

Interacting Effects of Climate Change Variables on Fish Aerobic Metabolism:  
Changing Exercise Performance in the Face of  
Deoxygenation, Hypercapnia, and Hyperthermia

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## **1. Introduction**

### **1.1. Abstract**

A large body of recent research has attempted to elucidate and forecast the impacts of climate change on marine teleosts. Although there have been many important contributions to our understanding, two key areas that have received relatively little attention are interactions between multiple stressors, and the development of a mechanistic, rather than purely statistical predictive framework. This thesis attempts to address these gaps by examining the effects of multi-stressor climate change perturbations on the capacity for aerobic metabolism in two marine teleosts – the yellowfin tuna (*Thunnus albacares*) and the Mozambique tilapia (*Oreochromis mossambicus*). The methods used in the investigation will range from theory-based physiological models (Chapter I), direct experimentation on captive animals (Chapter II) and empirical data-based forecasts of future performance (Chapter III). Put together, these chapters demonstrate a variety of complementary, physiology-based approaches to understanding and predicting the response of marine teleosts to complex environmental change. This introduction will seek to provide some background on the energetic terms used—particularly the concept, and methods of measurement for the aerobic scope—as well as considerations of different approaches to understanding population responses to future change.

### **1.2. Teleost Ecology in a Climate Change Future**

The oceans are expected to undergo substantial changes due to anthropogenic climate change in the near future. Estimates indicate that end-century global sea-surface warming is likely to reach close to 3°C under a RCP 8.5 greenhouse gas

emissions scenario (also known as the “business as usual” scenario), while sea surface pH is likely to fall by 0.35 pH units and global ocean oxygen content is likely to decline by close to 4% (Bopp et al., 2013). These changes are driven by atmospheric warming, increased dissolved inorganic carbon (DIC) due to the diffusion of anthropogenic CO<sub>2</sub> into the oceans (driving a decline in ocean pH), and by surface ocean stratification lowering the rate of transport of oxygenated, mixed-layer water into the oxygen minimum layer (Gruber, 2011). Furthermore, there is concern that strong interactions between these drivers could amplify future perturbations beyond what is predicted by the ensemble average of global climate models (Gruber, 2011). One such interaction that is not captured in many models is the potential impact of ocean pH on the speed and effectiveness of the biological carbon pump (Hoffman and Schellnhuber, 2009). Low pH (and the resultant decline in carbonate saturation) may reduce the mass of the mineral skeletons of many planktonic organisms by inhibiting biological calcification. The resulting slow-down in the rate of sinking of dead calcifying plankton would cause more decomposition to occur at shallower depths, intensifying the oxygen minimum layer and raising CO<sub>2</sub> concentrations at these depths due to the stoichiometry of the Redfield Ratio (Brewer and Peltzer, 2009).

Regardless of these uncertainties in the magnitude of these changes, it is clear that there will be profound impacts on marine biogeochemistry and ecosystem functioning (Gruber, 2011). While much research has focused on the potential impacts on calcifying organisms such as corals and pteropods (Hoegh-Guldberg et al., 2007), there is strong uncertainty as to how teleosts, which account for over half of the world’s vertebrate species and 6% of the total protein intake for the human population (Allison et al., 2009), will be affected (Le Quesne and Pinnegar, 2011). Studies have variously indicated strongly negative, positive, or no impacts of climate-

related variables on teleosts, sometimes even within the same species (Cruz-Neto and Steffensen, 1997; McKenzie et al., 2003; Melzner et al., 2009; Moran and Støttrup, 2011).

Two key issues that have received comparatively little attention in this field are the effect of interactions between multiple climate change variables on determining biological outcomes, and the development of a mechanistic framework for understanding and predicting climate change effects. The first issue is of critical importance because climate change in the marine environment is fundamentally a multi-stressor phenomenon, with changes in temperature, oxygen, and pCO<sub>2</sub> expected to co-occur across large regions of the oceans (Bopp et al., 2013). Furthermore, recent evidence suggests that simultaneous stressors rarely interact additively (Crain et al., 2008), thereby precluding the assumption that studies on organism responses to single stressors can be accurately applied to the multiple stressor case. Although this issue has long been identified as a key challenge across many fields of ecology and conservation biology (Breitburg et al., 1999; Crain et al., 2008), the complexity of the experiments necessary to investigate it have discouraged more scrutiny. Nevertheless, at a minimum the interactions between the three core variables of temperature, oxygen, and pCO<sub>2</sub> need to be better understood if we are to understand the potential impacts of climate change more fully.

The second issue of the need for a more mechanistic model is also a fundamental one. This is because climate change prediction is almost always an extrapolative exercise in that forecasts must deal with species' responses to future temperature, oxygen, or pCO<sub>2</sub> conditions that are not observed in their current habitats (Kearney and Porter, 2009). The most common methodology of predicting the impacts of climate change involve some form of statistical correlation between

current species distributions and observed environmental variables (Denny and Helmuth, 2009). Such correlations are singularly unsuited to extrapolation and thus must be treated with care in the climate change context (Kearney and Porter, 2009). Recently, the development of the aerobic scope concept as a coherent mechanistic framework for understanding the impacts of environmental change on marine species has become a promising avenue for moving beyond such correlative models (Pörtner, 2010). Unfortunately this approach is currently not widely used, and to our knowledge has been limited to simple single-stressor models that only predict presence or absence in a given future habitat.

### **1.3. Understanding Aerobic Scope**

The concept of a scope for aerobic metabolism was first suggested by Fry (1947) as an organism-level indicator of the effect of environmental factors on animal activity. Since then, it has been used by investigators to study the effects of temperature, oxygen, salinity, pCO<sub>2</sub>, disease burden, and a host of other factors on a variety of species ranging from mammals to teleosts (Khan et al., 2014; Killen et al., 2007; Lapointe et al., 2014; Nilsson et al., 2009; Pörtner, 2002; Weibel et al., 2004). More broadly speaking, the concept that an energy budget, including not just aerobic scope but scope for growth (Stumpff et al., 2011; Widdows et al., 1997), and mass-balance bioenergetics (Brandt et al., 2009; Kearney and Porter, 2009) influences population viability and distribution has found wide acceptance across many fields of ecology and conservation biology.

The key insight elucidated by Fry is that the capacity of an organism for aerobic metabolism can be partitioned into two parts. The first is the minimum metabolic expenditure necessary to sustain life, and this mandatory metabolic output

is commonly termed the standard metabolism (in analogy to the basal metabolism, which describes the minimum metabolic rate in endotherms under the special case where there is no metabolic cost of thermoregulation) (Hochachka et al., 2003). This standard metabolic rate is typically defined as including such activities as maintaining a constant osmolality of internal fluids, a minimum survivable level of ventilation and gas exchange, and maintenance of existing tissues. Activities that are excluded from the standard metabolic rate include growth (both somatic and gonadal), digestion, and locomotion. These optional activities contribute to the active metabolic rate, which is the maximum aerobic metabolism that can be sustained under maximum exercise (e.g. fast swimming) or a combination of exercise and digestion (Alsop and Wood, 1997). By subtracting the standard metabolic rate from the active metabolism, the remaining quantity, termed the aerobic scope, describes the total energy budget available for non-life-sustaining processes.

Due to years of study, a number of frameworks have been developed to describe the changes in aerobic scope in response to a gradient of environmental conditions. The most prominent of these is the oxygen-limited thermal tolerance hypothesis (Pörtner, 2001), which states that the power-law scaling of standard metabolic rates with temperature compared with the bell-shaped relationship for active metabolic rates necessarily results in a temperature optimum ( $T_{opt}$ ) for aerobic scope and an upper and lower critical temperature ( $T_{crit}$ ) at which aerobic scope falls to zero (Figure 1a). Within a few degrees of the optimum temperature there is typically a thermal tolerance zone where changes in temperature have little effect on the aerobic scope. The boundaries of this temperature range are described by the upper and lower *pejus* temperatures ( $T_{pejus}$ ) beyond which further thermal stress causes a rapid decline in aerobic scope (Pörtner et al., 2001). These parameters,  $T_{opt}$ ,

$T_{crit}$ , and  $T_{pejus}$  can be used to describe the impact of a range of ambient temperatures on aerobic fitness under a given set of environmental conditions (Farrell et al., 2009). The addition of simultaneous stressors impacting the oxygen uptake or demand of an organism can alter the thermal tolerance by changing the capacity for a high active metabolism. Such stressors include low oxygen (hypoxia) or high  $pCO_2$  (hypercapnia), both of which are thought to decrease active metabolism leading to a narrowing of the *pejus* and critical temperature limits (Figure 1b).

A second framework that has been used to understand changes in aerobic scope is the limiting oxygen concentration continuum (Claireaux and Lagardère, 1999), which states that any given metabolic rate has a limiting oxygen concentration below which that metabolic activity cannot be sustained. As a result, active metabolic rates are observed to rise monotonically with oxygen concentration, reaching an asymptote at a point very close to air saturation in fully active individuals (Figure 2). Under this framework, simultaneous stressors such as high or low temperature (hypo- or hyperthermia) or hypercapnia leads to a decline in the maximum active metabolic rate (the active metabolism under air saturation) and a narrowing of the critical limits (Claireaux et al., 2000).

#### **1.4. Metabolic Linkages Between Aerobic Scope and Fitness**

The basis for the focus on aerobic scope as a predictor of environmental impacts on organism physiology is the fact that it describes a fundamental and mechanistic constraint on population abundance and distribution. More specifically, the aerobic scope can be thought to define a “fundamental niche” (Hutchinson, 1957), which is to say that it describes the impact of abiotic factors, but not the biotic

interactions that determine the “realized” or observed niche (Kearney and Porter, 2009). This is because the activities that drive organismal fitness – growth, reproduction, foraging, digestion, and locomotion – are all fueled by the non-maintenance aerobic metabolism that makes up the aerobic scope (Pörtner et al., 2001) either directly in the case of a purely aerobic activity, or indirectly, through the excess post-exercise oxygen consumption (EPOC) that is required to compensate for anaerobic expenditures (Gaesser and Brooks, 1983). In the absence of biotic change (e.g. changes in forage availability) declines in aerobic scope due to adverse environmental conditions necessarily reduce the capacity for one or more of these activities (Farrell et al., 2009). This is reflected in the multiple studies that have found close and significant correlations between aerobic scope and growth, appetite, or locomotory ability (Table 1).

In essence, the aerobic scope describes an integrative bioenergetics budget for the activities most closely tied to fitness. This is a key difference when compared against other fitness indicators such as somatic growth rate, which may be conserved in organisms faced with environmental stressors at the expense of other fitness-related activities (e.g. gonadal growth or immune functioning) or modulated to maintain stable levels of these activities (Lambert and Dutil, 2000; Rijnsdorp, 1990). As a result, such indicators may exaggerate or fail to detect the impact of an environmental stressor on an organism even under conditions that elicit a marked decline in aerobic scope. Furthermore, predictions of future growth rate in such studies rely upon the implicit assumption that the proportion of available aerobic energy that is allocated to growth will remain stationary under future conditions. Transient biotic factors such as the short-term modulation of endogenous hormone levels (e.g. the GH/IGF axis), which are difficult to predict with any certainty, can strongly influence the

allocation of energy to different activities (Schreck, 2010). The ability to relax this assumption is especially important for applications such as climate change prediction, where changes in population or ecosystem dynamics could easily influence these factors on a population-scale.

### **1.5. Methods of Aerobic Scope Measurement**

Aerobic scope measurements require two components: (1) a method of measuring the aerobic metabolic rate of the study animal, and (2) a method of modulating its activity. While manual exhaustion (chasing of fishes by investigators with hands/sticks) and feeding-based metabolic activity has been used previously (Lefrançois and Claireaux, 2003; Vanella et al., 2010), the most popular and standardized methodology for aerobic scope measurement is the swim tunnel respirometer. This combines a sealed chamber for measuring the oxygen uptake resulting from an animal's metabolism with a recirculating system that encourages the fish to swim at the desired speed. Because most teleosts exhibit a rheotactic response, raising the speed of recirculation through a respirometer chamber forces the fish to match its swimming speed to the current to maintain its position (Dewar and Graham, 1994). At swimming speeds above the critical swimming speed ( $U_{crit}$ ) defined as the highest swimming speed that can be maintained over prolonged periods (Plaut, 2001), a fish must rely on some combination of maximal aerobic metabolism and supplemental anaerobic metabolism to meet the energetic demands of locomotion. The active metabolic rate measured during this speed of swimming can be compared against the metabolic rate achieved during prolonged quiescence in a respirometer chamber with little or no flow, typically with the subject resting against the bottom of the chamber. In the case of small study animals, the chamber is often designed as a

Blazka-type respirometer (Blazka, 1960) with internal recirculation, whereas for larger fishes a Brett-type design (Brett, 1964) with the recirculation flow routed through a system of pipes is often used.

Respirometer designs can be further subdivided into continuous-flow or intermittent-flow systems based on whether fresh oxygenated water is allowed to flow continuously into the swim chamber, thereby maintaining a constant internal dissolved oxygen concentration, or only intermittently, typically every ten to fifteen minutes. In the continuous flow case, the metabolic oxygen consumption must be measured based on the differential in dissolved oxygen concentration between points upstream and downstream of the fish, and the measurements depend upon strict assumptions of the flow regime within the swim chamber. In the intermittent flow case, the decline of dissolved oxygen over time during the period when the chamber is sealed (i.e. no new oxygenated water is allowed to flow in) is used to measure the metabolic oxygen consumption. While this results in a small oscillation in the oxygen concentration within the chamber and limits the time-resolution of metabolic measurements, this method is commonly accepted to be a more reliable mode of metabolic measurement (Steffensen, 1989).

Our aerobic scope measurements rely upon a custom-designed Blazka-type intermittent flow respirometer diagrammed in Figure 3. Our custom-built swim-tunnel respirometer is unique to our knowledge due to its ability to simultaneously modulate not only the oxygen concentration within the swim chamber but also the pCO<sub>2</sub> and temperature, allowing us to investigate the impacts of multi-stressor perturbations involving all of these variables in an experimental setting.

## **1.6. Predicting Aerobic Scope: A Diversity of Approaches**

The approaches for predicting the future aerobic scope of a species or population reside within the wide spectrum of species distribution and abundance models that currently exist. For example, the most heavily statistically driven models of species distribution rely on correlations between presence/absence data and observed environmental characteristics. These are called climate envelope, habitat suitability, or niche models, and are a powerful and generalizable method of predicting species responses to climate change and other complex environmental challenges (Hijmans and Graham, 2006; Hirzel et al., 2006; Thuiller, 2004). More mechanistic or experimental approaches include bioenergetics modeling, biophysical models, and the use of experimental data on environmental tolerance to construct estimates of a niche space (Homyack, 2010; Kearney and Porter, 2009). While these models suffer from the difficulties of scaling organism physiology to population-level changes and are less likely to fit the data used for evaluation, they offer the advantages of being more easily ecologically interpretable and also of being more robust to non-equilibrium conditions and extrapolation to novel environments (Denny and Helmuth, 2009). In addition, they are less likely to conflate explanatory variables by spuriously attributing a limit in species range imposed by the latitudinal distribution of its prey for example on a temperature limitation (Homyack, 2010). Many of these advantages and disadvantages are summarized in Table 2.

Current approaches to predicting environmental impacts on populations using aerobic scope have focused on direct interpolation/extrapolation of experimental results. This type of forecasting has been used repeatedly for example to predict the response of migrating salmon to increases in the water temperature of their native

streams (Eliason et al., 2011; Farrell et al., 2009, 2008). Such models typically measure variation in aerobic scope against a single environmental variable (most often temperature), interpolate a simple statistical function that outputs aerobic scope within a range of that environmental variable, and then predict future aerobic scope under different scenarios of unidimensional environmental change. This is despite the fact that the techniques for measuring the impact of multiple variables on aerobic scope have been demonstrated multiple times (Claireaux and Lagardère, 1999; Claireaux et al., 2000; Lapointe et al., 2014; Munday et al., 2009), and despite repeated efforts in the niche modeling community to point out the overwhelming importance of multiple-stressor challenges when predicting responses to large-scale environmental shifts such as climate change (Bopp et al., 2013; Crain et al., 2008). This thesis attempts to extend this methodology to more sophisticated environmental change scenarios, using three methods: (1) a novel theoretical aerobic scope model based on measured physiological traits (Chapter I), (2) a set of empirical measurements of aerobic scope on captive animals in response to multidimensional environmental change (Chapter II), and (3) an empirical model informed by the oxygen-limited thermal tolerance and limiting oxygen continuum frameworks described above to interpolate between the measurements obtained in Chapter II (Chapter III). Collectively, these three approaches share many of the advantages described in Table 2. We will use the theoretical model developed in Chapter I for example to directly assess the capacity for adaptation to novel environmental conditions due to physiological plasticity/selection of already extant traits, and we will evaluate this model against independent data on species occurrence. We will use the experimental results obtained in Chapter II to examine the possible impacts of a range of novel environments on aerobic fitness. Finally, we will use the empirical

model developed in Chapter III to interpret the possible consequences of changes in aerobic scope on organismal physiology.

### **1.7. Acknowledgements**

In order to test the questions posed in this thesis, a variety of approaches encompassing a large knowledge base were required. Chapter I required a deep dive into the world of physiological modeling as well as the ability to parse a complex CMIP5 dataset, while Chapter II led away from the computer screen and into the machine shop, where a great deal of engineering was required to develop the custom designed and built respirometry system. Because triple stressor experiments require more controls than traditional respirometry, I ended up writing my own software to control the system. Finally, Chapter III required the programming expertise to analyze large amounts of data. I would not have been able to do these things without the guidance of my advisor and my committee who helped to develop this thesis, especially Kevin's knowledge of tuna biology and the tuna-research community, Andre's expertise on tilapia, David's guidance on gas control and measurement, Anna's help on combing the literature, and Jeff's advice on electronics and especially on my early (aborted) ideas of an oxygen-sensing buoy. I would also like to thank my friends, especially my roommate Michelle, and my fiancé Xue who helped me to get over the many frustrations of coding and especially of plumbing a respirometer system.

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## 1.9. Figures and Tables

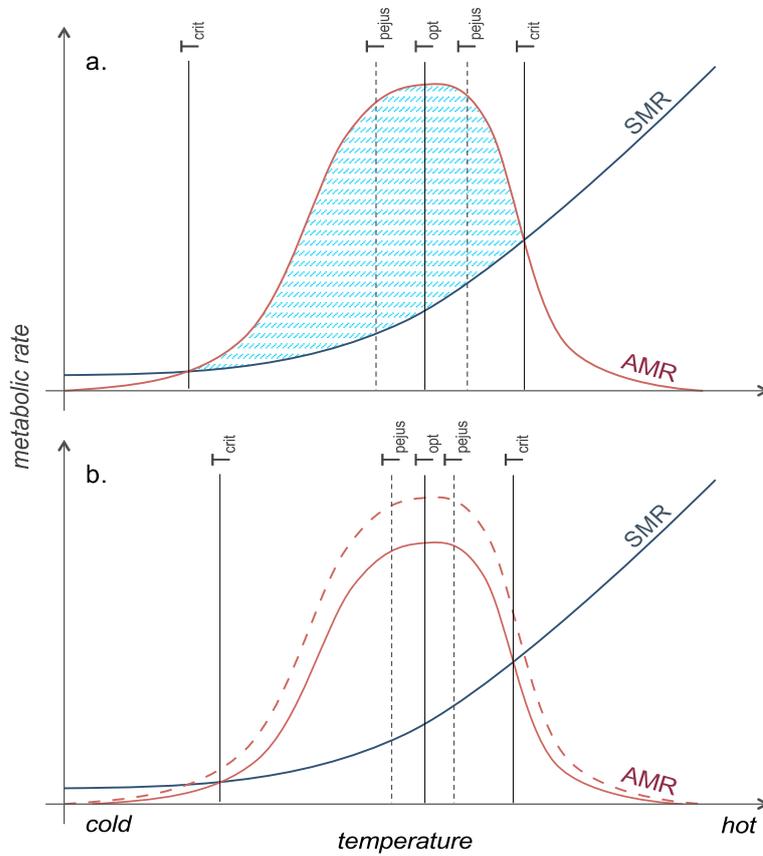


Figure 1: Schematic representation of changes in aerobic scope due to environmental perturbation. The temperature-limited thermal tolerance framework states that in case of temperature perturbation (a), standard metabolic rates (SMR; blue line) rise exponentially with increasing temperature (commonly described using a  $Q_{10}$  parameter of 2-3; Schurmann and Steffensen, 1997) while active metabolic rates (AMR; solid red line) follow a bell-shaped trend. The difference between the active and standard metabolisms is the aerobic scope (blue shaded region). Aerobic scope is largest at a temperature optimum ( $T_{opt}$ ) but exhibits a relatively stable trend up to a threshold on either side of the optimum ( $T_{pejus}$ ) where aerobic scopes become more acutely sensitive to temperature perturbation. Long-term habitation becomes

impossible under temperatures beyond the critical limits ( $T_{crit}$ ), which describe the limits of an above-zero aerobic scope. In the case of simultaneous stressors such as low oxygen (hypoxic stress) or high  $pCO_2$  (hypercapnic stress) **(b)**, a decline in active metabolic rate (dotted red line: unstressed; solid red line: stressed) often results in a narrowing of critical and pejus temperature limits and a decrease in the maximum active metabolic rate without changing the optimum temperature ( $T_{opt}$ ). Figure adapted from Pörtner (2001).

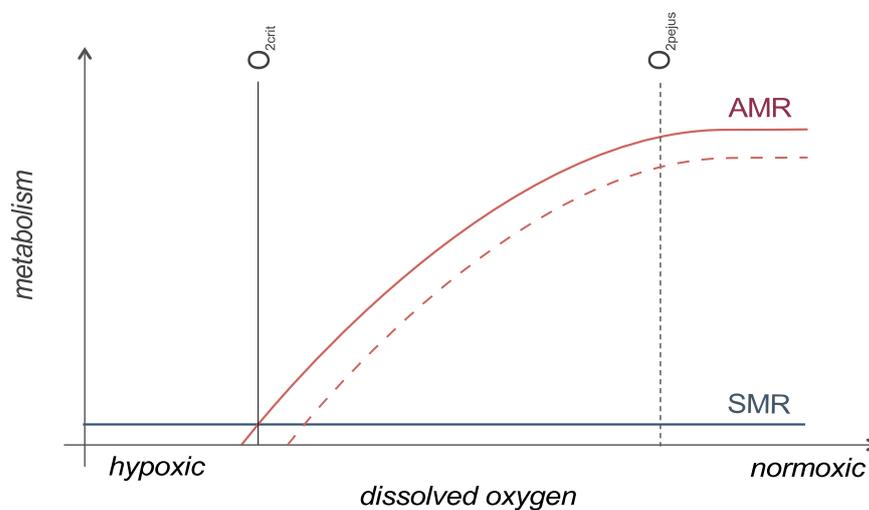


Figure 2: Change in aerobic scope in response to dissolved oxygen perturbation. The limiting oxygen concentration framework posits a maximum active metabolic rate (AMR; solid red line) at a high oxygen concentration (oxygen saturation = 100%) with declining oxygen concentrations resulting in an exponential decrease in active metabolism.  $O_{2pejus}$  and  $O_{2crit}$  are analogous to  $T_{pejus}$  and  $T_{crit}$  in Figure 1.

Simultaneous stressors such as high temperature (hyperthermia) or high  $pCO_2$  (hypercapnia) lead to a decline in maximum active metabolic rate (dotted red line) without changing the optimum dissolved oxygen concentration. Figure adapted from Claireaux and Lagardère (1999).

