1

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

Running Head: Geographic variation in warming vulnerability

Sarah E. Gilman[1*], Gordon T. Ober[1,3], Rhiannon L. Rognstad[1,4], Madeleine Bunnenberg-Ross[2],

and Tingyue Man[2]

[1]W. M. Keck Science Department of Scripps, Pitzer, and Claremont McKenna Colleges,

Claremont, CA USA

[2]Pitzer College, Claremont, CA USA

[3]Present Address: Environmental Science Department, Endicott College, Beverly, MA USA

[4]Present Address:  Data Lab, World Resources Institute, Washington DC, USA

*Corresponding author

E-mail: sgilman@natsci.claremont.edu

**Abstract**

Vulnerability to warming temperatures under climate change 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 (10-38°C).  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 emersion thermal physiology than those experienced during immersion.  We also

estimated vulnerability to future warming in two ways:  as the total annual energy demand and as

the number of days above 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.   The higher energetic costs in the

northernmost population may be explained in part by a strong latitudinal gradient in primary

productivity that is selecting 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.

Keywords: macrophysiology, climate change, intertidal, respirometry, *Balanus glandula*,

cirripede, thermal ecology, intraspecific variation, marine ecology

# 1   Introduction

Identifying the species, and populations within species, most vulnerable to warming temperatures under climate change is a pressing goal of ecology (Helmuth et al. 2002, Bennett et al. 2019).   A population's vulnerability is determined by the magnitude of expected warming relative to its current thermal physiology and to its capacity to alter that physiology through genetic adaptation or phenotypic plasticity.  Populations from different locations within a species' geographic range don't experience the same temperatures (Helmuth et al. 2002, Deutsch et al. 2008), are unlikely to experience the same amounts of warming (Gilman et al. 2006, Foulk et al. 2024), and likely exhibit different degrees of local adaptation and thermal plasticity (Gunderson et al. 2017, Gaitan-Espitia et al. 2017).  This is particularly true for ectotherms, whose body temperatures reflect a complex interaction between the individual organism and its environment (Porter & Gates 1969, Fey et al. 2019, Hayford et al. 2021).

Intertidal habitats present additional challenges to, and opportunities for, understanding inter-population variation in vulnerability to climate change.  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 (McMahon 1990, Sokolova & Portner 2001, Marshall et al. 2011).  Thus natural selection might favor a thermal physiology that is closely matched to the aquatic thermal environment (Sanford 2002, Iles 2014). 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 (Somero 2002,

Stillman 2003). Distinguishing the influences of emersion and immersion conditions on

population-level thermal physiology requires comparing multiple populations with contrasting

emersion and immersion thermal environments.

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

tides (Byun et al. 2023).  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

(Denny 2007).  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 conditions (Helmuth et al. 2002, 2006).  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. 1B).  This decoupling of emersion and immersion thermal

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

thermal physiology.  However, past studies of thermal physiology of NE Pacific intertidal

species provide conflicting results.  While physiological differences have been identified

between populations north and south of a known thermal break at Pt. Conception, CA (Fig. 1A;

e.g., Blanchette et al. 2007, Gleason & Burton 2013); intraspecific comparisons at higher

latitudes have variously shown associations with low tide temperatures (Sagarin & Somero 2006,

Kuo & Sanford 2009), water temperatures (Rao 1953), or patterns uncorrelated with both (Logan

et al. 2012).

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

*glandula* from single a high latitude northeastern Pacific population (Ober et al. 2019) 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 (Morris et al. 1980). It has also been introduced to Asia, Africa, and South America

(Geller et al. 2008, Simon-Blecher et al. 2008). For each study population, we measured

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

temperatures (10, 15, 20, 25, 30, 35, 38°C). We used long-duration trials to fully capture the

costs of low tide exposure, as metabolic rates can vary over the duration of a low tide exposure

(McGaw et al. 2015, Ober et al. 2019, Griffen et al. 2024).

Based on previous studies (Helmuth et al. 2002, 2006), 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 population, Alamitos 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 (Helmuth et

al. 2002).  These three sites also span a known genetic cline between two mitochondrial

genotypes in *Balanus glandula* (Wares & Skoczen 2019, Wares et al. 2021).  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).  There is some evidence to suggest that the genetic cline

reflects adaptation to different thermal regimes (Wares & Skoczen 2019, Wares et al. 2021).

To assess vulnerability to warming, we compared each population's thermal physiology

to intertidal and water temperature data collected for one year at each study site.  Hayford et al.

(2021) distinguish two general approaches to assessing a population's vulnerability to warming

under climate change.  Discrete or threshold approaches (e.g., Sunday et al. 2019, Bennett et al.

