INTRODUCTION

One in five vertebrate species is classified by the IUCN as vulnerable, endangered, or critically endangered (Hoffmann et al., 2010). These declines and losses are largely attributed to anthropocentric changes in the environment such as land conversion, climate change, and unsustainable natural resource harvesting and extraction (Brook et al., 2008). Ultimately, range contraction and biodiversity loss are the end product of extrinsic or intrinsic challenges leading to population decline, emigration, and local extirpation. Across a species’ range, populations occur...
along ecological gradients from optimal, or central, habitats, where conditions and resources lead to high population density or maximal reproduction and survival, to marginal habitats where population density, reproduction, and/or survival are much lower (Holt, 2009; Kawecki, 2008). Identifying and mitigating the causes of reduced reproduction, compromised survivorship, and emigration are key for predicting and arresting biodiversity loss (Chown & Gaston, 2008).

The simplest species models assume that fitness follows a unimodal distribution with high density and growth rates in the center of a range and low density or poorly performing marginal populations found at the range periphery (Guo et al., 2005). However, environmental characteristics and species’ responses are much patchier than this, such that geographic and ecological marginality are not equivalent (Pironon et al., 2017; Santini et al., 2019). Variation in carrying capacity across environmental gradients can lead to high-density “source” populations producing emigrants that disperse to low-density ‘sinks’ in marginal habitats (Pulliam & Danielson, 1991); however, if all populations reach carrying capacity, dispersal is likely to be balanced between high- and low-density populations (Fretwell, 1969, 1972). In natural populations, however, environmental and demographic stochasticity result in dynamic reproduction, survival, and dispersal rates (Holt, 2003), which can cause low-density populations to be less resilient, with higher rates of local extirpation and recolonization, than high-density populations. Moreover, environmental change can either increase carrying capacity, leading to population growth and colonization, or result in decreased carrying capacity, population decline, local extirpation, and range contraction (Gaillard et al., 2000). Where ecological conditions are extreme for a species, local extirpation occurs faster than recolonization, limiting viable ranges. Range contraction occurs where previously resilient populations become unviable as growth and immigration rates no longer sustain the population. Identifying and predicting these dynamics in marginal populations provides key insight into long-term dynamics.

Responses to environmental change can be predicted using climate envelope, population viability, and mechanistic distribution models. Climate envelope, or habitat suitability, models relate species occurrence to environmental variables to explain or predict species distribution (Pearson & Dawson, 2003) and can predict occupancy changes under different environmental scenarios. While they are widely applicable because they require limited information, climate envelope models have limited capacity to predict local occupancy changes as they do not incorporate population specific dynamics, carrying capacities, species interactions, or dispersal potential. Climate envelope models based solely on occupancy are often poor predictors of habitat suitability and patterns of population abundance across ranges (Ossorio-Olvera et al., 2019). Population viability models, by contrast, can reliably predict future trends for specific populations (Brook et al., 2000) but require accurate vital rates, which are labor- and time-intensive to collect. Thus, their ability to predict resilience and viability across taxa and at large scales is limited. A middle ground is combining range and population dynamics for large-scale assessments of occupancy based on factors associated with local resilience or vulnerability. Mechanistic population and distribution models can provide this link between local population viability and range dynamics (Kearney & Porter, 2009) by using key behavioral and ecophysiological factors as functional indicators of resilience. Such models are more widely applicable than population viability analyses, are more robust, can be extrapolated to other populations, and have more predictive value than climate envelope models. They can also provide rapid and large-scale population assessments of marginal habitats to produce spatially explicit, predictive distribution maps across ecological gradients.

Interpreting the relevance of functional indicators across ecological gradients requires understanding their relationship with population responses (Bonier et al., 2009). Here, we advocate a Functional Marginality framework using physiological and behavioral indicators to assess population resilience. First, we describe functional physiological and behavioral indicators in the context of key stressors and explain methods to incorporate multiple indicators in predictive models. Second, we describe how functional marginality can be used to identify predictive hypotheses for occupancy changes, range dynamics, responses to environmental change, and evaluate the efficacy of management interventions. Although we primarily focus on mammals, this approach could be applied to many other taxa.