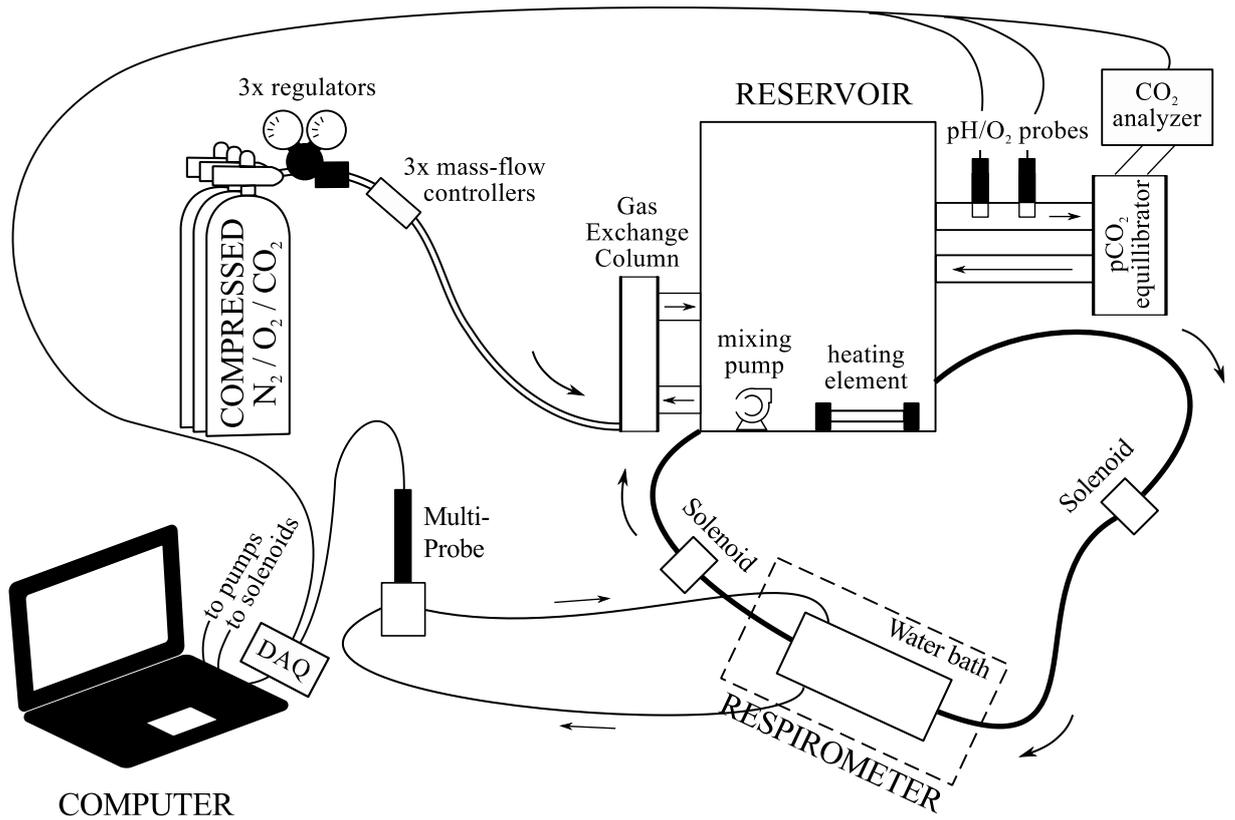


Figure 3: Schematic diagram of climate-controlled swim tunnel respirometer system.

A closed-feedback sensor loop controls the flow rate of N<sub>2</sub>, O<sub>2</sub> and CO<sub>2</sub> gas into a counter-current exchange column to set the gas tensions in a large UV-sterilized reservoir. Solenoids control the flow of water from the reservoir to the respirometers, which are maintained at constant temperature by a water bath. A dissolved oxygen optode, pH electrode, and temperature probe housed in a flow-through cell measures water conditions in the respirometer (Thermo Fisher Scientific, Waltham MA). The solenoids, gas flow rate, and sealing and unsealing of the respirometer are all fully automated through custom software implemented on a PC coupled with electronic controllers (Neptune Systems, Morgan Hill CA) with PID (proportional-integrative-derivate) software.

Table 1: Linear correlations between aerobic scope and fitness-related activities in captive teleosts. Significance denotes the p-value of a linear regression.

<b>Parameter</b>	<b>Species</b>	<b>Significance</b>	<b>R<sup>2</sup></b>	<b>Reference</b>
Specific Growth Rate	<i>Dicentrarchus labrax</i>	< 0.0001	0.975	(Claireaux and Lefrançois, 2007)
Specific Growth Rate	<i>Polyprion oxygeneios</i>	0.03	0.92	(Khan et al., 2014)
Specific Growth Rate	<i>Gadus morhua</i>	<0.01	0.847	(Claireaux et al., 2000)
Daily Feeding Rate	<i>Scophthalmus maximus</i>	<0.01	0.998	(Reidy et al., 2000)
Critical Swimming Speed	<i>Gadus morhua</i>	<0.05	0.525	(Mallekh and Lagardère, 2002)
Territorial Aggression	<i>Pomacentrus amboinensis</i>	0.001	0.267	(Killen et al., 2014)

Table 2: Comparison of predictive frameworks for forecasting the effect of environmental conditions on species occurrence, abundance, or performance. After Kearney and Porter (2009).

	<b>Mechanistic/experimental</b>	<b>Correlative/statistical</b>

	<b>Approaches</b>	<b>Advantages</b>	<b>Approaches</b>	<b>Advantages</b>
Model selection	Prescribed (e.g. energy balance equation)	Physiological traits provide common frame of reference	Flexible (e.g. generalized linear regression, maximum entropy)	More likely to fit available data
Generality	High across environments	Applies to non-equilibrium/novel environments	Local application	Less likely to overestimate potential range
Data requirement	Functional traits (physiology or behavioral responses)	Direct link between physiology and species range	Occurrence data (e.g. presence only, presence/absence, abundance)	Data source is commonly available
Scale	Individual organism	Less subjectivity in model selection	Flexible	Can capture processes across a range of scales
Genetic variability	Explicit	Can assess the degree of	Implicit	Geographic variation already

and physiological plasticity		adaptive capacity		incorporated in occurrence data
Output	Fitness components (e.g. growth, performance, reproductive capacity	Highly ecologically interpretable	Dimensionless habitat suitability indices or probability of occurrence or abundance	Single output indirectly represents many different processes
Model evaluation	Usually evaluated against independent datasets (e.g. occurrence/abundance in the field, field or laboratory measurements of physiology)	Can more easily evaluate against independent datasets	Evaluated against original input data, subsets of original data set aside for evaluation (i.e. cross- validation)	Convenient because model construction/evaluation relies on the same data source

## **2. An Aerobic Scope-Based Habitat Suitability Index for Predicting the Effects of Multi-Dimensional Climate Change Stressors on Marine Teleosts**

### **2.1. Abstract**

Climate change will expose many marine ecosystems to temperature, oxygen and CO<sub>2</sub> conditions that have not been experienced for millennia. Predicting the impact of these changes on marine fishes is difficult due to the complexity of these disparate stressors and the inherent non-linearity of physiological systems. Aerobic scope (the difference between maximum and minimum aerobic metabolic rates) is a coherent, unifying physiological framework that can be used to examine all of the major environmental changes expected to occur in the oceans during this century. Using this framework, we develop a physiology-based habitat suitability model to forecast the response of marine fishes to simultaneous CO<sub>2</sub>-induced ocean acidification, warming and deoxygenation, including interactions between all three stressors. We present an example of the model parameterized for *Thunnus albacares* (yellowfin tuna), an important fisheries species that is likely to be affected by climate change. We anticipate that if embedded into multispecies ecosystem models, our model could help to more precisely forecast climate change impacts on the distribution and abundance of other high value species. Finally, we show how our model may indicate the potential for, and limits of, adaptation to chronic stressors.

### **2.2. Introduction**

#### *2.2.1. Habitat quality is central to conservation biology*

Habitat alteration is the dominant driver of species extinction in our time (Homyack, 2010) and is likely to only increase in importance in the future due to the perturbations brought about by global climate change (Thomas et al., 2004; Travis, 2003). The ubiquity of habitat quality models (i.e. those that attempt to simulate species distribution based on habitat variables) across different disciplines of ecology reflects this importance. For example, one of the simplest forms of the habitat quality model is the climate envelope model (CEM), which correlates presence/absence data against the spatial distribution of environmental variables to try to define the limits of a species distribution (Thomas et al., 2004). A generalized form of the CEM is the habitat suitability model, which attempts to simulate not only the limits of a species' distribution but also its probability of occurrence/abundance within its range (Hirzel and Le Lay, 2008).

These habitat suitability models can be incorporated into more complex models either as a single-species add on to a fisheries assessment model (Link et al., 2011) or as part of a niche model that adds biotic factors (resource limitation or predation pressures that define a Grinnellian niche) through interactions with other species whose distribution and abundance is partially driven by habitat suitability models incorporating scenopoietic factors (e.g. temperature and oxygen limits that define an Eltonian niche) (Hutchinson, 1957). When enough niche models are aggregated to represent the full breadth of trophic levels existing in an ecosystem, these are then termed end-to-end ecosystem models (see Lehodey et al., 2008; Link et al., 2010; Maury, 2010) and may be used to perform large-scale simulations of ecosystem-responses to climatic or biogeochemical shifts (Lehodey et al., 2010).

Finally, a special application of habitat suitability models are those required by fisheries biologists who need to disentangle habitat effects from stock abundance in

order to constrain their estimates of optimum yield. These so-called habitat-based standardizations (Hinton and Nakano, 1996) generally take the form of correction factors to a “catchability” term that are used to ensure that the catch per unit effort at a given location and time is more reliably indicative of abundance (Maunder et al., 2006).

### 2.2.2. *Physiology is a key constraint for climate-change studies*

While all of the model types discussed above can be applied both to current and future environments, habitat suitability models that attempt to tackle the problem of climate change face fundamentally different challenges than those that are primarily concerned with current or historical environments. This is because applying habitat suitability models to a climate change future is inherently extrapolative – that is, future habitats are likely to involve either novel states (e.g. temperature or  $pCO_2$  levels not previously experienced in a given region) or novel combinations of states (e.g. unusually high temperatures combined with low oxygen) for which species distribution data is not currently available (Kearney and Porter, 2009). The inherent non-linearity of physiological control mechanisms (typically involving some tolerable range of environmental stressors bounded by *pejus* limits beyond which fitness rapidly declines (Nagarajan, 2002; Pörtner, 2002)) and the fact that the regulatory machinery involved in responding to distinct stressors often overlap (leading to complex interactions when multiple environmental variables are perturbed simultaneously; see (Claireaux and Lagardère, 1999; Kato et al., 2005; Lefrançois and Claireaux, 2003; Perry and Gilmour, 1996; Roch and Maly, 1979)) make these extrapolations highly uncertain. To make matters worse, many of the environmental variables used to inform habitat suitability models (e.g. temperature, oxygen,  $pCO_2$ )

exhibit high spatial correlation (Prince and Goodyear, 2006; Schmittner et al., 2008), compromising the ability to assign a gradient in species abundance to a specific variable. Indeed, many studies have demonstrated large variability in estimated abundance when extrapolating statistical models trained on current observations to future states (Thuiller, 2003; Thuiller, 2004; Thomas et al., 2004; Hijmans and Graham 2006).

One way to ameliorate these problems is to constrain the results of statistical habitat suitability models with information from organismal physiology (Denny and Helmuth, 2009; Hijmans and Graham, 2006). Physiological relationships provide a mechanistic (or semi-mechanistic) basis for understanding changes in abundance for which (in the absence of substantial genetic or epigenetic adaptation) we can assume stationarity into the relatively distant future (Kearney and Porter, 2004). Furthermore, because organismal physiology is principally modified by genetic or epigenetic processes, physiology-based models address a major weakness in statistical habitat suitability models by allowing us to ask questions such as (1) which physiological parameters could play a role in mitigating the changes to habitat suitability caused by climate change, and (2) is the currently observed variability in these parameters (indicative of the range of phenotypic plasticity) sufficient to adapt to future climate change even in the absence of genotypic or epigenetic modification (Kearney and Porter, 2009)?

### 2.2.3. *Aerobic scope is an integrative index of abiotic habitat suitability*

Implicit in the concept of linking organismal physiology to habitat suitability (and, after consideration of biotic factors, to abundance) is the criteria that the metric of physiological state used must be linked to fitness (Homyack, 2010). The most popular

methods utilize some measure of energetic state, such as the metabolic cost of key fitness-determining behaviors including digestion, hibernation, flight, and foraging; or of benchmark metabolic states such as standard (minimum) or active (maximum) metabolism (Braaten et al., 1997; Homyack, 2010; Rosenfeld et al., 2005). In this paper we utilize aerobic scope (the difference between maximal aerobic metabolism and the minimum metabolic expenditure needed for survival (Fry, 1947)) as an integrative index of habitat suitability with strong ties to fitness. Because the machinery that underlies aerobic metabolism is conserved across metazoan taxa (Thannickal, 2009), much of the discussion below is broadly applicable.

The correspondence between aerobic scope and the other energetic indicators discussed above can be seen by partitioning an organism's energy budget into four parts: (1) the proportion of total energy required to maintain essential life processes, (2) the proportion that can be devoted to foraging and somatic growth, (3) the proportion dedicated to reproduction either in the form of gonadal growth or activities such as maintaining a territory, and (4) the energy available for storage (Homyack, 2010). In vertebrates, almost all of these energetic costs must be paid for through aerobic metabolism (Pörtner, 2010), either by directly powering these activities or via excess post-exercise oxygen consumption (EPOC: Gaesser and Brooks, 1983; Lee et al., 2003). The aerobic scope can be thought of as the total potential metabolic expenditure available for fueling portions 2-4 in excess of that required for portion 1. Reductions in aerobic scope either through an increase in minimum metabolism (a rise in portion 1) or a decrease in maximum metabolism (a decline in the total available energy output) therefore imply a reduction in the maximum rate for one of the processes (growth, reproduction, or energy storage) that drives current and future fitness. Furthermore, aerobic scope defines physical limits to long-term habitation at

the critical limits (conditions under which aerobic scope is zero) beyond which maintenance of essential life processes is only possible by drawing down stored oxygen reserves or relying (temporarily) upon anaerobic capacity (Pörtner, 2010). This definition is only a slight variation on what is termed the fundamental niche: the space where, in the absence of biotic interactions, an organism can survive and reproduce (Kearney and Porter 2004). These properties make aerobic scope a strong metric on which to base a physiological habitat-suitability model.

#### 2.2.4. *Thunnus albacares* as a model organism for aerobic scope modeling

While a number of species could conceivably be used to parameterize an aerobic scope model, we chose the yellowfin tuna *Thunnus albacares* for this study due to the wealth of physiological data available on this species as well as the multiple environmental changes that are likely to befall portions of its habitat due to climate change. These include increased surface ocean temperatures, surface ocean and thermocline deoxygenation, and increased hypercapnia (high levels of  $pCO_2$ ) both in the surface ocean and especially in the thermocline and oxygen minimum zones where high levels of dissolved carbon dioxide have already substantially depleted the buffering capacity of seawater (Brewer and Peltzer, 2009; Cai et al., 2011; Gruber, 2011a). We rely on well-established physiological equations to create an aerobic scope estimator from already available laboratory data, and incorporate both the mean and variability of this data to create not just an estimate of how these complex environmental perturbations could affect habitat suitability for *T. albacares* but also some indications of the extent to which the observed variability in these parameters could compensate for future climate change.

While this paper provides one example of an aerobic-scope based habitat suitability model parameterized for a single species at a single location, we have designed this model to be applicable to a wide range of teleosts in many different marine environments with very little modification, and we hope broader aspects of the model can be useful for an even wider range of taxa.

### **2.3. Methods**

We developed a three-box model of a tuna consisting of a conceptually infinite seawater reservoir, a gill reservoir, and a reservoir for the working tissues (principally the muscles). The maximum metabolic rate model consists of ten coupled equations drawn from classical physiological relationships (e.g., the Fick Equation, the Van't Hoff Equation, the Bohr effect) and empirical regression equations (e.g., temperature and oxygen effects on cardiac output), from which we subtract an empirically derived estimate of minimum metabolic rate to obtain the aerobic scope. In the maximum metabolic rate model, the flux of blood between the gills and working muscles is determined by the cardiac output (the rate of blood flow out of the heart), and the rate of oxygen flux per unit of blood volume is determined by modeled red blood cell kinetics, which is in turn affected by oxygen concentration, temperature, and blood pH.

Vertebrate red blood cells are extremely complicated physiological systems with many effectors. The effects of allosteric molecules such as adenosine triphosphate (ATP) and guanosine triphosphate (GTP), as well as the effects of catecholamine release driving the activity of Na<sup>+</sup>/H<sup>+</sup> exchangers in the red blood cell membranes have not been explicitly modeled in these equations even though they can have strong effects on oxygen binding. Both nucleoside triphosphates and circulating

catecholamines can increase dramatically in fishes during maximal exercise and in other aerobically stressful conditions (Berenbrink and Bridges, 1994; Lowe and Brill, 1998). Our approach to this issue was to parameterize our equations as much as possible from data derived from experimental animals undergoing maximal exercise under the assumption that these parameters more closely reflected the kinetics of oxygen binding under conditions of maximal NTP and catecholamine release (e.g. Korsmeyer et al (1997)). The full list of parameter names, symbols, and chosen values are reported in Tables 1 and 2.

### 2.3.1. Red blood cell kinetics

To approximate the effects of oxygen partial pressure, temperature and pH on the oxygen content of the blood, we begin by expressing arterial and venous oxygen content ( $O_a$  and  $O_v$ , respectively) as a function of red blood cell hemoglobin oxygen saturation ( $S$ ) and the red blood cell oxygen capacity ( $\kappa$ ) (Korsmeyer et al., 1997)

$$O_a = S_a \kappa, \quad O_v = S_v \kappa \quad (2.3-1)$$

where  $\kappa$  is the product of the mean corpuscular hemoglobin concentration ( $\bar{H}$ ), the hemoglobin oxygen binding capacity ( $O_H$ ) and the hematocrit ( $Hct$ )

$$\kappa = \bar{H} \cdot O_H \cdot Hct \quad (2.3-2)$$

The red blood cell hemoglobin oxygen saturation can in turn be expressed as a function of the oxygen partial pressure ( $pO_2$ ), the half-saturation constant ( $p50$ ), and the hill coefficient of cooperativity ( $\eta$ ) (Hill, 1910)

$$S = \frac{pO_2^\eta}{pO_2^\eta + p50^\eta} \quad (2.3-3)$$

We then model the effect of temperature and pH on the half-saturation constant by combining the Bohr (Siggaard-Andersen and Garby, 1973) and Van't Hoff equations (Brill and Bushnell, 1991)

$$\log p50(\Delta T, \Delta pH) = \log p50_0 + \varepsilon_T + \varepsilon_{pH} \quad (2.3-4)$$

$$\varepsilon_T = \frac{H[\Delta(T^{-1})]}{2.303R} \quad \varepsilon_{pH} = \Delta pH \cdot \beta \quad (2.3-5)$$

where  $R$  is the ideal gas constant ( $1.987 \text{ cal K}^{-1} \text{ mol}^{-1}$ ),  $H$  is the apparent heat of hemoglobin oxygenation and  $\beta$  is the extracellular fixed-acid Bohr coefficient ( $\Delta \log p50 / \Delta pH$ ).

Solving equation (2.3-1) under the pH,  $pO_2$ , and temperature conditions of the arterial and the venous bloodstreams allows us to calculate the arterio-venous oxygen difference ( $\Delta O$ ) as simply

$$\Delta O = O_a - O_v \quad (2.3-6)$$

### 2.3.2. Blood carbonate chemistry

To solve the carbonate system in the arterial and venous blood, we began by making the following assumptions:

- i. The  $pCO_2$  of the arterial bloodstream ( $pCO_{2a}$ ) is equal to the  $pCO_2$  of the inspired seawater. This condition approximates the condition for water-breathing animals, which are unable to regulate  $pCO_{2a}$  through ventilatory processes (Cameron and Polhemus, 1974) and (in the absence of metabolic acidosis arising from significant anaerobic energy output) are generally unable to maintain substantial  $pCO_2$  gradients across the gill membranes (Ishimatsu et al., 2004; Toews et al., 1983).

- ii. The pH of the arterial bloodstream ( $pH_a$ ) is constant in the face of environmental hypercapnia. This assumption follows from experimental results showing a complete compensation of arterial pH within 12-24 hours in the face of acute hypercapnia up to and including lethal levels (Ishimatsu et al., 2005).
- iii. The temperature of the arterial bloodstream ( $T_a$ ) equals the temperature of the inspired water, while the temperature of the venous bloodstream ( $T_v$ ) equals the temperature of the arterial bloodstream plus a constant term ( $T_{excess}$ ) reflecting the effect of regional endothermy (in certain teleosts including some in the genus *Thunnus*) on the temperature of the working muscles.
- iv. The venous dissolved inorganic carbon concentration ( $C_v$ ) is equal to the arterial dissolved inorganic carbon concentration ( $C_a$ ) plus the moles of oxygen produced through metabolism (i.e.  $\Delta O$ ) due to the stoichiometry of the cellular respiration equation assuming a respiratory quotient ( $RQ$ ) of 1. An  $RQ$  of one, implying that metabolic output is dominated by carbohydrate oxidation (as opposed to fats or proteins) driving aerobic metabolism, is appropriate for states of maximal aerobic output where the rates of fat and protein catabolism are too slow to make a significant contribution (Kutty, 1968; Richardson, 1923).
- v. The venous oxygen partial pressure ( $pO_{2v}$ ) at maximal exercise approaches a constant value (the minimum venous oxygen partial pressure). This assumption derives from the fact that exercising venous  $pO_2$  is relatively invariant in a wide variety of animals (Farrell, 2002), as is the case for *T. albacares* (Bushnell, 1988). It is important to note however that this does not

preclude changes to venous oxygen *content* ( $O_v$ ) due to fluctuations in the pH or temperature of the venous bloodstream. In this study, we do not take into account the collapse of  $pO_{2v}$  beyond the minimum limits which can lead to acute cardiorespiratory collapse and reduced neural functioning (Farrell, 2002) because these values occur beyond the pejus ranges we are considering.

- vi. The arterial oxygen partial pressure ( $pO_{2a}$ ) is equal to the  $pO_2$  of the inspired water multiplied by a utilization constant ( $U$ ) where  $0 < U \leq 1$ .

With these assumptions, we solve for arterial DIC and total alkalinity ( $A_a$ ) given arterial pH,  $pCO_2$  and temperature using the R package seacarb (Lavigne et al., 2014).

### 2.3.3. Arterio-venous oxygen difference

We initially assume a constant DIC ( $C_v = C_a$ ) and the absence of an arterio-venous oxygen difference ( $\Delta O = 0$ ). We then iterate through the equations

$$\Delta O_i = O_a - O_{v_i}(C_{v_i}, T_v, pO_{2v}) \quad (2.3-7)$$

$$C_{v_{i+1}} = C_v + \Delta O_i \quad (2.3-8)$$

with the condition  $\Delta O_i \neq 0$ .

The stable value ( $\Delta O_n$ ), defined operationally as  $\Delta O_{i+1} - \Delta O_i < 10^{-7} \text{ mol kg}^{-1}$  is then taken to be the steady-state arterio-venous oxygen difference for a given set of environmental parameters.

### 2.3.4. Cardiac Output

To characterize the maximum cardiac output ( $\dot{Q}$ ) under environmental perturbation we first made the following assumptions about the temperature ( $T_h$ ) and oxygen ( $O_h$ ) state of the heart

$$T_h = T_a, \quad O_h = fO_a + (1 - f)O_v \quad (2.3-9)$$

where  $0 \leq f \leq 1$ . In all teleosts including regional endotherms such as tuna, the position of the heart upstream of the rete mirabile means that cardiac temperature is indistinguishable from the temperature of the arterial circulation (and therefore from the temperature of the inspired seawater). Oxygen supply to the heart however shows marked variation between taxa. In many teleosts, the myocardial oxygen supply derives entirely from the venous circulation (Farrell, 2002), corresponding in the model to a situation where  $f = 0$ . More athletic fishes typically possess a coronary circulation that supplies arterial blood to the compact myocardium. In salmonids for example it is estimated that the coronary vasculature supplies about 30% of peak cardiac demand ( $f \cong 0.3$ ) (Farrell, 2002). Tunas such as *T. albacares* have developed an even more extensive coronary circulation capable of supplying the majority of cardiac oxygen demand (Farrell, 1996), and we set  $f = 1$  in this case.