2019, Buckley et al. 2022) quantify 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. 2) of a trait related to fitness. Commonly used thresholds

include the critical thermal maximum and lethal limit (Sunday et al. 2010), but the peak of the

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

biologically (Tagliarolo & McQuaid 2015, Buckley et al. 2022) and easier to accurately estimate

(Low-Décarie et al. 2017).  Continuous approaches (Fly et al. 2012, Montalto et al. 2016)

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 can be

estimated from a TPC curve fitted for oxygen consumption, if one assumes that costs continue to

increase above $T_{peak}$ even as performance declines (Fig. 2; Sanford 2002, Dell et al. 2011,

Lemoine & Burkepile 2012, Ober et al. 2019).   Costs can also be incorporated into more

detailed energetic models that predict growth rates and/or energy balance (Mislan et al. 2014,

Monaco et al. 2014), although this requires knowledge of food availability and intake rates, both

of which also may change with temperature (Dell et al. 2011, Englund et al. 2011, Iles 2014).  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 (Buckley et al. 2022).  It is not

known if the two approaches produce similar predictions.

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 test for changes in respiration over time.  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.

## 2    Methods

### 2.1    Study sites and animal collection

The three study sites (Fig. 2) were selected based on their tidal regimes and accessibility

for ecological study. All three sites experience mixed semi-diurnal tidal cycles with two unequal

low tides on most days.  At all three sites animals were collected inside bays or fjords that were

protected from most wave action.  The northernmost site ("FH", 48.5457°N, 123.0127W) falls

within the Friday Harbor Marine Reserve, managed by the University of Washington, in the

Salish Sea.  Ober et al. (2019) collected barnacles by deploying 15cm x 15cm PVC settlement

plates intertidally off the Friday Harbor Labs dock.  The central site ("BB") was inside the mouth

of Bodega Bay CA ("BB", 38.3037N, 123.0532°W). Here, barnacles were collected both

growing on mussel shells within the bay and on PVC settlement plates attached to a rope hung

intertidally between pieces of iron rebar installed in the sediment (Quinn et al. 1989).  For the

southern population (southern California, "SC"), we collected barnacles on oyster shells growing

on an artificial jetty on the eastern side of the mouth of Alamitos Bay (33.7484N, 118.1155W).

**2.2    Temperature Measurements**

Intertidal temperature dataloggers were established near each collection site as part of

separate study of barnacle growth rates in the field (Gilman et al. *unpubl. data*).  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 Zspar epoxy and recorded

temperature every 15-20 minutes.  This method directly estimates barnacle temperatures

(Wethey 2002, Helmuth et al. 2016), 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), spaced several meters apart.  The FH datalogger site was

on a west-facing rock wall, approximately 100 m from the collection site.  The BB datalogger

site was on the jetty inside the mouth of Bodega Bay and within the area used to collect

barnacles.  For SC, two replicate ibutton dataloggers (DS-1921G or DS-1922L, Analog Devices

Inc., Wilmington MA, USA) 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.  Loggers were

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

Water temperature data was downloaded from established monitoring stations near each

site. The FH site contains a NOAA monitoring station (9449880,

https://tidesandcurrents.noaa.gov/stationhome.html?id=9449880). Water temperature data for BB

came from the Bodega Ocean Observing Node (https://boon.ucdavis.edu/data-access/products)

approximately 2km from the collecting site. Water temperature data for SC came from the

SCCOS Automated Shore station at Newport Pier (https://erddap.sccoos.org/erddap/index.html),

approximately 1.7 km from the sea wall with the intertidal dataloggers.

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 and is more physiologically relevant than the

absolute daily maximum, which may only be experienced for a minute (Helmuth & Hofmann

2001, Fitzhenry et al. 2004, Gilman et al. 2006, Helmuth et al. 2006).

**2.3   Laboratory acclimation conditions**

As described in Ober et al. (2019), 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 (Gilman et al. 2013).

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 local water

monitoring station. We acclimated each population to its local average water temp, rather than a

common water temperature amongst the three populations, in order to capture ecologically

relevant responses to the emersion treatments.  Additionally, there is little overlap in water

temperatures between SC and the other two sites (Fig. 3), meaning a common water temperatures

would be stressful to at least one population.  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.

### 2.4    Respiration experiment

Full methods for the respiration protocol are reported in Ober et al. (2019).  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 20ml glass 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 interior of the chambers reached the experimental temperature.  They remained

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.

While Ober et al. (2019) 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.

**2.5  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).

**2.6  Statistical analyses**

All statistical analyses were done in R 4.2.2, using the packages lme4 v. 3.1.160 (Bates et

al. 2015), nlme v. 1.1.31 (Pinheiro et al. 2022), car 3.1-1(Fox & Weisberg 2019), and emmeans

1.8.3 (Lenth 2022).  In all models that included size as covariate, we used (Operculum

Length)$^{1.8533}$, based on a previously reported allometric relationship (Gilman et al. 2013).  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 25%, 50%, and 75%ile

of the intertidal temperatures using the rq function of the quantreg package v. 5.94 (Koenker

2022).

All respirometry data were analyzed in mixed models following Zuur et al (2009).  First a

full model was built, with all possible random and fixed terms.  The random terms, which

included trial date, individual barnacle, and block, were selected using AIC and this model was

tested for departures from normality and heteroskedasticity.  The fixed effects were then selected

in backwards stepwise approach, starting with the highest order interaction terms, with type II

likelihood ratio $\chi^2$ tests.  Normality and heteroskedasticity were re-tested in this final model.