1.1 | Functional indicators

Functional traits are morphological, physiological, or behavioral traits that are fitness proxies via their effects on growth, reproduction, and survival (Viole et al., 2007), and indicate how a species perceives and responds to its environment (McGill et al., 2006). Positive functional indicators include relaxed time budgets, positive energy balance, and surplus energy stores manifesting in good body condition, good reproductive performance, and sound immune function or low disease burden. Negative indicators are those associated with a decline in condition in response to four types of challenges or stressors: physiological, psychological, chemical, and physical (Pottinger, 2003). Physiological stressors include resource, nutrient, or water restriction and disease. Psychological stressors include conflict, predation risk, and disturbance or persecution. Chemical stressors include altered pH, low dissolved oxygen, and exposure to pollutants, contaminants, or toxins. Finally, physical stressors encompass climate extremes and substrate as well as damage incurred by predation, conflict, or injury. Each class of stressor is associated with characteristic physiological and behavioral responses tied to pathways that maintain homeostasis (Madliger et al., 2018). Here, we discuss how physiological, psychological, chemical, and physical stressors can be manifest in physiological and behavioral indicators.

1.1.1 | Physiological stressors

Energetic and metabolic stress

Fitness is inextricably tied to maintaining sufficient energy reserves to support metabolism, invest in reproduction, and allow
changes can lead to changes of key microbiota, which impact on gut health (Gilbert et al., 2018; Sommer & Bäckhed, 2013). Diet can also be manifest within the gut, where microbial communities indicate acute fluctuations in energy mobilization. Mental factors on fitness than those that use glucocorticoids, which thyroid hormones as biomarkers to assess the impact of environmental change due either to climate or land use can have direct impacts on resource availability and seasonality. Behavioral changes can indicate energy budget challenges. For example, animals can adjust time budgets to spend more time traveling and feeding, and less time resting, to meet energy needs (Dunbar et al., 2009). Changes in habitat use or diet can also indicate energetic stress. For example, browsers becoming more dependent on grazing (Landman et al., 2013), or conversely grazers becoming more reliant on browse (Faith, 2012) suggesting a forced shift from preferred foods. The extent of temporary seasonal switching versus prolonged dependence on less preferred “fallback foods” can indicate significant resource stress in marginal habitats (Grueter et al., 2009). This is especially true when animals are pushed from an optimal diet to consume items that they are not physiologically adapted to handle (Ingala et al., 2019; Kitaysky et al., 2006). Thus, changes in the dynamics of seasonal and prolonged dietary shifts within and between populations could be used as a proxy for energetic stress. For terrestrial vertebrates, in addition to food limitations, water stress caused by abstraction or seasonality can lead to changes in space use, increased aggregations, and distance traveled.

Energy stress is also manifest in physiological responses. The hypothalamus–pituitary–thyroid axis (HPT) regulates metabolic rate by changing the amount of circulating thyroid hormone in response to metabolic requirements and responds to both thermal stress and food availability (Costa e Sousa & Hollenberg, 2012). Thyroid hormones and metabolic rate measures can identify how quickly animals are mobilizing and using energy; however, opposing responses to thermal and nutritional challenges can lead to a difficult to interpret metabolic trade-off between energy use and acquisition (Cristóbal-Azkarate et al., 2016). Large, longer-term differences in energy balance can be evaluated through changes in body condition, as the loss of muscle and fat reserves suggests a negative energy budget. Body condition scoring is routinely used in the management of wild mammals and standardized schemes have been developed for several species including black rhinos (Diceros bicornis) (Reuter & Adcock, 1998) and African buffalo (Syncerus caffer) (Ezenwa et al., 2009). Despite thyroid hormones offering a window into an individual’s energy balance (Behringer et al., 2018), fewer studies utilize thyroid hormones as biomarkers to assess the impact of environmental factors on fitness than those that use glucocorticoids, which indicate acute fluctuations in energy mobilization.

The impact of resource driven dietary shifts and external stressors can also be manifest within the gut, where microbial communities perform key functional roles in the host and contribute significantly to host health (Gilbert et al., 2018; Sommer & Bäckhed, 2013). Diet changes can lead to changes of key microbiota, which impact on gut function (Borbón-García et al., 2017). Beyond diet, microbiome communities are influenced by a range of factors including habitat, social network properties, and climatic conditions (Trevelline et al., 2019). Furthermore, primary and secondary acute stress responses such as glucocorticoids modulate the microbiome (Noguera et al., 2018). An imbalance of the microbial community, known as dysbiosis, can reduce digestive efficiency, increase inflammation, and susceptibility to infection (Amato et al., 2013; Dethlefsen et al., 2007; Gilbert et al., 2016). Signatures of dysbiosis will vary across hosts, as microbiome composition is sensitive to both diet and vertical transmission; however, dysbiosis or atypical microbiomes can be characterized by the degree of a displacement from a core microbiome composition (Zaneveld et al., 2017). Although the fitness consequences of changes in microbial community are poorly understood, microbiome composition has been linked to reproductive performance (Antwis et al., 2019) and cellular inflammation (Walshe et al., 2019).