We characterized the maximum  $\dot{Q}$  using an empirically determined two-piece linear function (analogous to that used by Bigelow and Maunder (2007)) where temperature and oxygen were assumed to have a constant effect on  $\dot{Q}$  below a threshold value ( $O_T, T_T$ ) and no effect above it.

$$\dot{Q}(O_h, T_h) = Q_{max} \left[ \frac{\delta \dot{Q}}{\delta O_h} (O_h - O_T) + \frac{\delta \dot{Q}}{\delta T_a} (T_h - T_T) \right] \quad (2.3-10)$$

$$\frac{\delta \dot{Q}}{\delta O_h} = \begin{cases} \lambda_1, & O_h < O_T \\ 0, & O_h \geq O_T \end{cases}, \quad \frac{\partial \dot{Q}}{\partial T_h} = \begin{cases} \lambda_2, & T_h < T_T \\ 0, & T_h \geq T_T \end{cases} \quad (2.3-11)$$

where  $Q_{max}$  is the maximum possible cardiac output under optimum conditions and  $\lambda_1$  and  $\lambda_2$  are the magnitudes of the oxygen and temperature sensitivity, respectively.

### 2.3.5. Aerobic scope

We calculate maximum aerobic metabolism ( $\dot{M}_{max}$ ) by combining equations (2.3-6) and (2.3-10) into the Fick equation (Selzer and Sudrann, 1958)

$$\dot{M}_{max} = \dot{Q}(\Delta O) \quad (2.3-12)$$

The minimum metabolic rate ( $\dot{M}_{min}$ ) is then estimated according to a commonly used empirical regression

$$\dot{M}_{min} = cF^{0.1\Delta T} \quad (2.3-13)$$

In the case of *T. albacares*, we used the regression parameters calculated by Dewar and Graham (1994) for tuna of mass 2.17 kg (fork length 51.3 cm) corrected to the minimum hydrostatic swimming speed calculated by Magnuson (1973) to obtain a routine metabolic rate. The mass was chosen to be similar to the electronically tagged fish for which depth distribution data could be obtained (Schaefer et al., 2007). The routine metabolic rate correction was used to provide a more realistic minimum survivable metabolic rate than would standard metabolism, which is defined as metabolism in the absence of growth, digestion or locomotion and would constitute a lethal metabolic expenditure in obligate ram ventilators such as tunas (Bushnell and Brill, 1991; Dotson, 1976).

Finally, aerobic scope ( $\dot{M}_{scope}$ ) is obtained by subtracting equation (2.3-13) from (2.3-12)

$$\dot{M}_{scope} = \dot{M}_{max} - \dot{M}_{min} \quad (2.3-14)$$

### 2.3.6. Statistical Methods

Uncertainty estimates for the aerobic scope were obtained using latin-hypercube sampling based Monte Carlo error propagation (samples per parameter = 1000).

Throughout this paper, uncertainty estimates reflect both the range of physiological

parameter values (e.g. variability in the measured Bohr coefficient  $\beta$ ) and the variability in environmental conditions (e.g. water temperature at depth). In all cases, input parameters were assumed to be normally distributed with mean and standard deviations calculated from experimental data.

Sensitivity metrics were estimated using Sobol global variance-based sensitivity analysis (Saltelli et al., 2010) using the Python package SALib (Herman et al., 2014) to determine both first-order main effects (the magnitude of uncertainty of the output variable resulting from uncertainty in the input parameter) and total effects (main effects + interaction effects).

### 2.3.7. Model Evaluation

To evaluate the aerobic scope estimate as an indicator of habitat suitability, we constructed a simple linear estimator of vertical habitat usage

$$f_i = y\dot{M}_{scope_i} + b \quad (2.3-15)$$

where  $f_i$  is the percentage of total time that a fish would be expected to spend at depth  $i$ ,  $\dot{M}_{scope_i}$  is the output of the aerobic scope model evaluated at the environmental conditions (*temperature*,  $pO_2$ ,  $pCO_2$ ) for that depth, and  $y$  and  $b$  are fitting parameters. To isolate the scenopoetic influence on habitat choice and minimize any bionomic effects (e.g. the availability of prey, competition from other predators) we focused on a study area with strong gradients in environmental variables driving pronounced vertical habitat compression, which for *T. albacares* was chosen to be the Eastern Tropical Pacific oxygen minimum zone, with further focus on the region bounded by the latitudes 5°S-5°N and the longitudes 90-105°W (the study area) for

which electronic-tagging data on vertical habitat usage was presented in Schaefer et al. (2007).

## 2.4. Results

### 2.4.1. Interactions between environmental variables

The aerobic scope model revealed multiple putative interactions between environmental variables (Figure 4). The effects of temperature perturbations on  $\dot{M}_{scope}$  showed the largest amplification under conditions of high  $pO_2$ , low temperature, and low  $pCO_2$ , with the sign of the temperature effect ( $\partial\dot{M}_{scope}/\partial T$ ) being positive below the threshold temperature  $O_T$  (increasing temperature increases  $\dot{M}_{scope}$ ) and negative above (increasing temperature decreases  $\dot{M}_{scope}$ ). Oxygen perturbations ( $\partial\dot{M}_{scope}/\partial pO_2$ ) were always positive and most influential at the threshold oxygen concentration ( $O_T$ ), threshold temperature ( $T_T$ ) and low  $pCO_2$ . Finally,  $pCO_2$  effects ( $\partial\dot{M}_{scope}/\partial pCO_2$ ) were always negative and relatively invariant across different  $pO_2$ , but amplified at high  $pCO_2$  and high temperature.

### 2.4.2. Defining projected climate change perturbations

Recent historical data for the study area were obtained from the World Ocean Atlas 2013 (Garcia et al., 2014; Locarnini et al., 2013) and the Global Ocean Data Analysis Project (Key et al., 2004) for the depths 0-325 m (Figure 5). These data showed substantial changes in temperature,  $pO_2$  and  $pCO_2$  within the upper 100 m, with the depth profiles being divisible into a surface mixed layer (0-35 m depth), a thermocline layer (35-80 m depth), and an oxygen minimum layer (80-300 m). Representative mean conditions in the three layers were: mixed layer (Temperature

296 K;  $pO_2$  0.19 atm;  $pCO_2$  410  $\mu\text{atm}$ ), thermocline (Temperature 291 K;  $pO_2$  0.12 atm;  $pCO_2$  510  $\mu\text{atm}$ ), and oxygen minimum layer (Temperature 287 K;  $pO_2$  0.07 atm;  $pCO_2$  620  $\mu\text{atm}$ ). Comparison of these values against projections for end-century environmental conditions obtained from the Meteorological Research Institute-Earth System Model Version 1 (MRI-ESM1; (Yukimoto et al., 2011)) under the RCP 8.5 emissions scenario suggest that our study area will undergo a suite of perturbations over the next century, including a 2-5 K increase in mean temperatures, an approximately 20,000  $\mu\text{atm}$  increase in mixed layer  $pO_2$  coupled with an up to 30,000  $\mu\text{atm}$  decline in the oxygen minimum layer, and a  $pCO_2$  increase ranging from 700  $\mu\text{atm}$  at the surface to 2,800  $\mu\text{atm}$  at depth.

#### 2.4.3. *Aerobic scope as a predictive variable*

Predicted aerobic scope was obtained for the study region by Monte Carlo sampling of both historical environmental variables (using the data sources described in section 2.4.2) and physiological parameters (the means and standard deviations are listed in Table 4). We tested the utility of the predicted aerobic scope as an indicator of habitat suitability by regressing predicted aerobic scope against the observed vertical distribution of 5 electronically tagged *T. albacares* in the same region (Schaefer et al., 2009). The resulting correlation (Figure 6) shows some autocorrelation of residuals but also a tight ( $R^2 = 0.91$ ) and highly significant ( $p < 0.0001$ ) linear fit, with the value of the regression coefficients in equation (3.7-1) estimated as  $y = 2.79$ ,  $b = 5.10$ .

#### 2.4.4. *Projected changes in vertical habitat*

Using the linear regression model of vertical distribution obtained in 2.4.3, we projected changes in vertical habitat between current and end-century conditions in

the Galapagos region (Figure 7). Our analysis showed vertical habitat compression driven by the changing environmental variables in the thermocline resulting in the shoaling of the critical limits (environmental conditions causing aerobic scope to fall to zero) from 90 m in the current ocean, to 60 m by the end of the century. By allowing the input physiological parameters to vary in a bootstrapping scheme (using the mean and standard deviations reported in Table 4), we were able to isolate the 95<sup>th</sup> percentile of results for which the estimated aerobic scope in the end-century scenario was closest to the present. This analysis showed that the observed range of physiological plasticity could mitigate the effects of climate change perturbations in the upper thermocline (40-65 m) but is unlikely to mitigate impacts in deeper waters.

#### 2.4.5. Sensitivity analysis

A Sobol global variance-based sensitivity analysis revealed a large range in model sensitivities to the input physiological parameters for *T. albacares* (Figure 8), with temperature and oxygen related parameters ( $O_{utilization}$ ,  $O_{slope}$ ) and the magnitude of regional endothermy ( $T_{excess}$ ) having the strongest effect. In all cases except for the temperature threshold of cardiac output ( $T_T$ ) and the slope of the temperature response ( $\lambda_2$ ), the analysis showed a substantially (3-9 fold) higher interaction effect compared to the main effect.

## 2.5. Discussion

### 2.5.1. Aerobic scope as a habitat suitability index

Our results demonstrate a habitat suitability model capable of integrating multivariate climate-change related perturbations into a unified physiological framework. Using input parameters representative of a generalized marine teleost, we

revealed that our model includes multiple putative interactions between simultaneous environmental stressors that are likely to be lacking in correlative habitat suitability models. This is particularly important when considering that climate change perturbations are generally multivariate (Gruber, 2011b; Prince and Goodyear, 2006) and that complex interactions between stressors (including synergisms and antagonisms) are ubiquitous in ecological systems (Crain et al., 2008). The complex changes projected to occur by the end of the century in our study region in the Eastern Tropical Pacific exemplify these trends, with each of the different depth zones (mixed layer, thermocline, and oxygen minimum layer) likely to experience qualitatively distinct perturbations over the next 100 years. Using input parameters specific to one teleost (*T. albacares*), we create a simple linear model to predict vertical habitat usage from aerobic scope and demonstrate that this estimator has high fidelity to the actual habitat usage observed through electronic tagging of 5 individuals in the study region. Furthermore, we project that these changes could manifest as a pronounced vertical habitat compression in this region. Through Monte Carlo sampling from the distribution of measured physiological parameters, we not only estimate uncertainty distributions for this vertical habitat usage, we also derive an estimator of adaptive capacity and predict that at least some of the projected habitat compression for *T. albacares* will be unaffected by adaptation operating through selection pressures alone.

### 2.5.2. *Consequences of vertical habitat compression*

Projected vertical habitat compression in the study area portends declining *T. albacares* abundance in the Eastern Tropical Pacific oxygen minimum zone in the

future. This declining abundance may be exacerbated by a rise in catchability as the squeezing of the tuna population into a shallower depth range makes them more accessible to the longline and purse-seine fisheries operating in this region (Prince and Goodyear, 2006). In particular, surface-floating fish-aggregating devices (FADs), which are widely used between 15°S-5°N (Schaefer and Fuller, 2004) could interact more frequently with increasingly surface-restricted tunas. While we focus on one section of the eastern tropical pacific oxygen minimum zone near the Galapagos islands for which tagging data happened to be available, similar environmental pressures are likely to lead to habitat compression across much of the rest of the oxygen minimum zone, including the nearby and highly productive Costa Rican Dome ecosystem. Any changes will have a disproportionate impact on global *T. albacares* landings because of the unusually large tuna yield from this region (Essington et al., 2002).

Interestingly, our model suggests that part of the current depth distribution of *T. albacares* may already lie outside the critical limits set by environmental variables. This is consistent with studies on other epipelagic predators associated with oxygen minimum zones such as *Dosidicus gigas* (Rosa and Seibel, 2008), and implies that *T. albacares* spend a substantial portion of their time (18%) making temporary forays into habitats that require them to rely on some combination of blood oxygen stores and anaerobic metabolism (paid for by elevated metabolic rates upon return to more favorable surface waters). The ability of some tagged *T. albacares* to move deep into physiologically inhospitable environments including at least one dive to 1160 m and an ambient water temperature of 5.8°C (Dagorn et al., 2006) indicates that these adaptations may be highly developed. Our analysis suggests that *T. albacares*

populations in the area will be forced to rely more frequently on these adaptations in the future.

### 2.5.3. *Methods for integrating bionomic factors into our model*

In our limited demonstration we assumed that aerobic scope (driven only by scenopoetic factors) was the dominant driver of habitat usage in the study region. This assumption was based on the existence of strong, persistent environmental gradients in the Eastern Tropical Pacific resulting in substantial vertical habitat limitation for tunas (Schaefer et al., 2009). While the remarkable tightness of fit of our model ( $R^2$  was greater than 0.9) against habitat usage patterns observed in tagged animals is consistent with this assumption, there is noticeable autocorrelation of residuals, and it is likely that a parabolic or other nonlinear estimator could be more appropriate in future studies. Other possible sources of uncertainty include differences between the actual environmental conditions experienced by the tagged fish and the assumed conditions drawn from the WOA and GLODAP climatology, or to the contribution of minor variables that were not accounted for in the model. The most obvious candidates for these minor variables are bionomic factors.

More sophisticated analyses attempting to account for these factors, especially in regions with permissive environmental conditions, could easily incorporate the effect of additional bionomic terms. One term that is particularly straightforward to implement is prey availability, which can be conceptualized as a limiting factor that converts between the potential metabolic output (set by the aerobic scope) and the portion of that output that is consistent with the rate of energy input (prey availability). For example, a simple habitat preference model incorporating a prey density term could take the form

$$H = y\dot{M}_{scope} \left( \frac{\rho}{\rho_{max}} \right) + b \quad (2.5-1)$$

where  $y$  and  $b$  are fitting parameters,  $\dot{M}_{scope}$  sets the potential (maximum) habitat preference, and the normalized prey density  $\rho/\rho_{max}$  is a scaling factor to obtain the realized preference  $H$ . A similar prey-availability index ranging from 0 to 1 is used in the SEAPODYM ecosystem model (Lehodey et al., 2008). In non-extrapolative applications (e.g. fisheries monitoring), this prey-availability term could be allowed to vary over space to estimate a time-varying abundance term. For example, the commonly estimated abundance effect ( $I_t$ ) in fisheries models could be derived using an equation of the form (modified from Maunder et al. (2006))

$$I_t = \frac{C_{n,t}}{H_{n,t}E_{n,t}} \quad (2.5-2)$$

where  $C_{n,t}$  is the observed catch at location  $n$  and time  $t$ ,  $H_{n,t}$  is the estimated habitat preference according to equation (2.5-1) (with  $y$  and  $b$  as constants,  $\dot{M}_{scope}$  allowed to vary over time and space and  $\frac{\rho}{\rho_{max}}$  only over space) and  $E_{n,t}$  is the observed fishing effort (e.g. number of hooks).

Conceptually, the methodology we present for linking the habitat preference  $H$  to species distribution is similar to the statistical habitat-based standardization method (statHBS) described by Maunder et al. (2006). In both cases, a function that takes environmental parameters as inputs and returns the estimated species distribution is assumed to take a certain form (our study assumes a substantially more complicated form for these equations than do Maunder et al.), and then a statistical process is used to fit this function to some dataset of observed species distributions. While in a typical statHBS model the fitting parameters determine the effect of each environmental

variable on the species distribution (e.g. Bigelow and Maunder (2007)), in our case the cumulative effect of all the environmental variables is calculated according to our aerobic scope model (driven by fixed physiological input parameters), and the fitting parameters try to match this cumulative effect against the observed distribution. Although the benefit of our method is that it is less liable to over-fitting, this comes at a cost of a reliance on a large volume of physiological data. Especially in complex models incorporating predator-prey interactions between multiple species, resource and data constraints will likely dictate that only a few focal species are parameterized with our aerobic scope model, while other species may need to rely on simpler CEM-type models.

#### *2.5.4. Predicting the outcome of selection pressures*

Hinton and Nakano (1996) described a deterministic habitat-based standardization for blue marlin where input parameters (habitat preferences) were drawn not from a statistical fitting scheme as in statHBS but through studies of electronically tagged fish. By allowing the input parameters to vary according to a multinomial distribution, the authors were able to estimate confidence intervals in their predicted abundance trends. The aerobic scope model we describe also allows us to use this procedure with the added benefit that the nature of our input variables allows to us decompose the variability in estimated aerobic scope between the physiological and environmental input parameters. The variability driven by the physiological input parameters is particularly interesting because it provides an estimate of how selective pressures acting to modify the frequency of phenotypes already present in wild populations could mitigate any fitness decrement expected under future climate change. For example, our analysis of *T. albacares* in the Eastern Tropical Pacific suggests that

these selective pressures could mitigate some, but not all of the projected habitat compression resulting from climate change. Specifically, at some depths such as the upper thermocline, these selective processes may reduce the effects of climate change-induced vertical habitat compression by as much as 50%, while climate change effects are likely to be unaffected by adaptive compensation at deeper depths. Increased data collection (e.g. more physiological measurements from *T. albacares* in the same or from different populations) could be used to further constrain the distribution of physiological phenotypes in the wild and improve the confidence in these predictions.

Another perspective on this problem is provided by the global sensitivity analysis. One interpretation of the sensitivity indices presented in Figure 8 is that parameters assigned large sensitivity values (such as the percent oxygen utilization  $O_{utilization}$ ) have a large impact on the estimated aerobic scope, and therefore more accurate measurements of these variables could contribute most to a narrowing of the confidence bounds of the model. A parallel interpretation that is relevant to the adaptation problem is that high sensitivity indices show the parameters that are most likely to respond to selective pressures because changes in these parameters (either through selective enrichment of more extreme phenotypes or through genetic/epigenetic evolution) can result in substantial changes in aerobic scope.

It is important to note however that our analysis assumes that all of the physiological input parameters are independent, and therefore fails to take into account some adaptive strategies that would necessitate modifications to multiple input parameters. For example, the large total sensitivity index for  $O_{utilization}$  (defined as the ratio of  $pO_2$  in the arterial bloodstream versus the  $pO_2$  in the inspired water) indicates that all else being equal, increasing the arterial  $pO_2$  could contribute

substantially to maintaining a high aerobic scope. One of the primary methods for increasing arterial  $pO_2$  in fishes is by increasing the residence time of arterial blood in the gills, which can be accomplished most easily by decreasing cardiac output. Cardiac output however is directly proportional to aerobic scope via the Fick equation (2.3-12), therefore increasing  $O_{utilization}$ , if accomplished via a decline in cardiac output, is likely to be counterproductive in this case. On the other hand,  $O_{utilization}$  can also be boosted without affecting cardiac output by modifying the architecture of the gills (e.g. by increasing lamellar surface area, (Sollid et al., 2003)), and such adaptations are much more likely to produce an effective response to future climate change.

#### 2.5.5. *Applicability to other species*

While we have presented only two parameterizations of our model (the generic teleost model using input parameters from Table 3 and the *T. albacares* model using the parameters from Table 4), our model was deliberately constructed to be compatible with a wide range of marine teleosts. In the case of well-studied species including *Gadus morhua* (Atlantic cod), *Oreochromis mossambicus* (Mozambique tilapia), *Mugil cephalus* (striped mullet) and *Katsuwonus pelamis* (skipjack tuna), most if not all of the input parameters necessary for this model can be gleaned from published laboratory experiments. For other less-studied species, the necessity for data collection can be mitigated somewhat by utilizing available information for congeners or other relatives in the case of variables to which the model has low sensitivity, such as the hill number or arterial pH (Figure 8, variables  $pH_a$  and  $\eta$ ). Predicting the range shifts of some species such as *T. albacares* and *G. morhua* is of particular interest because studies have already shown significant changes in

distribution due either to ongoing climate change (Rose, 2004) or episodic climate events such as the El Niño Southern Oscillation (Lehodey et al., 1997; Pearcy, 2002). Integrating our habitat suitability formulation as a single-species add-on to end-to-end ecosystem models such as SEAPODYM, APECOSM or ATLANTIS (Lehodey et al., 2008; Link et al., 2010; Maury, 2010) could allow researchers to improve their analysis of how climate change perturbations in the distribution of these species might interact with food web structure and biogeochemical cycling. We have also described in section 2.5.3 how fisheries assessment models could integrate our habitat suitability index. We anticipate that both these cases, the physiological basis of our model and the resulting robustness to extrapolation and over fitting will improve the quality of these efforts.

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## 2.8. Figures and Tables:

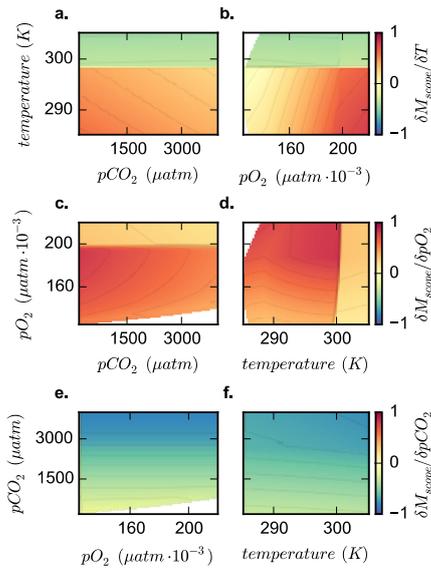


Figure 4: Aerobic scope model reveals complex putative interactions of environmental variables in a generic teleost fish. **(a)** Change in estimated  $\dot{M}_{scope}$  with increasing temperature (normalized to the range 0-1 and averaged across all possible  $pO_2$  values) is shown for all combinations of temperature and  $pCO_2$ . Colors indicate regions where  $\dot{M}_{scope}$  increases (red) or decreases (blue) with increasing temperature. Contours (black lines) help to visualize patterns in the effect size. The slant in the contour lines below temperatures of 298 K (the value of the cardiac temperature threshold  $T_T$ , see Table 3) for example reveals that increasing temperature has the largest effect on  $\dot{M}_{scope}$  ( $\partial\dot{M}_{scope}/\partial T$  is highest) at low temperatures and low  $pCO_2$ . **(b)**  $\partial\dot{M}_{scope}/\partial T$  across the temperature/ $pO_2$  space,  $\partial\dot{M}_{scope}/\partial pO_2$  across the **(c)**  $pO_2/pCO_2$  and **(d)**  $pO_2$ /temperature space, and  $\partial\dot{M}_{scope}/\partial pCO_2$  across the **(e)**  $pCO_2/pO_2$  and **(f)**  $pCO_2$ /temperature space. Physiological parameters for this model realization are shown in Table 3 with typical values from the literature.

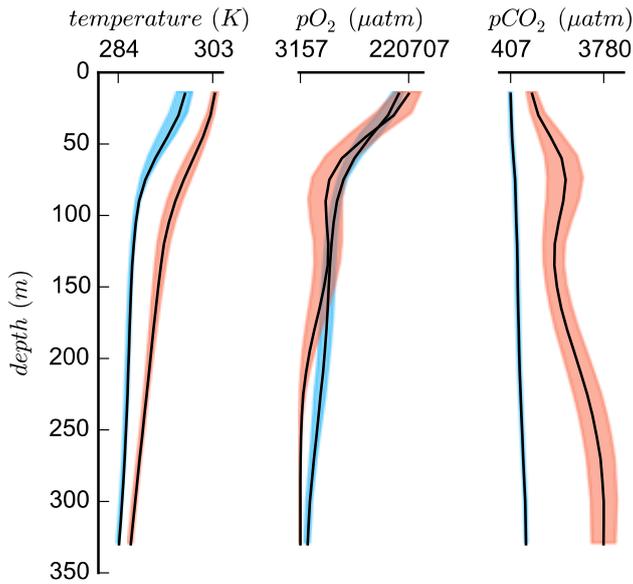


Figure 5: Projected multivariate perturbations in environmental variables for the study region in the Eastern Tropical Pacific oxygen minimum zone. Depth profiles of mean (black line) and  $\pm 1$  standard deviation (shaded) temperature,  $pO_2$  and  $pCO_2$  for the latitudes  $5^\circ\text{S}$  to  $5^\circ\text{N}$  and longitudes  $90^\circ\text{W}$  to  $105^\circ\text{W}$ . Current conditions (blue) are drawn from the World Ocean Atlas 2013 (temperature,  $pO_2$ ) and the Global Ocean Data Analysis Project ( $pCO_2$ ). Projected end-century conditions (orange) are drawn from the Meteorological Research Institute-Earth System Model Version 1 (MRI-ESM1) for the year 2100 under the RCP 8.5 scenario.