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 (Low-Décarie et al. 2017).  We included

only barnacle identity as a random factor in both the emersion and immersion models, as it had

better AIC values than any other combination of random terms.   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).

The long duration of the trials also allowed us to look for temporal pattern in respiration

rates. 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.

We tested for temporal patterns by fitting a separate mixed-model ANCOVA 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: $\mu mol\ O_2 = \text{\ss}_1 + \text{\ss}_2 P + \text{\ss}_3 H + \text{\ss}_4 T + \text{\ss}_5 PH + \text{\ss}_6 PT + \text{\ss}_7 HT +$

$\text{\ss}_8 PHT + \text{\ss}_9 Op^{1.8533}$. We included only barnacle identity as a random factor in both the emersion

and immersion models, as it had better AIC values than other combinations of random terms.

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.

**2.7   Vulnerability to warming**

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

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

estimated marginal means (EMM) from the quadratic models described above. For the cost

during emersion, we multiplied the 5-minute EMM values from the emersion model by 6 to scale

up to 30 minutes. For immersion, we subtracted off a baseline rate of oxygen consumption from

the 5-minute immersion EMMs, using the EMM for hour 6 of from the temporal analysis as the

baseline, 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

oxygen demand 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 one 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 declines (Fig. 1; Sanford 2002, Dell et al. 2011, Lemoine & Burkepile 2012,

Ober et al. 2019), we calculated an exponentially increasing total cost function based on the

metabolic theory of ecology equation (Brown et al. 2004, Dell et al. 2011, Ober et al. 2019),

excluding temperatures above the peak of the total cost curve (or below the minimum for SC).

**3    Results**

**3.1    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 which was significantly warmer than BB and FH in both

mediums (p < 0.001 in all comparisons). The mean intertidal DPM at the SC site (26.0, CI: 23.7-

28.4°C) exceed both the BB (19.5, CI:17.5-22.2°C) and FH (18.4, CI:16.0-20.7°C) sites by more

than 6°C. The mean water DPM for SC (19.8, CI: 17.9-21.7°C) was similarly more than 6°C

warmer than the mean water DPMs at BB (12.9, CI:10.0-14.8°C) and FH (9.8, CI:7.8-11.7°C).

The northernmost FH site had a 3°C cooler mean water DPM and a 1.5°C warmer mean

intertidal DPM than the BB site, but these differences were not significant (p >0.05). Similarly,

the quantile regression revealed no significant difference in the median intertidal DPM between

BB and FH (p=0.634). However, all three sites differed significantly in their the $25^{th}$ and $75^{th}$

percentiles of the intertidal DPMs (p<0.004 for all comparisons).  The southernmost (SC) site

had both the warmest 75%ile and 25%ile for intertidal temperatures, while the northernmost

(FH) site had the coldest 25%ile and the central BB site had the coldest 75%ile (Fig. 3).  Thus,

the FH site had the greatest range of intertidal DPMs (Fig. 3).

**3.2    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 was a significant main effect of

population, with FH showing significantly higher respiration than SC or BB at most temperatures

(p<0.05). 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).  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).

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).  These differences

were not statistically significant (p>0.05).

### 3.3 Temporal patterns of respiration by medium

#### 3.3.1 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). Most combinations of source population and temperature showed little variation across timepoints, with two exceptions. First, 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. Second, above 20 °C, both BB and FH showed linearly decreasing trends over time.

#### 3.3.2 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 time (Table 3). There were three distinct temporal patterns among the temperatures (Fig. 6). At the coolest temperatures we most frequently detected a significant quadratic trend in each population, 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

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

363    and 35°C for FH and 35°C for SC.  Finally, at 35 and 38°C for BB and at 38°C for the other two

364    sites, respiration was initially depressed and then increased over time.

365        **3.4    Total cost of emersion and vulnerability to warming**

366            The estimated total cost of a 30-minute emersion ranged from -0.07µmol for BB

367    barnacles at 38°C to 0.66µmol for FH barnacles at 25°C (Fig. 7A).  On average, immersion

368    respiration accounted for 18% of the total cost of exposure, but this value ranged from 0 to

369    nearly 50%, depending on the temperature and population.

370            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 95%CI) for FH and 21.3

(21.3-21.3 95%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 convex shape of the SC curves suggests we may not

have tested this population at temperatures above its $T_{peak}$ The MTE-based cost curves showed

increasing costs with latitude (Fig.  7A, Table 4).  When we these values to calculate the number

of days above $T_{peak}$ for each site (Fig. 7C), substituting 38°C for the SC $T_{peak}$, we found BB to be

most vulnerable, with 149 days over $T_{peak}$, followed by FH (93 days) and SC (4 days).

378            The exponentially increasing total cost functions (Fig. 7A, solid lines; Table 4) were

379    calculated from the 30-minute emersion costs using only temperatures below (BB, FH) or above

380    (SC) the $T_{peak}$ value.  FH had the largest total costs at all temperatures, followed by BB and SC.

381    When we used these equations to calculate the total annual emersion costs for each site (7B), the

382    costs followed the ordering of the cost curves (Fig. 7A), with FH showing the highest energy

383    costs and SC the lowest.

