradient in primary productivity throughout this region

of the northeastern Pacific, with coastal Washington State and British Columbia exhibiting 5-

fold greater chlorophyll concentrations than California [67,68], along with higher zooplankton

concentrations [67]. This gradient is attributed to greater terrestrial nutrient supply at higher

latitudes, particularly in the Salish Sea region [68], which includes the FH field site. A greater

food supply could select for greater energy use in the two higher latitude populations, while the

SC (southernmost) barnacles may be adapted to succeed in a lower resource environment. This

is an alternate explanation for previously described genetic cline in *B. glandula* [36,42].

501 Conclusion

The results presented herein provide strong evidence that thermal conditions experienced

during emersion do influence the thermal physiology of the intertidal barnacle *Balanus glandula*.

However, more empirical studies are needed to fully test the Climate Extremes and Climate

Variability Hypotheses in intertidal species. The possibility that *B. glandula* populations are

locally adapted or acclimated to both thermal regimes and food supplies complicates predictions

of their response to warming temperatures. While its well-established that food supply can limit

many ectotherms' tolerance of thermal stress [e.g., 69,70], few models of climate change

responses have considered both warming temperatures and changes in food supply at the same

time (but see [71]). While its higher overall energy demand suggests that the FH population is

the most vulnerable to warming temperatures, phytoplankton productivity is also predicted to

increase in this region under climate change [72]. Thus, in this case, the threshold approach may

provide a better estimate of vulnerability to future warming for *Balanus glandula*.

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