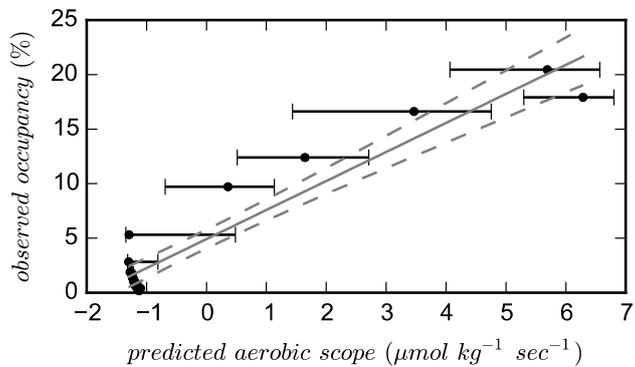


Figure 6: Estimated aerobic scope strongly predicts the observed distribution of *T. albacares*. The vertical distribution of *T. albacares* (% time spent at depth, bin size 15m) tracked by electronic tags in the study region (y axis, Schaefer et al., 2007) shows a strong correlation with aerobic scope (x axis) estimated by the model in the same region using environmental parameters drawn from the World Ocean Atlas 2013 and the Global Ocean Data Analysis Project. Error bars show the interquartile range of estimated aerobic scope (some error bars are too small to see). The grey line and dotted curves show the mean and 95% confidence bounds of a linear regression (N = 22 observations,  $R^2 = 0.91$ ,  $p < 0.0001$ ).

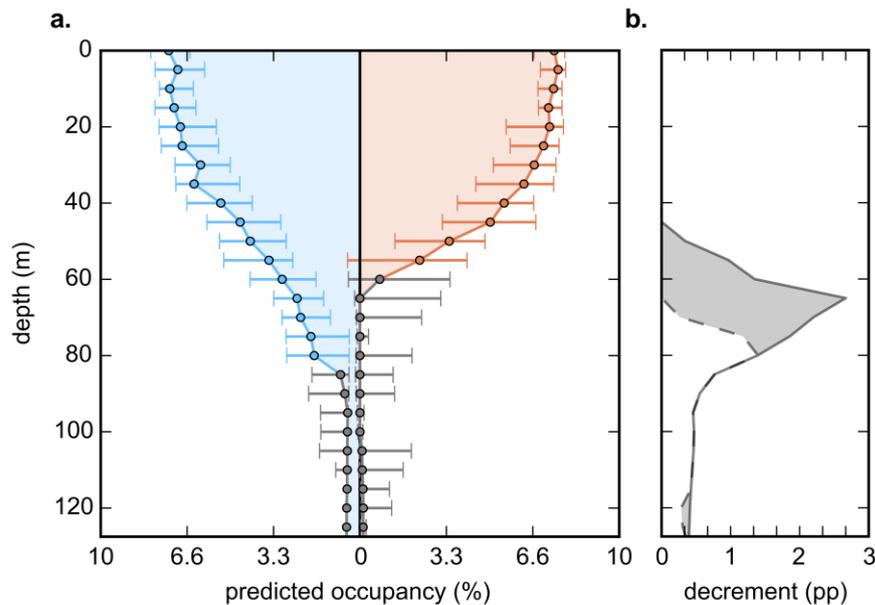


Figure 7: Climate-change driven reduction in vertical habitat for *T. albacares*. (a) Using the linear regression model shown in Figure 6, we predict the vertical habitat of *T. albacares* under the current (blue) and end-century (orange) conditions shown in Figure 5. Error bars show the interquartile range of estimated occupancy at a given depth obtained using a bootstrap sampling procedure of environmental and physiological variables ( $n = 1000$ ). Grey dots in both the current and future

projections show depths where estimated aerobic scope is below zero, indicating these depths are accessible only through temporary forays (such as for foraging or thermoregulation) and not long-term habitation. **(b)** The median decrement in occupancy between the current and future scenarios (solid line) is compared against the decrement (in percentage points) at the 5<sup>th</sup> percentile (dotted line) when estimates are obtained by bootstrap sampling of physiological parameters from their observed distributions ( $n = 1000$ ). The 5<sup>th</sup> percentile was chosen in this figure as a plausible value for the upper range of adaptive capacity. The distance between the median and 5<sup>th</sup> percentile decrement (grey shaded area) indicates the ability of currently existing physiological plasticity to mitigate changes in vertical habitat caused by climate change.

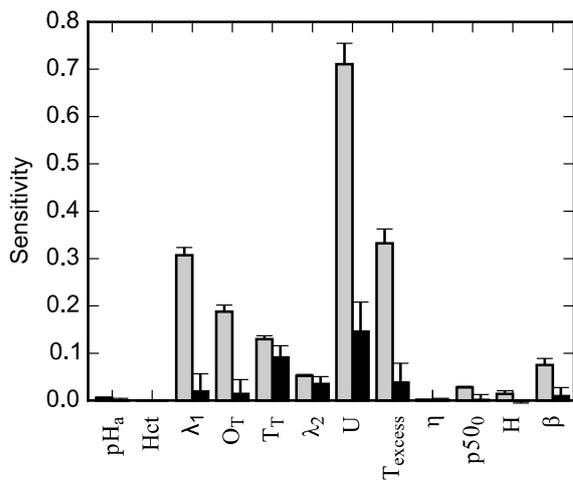


Figure 8: Sensitivity of estimated aerobic scope to input parameters. Bars show magnitude of Monte Carlo estimations of Sobol sensitivity indices. First order main effects (black bars) indicate the effect of uncertainty in the given parameter on the uncertainty of the estimated aerobic scope. Total effects (grey bars) show the sum of the main effects plus interactions between the given parameter and all other

parameters. Error bars indicate one standard deviation. Parameter labels correspond to those used in Table 3.

Table 3: List of input parameters for the generic marine teleost model

Parameter (units)	Symbol	Typical value (range)	Reference
<b>Blood Parameters</b>			
Extracellular Bohr Coefficient ( $\Delta \log p_{50} / \Delta \text{pH}$ )	$\beta$	-1.1 (-1.4, -0.2)	Brauner and Randall, 1998
Heat of Oxygenation (kcal/mole)	$H$	-15 (-17, -1)	Powers, 1980
Hemoglobin half-saturation constant ( $\mu\text{atm}$ )	$p_{50}$	15000 (2400, 26800) <sup>2</sup>	Bernier et al., 1996; Korsmeyer et al., 1997
Reference pH	$pH_0$	7.8	N/A
Reference Temperature (K)	$T_0$	293.15	N/A
Hemoglobin Oxygen Capacity (mol O <sub>2</sub> /g Hb)	$O_H$	$6.95 \times 10^{-5}$	Korsmeyer et al., 1997
Mean Cell Hemoglobin Concentration (g Hb/L)	$\bar{H}$	212 (125-381) <sup>1</sup>	Filho et al., 1992
Hematocrit (volume fraction)	$Hct$	0.41 (0.12, 0.76) <sup>1</sup>	Filho et al., 1992
Arterial pH	$pH_a$	7.8 (7.74, 8.08)	Burton, 1996
Oxygen Utilization (partial pressure fraction)	$U$	0.7 (0.25, 0.85)	Wood et al., 1979
Hill Coefficient	$\eta$	1.5 (0.6, 2.0)	Decker and Nadja, 2007
Minimum venous pO <sub>2</sub> ( $\mu\text{atm}$ )	$pO_{2v}$	75000 (50000, 100000) <sup>3</sup>	Korsmeyer et al., 1997; Kunzmann, 1991

### Cardiac Parameters

O <sub>2</sub> Threshold (mol O <sub>2</sub> /L blood)	$O_T$	0.00475 ±0.0005	Bushnell, 1988
O <sub>2</sub> Slope (L <sup>2</sup> blood/ kg body sec mol O <sub>2</sub> )	$\lambda_1$	330 (0, 340)	Bushnell, 1988
Temperature Threshold (K)	$T_T$	298.15 (276.15, >298.15) <sup>3</sup>	Blank et al., 2002
Temperature Slope (L blood/kg body sec K)	$\lambda_2$	0.044 (0, 0.05)	Blank et al., 2002
Reference Cardiac Output (L blood /kg body sec)	$Q_{max}$	0.001 (0.0004, 0.00196) <sup>2</sup>	Forster et al., 1991; Korsmeyer et al., 1997
Coronary oxygen supply (fraction)	$f$	0.3 (0, 1.0)	Farrell, 1996
Excess Muscle Temperature (K)	$T_{excess}$	0 (0, 20)	Brill et al., 1999

N/A: not applicable

<sup>1</sup> Excludes white-blooded Antarctic Notothenioids (*Channichthyidae*) that lack hemoglobin or red blood cells ( $\bar{H}, Hct = 0$ )

<sup>2</sup> Range constructed using hagfish and tunas as endpoints

<sup>3</sup> Range constructed using white-blooded Antarctic Notothenioids (*Channichthyidae*) and tunas as endpoints

Table 4: List of input parameters for *Thunnus albacares* model

Parameter (units)	Symbol	Values (mean ± SD)	Reference
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**Blood Parameters**

Extracellular Bohr Coefficient ( $\Delta \log p_{50} / \Delta \text{pH}$ )	$\beta$	$-0.87 \pm 0.15$	Bushnell, 1988
Heat of Oxygenation (kcal/mole)	$H$	$-1.0 \pm 2.0$	Bushnell, 1988
Hemoglobin half-saturation constant ( $\mu\text{atm}$ )	$p_{50}$	$26800 \pm 1320$	Korsmeyer et al., 1997
Reference pH	$\text{pH}_0$	7.893	Korsmeyer et al., 1997
Reference Temperature (K)	$T_0$	293.15	Korsmeyer et al., 1997
Hemoglobin Oxygen Capacity (mol O <sub>2</sub> /g Hb)	$O_H$	$6.95 \times 10^{-5}$	Korsmeyer et al., 1997
Mean Cell Hemoglobin Concentration (g Hb/L)	$\bar{H}$	370	Korsmeyer et al., 1997
Hematocrit (volume fraction)	$H_{ct}$	$0.283 \pm 0.01$	Korsmeyer et al., 1997
Arterial pH	$\text{pH}_a$	$7.73 \pm 0.01$	Korsmeyer et al., 1997
Oxygen Utilization (partial pressure fraction)	$U$	$0.45 \pm 0.06$	Korsmeyer et al., 1997
Hill Coefficient	$\eta$	$1.72 \pm 0.06$	Bushnell, 1988
Minimum venous pO <sub>2</sub> ( $\mu\text{atm}$ )	$pO_{2v}$	50000	Korsmeyer et al., 1997

**Cardiac Parameters**

O <sub>2</sub> Threshold ( $\mu\text{atm}$ )	$O_T$	$0.00475 \pm 0.0005$	Bushnell, 1988
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O <sub>2</sub> Slope (L blood/ kg body sec $\mu$ atm O <sub>2</sub> )	$\lambda_1$	337.7 $\pm$ 120.9	Bushnell, 1988
Temperature Threshold (K)	$T_T$	298.15 $\pm$ 2.0	Blank et al., 2002
Temperature Slope (L blood/kg body sec K)	$\lambda_2$	0.044 $\pm$ 0.007	Blank et al., 2002
Reference Cardiac Output (L blood /kg body sec)	$Q_{max}$	0.0196	Korsmeyer et al., 1997
Coronary oxygen supply (fraction)	$f$	1.0	Farrell, 1996
Excess Muscle Temperature (K)	$T_{excess}$	5.0 $\pm$ 3.0	Brill et al., 1999

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\*Values given without standard deviations were treated as constants



### 3. Synergistic Impact of Simultaneous Climate-Related Parameters on Aerobic Scope in a Model Teleost, *Oreochromis mossambicus*

#### 3.1. Abstract

Multiple-stressor perturbations, where simultaneous environmental changes act in concert to disrupt physiological functioning, will characterize much of the impacts of climate change on marine teleosts. Three stressors that are expected to become especially prevalent by the end of the century are CO<sub>2</sub>-induced ocean acidification, surface-ocean warming, and deoxygenation. Using the euryhaline Mozambique tilapia (*Oreochromis mossambicus*) as a model teleost, we explore the effects of these stressors both in isolation and in combination on aerobic scope, a key physiological indicator of organism fitness, and blood-based stress indicators including glucose, base excess, pCO<sub>2</sub>, and venous oxygen saturation. Our results show that while neither increased temperature nor pCO<sub>2</sub> alone are sufficient to cause substantial declines in the aerobic scope of *O. mossambicus*, these same stressors have a significant effect when combined with simultaneous deoxygenation. A combination of stressors was also more effective than single stressors at eliciting responses in most blood stress parameters. In particular, both deoxygenation and acidification interact synergistically with warming, suggesting that the total impact of climate-change related stresses may be substantially higher than what is indicated by single-variable studies.

#### 3.2. Introduction

Understanding the cumulative impact of multiple simultaneous stressors on organism functioning and population distribution remains one of the most pressing

1 challenges for conservation biology (Crain et al., 2008). This issue is particularly  
2 important in the marine context, where a variety of anthropogenic stressors including  
3 pollutants, habitat alteration, and nutrient input affect organisms across large areas of  
4 the ocean (Adams, 2005). Whereas it is often assumed that the impact of multiple  
5 stressors can be approximated as a simple additive accumulation of the impacts  
6 associated with single stressors, recent studies have revealed the prevalence of  
7 complex synergistic or compensatory interactions (Crain et al., 2008). Further  
8 experimental data on organism responses to simultaneous stressors is necessary to  
9 contribute to a more general understanding of these effects.

10       Anthropogenic climate change will become an increasingly important driver  
11 of simultaneous environmental perturbation of the marine environment during this  
12 century. Ocean acidification (and attendant high dissolved CO<sub>2</sub>), ocean  
13 deoxygenation, and surface ocean warming are three of the most important physical  
14 changes that are likely to co-occur in much of the ocean (Bopp et al., 2013). All three  
15 are known to strongly affect the oxygen transport system of marine teleosts, which in  
16 turn can have profound effects on metabolism and fitness (Farrell et al., 2008; Pörtner,  
17 2010; Pörtner et al., 2010). For example, recent studies portend declines in migratory  
18 success (Farrell et al., 2008), increased mortality (Munday et al., 2009), and changes  
19 in community composition (Nilsson et al., 2009) among various marine teleosts as a  
20 result of climate-driven changes in these environmental parameters.

21       Whereas the theoretical and empirical basis for interactions between  
22 temperature and oxygen impacts on water breathing organisms has been fairly well  
23 established (Cooke et al., 2008; Lapointe et al., 2014; Pörtner, 2010, 2001), there is  
24 much less information available on interactions with hypercapnia (elevated CO<sub>2</sub>).  
25 Munday et al. (2009) found that although both warming and hypercapnia affect

1 aerobic scope in two species of coral reef fish, interactions between the two stressors  
2 were strictly additive. By contrast, Cruz-Neto and Steffensen (1997) revealed  
3 increased sensitivity of routine metabolism to hypoxia (low oxygen) in the eel  
4 *Anguilla anguilla* when exposed to simultaneous hypercapnia. Similarly, Ginneken et  
5 al. (1996) examined the effects of simultaneous hypoxia and acidification (where the  
6 pH-perturbation was achieved via acid titration, not CO<sub>2</sub> addition) on the tilapia  
7 *Oreochromis mossambicus* and found synergistic effects on intracellular pH and  
8 phosphocreatine levels, but not on routine metabolism. Finally, Rosa and Seibel  
9 (2008) revealed synergistic interactions between warming, hypoxia, and hypercapnia  
10 on routine metabolic rates in the jumbo squid *Dosidicus gigas*. To our knowledge, no  
11 study to date has examined interactions between temperature, oxygen, and CO<sub>2</sub>  
12 simultaneously on a marine teleost.

13 An important distinction that deserves discussion here is the disparate impact  
14 of pH and pCO<sub>2</sub> on marine organisms. Marine teleosts have a robust ability to  
15 regulate arterial pH and are therefore likely to be isolated from the effects of  
16 environmental pH perturbations (Ishimatsu et al., 2004; Melzner et al., 2009). One  
17 potential exception is that increased bicarbonate uptake, which is required to maintain  
18 high arterial pH in an acidified seawater environment, can lead to notable pathologies  
19 in teleosts including hypercalcification (Melzner et al., 2012). Furthermore, temporary  
20 blood-pH perturbations, if they propagate through the red blood cell membrane, may  
21 have a direct effect on blood-oxygen binding due to the binding of free protons (H<sup>+</sup>)  
22 to an allosteric site on the hemoglobin molecule (Jensen, 2004).

23 By contrast, teleost fishes have a low capacity to ameliorate pCO<sub>2</sub>  
24 perturbations. While ocean acidification and environmental hypercapnia often go  
25 hand in hand, the ability of the pCO<sub>2</sub> molecule to diffuse across cell membranes

1 means that hypercapnia is often the driving factor for systemic effects. While the  
2 specific mechanisms for these effects are not well understood, experimental studies  
3 have clearly shown that acidification accompanied by environmental hypercapnia  
4 shows substantially more lethality in a variety of teleosts as compared to acidification  
5 alone (K.-S. Lee et al., 2003).  $p\text{CO}_2$  regulation in teleosts is particularly coarse due to  
6 the unavailability of a number of strategies used by other taxa. Elasmobranches, for  
7 example, carry carbonic anhydrase (an enzyme that catalyzes the conversion of  
8 carbonic acid to aqueous  $\text{CO}_2$ ) in the blood plasma to allow  $\text{CO}_2$  produced during  
9 metabolic processes to be offloaded efficiently during passage through the gills.  
10 Teleosts, by contrast, only carry carbonic anhydrase in the red blood cell, with the  
11 result that the rate of  $\text{CO}_2$  offloading is primarily limited by diffusion through the red  
12 blood cell membrane (Evans et al., 2003). Similarly, while air-breathers are able to  
13 raise ventilatory rates to boost  $\text{CO}_2$  loss through the gills, high ventilatory rates in  
14 water-breathers are limited by the viscosity of the medium (Toews et al., 1983). The  
15 net result is that even as arterial pH typically remains stable after an adaptive period  
16 of 12-24 hours even under extreme pH perturbation,  $p\text{CO}_2$  homeostasis is never  
17 achieved in a hypercapnic environment (Ishimatsu et al., 2004). This prolonged  $p\text{CO}_2$   
18 perturbation can lead not only to metabolic pathologies but also to a wide range of  
19 potential issues including neuronal dysfunction (Nilsson et al., 2012).

20 Our study examines the impact of temperature, oxygen, and  $p\text{CO}_2$   
21 perturbations reflective of the upper range of those expected under end-century  
22 climate change on the Mozambique tilapia, *Oreochromis mossambicus*. This species  
23 has a native range that is limited to the eastward flowing rivers of central and southern  
24 Africa (Russell et al., 2012) but is currently found in a wide variety of estuarine and  
25 coastal marine environments across the tropics and subtropics (Canonico et al., 2005).

1 It is notable not only for its high salinity tolerance (Whittfield and Blaber, 1979) but  
2 also for its extreme tolerance of hypoxia (Lague et al., 2012), high temperature  
3 (Philippart and Ruwet, 1982), and low pH (Krishna Murthy et al., 1984). Despite this  
4 eurytolerance, previous studies have revealed sublethal effects including decreased  
5 immune functioning, prolonged metabolic impairment, and intracellular pH declines  
6 resulting from environmental perturbations that fall well inside the lethal limits  
7 (Evans et al., 2003; Van Dijk et al., 1993; Van Ginneken et al., 1999, 1996; Waarde et  
8 al., 1990).

9 We measured the response of this species to environmental perturbation at two  
10 levels of organization: blood-borne indicators (blood glucose, base excess, pCO<sub>2</sub>, and  
11 hemoglobin oxygen saturation) and whole-animal performance (resting and maximum  
12 oxygen consumption rates). While blood glucose levels are a commonly used stress  
13 indicator (Van Ham et al., 2003), base excess, pCO<sub>2</sub>, and hemoglobin oxygen  
14 saturation rates track various mechanisms by which changing environmental  
15 parameters could impact whole-organism processes. Base excess for example is an  
16 indicator of metabolic (i.e. non-respiratory) acid-base disturbance (Siggaard-Andersen  
17 and Fogh-Andersen, 1995). Stressors that reduce the capacity for aerobic metabolism,  
18 including high temperature or low oxygen, can induce metabolic acidosis by  
19 encouraging the production of acid species from anaerobic metabolism (Pörtner,  
20 2002). Blood pCO<sub>2</sub>, on the other hand, is an indicator of respiratory acid-base  
21 disturbance (Irizarry and Reiss, 2009; Perry and Gilmour, 2002). Increased rates of  
22 CO<sub>2</sub> transfer between the blood and environment through the gills (such as during  
23 environmental hypercapnia) can alter pCO<sub>2</sub> while leaving base excess relatively  
24 unchanged. Both metabolic and respiratory acidoses can alter hemoglobin oxygen  
25 affinity through the Bohr effect (Nikinmaa, 2001), while high levels of blood pCO<sub>2</sub>

1 can have more systemic effects including altered neuronal functioning (Chung et al.,  
2 2014; Nilsson et al., 2012). Finally, changes in acid-base balance, oxygen utilization,  
3 or environmental hypoxia can alter venous hemoglobin oxygen saturation rates, which  
4 in turn can affect the capacity for aerobic metabolism (Pörtner et al., 2010).

5 Many of these perturbations are likely to be reflected in changes to a quantity  
6 known as aerobic scope, which is calculated as the difference between maximum and  
7 resting aerobic metabolic rates (Fry, 1971). Various investigators have used aerobic  
8 scope as an integrative measure of environmental impacts on whole-organism  
9 performance, and numerous studies have linked aerobic scope to the ability to  
10 perform activities including growth, sustained swimming, feeding, and territorial  
11 aggression in a range of active and sedentary species (Claireaux et al., 2000;  
12 Claireaux and Lefrançois, 2007; Khan et al., 2014; Killen et al., 2014; Mallekh and  
13 Lagardère, 2002; Reidy et al., 2000). While there has been some recent controversy as  
14 to the extent to which aerobic scope can be considered a master limiting variable on  
15 all physiological performance (Clark et al., 2013), it remains a useful measure of the  
16 total metabolic energy available for activities including growth and reproductive  
17 processes (Farrell, 2013).

18 We used aerobic scope as an integrative measure of the impact of  
19 simultaneous climate-driven stressors on *O. mossambicus*, and supplemented this  
20 metric with the measurement of blood parameters. Our aim was to determine whether  
21 temperature, oxygen, and CO<sub>2</sub> impacts interact in an additive, synergistic, or  
22 compensatory manner for the magnitudes of perturbation that can be expected to  
23 result from end-century climate change.

### 3.3. Methods

#### 3.3.1. Study Animals

Male seawater-spawned and reared (i.e. freshwater naïve) *Oreochromis mossambicus* (N = 28) of mass 40–80g (mean =  $59.9 \pm 6.7$ g) were selected from outdoor tanks maintained at the Hawaii Institute of Marine Biology, University of Hawaii. Although *O. mossambicus* is an estuarine species that is capable of surviving in both seawater and freshwater environments, all experiments were conducted in seawater. Fish were kept under natural photoperiod and ambient seawater temperatures of approximately 24°C.

#### 3.3.2. Custom Climate-Controlled Respirometer Construction

Acrylic Blazka-type swim-tunnel respirometers of 7L internal volume were constructed for this study, following the design employed by Febry and Lutz (1987). Respirometers were submerged in heated water baths to maintain a constant temperature. Intermittent water exchange (i.e. intermittent flow respirometer; Steffensen, 1989) with a UV-sterilized 100L reservoir allowed us to manipulate the temperature, oxygen, and pCO<sub>2</sub> in the respirometer. A Neptune Systems controller combined with custom software automated the release of compressed O<sub>2</sub>, N<sub>2</sub>, and CO<sub>2</sub> through a gas-exchange column to change gas partial pressures. Two 400W thermostat-controlled submersible heaters were used to set water temperatures in the reservoir and optical oxygen, pH, and temperature probes (Thermo Fisher Scientific, Waltham, MA) were used to monitor the conditions both inside the respirometer and the reservoir. Reservoir pCO<sub>2</sub> was determined using a non-dispersive infrared (NDIR) CO<sub>2</sub> monitor (CO2Meter Inc., Ormond Beach, FL) in the confined headspace at the top of the reservoir.