## 4   Discussion

Intertidal organisms experience two very different thermal regimes during emersion and

immersion.  Understanding the influence of each environment on their thermal physiology is

essential for identifying which species and populations are most vulnerable to warming

temperatures under climate change.  We compared the metabolic costs of emersion for the

intertidal barnacle *Balanus glandula* from three populations with contrasting emersion and

immersion temperatures.  Our results suggest a greater overall influence of local emersion

temperatures than immersion temperatures.  Specifically, the peaks ($T_{peak}$) of the respiration

curves in all three populations exceeded mean aquatic DPM temperatures by 10°C or more and

are better matched to the geographic ordering of daily intertidal DPMs. However, this conclusion

is tempered by the small thermal differences in intertidal temperatures between our two

northernmost sites which limited the statistical significance of some comparisons.  Additionally,

in using these results to predict the vulnerability of each population to climate change, we 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 nearshore ocean productivity and food supply.

Two lines of evidence in our results suggest that the extreme temperatures experienced

during low tide conditions drive the thermal physiology of *B. glandula* during emersion more

than immersion temperatures.  First, the identified $T_{peak}$ of the total energy cost parabola for each

population is much greater than its site's average aquatic DPM, exceeding these values by at

least 10°C at all three sites. While, a population's $T_{peak}$ should exceed its average DPM to

minimize the amount of time spent at stressful temperatures (Martin & Huey 2008), the

differences here are much larger than would be expected. 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. Specifically, FH had both warmer emersion temperatures and colder

immersion temperatures than BB, but its $T_{peak}$ was warmer than BB. While the temperature

differences between the FH and BB sites were small, and not always statistically significant,

temperature differences on this scale can greatly influence intertidal community composition and

population dynamics (e.g., Barry et al. 1995, Harley 2011).

We also note that the $T_{peak}$ value for the southernmost (SC) population is greater than the

highest temperature we tested (38°C), placing it well above the $T_{peak}$ values for BB and FH. We

did not expect this result, given that a 40°C emersion is lethal for FH barnacles (Ober et al.

2019). However, it is consistent with the warmer overall temperatures at this site and with

previously identified genetic differences among populations of *B. glandula* (Wares & Skoczen

2019, Wares et al. 2021). There is a known thermal and biogeographic breakpoint at Point

Conception, California (Wares et al. 2001, Blanchette et al. 2007), which falls between our SC

and BB study sites (Fig. 1). Many intertidal species exhibit physiologically distinct populations

north and south of this break (e.g., Sagarin & Somero 2006, Gleason & Burton 2013).

Unfortunately, because both the aquatic and intertidal thermal environments at the southern SC

site are 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.

Our finding that emersion conditions drive thermal physiology at BB and FH matches

past studies of thermal physiology among NE Pacific intertidal congeners that occupy different

intertidal shore heights.  Despite experiencing the same immersion conditions, higher shore

congeners typically show a more warm-adapted thermal physiology than their low shore

congeners (Somero 2002, 2010).  But inter-latitudinal studies of variation in thermal physiology

with NE Pacific intertidal species, particularly among sites north of Point Conception, have not

consistently shown a role for emersion conditions in thermal physiology (Rao 1953, Sagarin &

Somero 2006, Kuo & Sanford 2009, Logan et al. 2012).  One explanation for this discrepancy

could stem from differences among species in the degree to which they are metabolically active

during low tide. *B. glandula* displayed high rates of emersion oxygen consumption, often higher

than they exhibited in water at the same temperature (Ober et al. 2019), but many other species

exhibit metabolic depression at low tide (e.g., Sokolova & Portner 2001, Marshall et al. 2011).  If

species are not metabolically active during emersion, then their physiology may be under

stronger natural selection during immersion than emersion.  Alternately, thermal physiology may

vary by trait (Dell et al. 2011, Rose et al. 2024) and if the study measures a trait that is not used

during emersion (e.g. feeding rates, locomotion, or pumping rates) then it may only be under

natural selection during immersion.  In the case of *B. glandula*, however, a separate study of

aquatic respiration rates in the FH and LA populations (Ober et al. *unpubl. data*) still found $T_{peak}$

values at water temperatures well over the local maxima.

We used long-duration respiration trials to fully capture the temporal pattern of

respiration during emersion and immersion (McGaw et al. 2015, Ober et al. 2019, Griffen et al.

2024).  We found significant effects of emersion temperature on the temporal patterns of

respiration during both the emersion and immersion phases of the trials.  We detected three

distinct temporal patterns of respiration during immersion.  At the coolest temperatures,

barnacles in all three populations exhibited parabolic patterns with initially low respiration rates

that increased in the second or third hour of immersion and then declined.  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 (Ober et al. 2019) 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 (Davenport & Irwin

2003).  The concentration of oxygen in air is more than 20-fold higher than in seawater, so

oxygen may be more accessible during emersion at non-stressful temperatures then during

immersion (Hughes & Otto 1999).  By staying closed barnacles may also reduce to vulnerability

predation during immersion (Horn et al. 2021).