1           3.3.3. *Selection of Treatment Conditions*

2           Experimental conditions within the respirometer were chosen to represent the  
3 upper range of plausible oxygen, temperature, and pCO<sub>2</sub> perturbations that may  
4 become prevalent in the oceans due to end-century climate change. Control treatments  
5 were chosen based on ambient environmental conditions. The precise target values for  
6 each treatment parameter are listed in Table 1. Justifications for these targets are as  
7 follows:

8           Temperature: End-century global sea surface warming projected by Coupled  
9 Model Intercomparison Project 5 (CMIP5) models under Representative  
10 Concentration Pathway (RCP) 8.5 emissions scenarios (similar to “business as usual”  
11 emissions projections) range from 2-4°C (Bopp et al., 2013). We elected for the  
12 higher range of temperature change: an increase of 4°C (control temperature 25°C,  
13 high temperature stress 29°C). This reflects the fact that although RCP 8.5 was  
14 originally intended as a high-emissions scenario, current trends in global emissions  
15 match or exceed this trajectory (Sanford et al., 2014).

16           Oxygen: Shoaling of oxygen minimum layers in the open ocean as well as  
17 intensification of coastal hypoxic zones are expected to result from climate-change  
18 induced surface-ocean stratification and the intensification of coastal upwelling  
19 (Bakun, 1990; Diaz and Rosenberg, 2008; Gilly et al., 2013; Stramma et al., 2011).  
20 Though levels of deoxygenation are likely to vary wildly by depth and location,  
21 movement of hypoxic waters into previously well-oxygenated habitats could cause  
22 substantial declines in oxygen concentration (>50%) in certain sections of both  
23 coastal and open waters (Diaz and Rosenberg, 2008; Hofmann and Schellnhuber,  
24 2009; Justin et al., 1996; Oschlies et al., 2008). We chose approximately 50% oxygen  
25 saturation (100 μmol O<sub>2</sub>/kg) as our low-oxygen treatment. Although this is a fairly

1 drastic decline, we believe it is a useful point of comparison for our study and reflects  
2 possible contributions of coastal eutrophication in addition to climate change driven  
3 deoxygenation.

4 CO<sub>2</sub>: pCO<sub>2</sub> levels equilibrated to 2x pre-industrial pCO<sub>2</sub> in oxygen minimum  
5 zone waters have been predicted to rise in many cases to over 2,000 μatm (Brewer  
6 and Peltzer, 2009). Similarly, CMIP5 model results suggest end-century pCO<sub>2</sub>  
7 increases in some open-ocean oxygen minimum zones of a similar degree (see  
8 Chapter I). Analogously large pCO<sub>2</sub> values in hypoxic coastal zones are also expected  
9 to be exacerbated by climate change in conjunction with anthropogenic eutrophication  
10 (Cai et al., 2011; Melzner et al., 2012). We chose this range of pCO<sub>2</sub> (2,000 μatm) for  
11 our experiments.

#### 12 3.3.4. *Respirometry Protocol*

13 A standard respirometry protocol (Reidy et al., 2000) was used to measure  
14 maximum (active) and minimum metabolic rates. Although minimum metabolic rates  
15 are often referred to as standard rates, because our study animals were juveniles our  
16 minimum metabolic rate measurements likely included some expenditure for growth.  
17 Therefore, we use the terminology “minimum metabolism” to distinguish from the  
18 strict definition of standard metabolism, which does not include a growth term.

19 Tilapia were transferred to the respirometer and maintained at rest for twenty-  
20 four hours, with a further overnight period of metabolic measurement at rest to  
21 determine the minimum metabolism. Current speeds within the respirometer were  
22 then slowly increased to 5 body-lengths per second for up to two hours until the fish  
23 tired, which was judged to occur when the fish were unable to maintain position in the  
24 current and fell to a retaining screen at the back of the swim tunnel. Flow within the  
25 swim tunnel was then briefly reversed and restarted to confirm that exhaustion had

1 occurred. The rate of oxygen consumption during the final measurement period prior  
2 to exhaustion was taken to be the maximal active metabolic rate, and was calculated  
3 according to the following formula:

$$\dot{M}_{O_2} = \left( \frac{\Delta O_{2w}}{\Delta T} \right) \times V \quad \text{Eq. 1.}$$

4  
5 where  $\Delta O_{2w}$  is the change in oxygen concentration during the measurement period,  
6  $\Delta T$  is the duration of the measurement period, and  $V$  is the volume of the  
7 respirometer. This protocol was repeated for each fish for all treatments in a  
8 randomized sequence. Measured metabolic rates were corrected for body mass scaling  
9 effects using a standard scaling exponent of 0.8 (Ros et al., 2006), which is consistent  
10 with the seawater mass scaling exponent measured in *O. mossambicus* by Febry and  
11 Lutz (1987) and Yamamoto (1992). Bacterial oxygen consumption in the respirometer  
12 system measured at the beginning and end of the experimental period by measuring  
13 oxygen uptake in the empty respirometer showed negligible bacterial consumption  
14 (possibly due to continuous UV-sterilization of the reservoir) and was therefore  
15 assumed to be zero.

### 16 3.3.5. Blood Sampling

17 Fish were anesthetized with 0.3 ml/L 2-phenoxyethanol and non-lethal blood  
18 draws with heparinized syringes were conducted from the caudal vein. Blood was  
19 sampled at the conclusion of the exercise trials and compared against resting controls.  
20 Analysis was performed on an i-STAT PCA with the CG4+ cartridge (Abbott  
21 Laboratories, Abbott Park, Illinois). All measurements were conducted with blood  
22 warmed to 37°C.

### 23 3.3.6. Experimental Design



1 while base excess (BE) ranged from  $-22 \pm 1$  mmol/L (resting) to  $-29 \pm 1$  mmol/L  
2 (high-temperature) (Figure 1, a and b). Due to limits in the measurement range of our  
3 instrument, base excess readings of  $-30$  (9 out of 28 measurements) could not be  
4 distinguished from values below  $-30$ . Blood pCO<sub>2</sub> ranged from  $22800 \pm 5400$   $\mu$ atm  
5 (resting) to  $11000 \pm 3100$   $\mu$ atm (high temperature), and venous oxygen saturation  
6 (svO<sub>2</sub>) ranged from  $54 \pm 11\%$  (resting) to  $17 \pm 8\%$  (high-temperature, high CO<sub>2</sub>)  
7 (Figure 1, c and d).

8           Comparison of blood parameters between aggregated treatment groups using  
9 Tukey's HSD showed that blood glucose was significantly elevated under control and  
10 single-stressor treatments relative to resting conditions, and under multiple-stressor  
11 treatments relative to control or single-stressor conditions (Figure 2a). Similarly, there  
12 was a significant decline in base excess observed between resting and exercise  
13 treatments (Figure 2b), as well as a decline in blood pCO<sub>2</sub> under normocapnic  
14 exercise compared to resting or hypercapnic treatments (Figure 2c). Finally, venous  
15 oxygen saturation was highest at rest, and significantly lower under both hypoxic and  
16 normoxic exercise (Figure 2d).

17           The mean minimum overnight metabolic rate achieved under control  
18 conditions was  $2.0 \pm 0.2$  mmol O<sub>2</sub> kg<sup>-0.8</sup> min<sup>-1</sup> and was  $2.86 \pm 0.13$  mmol O<sub>2</sub> kg<sup>-0.8</sup>  
19 min<sup>-1</sup> under elevated temperatures (treatments *Control* and *+T*; Figure 3). Active  
20 metabolic rates were  $10.28 \pm 1.08$  O<sub>2</sub> kg<sup>-0.8</sup> min<sup>-1</sup> under control conditions and  $10.91 \pm$   
21  $1.49$  mmol O<sub>2</sub> kg<sup>-0.8</sup> min<sup>-1</sup> under elevated temperatures (treatments *Control* and *+T*;  
22 Figure 3). Mean aerobic scope was highest under control conditions at  $8.27 \pm 0.97$   
23 mmol O<sub>2</sub> kg<sup>-0.8</sup> min<sup>-1</sup> and declined to  $3.51 \pm 0.77$  mmol O<sub>2</sub> kg<sup>-0.8</sup> min<sup>-1</sup> under  
24 maximally stressful, high-temperature low-oxygen high CO<sub>2</sub> conditions (treatments  
25 *Control* and *+CO<sub>2</sub>/O<sub>2</sub>/T*; Figure 4).

1 Comparison of aerobic scope between treatments using Tukey's HSD revealed  
2 statistically significant reductions in aerobic scope between the control treatment and  
3 all multi-stressor treatments (+CO<sub>2</sub>/T, +CO<sub>2</sub>/O<sub>2</sub>, +CO<sub>2</sub>/O<sub>2</sub>/T) at p < 0.05, but not  
4 between the control and single-stressor treatments (+CO<sub>2</sub>, +T; Figure 4). All  
5 comparisons of aerobic scope treatments are reported in Table 3.

6 The mean decrement in aerobic scope resulting from 4°C warming was 0.22  
7 mmol O<sub>2</sub> kg<sup>-0.8</sup> hr<sup>-1</sup> in the single-stressor case (comparison of treatments *Control* and  
8 +T), 0.64 mmol O<sub>2</sub> kg<sup>-0.8</sup> hr<sup>-1</sup> in the two-stressor case (comparison of treatments  
9 +CO<sub>2</sub> and +CO<sub>2</sub>/T), and 0.93 mmol O<sub>2</sub> kg<sup>-0.8</sup> hr<sup>-1</sup> in the three-stressor case  
10 (comparison of treatments +CO<sub>2</sub>/O<sub>2</sub> and +CO<sub>2</sub>/O<sub>2</sub>/T). The change in the magnitude  
11 of the temperature effect for the single, double, and three-stressor case was quantified  
12 for each individual fish using linear regressions, and a single-sample t-test of the  
13 resulting distribution of slopes was found to be significantly greater than zero (N = 7,  
14 p = 0.024; Figure 5).

## 15 **3.5. Discussion**

### 16 *3.5.1. Overview*

17 Our results indicate a substantial disparity between individual climate change  
18 stressors and multiple simultaneous stressors on the physiological performance of *O.*  
19 *mossambicus*. Aerobic scope values declined significantly in all multiple-stressor  
20 trials compared to control conditions but did not decline or showed non-significant  
21 declines under both high temperature and high pCO<sub>2</sub> challenge (Figure 4). While high  
22 aerobic scope was maintained due to compensatory changes in minimum and active  
23 metabolism in the single-stressor trials (i.e. high minimum metabolism in the +T trials  
24 were compensated by a non-significant rise in active metabolism, low active

1 metabolism in the +CO<sub>2</sub> trials by a decline in minimum metabolism), this  
2 compensation appeared to break down under multiple stressors (Figure 3). The ability  
3 to resist aerobic scope impairment in response to 4°C temperature warming in  
4 particular was significantly degraded as the number of stressors increased—the  
5 magnitude of the change is exemplified by an almost 4-fold rise in the effect of  
6 temperature in the presence of hypercapnic and hypoxic stress (the difference between  
7 +CO<sub>2</sub>/O<sub>2</sub> and +CO<sub>2</sub>/O<sub>2</sub>/T trials) as compared to the single-stressor case (Control  
8 versus +T) (Figure 5).

9         Notably, this synergistic interaction between hypercapnic/hypoxic stress and  
10 temperature was undetectable when minimum or active metabolic rates were  
11 compared instead of aerobic scope. This is important in view of previous work using  
12 other metabolic indicators such as routine metabolic rate, which failed to find  
13 significant interactions between multiple stressors (pH and hypoxia) on *O.*  
14 *mossambicus* (Van Ginneken et al., 1996). There is no straightforward way to infer  
15 aerobic capacity from minimum, routine, or active metabolic rates, and analyses of  
16 routine metabolism in particular are complicated by the fact that factors such as  
17 behavior or endogenous hormones can have a large impact on the measured  
18 metabolism (see Ros et al., 2004 for an example in *O. mossambicus*). Although  
19 routine and minimum metabolic rate measurements are a popular metric for  
20 evaluating the impact of environmental stressors on marine teleosts, our results  
21 emphasize the importance of aerobic scope as potentially a more sensitive metric of  
22 environmental stress.

### 23         3.5.2. *Comparison of Metabolic Rates With Previous Studies*

24         The metabolic rates we measured are broadly consistent with those obtained  
25 from previous studies. The minimum metabolic rate we measured under control

1 conditions ( $2.0 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$  at  $25^\circ\text{C}$ ) is equivalent to the standard metabolic  
2 rate reported by Zikos et al. (2014) ( $2.4 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$  at  $25^\circ\text{C}$ ) and slightly  
3 above those measured using *O. mossambicus* X *hornorum* hybrids by Febry and Lutz  
4 (1987) ( $1.8 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$  at  $28^\circ\text{C}$ ). Measurements of maximum metabolic rate  
5 for this species, on the other hand, are scarce, and a “maximum routine metabolic  
6 rate” reported by Ros et al. (2004) ( $4.4 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$  at  $26^\circ\text{C}$ ) should be  
7 considered to be an underestimate of maximum metabolism because it was measured  
8 during spontaneous activity rather than during a swimming or exhaustion trial.  
9 Similarly, while the “recovery metabolism” reported by Schnell and Seebacher (2008)  
10 was taken post-exhaustion, the fact that their reported recovery metabolic rates ( $1.9$   
11  $\text{mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$  at  $26^\circ\text{C}$ ) often fall below what they report to be minimum  
12 metabolic rates strongly suggests that their measurements missed the peak excess  
13 post-exercise oxygen consumption (EPOC) and are therefore substantial  
14 underestimates of peak recovery metabolism. A swim-tunnel respirometry trial of *O.*  
15 *mossambicus* X *hornorum* hybrids by McKenzie et al. (2003) did generate  
16 substantially higher maximum metabolic rates ( $19.2 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$  at  $25^\circ\text{C}$ ) than  
17 in our study, however it is difficult to determine whether this higher rate was due to  
18 differences in the fish stock (hybrid tilapia versus *O. mossambicus*), water parameters  
19 (McKenzie et al. conducted their experiments in freshwater), or other factors. The  
20 unusually high minimum metabolic rates reported in the McKenzie et al. study ( $3.2$   
21  $\text{mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$  at  $25^\circ\text{C}$ ) do suggest however that their fish may have had a more  
22 active aerobic metabolism than those used in ours. Moreover, an earlier study (also in  
23 freshwater) on the same *mossambicus* X *hornorum* hybrids by Febry and Lutz (1987)  
24 measured active metabolic rates between  $8\text{--}13 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}$  at  $28^\circ\text{C}$ , which  
25 closely matches our results.

1           Recent studies have shown that some teleosts may achieve higher metabolic  
2 rates when exhausted under postprandial conditions than in a fasted state (Clark et al.,  
3 2013; Fu et al., 2009), although these metabolic rates may converge under aerobic  
4 stress (Jourdan-Pineau et al., 2010). Unfortunately, no measurements of postprandial  
5 maximum metabolic rate have been reported in *O. mossambicus* to date, and therefore  
6 it is untenable to compare our measurements with those obtained from fed fish.  
7 Measurements do exist in similarly sized (40–80 g) postprandial males of the  
8 congeneric Nile tilapia (*Oreochromis niloticus*) however (Mamun et al., 2015), and  
9 these suggest that postprandial maximum metabolic rates ( $9 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ min}^{-1}$ ) are  
10 comparable to those obtained in our study.

### 11           3.5.3. *Blood-Borne Stress Indicators*

12           Our blood analysis bolsters our metabolic rate analysis by revealing additional  
13 evidence of physiological stress under increasing environmental challenge. Elevated  
14 blood glucose levels suggest significantly higher stress under multiple-stressor as  
15 opposed to single-stressor and control treatments (Figure 2). The magnitude of the  
16 change in blood glucose between resting and maximal stress conditions is consistent  
17 with previous research on the effect of exercise and environmental stressors on teleost  
18 glucose levels including interactions between temperature and exercise in turbot  
19 (*Scophthalmus maximus*; Van Ham et al., 2003), temperature and salinity in *O.*  
20 *mossambicus* (Fiess et al., 2007; Zaragoza et al., 2008) and dissolved oxygen (Evans  
21 et al., 2003). Significantly elevated blood pCO<sub>2</sub> under hypercapnic as opposed to  
22 normocapnic exercise indicates a respiratory acidosis caused by increased CO<sub>2</sub>  
23 diffusion from the environment, while significantly lower pCO<sub>2</sub> under normocapnic  
24 exercise as opposed to resting conditions can be attributed to respiratory alkalosis  
25 caused by increased ventilatory frequency. The high blood pCO<sub>2</sub> observed under

1 hypercapnia is also likely to have been driven in part by the metabolic acidosis caused  
2 by exhaustive exercise, which is indicated by more negative base excess under all  
3 exercise treatments. It is important to point out that the common measurement  
4 temperature of 37°C (regulated by the i-STAT analyzer) means that our dataset is not  
5 likely to be reflective of *in vivo* conditions. Previous measurements of pCO<sub>2</sub> in  
6 freshwater-reared *O. mossambicus* for example show partial pressures of  
7 approximately 4,000 μatm (although the text does not specify whether measurements  
8 were performed in the dorsal or ventral aorta; Smit et al., 1981). However, studies  
9 across a variety of teleosts have indicated that relative differences in measured values  
10 are meaningful even under this elevated temperature (see Stoot et al., 2014, for a  
11 review).

#### 12 3.5.4. *Implications for Climate Change Impacts*

13 Our experimental treatments were limited to acute (<36 hours) stressors,  
14 whereas climate change stressors are chronic, multi-generational perturbations that  
15 allow affected species to mitigate the impacts on physiological functioning through  
16 gene regulatory responses, epigenetic inheritance, or genetic evolution (Somero,  
17 2011). Studies on *O. mossambicus* have shown substantial ability to acclimate to large  
18 (8°C) temperature shifts over a period of several weeks (Schnell and Seebacher,  
19 2008). Conversely, declining *O. mossambicus* populations in the Salton Sea in  
20 response to incremental salinity increases (approx. 0.25ppt/year) over a forty-year  
21 period attest to the limits of these adaptive processes (Caskey et al., 2007; Hurlbert et  
22 al., 2007). Similarly, Ginneken et al. (1996) reported that *O. mossambicus* exhibited  
23 no adaptation response to 6-weeks of environmental acidosis. Unfortunately,  
24 logistical limitations prevented us from evaluating the ability of our animals to  
25 acclimate to the treatment conditions we imposed in this study.

1           Nevertheless, it is possible to draw some general conclusions from our results  
2 for predicting teleost responses to climate change. The significant impact of  
3 simultaneous stressors on our study animals was particularly surprising in the context  
4 of the resilience of this species to single-stressor perturbations, both in our  
5 experiments and in the field. The thermal tolerance zone of *O. mossambicus* in the  
6 hypersaline environment of the Salton Sea is reported to range from 10–35°C (Costa-  
7 Pierce and Doyle, 1997) for example, with some studies reporting lethal limits as low  
8 as 8°C (Stauffer, 1986) and as high as 42°C (Philippart and Ruwet, 1982).  
9 Hypercapnia and hypoxia tolerance is likewise very high, with acidic lethal limits  
10 (measured in acid-titration experiments, not CO<sub>2</sub>-addition) in freshwater as low as pH  
11 3.4 (Krishna Murthy et al., 1984), and a survivable oxygen limit down to 20%  
12 saturation (Rappaport et al., 1976). The existence of measurable sublethal effects in  
13 our experiments on such a putatively “hardy” organism portends similar synergistic  
14 impacts in other marine teleosts. CMIP5 simulations of end-century climate change  
15 underscore the correlation between these climate-related parameters, with projections  
16 under an RCP 8.5 emissions scenario showing a 0.127 unit decline in global mean  
17 surface pH and a 1.3% decline in global oxygen content for every °C increase in sea  
18 surface temperature (Bopp et al., 2013).

### 19           3.5.5. *Conclusions and Directions for Further Study*

20           The existing literature has made large strides in understanding the impact of  
21 one or two stressors in combination (Claireaux et al., 2000; Ginneken et al., 1996;  
22 Ishimatsu et al., 2008; Lefrançois and Claireaux, 2003; Munday et al., 2009) as well  
23 as on integrating species-level changes to an ecosystem context (Lehodey et al., 2008;  
24 Link et al., 2010). While these are important endeavors, our results show that the  
25 impact of temperature, oxygen, and CO<sub>2</sub> together may exceed even that predicted by

1 these studies. More experiments on a finer range of environmental conditions and a  
2 variety of teleost species will further contribute to our understanding. In particular, we  
3 hope that our results may serve as a baseline against which the effect of chronic  
4 stressors (i.e. long-term acclimation or the rearing of multiple generations under  
5 environmental challenge) could be compared to evaluate the capacity for adaptation  
6 driven by short-term physiological plasticity and epigenetic change.

7

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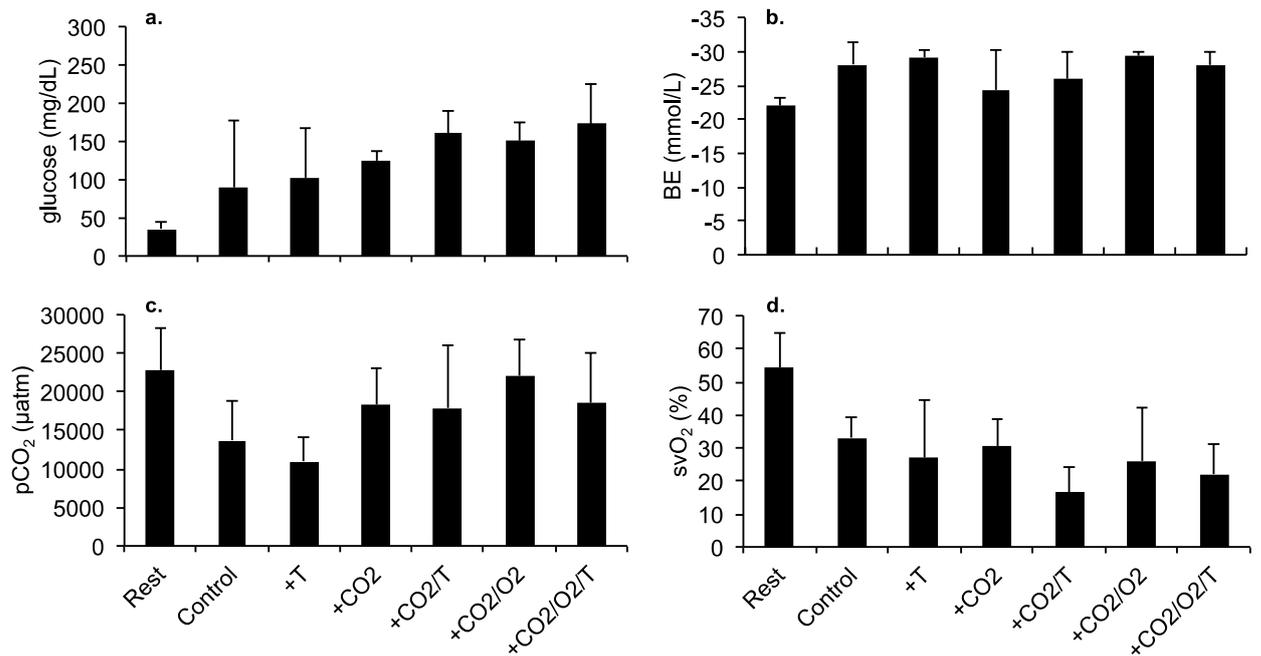
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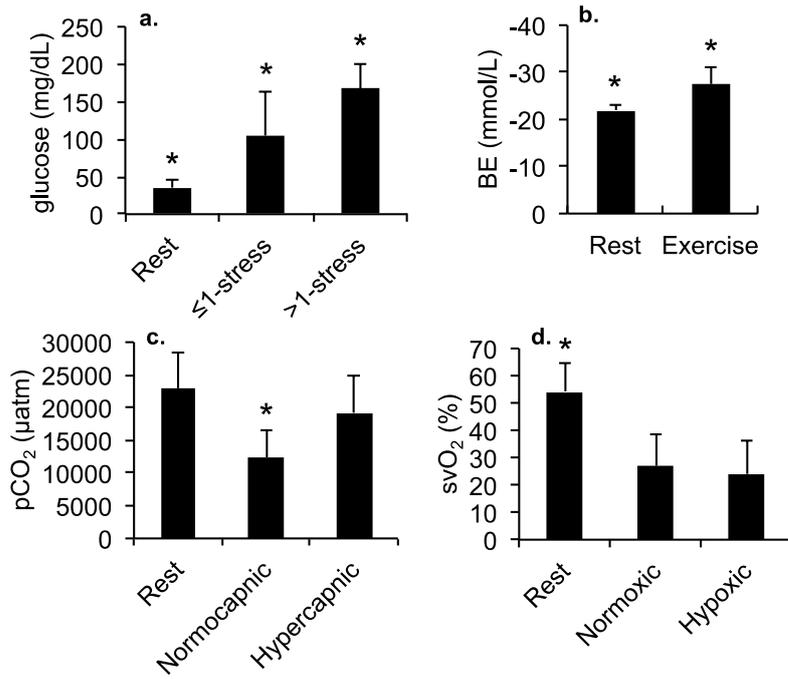
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### 3.8. Figures and Tables



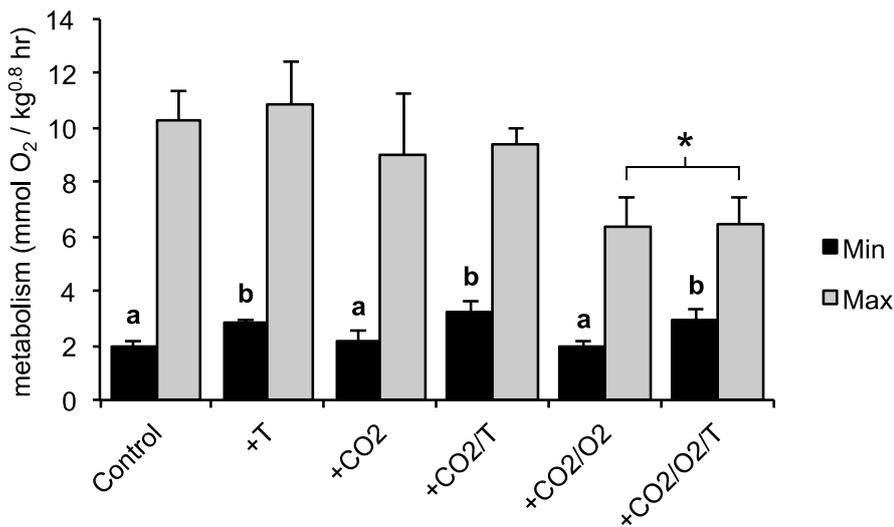
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3 Figure 1: Comparison of blood parameters across experimental treatments. All  
4 measurements performed with i-STAT clinical analyzer on venous blood at  
5 temperature 37°C. BE = base excess. svO<sub>2</sub> = venous oxygen saturation. Error bars  
6 show 1 standard deviation. N = 4 for each treatment.



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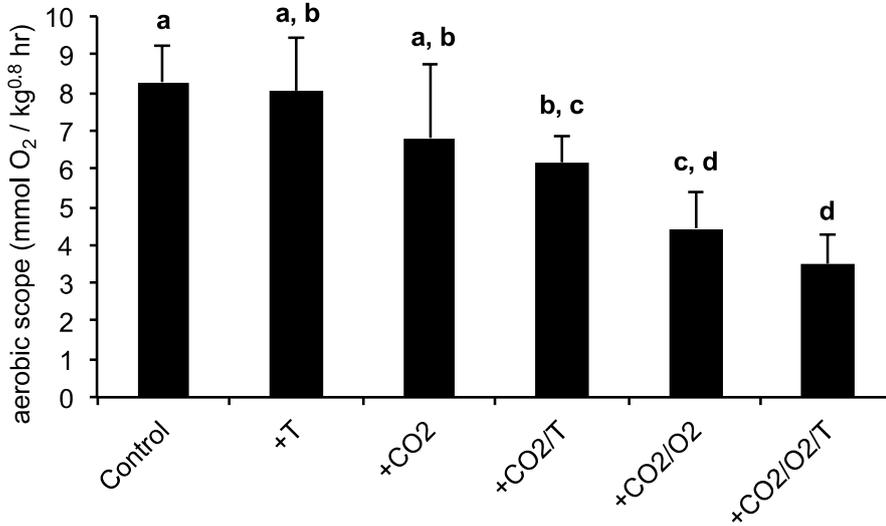
2 Figure 2: Comparison of blood parameters grouped for statistical comparison. All  
 3 measurements performed with i-STAT clinical analyzer on venous blood at  
 4 temperature 37°C. BE = base excess. svO<sub>2</sub> = venous oxygen saturation. Error bars  
 5 show 1 standard deviation. Categories marked by asterisks are significantly different  
 6 at p < 0.05.



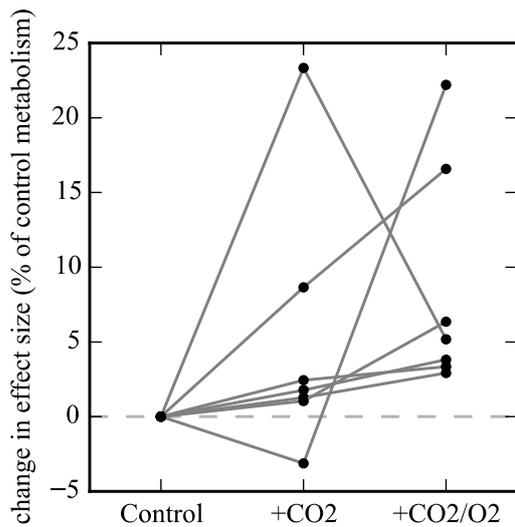
7

8 Figure 3: Impact of temperature, oxygen and CO<sub>2</sub> perturbations on active (grey) and  
 9 minimum (black) metabolic rates of *O. mossambicus*. Means with different letters (a-

1 b, \*) are significantly different (Tukey's HSD,  $p < 0.05$ ). Error bars are  $\pm 1$  standard  
 2 deviation.  $N = 7$  for each treatment. Treatment labels correspond to those in Table 1.



3  
 4 Figure 4: Impact of temperature, oxygen, and CO<sub>2</sub> perturbations on aerobic scope of  
 5 *O. mossambicus*. Means with different letters (a-d) are significantly different (Tukey's  
 6 HSD,  $p < 0.05$ ; see Table 3). Error bars are  $\pm 1$  standard deviation.  $N = 7$  for each  
 7 treatment. Treatment labels correspond to those in Table 1.



8  
 9 Figure 5: Effect of simultaneous stressors on in the effect of a 4°C warming on  
 10 aerobic scope. Negative numbers indicate a smaller magnitude effect while positive

1 numbers indicate increased magnitude. X-axis labels correspond to the treatment  
 2 labels as in Table 1, and indicate a comparison between high and normal temperature  
 3 under normocapnia and normoxia (Control), in the presence of hypercapnia (+CO<sub>2</sub>)  
 4 and under hypercapnia and hypoxia (+CO<sub>2</sub>/O<sub>2</sub>), respectively. Lines indicate  
 5 individual fishes. In 6 out of 7 cases, the change in aerobic scope due to 4°C warming  
 6 was greater in the presence of simultaneous CO<sub>2</sub> stress (+CO<sub>2</sub>) than in the control  
 7 case. In all cases, the change in aerobic scope due to warming was greater in the  
 8 presence of simultaneous CO<sub>2</sub> and oxygen stress (+CO<sub>2</sub>/O<sub>2</sub>) than in the control case.  
 9 The distribution of the slopes of linear regressions performed on each fish are  
 10 significantly greater than zero by single-sample t-test ( $p = 0.024$ ,  $N = 7$ ).

11 Table 1: Target environmental conditions for experimental treatments.

Treatment	Temperature (°C)	DO (μmol/kg)	pCO <sub>2</sub> (μatm)
Control	25.0	200	390
+T	29.0	200	390
+CO <sub>2</sub>	25.0	200	2000
+CO <sub>2</sub> /T	29.0	200	2000
+CO <sub>2</sub> /O <sub>2</sub>	25.0	100	2000
+CO <sub>2</sub> /O <sub>2</sub> /T	29.0	100	2000

12

13 Table 2: Measured environmental conditions for experimental treatments. Errors

14 indicate one standard deviation.

Parameter	Routine	Stressful
Temperature (°C)	25.1 ± 0.1	29.1 ± 0.1

DO ( $\mu\text{mol/kg}$ )	$201.3 \pm 6.7$	$102.0 \pm 9.5$
pCO <sub>2</sub> ( $\mu\text{atm}$ )	$381.4 \pm 46.0$	$2034 \pm 54.6$

1

2 Table 3: Significance testing of pairwise comparisons using Tukey's HSD. Treatment

3 labels correspond to column labels in Table 1. \* Indicates  $p < 0.05$ . \*\* Indicates  $p <$

4 0.01.

	+T	+CO <sub>2</sub>	+CO <sub>2</sub> /T	+CO <sub>2</sub> /O <sub>2</sub>	+CO <sub>2</sub> /O <sub>2</sub> /T
Control	0.90	0.30	*0.04	**0.00	**0.00
+T		0.48	0.09	**0.00	**0.00
+CO <sub>2</sub>			0.90	*0.02	**0.00
+CO <sub>2</sub> /T				0.15	*0.01
+CO <sub>2</sub> /O <sub>2</sub>					0.73

5

6

7

8

## 1 **4. An Empirically Derived Index for Predicting the Impacts of Multi-** 2 **Dimensional Environmental Stress on Marine Teleosts**

### 3 4 **4.1.1. Abstract**

5 Habitat suitability indices play a key role in many fields of conservation  
6 biology from fisheries models to ecosystem modeling. Yet commonly used indices  
7 often suffer from weaknesses in predicting habitat suitability in novel or uncommon  
8 environments, as well as in disentangling the effects of highly correlated  
9 environmental variables. We derive a new index from measurements of a key  
10 physiological indicator (aerobic scope) and demonstrate its use in a model teleost,  
11 *Oreochromis mossambicus*. We show how this new index can be used to evaluate  
12 previously developed theory-based models and as well as to project the possible  
13 impacts of anthropogenic environmental change on an area of *O. mossambicus*  
14 habitat. Our results are promising for predicting the responses of a wide range of  
15 teleosts to future environmental perturbations such as climate change and coastal  
16 eutrophication.

### 17 **4.1.2. Introduction**

#### 18 *4.1.3. Habitat Suitability Index: A Tool for Managers and Forecasters*

19 The relationship between a species' distribution and its environment is central  
20 to the study of ecology (Guisan and Zimmermann, 2000). It is no surprise then, that  
21 models that attempt to predict the spatial distribution of populations in relation to  
22 environmental factors are fundamental to many fields (Hirzel et al., 2006; Peterson,  
23 2006). Many of these models take the approach of a "habitat suitability index": an  
24 index that takes environmental conditions as inputs and outputs a probability of

1 habitation under those conditions (Hirzel and Le Lay, 2008). Habitat suitability  
2 indices are used, among other purposes, for managing exploited species, assessing  
3 biotic impacts of pollutants and climate change, analyzing population viability,  
4 assessing invasion risk, and predicting human-wildlife conflicts (Brooks, 1997; Hirzel  
5 et al., 2001, 2006; Peterson, 2006).

6 In the marine environment, variants of the habitat suitability index are  
7 widespread in fisheries models that attempt to estimate stock abundance from catch  
8 and effort data. Maunder, Hinton and Bigelow (Maunder et al., 2006) for example,  
9 use a “habitat effect” variable in their fisheries model that describes the effect of  
10 environmental conditions on the catchability of a species at a given hook:

$$C = q_{base}H(t_{water})EN_t$$

Eq.  
1.

11 In this equation,  $H$  is in effect a habitat suitability index that uses physical  
12 parameters (water temperature in this example) to predict the likelihood of habitation  
13 in a particular water mass, thereby correcting  $q_{base}$  (the default catchability across all  
14 habitats) to predict  $C$  (the catch at a given hook) by taking into account fishing effort  
15  $E$  and a time-dependent overall species abundance term  $N_t$ .

16 Similarly, end-to-end ecosystem models, which are rapidly gaining traction as  
17 a primary method of predicting climate and fisheries impacts on higher trophic level  
18 species (Travers et al., 2007), often rely on habitat suitability indices to relate  
19 projected environmental conditions to local population abundance. Lehodey, Senina  
20 and Murtugudde (Lehodey et al., 2008) for example, use a “feeding habitat index” in  
21 their SEAPODYM (Spatial Ecosystem and Populations Dynamics) model for pelagic  
22 tunas that indicates the most favorable areas for habitation outside of the spawning  
23 season:

$$H = \Theta(t_{water}, O_{2water})\rho$$

Eq.  
2.

1 where  $H$  is the feeding habitat index,  $\Theta$  is the accessibility index, and  $\rho$  is the prey  
2 density. In this case,  $\Theta$  is a habitat suitability index that describes the likelihood of  
3 habitation of a species as a function of water temperature and dissolved oxygen.

4 This ubiquity of habitat suitability indices means that endeavors ranging from  
5 fisheries assessment to ecosystem modeling depend on certain basic assumptions, and  
6 in some cases, share certain weaknesses. Our objective is to develop a habitat  
7 suitability index that addresses some of these weaknesses: specifically, those that arise  
8 with regards to predicting species distributions under scenarios of future  
9 environmental change.

#### 10 *4.1.4. Constructing an Index for Future Habitats*

11 Habitat suitability modeling in the context of environmental change presents a  
12 unique set of challenges. Two of the most fundamental are (1) correlations between  
13 explanatory variables (Prince and Goodyear, 2006; Schmittner et al., 2008) and (2)  
14 model extrapolation to novel habitats (Kearney and Porter, 2009; Kearney et al.,  
15 2008). Current habitat suitability indices, which largely consist of statistical  
16 correlations between presence/absence data and observed environmental parameters,  
17 are ill equipped to deal with these challenges (Hirzel and Le Lay, 2008; Homyack,  
18 2010; Kearney and Porter, 2009; Thuiller, 2004).

19 The problem of the correlation in explanatory variables is strongly related to  
20 the fact that common modes of environmental change in the marine realm such as  
21 climate change and coastal eutrophication are multivariate phenomena. For example,  
22 increasing anthropogenic CO<sub>2</sub> emissions are expected to raise surface ocean pCO<sub>2</sub>,  
23 raise mixed-layer temperatures, and lower dissolved oxygen content simultaneously

1 (Gruber, 2011). Similarly, increasing nutrient efflux in a coastal environment drives  
2 both declining dissolved oxygen content and substantial spikes in pCO<sub>2</sub> (Diaz and  
3 Rosenberg, 2008). Each of these changes requires distinct physiological adaptations  
4 on the part of marine teleosts, with different combinations of each stressor likely to  
5 lead to complex synergies or antagonisms (Pörtner et al., 2004). Yet the power of  
6 statistically derived habitat suitability indices is often limited by strong correlations  
7 between these variables (Rodda et al., 2011). Consequently, there is a high risk of  
8 overfitting the habitat space or misattributing patterns in distribution or abundance to  
9 spurious variables (Kearney et al., 2008; Rodda et al., 2011).

10         Similarly, the applicability of statistically derived habitat suitability indices is  
11 limited with regards to novel environmental conditions. Because such indices rely on  
12 presence/absence data, there is a lack of data on environments that are not currently  
13 extant in the field (Hirzel and Le Lay, 2008; Kearney et al., 2008). Climate change for  
14 example is a widespread forcing that is likely to push certain environmental variables  
15 such as pCO<sub>2</sub> beyond the levels currently seen in most habitats. Furthermore, novel  
16 combinations of variables are also difficult to predict (Kearney and Porter, 2009). The  
17 ability to explore beyond the naturally observable climate space without relying on  
18 extrapolation is a key challenge for improving the performance of habitat suitability  
19 indices.

20         In some cases, a physiological perspective can provide the basis for a more  
21 robust mechanistic framework (Claireaux and Lefrançois, 2007; Homyack, 2010;  
22 Kearney et al., 2008). Namely, a measured quantity known as the aerobic scope,  
23 which describes the difference between maximal aerobic metabolism and the  
24 minimum metabolic expenditure necessary for life-sustaining processes (Fry, 1971),  
25 is a promising integrative indicator of habitat quality (Lefrançois and Claireaux, 2003;

1 Pörtner and Knust, 2007). The magnitude of the aerobic scope in a given environment  
2 has been directly linked to the capacity for fitness-related activities such as growth,  
3 reproduction, migration, and foraging (Pörtner, 2001) both for active, highly aerobic  
4 swimmers such as pelagic piscivores for which locomotoray capacity is typically  
5 limited by aerobic scope as well as for comparatively sedentary benthic species with  
6 high anaerobic capacity for which digestive ability and burst swimming recovery are  
7 key (Clark et al., 2013). This provides the opportunity to avoid both of the two main  
8 weaknesses of correlative habitat suitability indices by basing a novel index on a  
9 factor that is directly linked to fitness under different environmental conditions. While  
10 it is possible in some cases to construct a theoretical model based on prior knowledge  
11 of a number of physiological parameters in a well-studied species (Del Raye and  
12 Weng, 2015), such models are vulnerable to modeling error due to gaps in our  
13 knowledge of physiological adaptations. Direct measurements of the output variable  
14 (aerobic scope) are a valuable opportunity to evaluate such theory-based models, and  
15 present an alternate approach to building a robust predictor of future aerobic  
16 performance.

#### 17 4.1.5. Two Approaches to Constructing a Physiology-based Index

18 Our study uses a previously constructed data set of aerobic scope (Del Raye  
19 and Weng; *in preparation*) in the eurytolerant teleost *Oreochromis mossambicus* under  
20 varying oxygen, temperature, and pCO<sub>2</sub> conditions to fit a novel habitat suitability  
21 model. We use this fitted empirical model to evaluate a theoretical physiology-based  
22 modeling framework previously developed in Del Raye and Weng (2015). We  
23 demonstrate the use of the model in a coastal habitat within the range of *O.*  
24 *mossambicus*, and explore a series of future scenarios corresponding to the types of

1 environmental perturbation that are expected to result from climate change and  
2 coastal eutrophication.

### 3 **4.1.6. Methods**

#### 4 *4.1.7. Empirical Model Construction*

5 In creating our model we followed the example of Claireaux and Lagardère  
6 (1999) in adopting the following basic assumptions: (1) the aerobic scope follows an  
7 asymptotic curve with respect to dissolved oxygen where the asymptote (maximum  
8 aerobic scope) is attained at 100% air saturation (roughly 200  $\mu\text{mol O}_2 \text{ kg}^{-1}$  at 30°C),  
9 and (2) declines in oxygen concentration result in an exponential decline in aerobic  
10 scope. In addition, we assumed a similar form of the function with respect to  
11 temperature (as in Lefrançois and Claireaux (2003)) where the optimum temperature  
12 in this case is the mean water temperature of the habitat throughout the year. This  
13 assumption is consistent with the strong correlation between the mean historic  
14 temperature experienced in a habitat and the thermal optima of the aerobic scope of  
15 populations that reside there (Eliason et al., 2011; Nilsson et al., 2009). The resulting  
16 form of the model is as follows:

$$M_{scope} = M_{peak} - e^{(T_{opt}-T)\lambda_T} - e^{(O_{2opt}-O_2)\lambda_O} \quad \text{Eq. 3.}$$

17 where  $M_{scope}$  is the predicted aerobic scope,  $T$  and  $O_2$  are the ambient temperature and  
18 dissolved oxygen, respectively,  $T_{opt}$  and  $O_{2opt}$  are the optimum temperature and  
19 oxygen, and  $\lambda_T$  and  $\lambda_O$  are the curvature coefficients for temperature and oxygen  
20 sensitivity. Higher  $\lambda_T$  or  $\lambda_O$  increases the sensitivity of the predicted aerobic scope to  
21 environmental conditions, narrowing the *pejus* limits (Pörtner et al., 2001) beyond  
22 which substantial declines in physiological performance begin to be seen. These  
23 model variables and symbols are as described in Table 2. While a number of studies

1 have explored the impact of pCO<sub>2</sub> on aerobic performance in marine teleosts  
2 (Mckenzie et al., 2003; Melzner et al., 2009; Nilsson et al., 2009) and the interactions  
3 between pCO<sub>2</sub> and the cellular or molecular constituents that underlie many of the key  
4 processes driving aerobic scope have also been elucidated, this literature is still quite  
5 sparse in comparison with our understanding of temperature and oxygen effects. As a  
6 result, we chose not to explicitly model pCO<sub>2</sub> effects but to explore this effect purely  
7 through the empirical data.

#### 8 *4.1.8. Model Fitting*

9 We trained our model on empirical measurements of aerobic scope in  
10 seawater-reared captive *O. mossambicus* under varying temperature, oxygen, and  
11 pCO<sub>2</sub> conditions reported previously in Del Raye and Weng (2015; [in preparation](#)).  
12 Briefly, seven measurements each of aerobic scope were conducted under  
13 temperatures of 25 and 29°C, oxygen concentrations of 100 and 200 μmol O<sub>2</sub> kg<sup>-1</sup>,  
14 and pCO<sub>2</sub> of 390 and 2,000 μatm, as well as various combinations of these variables.  
15 Two separate models were constructed to fit the normocapnic (pCO<sub>2</sub> = 390 μatm) and  
16 hypercapnic (pCO<sub>2</sub> = 2,000 μatm) data. Fitting was done by minimization of non-  
17 linear least squares using the Levenberg-Marquardt algorithm.

#### 18 *4.1.9. Theoretical Model Construction*

19 The theoretical model followed the basic construction previously outlined in  
20 Del Raye and Weng (2015). Briefly, we created a three-box model consisting of a  
21 conceptually infinite seawater reservoir, a gill reservoir, and a reservoir for the  
22 working tissues. The maximum metabolic rate component consists of ten coupled  
23 equations drawn from classical physiological relationships (e.g., the Fick Equation,  
24 the Van't Hoff Equation, the Bohr effect) and empirical regression equations (e.g.,  
25 temperature and oxygen effects on cardiac output). The estimated maximum

1 metabolism is a function of the flux of blood between the gills and working muscles  
2 (determined by the cardiac output), and the rate of oxygen flux per unit of blood  
3 volume (determined by modeled red blood cell kinetics). The minimum metabolic rate  
4 component is based on empirical regression equations. Subtracting the minimum from  
5 maximum metabolism allows us to estimate the aerobic scope.

6 The models presented in Del Raye and Weng (2015) were parameterized for  
7 yellowfin tuna (*Thunnus albacares*) and a generic marine teleost using measurements  
8 of physiological parameters gleaned from the available literature. Our *O.*  
9 *mossambicus* theoretical model used the same approach to gather parameters specific  
10 to this species.

#### 11 4.1.10. Model Evaluation

12 We evaluated the empirical and theoretical aerobic scope models against our  
13 measured aerobic scope dataset using the Nash-Sutcliffe model efficiency coefficient  
14 (Nash and Sutcliffe, 1970), which is an integrative measure of the correlation between  
15 the observed and predicted data as well as the degree of bias (Krause et al., 2005).  
16 Possible values for the model efficiency coefficient range from  $-\infty$  to 1, with a value  
17 of zero indicating that the model output is as predictive as the mean of the observed  
18 dataset, and a value of 1 indicating perfect agreement (Krause et al., 2005). To help  
19 interpret the model efficiency values, we also calculated the coefficient of  
20 determination ( $r^2$ ) as an indicator of the degree of correlation, and the percent bias.

#### 21 4.1.11. Study Site

22 To demonstrate the use of our model we chose a location on the south shore of  
23 the island of Oahu in Hawai'i (Figure 1). *O. mossambicus* were introduced to the  
24 island in 1952 (S. L. Coles et al., 1999) and became widely distributed across multiple

1 estuarine regions, although substantial displacement by and hybridization with the  
2 invasive blackchin tilapia (*Sarotherodon melanotheron*) has been observed in recent  
3 years (Wu and Yang, 2012). In addition to being a known habitat of *O. mossambicus*  
4 as has been confirmed by multiple surveys (S L Coles et al., 1999; United States  
5 Geological Survey, 2012; Williams and Hawaii State Department of Land and Natural  
6 Resources, 2010), our study site has also been the subject of high-frequency  
7 monitoring with moored sensors on a surface buoy since 2008 by the Pacific Islands  
8 Ocean Observing System (PacIOOS, 2015) and Sutton et al. (2014). The available  
9 record of seawater pCO<sub>2</sub>, dissolved oxygen, temperature, and salinity spans six years  
10 with pCO<sub>2</sub> measurements taken every 30 minutes (with the exception of breaks in the  
11 data) and temperature, oxygen, and salinity every 20 minutes.