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

that declined gradually over the 6-hr immersion period.  The high initial immersion respiration

coincides with high initial activity in the video data recorded from FH barnacles (2019) and

suggests that high energy demand during emersion created an immediate need for respiration

upon resubmersion.  At these same temperatures, barnacles often also showed a decline in

emersion respiration over time, suggesting an oxygen debt was forming (Ellington 1983, McGaw

et al. 2015).  Finally, at the warmest temperatures, immersion 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.  Surprisingly, our assessment of vulnerability depended on

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 vary with the metric used.

Greater energetic costs only reflect a vulnerability to warming temperatures if energy

intake cannot meet demand.  The FH site showed greater overall energy demand at most

temperatures, particularly during emersion, which could reflect difference in overall metabolism,

driven by greater food availability at higher latitudes.  *Balanus glandula*'s exact diet is unknown

(Sanford & Menge 2001), but they likely consume a range of phytoplankton and zooplankton, as

adults can thrive on either in the laboratory (Hines 1978, Geierman & Emlet 2009, Gilman et al.

2013).  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 (Ware & Thomson 2005, Hickey & Banas

2008), along with higher zooplankton concentrations (Ware & Thomson 2005).  This gradient is

attributed to greater terrestrial nutrient supply at higher latitudes, particularly in the Salish Sea

region (Hickey & Banas 2008), which includes the FH field site.  A greater food supply could

select for greater overall 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* (Sotka et al. 2004,

Wares et al. 2021). If true, this suggests that energy costs are a less useful metric of vulnerability

to warming than threshold temperatures for NE Pacific suspension feeders such as *B. glandula*.

## 5  Conclusion

The results presented herein provide strong evidence that the thermal conditions

experienced during emersion influence the thermal physiology of the intertidal barnacle *Balanus*

*glandula*.   However, more empirical studies are needed to predict vulnerability to warming

temperatures under climate change.  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., Gilman 2006, Schneider et al. 2010), few models of

climate change responses have considered both warming temperatures and changes in food

supply at the same time (but see Huey & Kingsolver 2019). 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

(Khangaonkar et al. 2019).  Thus, in this case, the central BB population may be the most

vulnerable to future warming under climate change.

## Acknowledgements

513    This work was supported by the National Science Foundation (grant IOS-1351445 to S.E.G.)

514    with additional support from the Keck Science Department of Claremont McKenna, Pitzer, and

515    Scripps Colleges.   We thank E. Roberts for commenting on an earlier draft of this manuscript.

516    The present work is available as preprint at

517    https://www.biorxiv.org/content/10.1101/2024.03.26.586848v1.

# 6 References

Barry JP, Baxter CH, Sagarin RD, Gilman SE (1995) Climate-related, long-term faunal changes in a California rocky intertidal community. Science 267:672–675.

Bates D, Maechler M, Bolker B, Walker S (2015) Fitting linear mixed-effects models using lme4. J Stat Softw 67:1–48.

Bennett S, Duarte CM, Marbà N, Wernberg T (2019) Integrating within-species variation in thermal physiology into climate change ecology. Philos Trans R Soc B Biol Sci 374:20180550.

Blanchette C, Helmuth B, Gaines S (2007) Spatial patterns of growth in the mussel, *Mytilus californianus*, across a major oceanographic and biogeographic boundary at Point Conception, California, USA. J Exp Mar Biol Ecol 340:126–148.

Brown JH, Gillooly JF, Allen AP, Savage VM, West GB (2004) Toward a metabolic theory of ecology. Ecology 85:1771–1789.

Buckley LB, Huey RB, Kingsolver JG (2022) Asymmetry of thermal sensitivity and the thermal risk of climate change. Glob Ecol Biogeogr 31:2231–2244.

Byun D-S, Hart DE, Kim S, Ha J (2023) Classification of monthly tidal envelopes in mixed tide regimes. Sci Rep 13:4786.

Davenport J, Irwin S (2003) Hypoxic life of intertidal acorn barnacles. Mar Biol 143:555–563.

Dell AI, Pawar S, Savage VM (2011) Systematic variation in the temperature dependence of physiological and ecological traits. Proc Natl Acad Sci 108:10591–10596.

Denny MW (2007) Tides. In: *Encyclopedia of Tides and Rocky Shores*, 1st ed. Denny MW, Gaines SD (eds) University of California Press, p 588–595

Deutsch CA, Tewksbury JJ, Huey RB, Sheldon KS, Ghalambor CK, Haak DC, Martin PR (2008) Impacts of climate warming on terrestrial ectotherms across latitude. Proc Natl Acad Sci 105:6668–6672.

Ellington WR (1983) The recovery from anaerobic metabolism in invertebrates. J Exp Zool 228:431–444.

Englund G, Ohlund G, Hein CL, Diehl S (2011) Temperature dependence of the functional response. Ecol Lett 14:914–921.

Fey SB, Vasseur DA, Alujevic K, Kroeker KJ, Logan ML, O'Connor M, Rudolf VHW, DeLong JP, Peacor S, Selden RL, Sih A, Clusella-Trullas S (2019) Opportunities for behavioral rescue under rapid environmental change. Glob Change Biol 25:3110–3120.