12         The proximity of the site to the Ala Wai Canal outfall (a distance of  
13 approximately 300m) underscores its potential vulnerability to anthropogenic  
14 eutrophication. The canal is an artificial tropical estuary that has historically  
15 experienced chronic problems with residential and light industrial runoff, and it is  
16 estimated that 12% of total allochthonous carbon inputs are discharged into the  
17 surrounding ocean during non-storm periods (Laws et al., 1993). Intensification of  
18 anthropogenic inputs or changes in circulation at the study site could lead to more  
19 severe problems in the future.

#### 20         4.1.12. *Environmental Change Scenarios*

21         The model was evaluated using the time series of environmental conditions  
22 derived from the moored sensors in the study site. We chose four environmental  
23 change scenarios derived from the sensor data. (1) Current conditions, reflecting  
24 persistent high dissolved oxygen and low pCO<sub>2</sub> and including seasonally driven  
25 temperature oscillations. (2) A high CO<sub>2</sub> scenario with the conditions as in 1 but mean

1 pCO<sub>2</sub> raised to 2,000 μatm. (2) A high CO<sub>2</sub>, mild hypoxia scenario with conditions as  
2 in 2 but mean dissolved oxygen lowered to 150 μmol kg<sup>-1</sup>. (3) A high CO<sub>2</sub>, moderate  
3 hypoxia scenario with conditions as in 2 but mean dissolved oxygen lowered to 100  
4 μmol kg<sup>-1</sup>. Scenarios 2-4 reflect possible future changes of the kind that could be  
5 caused by coastal eutrophication (due to factors such as nutrient runoff or increased  
6 outflow from the nearby canal) alone or in combination with climate change forcing  
7 (e.g. acidification due to the uptake of anthropogenic CO<sub>2</sub>, hypoxia due to increased  
8 rainfall-induced runoff). A summary of the environmental change scenarios is shown  
9 in Table 1.

#### 10 **4.1.13. Results**

##### 11 *4.1.14. Empirical Model Fitting*

12 The fitted empirical model exhibited the expected behavior with distinct  
13 limiting oxygen curves characterized by changes in the maximum aerobic scope  
14 ( $M_{peak}$ ) and curvature ( $\lambda_o$ ) (Figure 2). While mean  $M_{peak}$  (maximum aerobic scope)  
15 was similar in the normocapnic and hypercapnic fit,  $\lambda_T$  (the temperature sensitivity  
16 coefficient) was 4-fold larger in magnitude when the model was fitted using the  
17 hypercapnic dataset, indicating a higher temperature sensitivity. Full parameter means  
18 and standard deviations are reported in Table 2

19 Evaluating the performance of the fitted empirical model against the observed  
20 dataset from Del Raye and Weng (2015; *in preparation*) revealed a Nash-Sutcliffe  
21 model efficiency coefficient of 0.71 (Figure 3). The correlation between the predicted  
22 and observed values, as indicated by the coefficient of determination ( $r^2$ ), was 0.71,  
23 and the percent bias was  $-3.2 \times 10^{-8}\%$ .

##### 24 *4.1.15. Theoretical Model*

1           The theoretical aerobic scope model for *O. mossambicus* was specified using  
2 physiological parameters gleaned from the literature. Due to a lack of available data, a  
3 number of cardiac parameters (e.g.  $\lambda_1$ ,  $\lambda_2$ ,  $Q_{max}$ ) were drawn from studies performed  
4 on *Oreochromis* hybrids (*niloticus* X *mossambicus* X *hornorum*) (see note to Table  
5 3).

6           Evaluating the performance of the theoretical model against the observed  
7 dataset from Del Raye and Weng (2015; *in preparation*) revealed a slightly lower  
8 Nash-Sutcliffe model efficiency coefficient (0.54) compared to the empirical model  
9 (Figure 4). The correlation between the predicted and observed values, as indicated by  
10 the coefficient of determination ( $r^2$ ), was 0.56, and the percent bias was 0.54%.

#### 11           4.1.16. *The Environment of the Study Site*

12           The six-year temperature, dissolved oxygen, and pCO<sub>2</sub> time series from the  
13 study site revealed a predominantly high oxygen, low pCO<sub>2</sub> regime with temperatures  
14 ranging seasonally from 23 to 28°C (Figure 5). Salinities were high and stable at or  
15 near 35 ppt. The oxygen record showed slight seasonal fluctuations with lower  
16 oxygen concentrations predominantly in the fall and higher concentrations in winter.  
17 One notable hypoxic event was recorded in September of 2013, with minimum  
18 dissolved oxygen concentrations reaching 51  $\mu\text{mol}/\text{kg}^{-1}$ . Summary statistics are  
19 reported in Table 4.

#### 20           4.1.17. *Evaluating Environmental Change Scenarios*

21           Running the fitted model on the time series of current environmental  
22 conditions in the study site (scenario 1) showed a mean predicted aerobic scope of  
23  $7.20 \pm 0.90 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$  (Figure 6) with seasonal temperature-driven  
24 oscillations of  $0.34 \pm 0.51 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$  (Figure 7). Under the high pCO<sub>2</sub>  
25 simulation (scenario 2), mean aerobic scope fell to  $6.88 \pm 0.74 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$ .

1 Seasonal oscillations also became more pronounced, with the magnitude of the  
2 oscillation reaching  $0.63 \pm 0.54 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$ . Further declines in aerobic scope  
3 occurred under the simulated high pCO<sub>2</sub>, mild hypoxia (scenario 3) and moderate  
4 hypoxia (scenario 4) conditions, with means of  $5.95 \pm 0.93$  and  $4.16 \pm 1.56 \text{ mmol O}_2$   
5  $\text{kg}^{-0.8} \text{ hr}^{-1}$  respectively, and seasonally-driven oscillations of  $0.86 \pm 0.99 \text{ mmol O}_2 \text{ kg}^{-$   
6  $0.8 \text{ hr}^{-1}$  and  $1.30 \pm 1.87 \text{ mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$ . Aerobic scope under both scenarios 3 and  
7 4 were significantly different from scenario 1 by  $p < 0.05$ .

#### 8 **4.1.18. Discussion**

9 Our results demonstrate the usage of an empirically derived, physiology-based  
10 model for predicting habitat suitability in a seawater environment for the euryhaline  
11 teleost *O. mossambicus*. We describe a simple set of equations based on published  
12 relationships between temperature and oxygen on aerobic scope and fit them against a  
13 dataset of experimental measurements obtained in a controlled laboratory setting. The  
14 fitted equations are a nearly unbiased estimator of the experimental values  
15 (percentage bias =  $-3.2 \times 10^{-8}\%$ ) and reproduce the following key characteristics of  
16 the measured aerobic scope:

- 17 1. Low normocapnic  $\lambda_T$  and high hypercapnic  $\lambda_T$ , leading to high temperature  
18 sensitivity only at high pCO<sub>2</sub>. This reproduces the synergistic interaction  
19 between temperature and pCO<sub>2</sub> observed in the experimental data, whereby  
20 high temperatures only depressed aerobic scope at pCO<sub>2</sub> = 2,000 $\mu$ atm.
- 21 2. Similar  $M_{peak}$  at high and low pCO<sub>2</sub>. As a result, pCO<sub>2</sub> has little effect as a  
22 single variable, but is important in combination with temperature due to the  
23 effect outlined above.

1           These characteristics have important implications when applied to the high-  
2 frequency time series of current environmental conditions and simulated  
3 environmental change, as will be discussed in subsequent sections.

#### 4           4.1.19. Comparison with Theoretical Aerobic Scope Model

5           While the theoretical aerobic scope model also produced biologically realistic  
6 estimates of *O. mossambicus* aerobic scope, it performed less well against the  
7 experimental dataset than did the empirical model. Nash-Sutcliffe modeling efficiency  
8 coefficients indicated that while both models performed substantially better than the  
9 mean of the observed data (i.e. the case where modeling efficiency = 0), the empirical  
10 model was a closer fit (modeling efficiency = 0.71 versus 0.54), partially due to  
11 higher bias in the theoretical model (percentage bias = 0.54%). In particular, the  
12 highest aerobic scope values in the experimental dataset (aerobic scope = 9.7, 9.9,  
13 10.5, and 10.6 mmol O<sub>2</sub> kg<sup>-0.8</sup> hr<sup>-1</sup>, respectively) were poorly captured in the  
14 theoretical model.

#### 15           4.1.20. Implications for *O. mossambicus* Under Environmental Change

16           The high frequency time series of environmental conditions at our study site  
17 revealed high and stable aerobic scope under current conditions (scenario 1),  
18 consistent with an area of favorable *O. mossambicus* habitat (Figure 6). Minimum  
19 aerobic scope even during the hypoxic event of 2013 never fell below 2.0 mmol O<sub>2</sub>  
20 kg<sup>-0.8</sup> hr<sup>-1</sup>, and the aerobic scope at the 10<sup>th</sup> percentile was 6.9 mmol O<sub>2</sub> kg<sup>-0.8</sup> hr<sup>-1</sup>.  
21 This is indicative of a eurytolerant organism that is often exposed to highly variable  
22 environmental conditions (Sardella et al., 2004). Consistent with this view, aerobic  
23 scope was not substantially affected by hypercapnia, with minimum and 10<sup>th</sup>  
24 percentile aerobic scope under scenario 2 (pCO<sub>2</sub> = 2,000 μatm) only slightly lower  
25 than under current conditions (1.5 and 6.5 mmol O<sub>2</sub> kg<sup>-0.8</sup> hr<sup>-1</sup>, respectively).

1           Despite this robust environmental tolerance, the suitability of the study site as  
2 an *O. mossambicus* habitat is likely to decline substantially under our hypothetical  
3 environmental change scenarios. Scenario 3 (moderate hypoxia) caused a 1.3 mmol  
4 O<sub>2</sub> kg<sup>-0.8</sup> hr<sup>-1</sup>, or 17% decline in mean aerobic scope, with minimum aerobic scope  
5 during the hypoxic event of 2013 falling below zero for a total of 3.3 hours. The  
6 environmental conditions under which aerobic scope reaches zero are termed the  
7 critical points, and indicate thresholds beyond which long-term habitation becomes  
8 impossible and animals must rely on anaerobic expenditures to survive (Pörtner et al.,  
9 2010). While *O. mossambicus* has a high anaerobic capacity (Sukumaran, 1986) and  
10 can carry out aquatic surface respiration to relieve short periods (several days) of  
11 hypoxic stress (Teichert-Coddington and Green, 1993), these conditions are likely to  
12 have strong sublethal effects. In particular, the shape of the limiting-oxygen curve  
13 (Claireaux and Lagardère, 1999) dictates that sensitivity to oxygen perturbation  
14 increases at low oxygen concentrations, and therefore even relatively small hypoxic  
15 excursions in excess of what is proposed for scenario 3 could dramatically increase  
16 the adverse outcomes. This is demonstrated by comparison with scenario 4, where  
17 mean aerobic scope was 3.1 mmol O<sub>2</sub> kg<sup>-0.8</sup> hr<sup>-1</sup>, or 42% lower than under current  
18 conditions. Furthermore, environmental conditions exceeded the critical thresholds for  
19 a total of 675 data points, broken into two segments such that 35% of the points  
20 within a 41-day period, and 23% of the points within a subsequent 38-day period  
21 exceeded the critical threshold.

22           One of the consequences of the increased sensitivity of the model to  
23 temperature and oxygen at high pCO<sub>2</sub> and low oxygen, respectively, is the emergence  
24 of substantial seasonal dynamics in the higher stress scenarios. In particular, the  
25 absolute difference in mean aerobic scope between the highest month (January) and

1 the lowest month (September) increased by 3.8-fold between scenarios 1 and 4  
2 (Figure 7), whereas the magnitude of this seasonal fluctuation as a proportion of mean  
3 aerobic scope increased from 5% (scenario 1) to 31% (scenario 4). Temperature  
4 fluctuations in higher latitude populations are thought to drive the shift from  
5 continuous breeding as is observed in the Hawaiian population to a summer breeding  
6 season at higher latitudes (Cornish, 1998). The high degree of variability in habitat  
7 suitability indicated by the fluctuations in simulated aerobic scope could portend a  
8 similar curtailing of the breeding season under conditions analogous to scenario 4.

9         It is important to note that although the integrative nature of aerobic scope  
10 makes it a powerful indicator of physiological change, this same quality creates  
11 inherent uncertainties when attempting to predict specific impacts on performance or  
12 behavior (Clark et al., 2013). This is apparent when it is considered that the aerobic  
13 scope essentially describes the total available budget for aerobic energy output at a  
14 given instant. Therefore, as in any energetic budget, the impact of an overall decline  
15 in available energy on a specific activity depends upon the apportionment of energy  
16 between activities. Some populations may seek to maintain constant energy allocation  
17 for somatic growth at the expense of fluctuating gonadal growth/immune function for  
18 example, meaning that growth rates will remain unaffected even under substantial  
19 declines in aerobic scope, whereas others may seek to maintain gonadal growth at the  
20 expense of somatic growth (Lambert and Dutil, 2000; Rijnsdorp, 1990). The  
21 proportion of energy allocation can also change within populations due to biotic,  
22 environmental, or endocrine factors such as in the closely related Nile tilapia  
23 (*Oreochromis niloticus*), where a bioenergetics model indicated that the proportion of  
24 aerobic metabolism allocated to biosynthesis ranged between 14–36% depending on  
25 the rate of feeding (van Dam and Pauly, 1995).

1           Despite these limitations, simultaneous measurements of aerobic scope and  
2 other key activities such as growth or locomotion can provide an indication of the  
3 sensitivity of specific functions to changes in aerobic scope. Unfortunately, this type  
4 of disaggregation of the total aerobic scope into its constituent activities is hampered  
5 in our study species due to the paucity of such measurements in the literature.  
6 McKenzie et al. (2003) found evidence of an association between aerobic scope and  
7 growth rate in wild type and growth hormone transgenic tilapia hybrids (*O.*  
8 *mossambicus* X *hornorum*), where higher growth rates in the transgenic hybrids was  
9 found to correspond with higher metabolism and aerobic scope. Similarly, Schnell  
10 and Seebacher (2008) found that critical swimming speeds in *O. mossambicus*  
11 correlated with the aerobic scope of fishes acclimated to different temperatures. Only  
12 three temperatures were tested in this study and therefore the slope of the correlation  
13 is unlikely to be reliable. However, if we assume a linear relationship between aerobic  
14 scope and critical swimming speed (as has been found in other species, e.g. Reidy et  
15 al. (2000)), this would imply a 6% reduction in critical swimming speed under  
16 scenario 2 (hypercapnia) compared to scenario 1 (control conditions), and an 18%  
17 reduction under scenario 4 (hypercapnia and moderate hypoxia).

18           In contrast to the paucity of published data for *O. mossambicus*, robust  
19 correlations between aerobic scope and fitness-related activities are available for other  
20 species. Such correlations cannot be taken to have predictive value for *O.*  
21 *mossambicus*, however, they do provide context for our results. For example, the  
22 magnitude of aerobic scope decline between scenario 1 and scenario 2 that our model  
23 predicts for *O. mossambicus*, if applied to seawater-acclimated European sea bass  
24 (*Dicentrarchus labrax*), would be sufficient to drive a 9.8% decline in specific growth  
25 (Claireaux and Lefrançois, 2007) (Table 5). In Atlantic cod (*Gadus morhua*) the same

1 change in aerobic scope would translate to a 27% decline in growth (Claireaux et al.,  
2 2000) and a 5% decline in the maximum sustainable swimming speed (Reidy et al.,  
3 2000), while in turbot (*Scophthalmus maximus*) we would expect a 27% decline in  
4 daily feeding rate (Mallekh and Lagardère, 2002). More severe declines in aerobic  
5 scope, such as those we predict for *O. mossambicus* in scenario 4 would correspond to  
6 a 39% decline in growth rate in *D. labrax*, a 79% decline in growth rate and a 15%  
7 swimming speed decline in *G. morhua*, and an 81% feeding rate decline for *S.*  
8 *maximus*. These substantial impacts are indicative of a magnitude of aerobic scope  
9 reduction that would have large effects on the *O. mossambicus* population of our  
10 study area.

#### 11 4.1.21. Prospects for Aerobic Scope-Based Habitat Suitability Indices

12 Our empirical model is a computationally cheap, physiology-based index that  
13 is well suited for use in fisheries, distribution, or end-to-end ecosystem models. Direct  
14 substitution into use cases such as Eq. 1 or 2 is made possible by dividing the  
15 modeled aerobic scope by  $M_{peak}$  to constrain the range to between 0–1. Many of the  
16 remaining limitations of our method could be addressed in future work.

17 One of the most significant limitations for the use of both the empirical and  
18 theoretical approaches is the large data burden required to specify the model. Climate  
19 envelope models and other statistical approaches have typically relied on abundance  
20 or presence-absence data coupled with publically available environmental data. For  
21 example, temperature and oxygen values in the World Ocean Atlas or World Ocean  
22 Database have been widely used in the marine realm (e.g. Prince and Goodyear,  
23 2006), whereas atmospheric parameters such as mean and minimum temperature and  
24 precipitation have often been preferred in freshwater environments (e.g. Zengeya et  
25 al., 2015). By contrast, the empirical model relies on intensive swim-tunnel

1 respirometry on live animals, and can only be carried out on species that are amenable  
2 to being kept in captivity (with rare exceptions, e.g. Farrell et al. (2003)). Similarly,  
3 the theoretical model relies upon a large volume of physiological data that has not yet  
4 been collected for all but a few exceptionally well-studied species such as Atlantic  
5 cod (*Gadus morhua*) and yellowfin and bigeye tuna (*Thunnus albacares* and *Thunnus*  
6 *obesus*) (see e.g. Del Raye and Weng, 2015). Creating such a physiological dataset for  
7 a poorly studied species would also require a substantial commitment of time and  
8 resources and is likely to be infeasible to carry out on more than a few species at a  
9 time. Applications such as end-to-end ecosystem models, which examine dozens of  
10 species or more, will therefore benefit by using our detailed theoretical or empirical  
11 model only on select species of particular interest. In the case of the theoretical model,  
12 prudent use of sensitivity analyses such as in Figure 8 will help to identify parameters  
13 that can be omitted or substituted with generic values without having a noticeable  
14 impact on the output values.

15         In cases where enough resources are available to collect more data, further  
16 refinements to the empirical model would help to buttress some important sources of  
17 uncertainty. For example, changes in the response of aerobic scope to temperature in  
18 conspecifics of different body mass have been suggested to drive population-level  
19 shifts in abundance or size structure (Pörtner et al., 2001). Conducting aerobic scope  
20 measurements across size classes could illuminate these differences and also prove  
21 important for applications such as fisheries models where the differential abundance  
22 of different size classes is extremely important (Hampton, 2000). Similarly, raising  
23 and/or breeding individuals under different environmental conditions would allow the  
24 model to take the effects of physiological plasticity and epigenetic adaptation into  
25 account. Studies on thermal acclimation of *O. mossambicus* for example have

1 revealed substantial capacity for the adaptation of juvenile fish to chronic  
2 perturbations in temperature (Schnell and Seebacher, 2008), and the high degree of  
3 variability in the physiological input parameters for the theoretical model (e.g.  $H$  and  
4  $p50$ ), which gives rise to the broad distribution of predicted aerobic scope, also  
5 suggest the potential for phenotypic variability in *O. mossambicus* populations to  
6 drive future adaptation. Such additional work would go a considerable way toward  
7 improving the precision of our predictions and our confidence in our understanding of  
8 the impacts of future environmental change.

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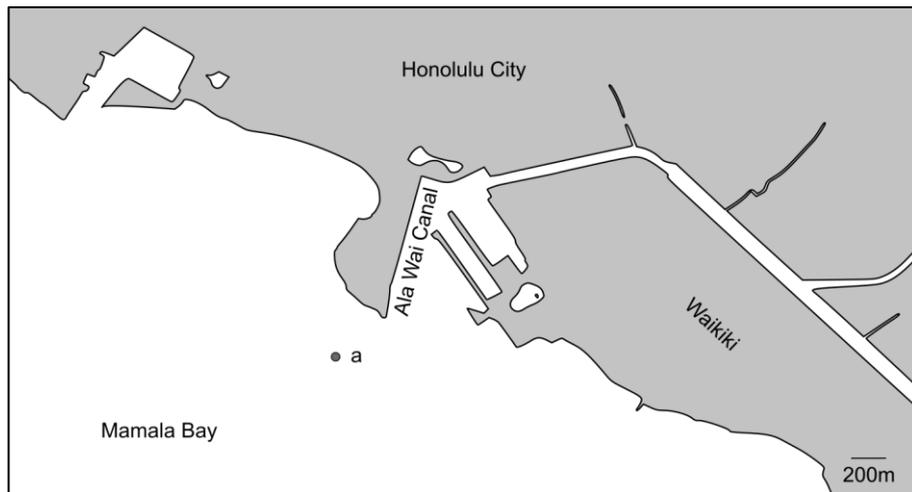
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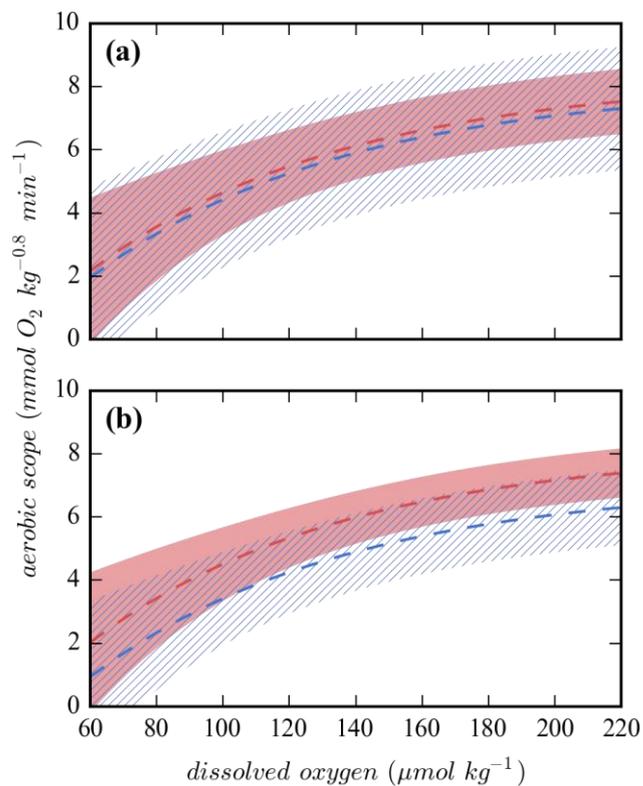
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#### 4.1.23. Figures and Tables



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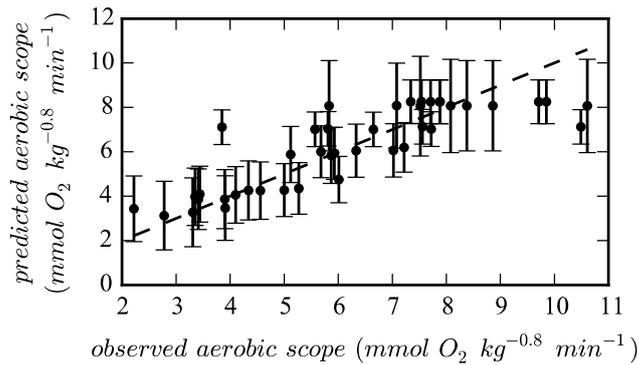
3 Figure 1: Map of study site (a) located at 21.280°N, 157.848°W, showing the  
4 proximity to the Ala Wai Canal outfall. Water depth at point a = 7.6 m. Map data  
5 from Google (2015).