Fitzhenry T, Halpin PM, Helmuth B (2004) Testing the effects of wave exposure, site, and behavior on intertidal mussel body temperatures: applications and limits of temperature logger design. Mar Biol 145:339–349.

Fly EK, Monaco CJ, Pincebourde S, Tullis A (2012) The influence of intertidal location and temperature on the metabolic cost of emersion in *Pisaster ochraceus*. J Exp Mar Biol Ecol 422:20–28.

Foulk A, Gouhier T, Choi F, Torossian JL, Matzelle A, Sittenfeld D, Helmuth B (2024) Physiologically informed organismal climatologies reveal unexpected spatiotemporal trends in temperature. Conserv Physiol 12:coae025.

Fox J, Weisberg S (2019) An R companion to applied regression, 3rd ed. Sage, Thousand Oaks, California.

Gaitan-Espitia JD, Bacigalupe LD, Opitz T, Lagos NA, Osores S, Lardies MA (2017) Exploring physiological plasticity and local thermal adaptation in an intertidal crab along a latitudinal cline. J Therm Biol 68:14–20.

Geierman C, Emlet R (2009) Feeding behavior, cirral fan anatomy, Reynolds numbers, and leakiness of *Balanus glandula*, from post-metamorphic juvenile to the adult. J Exp Mar Biol Ecol 379:68–76.

Geller J, Sotka EE, Kado R, Palumbi SR, Schwindt E (2008) Sources of invasions of a northeastern Pacific acorn barnacle, *Balanus glandula*, in Japan and Argentina. Mar Ecol Prog Ser 358:211–218.

Gilman SE (2006) Life at the edge: an experimental study of a poleward range boundary. Oecologia 148:270–279.

Gilman SE, Chen S, Wong JWH (2013) Oxygen consumption in relation to body size, wave exposure, and cirral beat behavior in the barnacle *Balanus glandula*. J Crustac Biol 33:317–322.

Gilman SE, Wethey DS, Helmuth B (2006) Variation in the sensitivity of organismal body temperature to climate change over local and geographic scales. Proc Natl Acad Sci U S A 103:9560–9565.

Gleason LU, Burton RS (2013) Phenotypic evidence for local adaptation to heat stress in the marine snail *Chlorostoma* (formerly *Tegula*) *funebralis*. J Exp Mar Biol Ecol 448:360–366.

Griffen BD, Ashton IB, Dominguez-Villalobos A, Fletcher LS, Marlow BN, Raj V, Reese LA, Reese TC, Richardson HC (2024) Variation in metabolic rate during low tide aerial exposure in the Asian shore crab *Hemigrapsus sanguineus*. Mar Biol 171:129.

Gunderson AR, Dillon ME, Stillman JH (2017) Estimating the benefits of plasticity in ectotherm heat tolerance under natural thermal variability. Funct Ecol 31:1529–1539.

Harley CDG (2011) Climate change, keystone predation, and biodiversity loss. Science 334:1124–1127.

Hayford HA, Gilman SE, Carrington E (2021) Tidal cues reduce thermal risk of climate change in a foraging marine snail. Clim Change Ecol 1:100003.

Helmuth B, Broitman BR, Blanchette CA, Gilman S, Halpin P, Harley CDG, O'Donnell MJ, Hofmann GE, Menge B, Strickland D (2006) Mosaic patterns of thermal stress in the rocky intertidal zone: Implications for climate change. Ecol Monogr 76:461–479.

Helmuth B, Choi F, Matzelle A, Torossian JL, Morello SL, Mislan K a. S, Yamane L, Strickland D, Szathmary PL, Gilman SE, Tockstein A, Hilbish TJ, Burrows MT, Power AM, Gosling E, Mieszkowska N, Harley CDG, Nishizaki M, Carrington E, Menge B, Petes L, Foley MM, Johnson A, Poole M, Noble MM, Richmond EL, Robart M, Robinson J, Sapp J, Sones J, Broitman BR, Denny MW, Mach KJ, Miller LP, O'Donnell M, Ross P, Hofmann GE, Zippay M, Blanchette C, Macfarlan JA, Carpizo-Ituarte E, Ruttenberg B, Pena Mejia CE, McQuaid CD, Lathlean J, Monaco CJ, Nicastro KR, Zardi G (2016) Long-term, high frequency in situ measurements of intertidal mussel bed temperatures using biomimetic sensors. Sci Data 3:160087.

Helmuth B, Harley CDG, Halpin PN, O'Donnell M, Hofmann GE, Blanchette CA (2002) Climate change and latitudinal patterns of intertidal thermal stress. Science 298:1015–1017.

Helmuth BST, Hofmann GE (2001) Microhabitats, thermal heterogeneity, and patterns of physiological stress in the rocky intertidal zone. Biol Bull 201:374–384.

Hickey B, Banas N (2008) Why is the northern end of the California Current System so
productive? Oceanography 21:90–107.
Hines AH (1978) Reproduction in three species of intertidal barnacles from central California.
Biol Bull 154:262–281.
Horn KM, Fournet MEH, Liautaud KA, Morton LN, Cyr AM, Handley AL, Dotterweich MM,
Anderson KN, Zippay ML, Hardy KM (2021) Effects of intertidal position on
metabolism and behavior in the acorn barnacle, *Balanus glandula*. Integr Org Biol
3:obab010.
Huey RB, Kingsolver JG (2019) Climate warming, resource availability, and the metabolic
meltdown of ectotherms. Am Nat 194:E140–E150.
Hughes JS, Otto SP (1999) Ecology and the evolution of biphasic life cycles. Am Nat 154:306–
320.
Iles AC (2014) Toward predicting community-level effects of climate: relative temperature
scaling of metabolic and ingestion rates. Ecology 95:2657–2668.
Khangaonkar T, Nugraha A, Xu W, Balaguru K (2019) Salish Sea response to global climate
change, sea level rise, and future nutrient loads. J Geophys Res Oceans 124:3876–3904.
Koenker R (2022) _quantreg: Quantile Regression_. R package version 5.94. https://CRAN.R-
project.org/package=quantreg
Kuo ESL, Sanford E (2009) Geographic variation in the upper thermal limits of an intertidal
snail: implications for climate envelope models. Mar Ecol Prog Ser 388:137–146.
Lemoine NP, Burkepile DE (2012) Temperature-induced mismatches between consumption and
metabolism reduce consumer fitness. Ecology 93:2483–2489.
Lenth RV (2022) _emmeans: Estimated Marginal Means, aka Least-Squares Means_. R
package version 1.8.3. https://CRAN.R-project.org/package=emmeans
Logan CA, Kost LE, Somero GN (2012) Latitudinal differences in *Mytilus californianus* thermal
physiology. Mar Ecol Prog Ser 450:93–105.
Low-Décarie E, Boatman TG, Bennett N, Passfield W, Gavalás-Olea A, Siegel P, Geider RJ
(2017) Predictions of response to temperature are contingent on model choice and data
quality. Ecol Evol 7:10467–10481.
Marshall DJ, Dong YW, McQuaid CD, Williams GA (2011) Thermal adaptation in the intertidal
snail *Echinolittorina malaccana* contradicts current theory by revealing the crucial roles
of resting metabolism. J Exp Biol 214:3649–3657.
Martin TL, Huey RB (2008) Why "suboptimal" Is optimal: Jensen's Inequality and ectotherm
thermal preferences. Am Nat 171:E102–E118.
McGaw LJ, Clifford AM, Goss GG (2015) Physiological responses of the intertidal starfish
*Pisaster ochraceus*, (Brandt, 1835) to emersion at different temperatures. J Exp Mar Biol
Ecol 468:83–90.
McMahon RF (1990) Thermal tolerance, evaporative water loss, air-water oxygen consumption
and zonation of intertidal prosobranchs: a new synthesis. Hydrobiologia 193:241–260.
Mislan KAS, Helmuth B, Wethey DS (2014) Geographical variation in climatic sensitivity of
intertidal mussel zonation. Glob Ecol Biogeogr 23:744–756.
Monaco CJ, Wethey DS, Helmuth B (2014) A dynamic energy budget (DEB) model for the
keystone predator *Pisaster ochraceus*. Plos One 9:e104658.
Montalto V, Helmuth B, Ruti PM, Dell'Aquila A, Rinaldi A, Sara G (2016) A mechanistic
approach reveals non linear effects of climate warming on mussels throughout the
Mediterranean sea. Clim Change 139:293–306.

Morris RH, Abbott DP, Haderlie EC (1980) Intertidal invertebrates of California, 1st ed.
Stanford University Press, Stanford, Calif.

Ober GT, Rognstad RL, Gilman SE (2019) The cost of emersion for the barnacle *Balanus*
*glandula*. Mar Ecol Prog Ser 627:95–107.

Pinheiro J, Bates D, Team RC (2022) _nlme: Linear and Nonlinear Mixed Effects Models_.
https://CRAN.R-project.org/package=nlme

Porter WP, Gates DM (1969) Thermodynamic equilibria of animals with environment. Ecol
Monogr 39:245–270.

Quinn JF, Wolin CL, Judge ML (1989) An experimental analysis of patch size, habitat
subdivision, and extinction in a marine intertidal snail. Conserv Biol 3:242–251.

Rao KP (1953) Rate of water propulsion in *Mytilus californianus* as a function of latitude. Biol
Bull 104:171–181.

Rose K, Holsman K, Nye J, Markowitz E, Banha T, Bednaršek N, Bueno-Pardo J, Deslauriers D,
Fulton E, Huebert K, Huret M, Ito S, Koenigstein S, Li L, Moustahfid H, Muhling B,
Neubauer P, Paula J, Siddon E, Skogen M, Spencer P, Van Denderen P, Van Der Meeren
G, Peck M (2024) Advancing bioenergetics-based modeling to improve climate change
projections of marine ecosystems. Mar Ecol Prog Ser 732:193–221.

Sagarin RD, Somero GN (2006) Complex patterns of expression of heat-shock protein 70 across
the southern biogeographical ranges of the intertidal *mussel Mytilus californianus* and
snail *Nucella ostrina*. J Biogeogr 33:622–630.