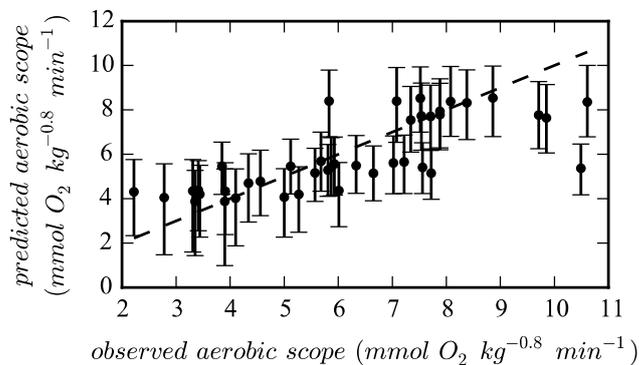
6

7 Figure 2: Demonstration of empirical model aerobic scope model for *Oreochromis*  
8 *mossambicus*. (a) Expected aerobic scope under normocapnic conditions ( $\text{pCO}_2 = 390$

1  $\mu\text{atm}$ ) at 25°C (red dotted line) and 29°C (blue dotted line). The red shaded and blue  
 2 hatched regions indicate the 95% confidence intervals of the fitted model at 25°C and  
 3 29°C, respectively. (b) Mean aerobic scope under hypercapnic conditions ( $\text{pCO}_2 =$   
 4 2,000  $\mu\text{atm}$ ) at 25°C (red dotted line) and 29°C (blue dotted line). Color and hatching  
 5 of the 95% confidence are as in (a).

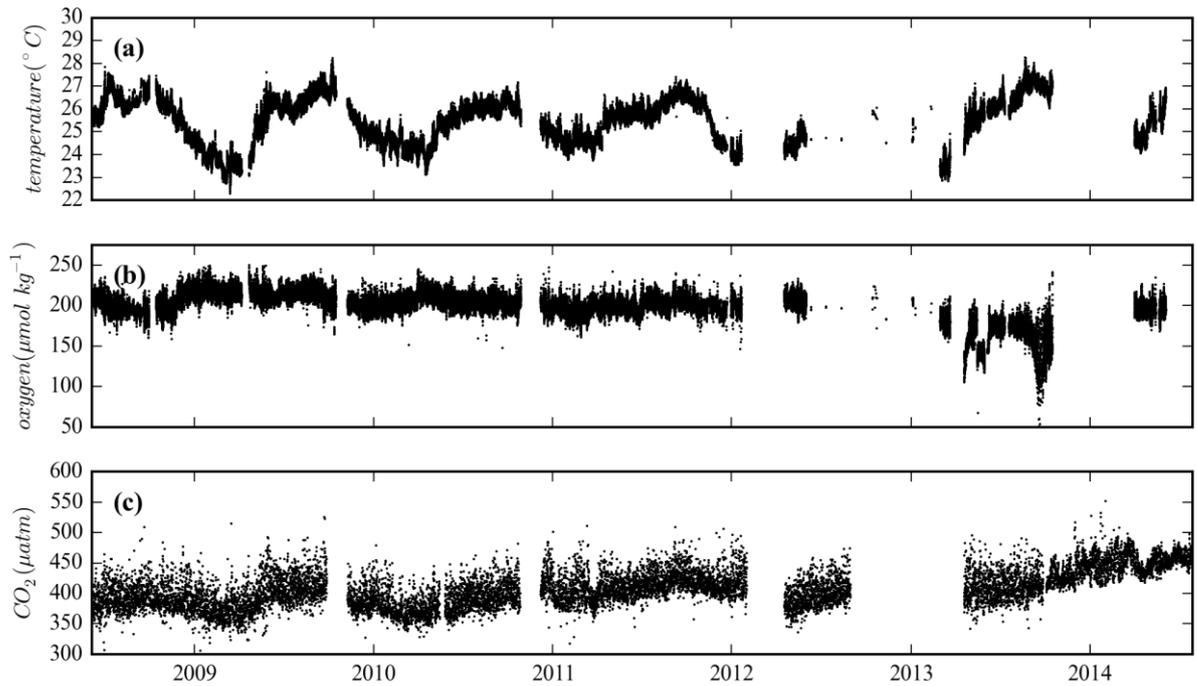


6  
 7 Figure 3: Observed versus predicted aerobic scope for *Oreochromis mossambicus*  
 8 using an empirical aerobic scope model. Error bars indicate the 95% confidence  
 9 interval of the fitted model. The dotted line shows the 1:1 line of perfect agreement.  
 10 Nash-Sutcliffe model efficiency coefficient = 0.71,  $r^2 = 0.71$ , and absolute percent  
 11 bias < 0.01%. N = 42 points.



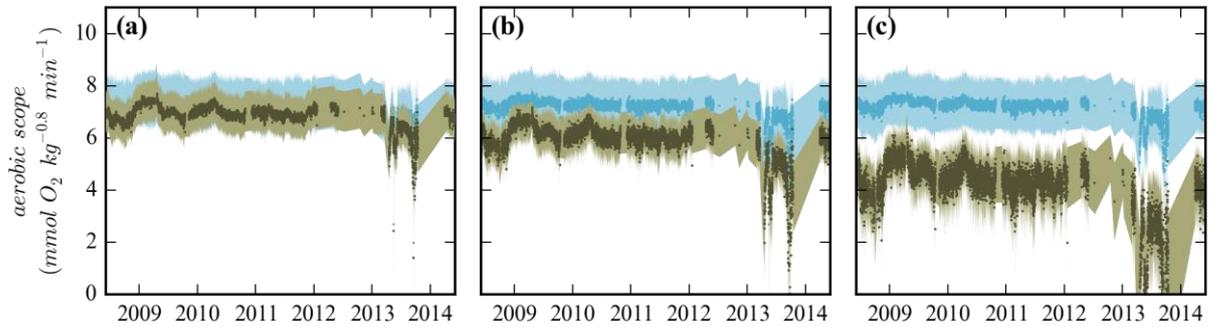
12  
 13 Figure 4: Observed versus predicted aerobic scope for *Oreochromis mossambicus*  
 14 using a theoretical aerobic scope model. Error bars indicate 95% confidence intervals,

1 where the magnitude of the uncertainty is a product of the observed variability in  
2 input physiological parameters. The dotted line shows the 1:1 line of perfect  
3 agreement. Nash-Sutcliffe model efficiency coefficient = 0.54,  $r^2 = 0.56$ , and absolute  
4 percent bias = 0.54%. N = 42 points.



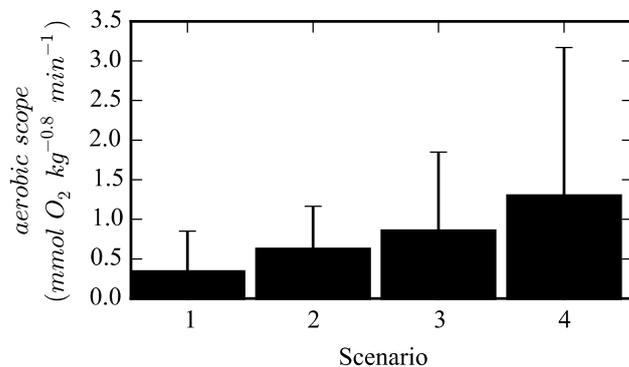
5  
6 Figure 5: Time series of observed environmental parameters at the study site. The  
7 number of temperature (a) and oxygen (b) records shown is 98,568. The number of  
8 pCO<sub>2</sub> (c) records is 14,670. Data from Sutton et al. (2014) and the Pacific Islands  
9 Ocean Observing System (PacIOOS, 2015)

10



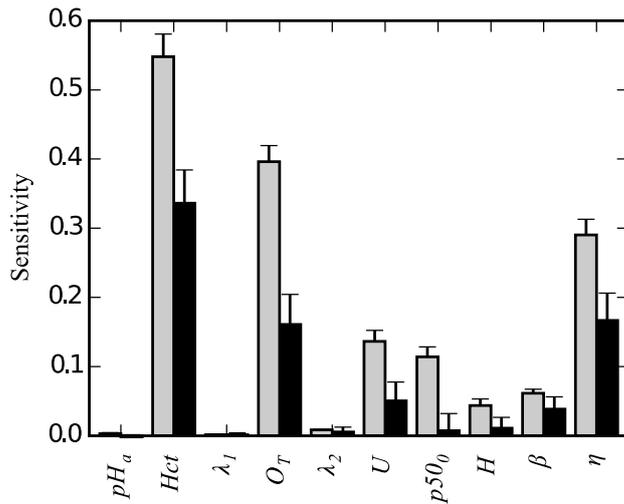
1

2 Figure 6: Simulated aerobic scope under four environmental change scenarios.  
 3 Current conditions (scenario 1, blue) versus future conditions (green) under (a)  
 4 simulated hypercapnia (scenario 2), (b) simulated hypercapnia and mild hypoxia  
 5 (scenario 3), and (c) simulated hypercapnia with moderate hypoxia (scenario 4). See  
 6 Table 1 for a summary of the environmental change scenarios. Lines show mean  
 7 values and shaded regions indicate 95% confidence.



8

9 Figure 7: Emergence of seasonal dynamics under high stress environmental change  
 10 scenarios. The difference in simulated aerobic scope between the peak (January) and  
 11 lowest (September) months are shown for each environmental change scenario. Black  
 12 bars indicate the mean difference across  $n = 6$  years. Error bars show one standard  
 13 deviation.



1

2 Figure 8: Sensitivity of the theoretical aerobic scope model to selected input  
 3 parameters. Bars show the magnitude of Monte Carlo estimates of Sobol sensitivity  
 4 indices. First order main effects (black bars) indicate the effect of uncertainty in the  
 5 given parameter on the uncertainty of the estimated aerobic scope. Total effects (grey  
 6 bars) show the sum of the main effects plus interactions between the given parameter  
 7 and all other parameters. Error bars indicate one standard deviation. Parameter labels  
 8 correspond to those shown in Table 3.

9

10 Table 1: Summary of environmental change scenarios.

Scenario	Description	Temperature range (°C)	Mean [O <sub>2</sub> ] (μmol kg <sup>-1</sup> )	Mean pCO <sub>2</sub> (μatm)
1	Present-day	22.3–28.3	200	407
2	Hypercapnic	22.3–28.3	200	2,000
3	Mild hypoxia	22.3–28.3	100	2,000
4	Severe hypoxia	22.3–28.3	50	2,000

11

- 1 Table 2: Fitting parameters for aerobic scope model under normocapnic ( $p\text{CO}_2 = 400$   
 2  $\mu\text{atm}$ ) and hypercapnic ( $p\text{CO}_2 = 2000 \mu\text{atm}$ ) conditions.

Symbol	Parameter	Units	Mean (sd) for Normocapnia	Mean (sd) for Hypercapnia
$M_{peak}$	Peak aerobic scope	$\text{mmol O}_2 \text{ kg}^{-0.8} \text{ hr}^{-1}$	9.26 (0.50)	9.01 (0.39)
$\lambda_T$	Temperature sensitivity coefficient	<i>n/a</i>	-0.042 (0.143)	-0.165 (0.056)
$\lambda_O$	Oxygen sensitivity coefficient	<i>n/a</i>	0.013 (0.001)	0.013 (0.001)

3

4

- 5 Table 3: List of input parameters for theoretical aerobic scope model. Values given  
 6 without standard deviations were treated as constants. Data from Lague et al. (2012)  
 7 (marked with an asterisk) was collected using a tilapia hybrid (*Oreochromis niloticus*  
 8 *x mossambicus x hornorum*).

Parameter (units)	Symbol	Mean (sd)	Reference
Reference temperature (K)	$T_0$	293.15	Perez and Maclean, 1976
Reference pH (K)	$pH_0$	7.5	Perez and Maclean, 1976
Mean cell hemoglobin concentration (g Hb/L)	$\bar{H}$	1000	Buthelezi et al., 2000
Reference cardiac output (L blood/kg body	$Q_{max}$	$6.3 \times 10^{-4}$	Lague et al., 2012*

sec K)			
Minimum venous pO <sub>2</sub> (μatm)	$pO_{2v}$	3950	Wu and Deng, 2006
Arterial pH	$pH_a$	7.78 (0.06)	Waarde et al., 1990
Hematocrit (volume fraction)	$Hct$	0.50 (0.05)	Buthelezi et al., 2000
Cardiac pO <sub>2</sub> sensitivity (L blood/kg body sec μatm O <sub>2</sub> )	$\lambda_1$	3300 (470)	Lague et al., 2012*
Cardiac pO <sub>2</sub> threshold (μatm)	$O_T$	$8.1 \times 10^{-5}$ ( $2.7 \times 10^{-5}$ )	Lague et al., 2012*
Cardiac temperature threshold (K)	$T_T$	303.15 (12.38)	Schnell and Seebacher, 2008
Cardiac temperature sensitivity (L blood/kg body sec K)	$\lambda_2$	0.05 (0.009)	Schnell and Seebacher, 2008
Oxygen utilization (partial pressure fraction)	$U$	0.874 (0.303)	Yamamoto, 1992
Hemoglobin half-saturation constant (μatm)	$p50_0$	17600 (2800)	Perez and Maclean, 1976
Hemoglobin heat of oxygenation (kcal/mole)	$H$	-27.2 (9.4)	Perez and Maclean, 1976
Extracellular Bohr coefficient ( $\Delta \log p50 /$ $\Delta pH$ )	$\beta$	-0.16 (0.05)	Perez and Maclean, 1976
Hill coefficient	$\eta$	2.1 (0.4)	Perez and Maclean, 1976

1

2 Table 4: Observed environmental conditions in the study site between June 2008 and

3 July 2014. N = 98,568 for temperature, salinity, and oxygen. N = 14,670 for pCO<sub>2</sub>.

- 1 Data from Sutton et al. (2014) and the Pacific Islands Ocean Observing System
- 2 (PacIOOS, 2015).

Parameter (units)	Mean (sd)
Temperature (°C)	25.4 (1.0)
Salinity (ppt)	34.9 (0.3)
Oxygen ( $\mu\text{mol kg}^{-1}$ )	199.6 (17.2)
pCO <sub>2</sub> ( $\mu\text{atm}$ )	407.7 (30.3)

- 3
- 4 Table 5: Predicted change in aerobic scope ( $M_{scope}$ ) of *Oreochromis mossambicus*
- 5 under scenarios 2-4 compared to the control scenario (scenario 1), and comparisons to
- 6 the expected magnitude of change in fitness-related activities (specific growth rate
- 7 ( $SGR$ ), critical swimming speed ( $U_{crit}$ ), and daily feeding rate ( $DFR$ )) for three
- 8 different species under analogous aerobic scope declines. Columns 4-7 were
- 9 calculated based on data from Claireaux and Lefrançois (2007), Claireaux et al.
- 10 (2000), Mallekh and Lagardère (2002), and Reidy et al. (2000).

Scenario	$\Delta M_{scope}$ ( $\text{mmol O}_2$ $\text{kg}^{-0.8} \text{hr}^{-1}$ )	$\Delta M_{scope}$ (%)	$\Delta SGR$ ( <i>Dicentrarchus</i> <i>labrax</i> ; %)	$\Delta SGR$ ( <i>Gadus</i> <i>morhua</i> ; %)	$\Delta U_{crit}$ ( <i>Gadus</i> <i>morhua</i> ; %)	$\Delta DFR$ ( <i>Scophthalmus</i> <i>maximus</i> ; %)
2	-1.31	-16	-9.8	-27	-5	-27
3	-2.22	-27	-19.5	-43	-9	-46
4	-3.96	-48	-39.0	-79	-16	-81

11

12

## 1 5. Summary and Conclusions

2

3

### 5.1. Summary

4 This thesis demonstrated three methods of forecasting the impact of climate-  
5 change driven environmental change on marine teleosts using a mechanistic,  
6 physiological framework—theoretical modeling (Chapter 2), direct experimentation  
7 (Chapter 3), and empirical models (Chapter 4). Overall, all three methods indicate that  
8 a combination of stressors, namely CO<sub>2</sub>-induced ocean acidification, surface  
9 warming, and deoxygenation, are likely to interact in complex ways to impair aerobic  
10 performance in marine teleosts.

11 In Chapter 2 for example, we showed that yellowfin tuna (*Thunnus albacares*)  
12 are likely to experience pronounced vertical habitat compression in Eastern Tropical  
13 Pacific due to a combination of high temperature, deoxygenation in the oxygen  
14 minimum layer, and high pCO<sub>2</sub>. This shoaling of the vertical habitat is particularly  
15 important because it is likely to increase the catchability of *T. albacares* in this region  
16 by bringing the tuna closer to surface-based fishing gear and also by decreasing the  
17 total habitat volume. Without correcting for this change in catchability, future  
18 managers run the risk of overestimating the stock of tuna in the region and  
19 consequently allowing the population to become overfished. Incorporating our work  
20 into end-to-end ecosystem models of the type exemplified by SEAPODYM and  
21 APECOSM (Lehodey et al., 2008; Maury, 2010) would also allow researchers to  
22 explore the ramifications of this vertical habitat compression on the interaction  
23 between comparatively surface-bound *T. albacares* and forage species that reside in  
24 the deeper oxygen minimum layer. The potential implications of such a study include

1 forecasting a shift in the competitive advantage between tunas, whose ability to make  
2 forays into the oxygen minimum zone depends partly on drawing oxygen from the  
3 surrounding water, and marine mammals such as cetaceans and elephant seals, which  
4 are only constrained by the distance of travel. To our knowledge, no work examining  
5 such a trend has been carried out to date.

6 In Chapter 3, we showed that deoxygenation and hypercapnia interact  
7 synergistically with high temperature to impair aerobic scope in Mozambique tilapia  
8 (*Oreochromis mossambicus*). This is an important result because while previous  
9 research had strongly suggested that the tolerance of this species to temperature,  
10 oxygen, and pH stress would insulate it against future change (Costa-Pierce and  
11 Doyle, 1997; Krishna Murthy et al., 1984; Rappaport et al., 1976), we showed that  
12 even though aerobic performance is unaffected by single-stressor perturbations,  
13 multiple simultaneous stressors had a significant effect. While *O. mossambicus* is a  
14 coastal species that is unlikely to experience the kinds of open-water oxygen  
15 minimum zone expansion that we explored in *T. albacares*, there is still a strong  
16 possibility that coastal hypoxia driven by anthropogenic eutrophication combined  
17 with climate-change induced water column stratification could expose it to  
18 increasingly severe hypoxia, hypercapnia, and hyperthermia (Cai et al., 2011; Diaz  
19 and Rosenberg, 2008; Melzner et al., 2012). Furthermore, our results provide an  
20 indication that relatively stenotolerant species that are not already pre-adapted to  
21 declines in water quality could suffer more drastic constraints on their aerobic  
22 capacity. Comparisons of our results against future studies on a variety of species  
23 would help to elucidate this difference. A particularly fascinating comparison would  
24 be between our study species, *O. mossambicus* and the variety of tilapia hybrids that  
25 have been developed for saltwater aquaculture, enabling us to explore any links

1 between aerobic performance (i.e. active versus sedentary species) or tolerance to  
2 single-stressor perturbations on climate change resilience.

3         In Chapter 4 we developed a novel form of an empirical model that combined  
4 a widely used framework for understanding temperature effects on aerobic scope  
5 (oxygen limited thermal tolerance (Pörtner, 2002)) and a similar framework for  
6 oxygen effects (limiting oxygen continuum (Claireaux and Lagardère, 1999)). By  
7 fitting the model to the experimental results obtained in Chapter 3 and inputting high-  
8 frequency observations of the environmental conditions in an actual *O. mossambicus*  
9 habitat, we were able to project the ways in which aerobic scope could be affected  
10 under a variety of possible environmental change scenarios. Our results indicate a  
11 large aerobic scope decline that is likely to result in substantial reductions in the  
12 capacity for growth, feeding, or locomotory ability. Unfortunately, due to a lack of an  
13 equivalently well-developed theory for understanding the effects of hypercapnia on  
14 aerobic scope, we were unable to incorporate a pCO<sub>2</sub> response in this model.  
15 Comparisons of aerobic performance across a larger number of hypercapnic  
16 treatments would help to identify a plausible form for such an aerobic scope versus  
17 pCO<sub>2</sub> curve, and is a promising area of study for the future. Because *O. mossambicus*  
18 is a euryhaline species that is capable of robust growth both in seawater and  
19 freshwater, comparisons of the pCO<sub>2</sub> response across a salinity gradient would be  
20 especially interesting because the difference in the availability of carbonate ions in  
21 seawater versus freshwater could reveal important mechanistic differences in the  
22 physiological response (Perry and Gilmour, 2006).

## 5.2. Suggestions for Future Work

1  
2           Going forward, more complex experimental techniques could lead to an  
3 expanded understanding of the sub-organism level physiological processes  
4 underpinning the patterns that we revealed in this thesis. For example, *in vivo*  
5 magnetic resonance imaging (MRI) and nuclear magnetic resonance spectroscopy  
6 (NMR) are two methods that have been successfully demonstrated in un-anaesthetized  
7 marine teleosts but have only been rarely used to date (Pörtner et al., 2010; Van der  
8 Linden et al., 2004). <sup>31</sup>P-NMR in particular can be performed on fish undergoing  
9 exercise in a swim-tunnel respirometer of the kind we utilized in this thesis (Bock et  
10 al., 2007), making it possible to continuously monitor intracellular pH and the  
11 turnover of high-energy phosphates such as ATP and phosphocreatine in fishes at  
12 both standard and active metabolic rates. Complementary measurements of blood  
13 flow and tissue oxygenation in restrained, sedentary fishes can be monitored using  
14 MRI (Mark et al., 2002). While the temporal resolution of MRI precludes its use on  
15 exercising fishes, the phenomena of excess post-exercise metabolism (Lee et al.,  
16 2003) or, alternatively, specific dynamic action (Fitzgibbon et al., 2007) could be used  
17 to monitor restrained animals with heightened metabolic rates due to previous  
18 exercise or food consumption, respectively. The combination of cardiac performance,  
19 tissue oxygenation, and pH status would provide powerful new insights into many of  
20 the constituent processes that determine the overall metabolic rate under conditions of  
21 aerobic stress. Such insights could be instrumental in understanding broader inter-  
22 species trends including the possible effects of climate change on species interactions.  
23 For example, an MRI analysis of *in vivo* physiology could be used to isolate a limiting  
24 factor for maintaining high aerobic scope under a given set of environmental  
25 challenges. In the hypothetical case that oxygen delivery to the cardiac muscles was

1 found to limit aerobic scope among high-energy pelagic predators for example, this  
2 could be used to generate the prediction that tunas with highly developed coronary  
3 arteries (e.g. bluefin tunas) would be more resilient to the given set of environmental  
4 changes than competing pelagic predators with no coronary vasculature, and therefore  
5 may gain a competitive advantage.

6 An equally intriguing avenue for future work is the extension of metabolic rate  
7 measurements into the field. While there have been some attempts to perform  
8 respirometry in the wild (Farrell et al., 2003), the most versatile methods involve the  
9 use of metabolic rate proxies such as heart rate and overall dynamic body acceleration  
10 (ODBA) (Lucas et al., 1993; Shepard et al., 2008). Heart-rate monitoring tags affixed  
11 to free-swimming fish provide an indicator of metabolic rate via the Fick equation,  
12 which relates cardiac output (a product of heart rate and stroke volume) to metabolism  
13 (Brodeur, 2001). While many teleosts are able to modify stroke volume to an  
14 equivalent or larger degree than heart rate (Farrell, 1996), repeated studies have  
15 shown that heart rate can be a good indicator of metabolism (Clark et al., 2010), as  
16 well as of temperature stress (Heath and Hughes, 1973) and poor water quality  
17 (Milligan and Wood, 1982) in many species. Similarly, strong correlations have been  
18 demonstrated between ODBA and field metabolic rate (Halsey et al., 2009). ODBA is  
19 an especially interesting index of metabolic activity because, rather than reflecting a  
20 physiological component of overall metabolism, it reflects a functional component –  
21 namely the metabolic cost of movement (Gleiss et al., 2011). Although this has  
22 mainly been treated in the literature as being equivalent to overall metabolic rate in  
23 ectothermic animals where the locomotory component accounts for the vast majority  
24 of the total metabolic expenditure (Lucas et al., 1993), an alternative interpretation  
25 would be to treat it as analogous to the scope for growth concept (Widdows et al.,

1 1997), which attempts to isolate the investment of metabolic energy into a specific  
2 activity – namely somatic growth. Used together, a general metabolic rate indicator  
3 (heart rate) and a proxy for the locomotory component (ODBA) could be examined  
4 across natural gradients in temperature, oxygen, or pCO<sub>2</sub> to determine how maximum  
5 and minimum metabolic rates are affected by these parameters in the wild. Examining  
6 stable environmental gradients in the field is an important method of accounting for  
7 trans-generational adaptation of the type that is likely to accompany future climate  
8 change perturbations (Johnson et al., 2012). While the technology for doing this  
9 exists, few studies appear to be taking advantage of it to produce this valuable  
10 counterpoint to laboratory experiments.

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