Sanford E (2002) Water temperature, predation, and the neglected role of physiological rate
effects in rocky intertidal communities. Integr Comp Biol 42:881–891.

Sanford E, Menge BA (2001) Spatial and temporal variation in barnacle growth in a coastal
upwelling system. Mar Ecol Prog Ser 209:143–157.

Schneider KR, Van Thiel LE, Helmuth B (2010) Interactive effects of food availability and aerial
body temperature on the survival of two intertidal *Mytilus* species. J Therm Biol 35:161–
166.

Simon-Blecher N, Granevitze Z, Achituv Y (2008) *Balanus glandula*: from North-West America
to the west coast of South Africa. Afr J Mar Sci 30:85–92.

Sokolova IM, Portner HO (2001) Physiological adaptations to high intertidal life involve
improved water conservation abilities and metabolic rate depression in *Littorina saxatilis*.
Mar Ecol Prog Ser 224:171–186.

Somero GN (2010) The physiology of climate change: how potentials for acclimatization and
genetic adaptation will determine 'winners' and 'losers'. J Exp Biol 213:912–920.

Somero GN (2002) Thermal physiology and vertical zonation of intertidal animals: Optima,
limits, and costs of living. Integr Comp Biol 42:780–789.

Sotka EE, Wares JP, Barth JA, Grosberg RK, Palumbi SR (2004) Strong genetic clines and
geographical variation in gene flow in the rocky intertidal barnacle *Balanus glandula*.
Mol Ecol 13:2143–2156.

Stillman JH (2003) Acclimation capacity underlies susceptibility to climate change. Science
301:65–65.

Sunday J, Bennett JM, Calosi P, Clusella-Trullas S, Gravel S, Hargreaves AL, Leiva FP, Verberk
WCEP, Olalla-Tárraga MÁ, Morales-Castilla I (2019) Thermal tolerance patterns across
latitude and elevation. Philos Trans R Soc B Biol Sci 374:20190036.

Sunday JM, Bates AE, Dulvy NK (2010) Global analysis of thermal tolerance and latitude in
ectotherms. Proc R Soc B Biol Sci 278:1823–1830.

Tagliarolo M, McQuaid CD (2015) Sub-lethal and sub-specific temperature effects are better
predictors of mussel distribution than thermal tolerance. Mar Ecol Prog Ser 535:145–159.
Ware D, Thomson R (2005) Bottom-up ecosystem trophic dynamics determine fish production
in the northeast Pacific. Science 308:1280–1284.
Wares JP, Gaines S D, Cunningham CW (2001) A comparative study of asymmetric migration
events across a marine biogeographic boundary. Evolution 55:295–306.
Wares JP, Skoczen KM (2019) Maintenance of a genetic cline in the barnacle *Balanus glandula*.
Biol Bull 236:199–206.
Wares JP, Strand AE, Sotka EE, Bernardi G (2021) Diversity, divergence and density: How
habitat and hybrid zone dynamics maintain a genomic cline in an intertidal barnacle. J
Biogeogr 48:2174–2185.
Wethey DS (2002) Biogeography, competition, and microclimate: The barnacle *Chthamalus*
*fragilis* in New England. Integr Comp Biol 42:872–880.
Zuur A, Ieno EN, Walker N, Saveliev AA, Smith GM (2009) Mixed Effects Models and
Extensions in Ecology with R. Springer, New York, NY.
715

717    **Tables**

719    **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 | |

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

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

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

| A) | Emersion | | | Immersion | | |
|---|---|---|---|---|---|---|
| **Term** | $\chi^2$ | df | p | $\chi^2$ | 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) |

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

727 **and immersion.**

| Term | Emersion $\chi^2$ | df | p | Immersion $\chi^2$ | 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 |

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

|  | **Friday Harbor** | **Bodega Bay** | **Southern California** |
|---|---|---|---|
| **Parabola** |  |  |  |
| 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** |  |  |  |
| -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) |

**Figures**

**Figure 1. 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 (Morris et al.

1980). The flat-headed line indicates the region of genetic sampling in Sotka et al. (2004) and

Wares and Skozen (2019), 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)

**Figure 2. 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. (2021).

**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 breadth indicates 95% CIs around the median.

**Figure 4. Oxygen consumption by temperature for each population during emersion (A-C) and immersion (D-F) phases**. Parabolas represent the predicted respiration rates from the mixed model, calculated at the mean overall barnacle size (5.05 mm operculum length). Dashed lines are the estimated $T_{peak}$ values. The individual points are mean observed respiration rates for individual barnacles, with the darkness of the fill indicating size.

**Figure 5. Average emersion oxygen consumption by temperature and timepoint 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)

**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).

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

**cost for each population (B), and number of days over T$_{Peak}$ (C).** Exponential cost curves

(solid lines) in (A) are fit on the rising portion of the cost curve (solid dots) and used to calculate

total annual costs (B). Both hollow and solid dots in (A) are used to calculate T$_{peak}$ values for

(C). See text for additional explanation of cost calculations.