Acute challenges: predation, disturbance, and social instability

A primary response to acute stressors such as predation, persecution, or disturbance is the activation of the hypothalamic–pituitary–adrenal (HPA) axis in birds and mammals or the hypothalamic–pituitary–interrenal (HPI) axis in fish, amphibians, and reptiles, which leads to the release of glucocorticoids (GCs) and catecholamines (Beehner & Bergman, 2017; Sopinka et al., 2016). Thus, the HPA/HPI axes and epinephrine stress responses are coupled with metabolism and metabolic rates, as both increase the body’s ability to mobilize energy for acute challenges. GCs have been used as an indicator of stress, commonly under the assumption that chronic elevation compromises health and ultimately fitness (Millsapgh & Washburn, 2004). However, short-term activation of the HPA is an adaptive response to allow individuals to effectively respond to acute challenges such that relationships between GCs and fitness are not straightforward (Moberg, 2000). The relationship between GC levels, GC reaction potential, and individual fitness (Bonier et al., 2009; Breuner et al., 2008) is context-dependent, such that during good conditions a high GC responsiveness is associated with poor survivorship and recruitment, whereas during poor conditions the relationship may be reversed when individuals in poor condition become unable to mount significant GC responses (Blas et al., 2007). There is so much variation in how individuals and species respond to chronic stress that there is no consistent profile to identify chronic stress across species (Dickens & Romero, 2013). What GCs do provide is evidence for a perceived challenge or stressor.

Combining physiological responses with other functional traits can help identify where these responses may lead to reduced fitness. For example, behavioral and endocrine profiles can be supplemented with direct physiological measures such as blood pressure, heart, and respiratory rate, if these are feasible for the study species, or proxies for these metrics if they are not (Madliger et al., 2018; Sopinka et al., 2016). Social instability also interacts with physiology (Gersick & Rubenstein, 2017; Seebacher & Krause, 2017) and is associated with elevated GCs in spotted hyenas (Crocuta crocuta) (Van Meter et al., 2009), Barbary macaques (Macaca Sylvanus) (Edwards et al., 2016).
Disease burdens

Heavy disease or parasite burden have fitness consequences (Pedersen & Fenton, 2007) including survival and fecundity that directly impact on population dynamics (Hillegass et al., 2010; Hudson, 1986; Hudson et al., 1998). Gastrointestinal nematode communities, or the nemabio, can directly affect host fitness but also have the potential to influence resistance and susceptibility to other infecting species (Supali et al., 2010). However, parasite infections are not universally harmful, removing helminths induces a strong inflammatory response (Walshe et al., 2019) and can potentially trigger autoimmune diseases (McKay, 2009). In addition to direct transmission risk, widespread anthropogenic disturbance can exacerbate disease risk through stress-induced immunosuppression. Although there is limited causal evidence between human impacts, stress, and disease occurrence, it is widely assumed that stress may be a major cause of increased susceptibility to wildlife disease (Hing et al., 2016). This may be due to the suppression of reproduction and immune function by the HPA axis as evidence for direct relationships between elevated GCs and parasite burden is well established (O’Dwyer et al., 2020).

Immune responses are also molecular indicators of physiological challenge or stress (Celli et al., 2019; Madliger et al., 2018; Sopinka et al., 2016). Immunoglobulins, or “antibodies” (e.g., IgA, IgG, IgM), form a critical part of the immune response by recognizing, binding to and neutralizing antigens, such as bacteria or viruses (Schroeder & Cavacini, 2010). Fecal antibody assays have been used to measure the immune response to parasites (Watt et al., 2016), which in turn correlate with survival (Sparks et al., 2018). Additional biomarkers that are associated with short-term and long-term responses to external challenges and stressors are blood parameters such as hematocrit levels and white blood cell counts (Madliger et al., 2018; Sopinka et al., 2016). Reduced hematocrit levels in birds are associated with a range of challenges including disease burden and nutritional status (Fair et al., 2007). Heterophil or neutrophil to lymphocyte ratios can indicate chronic stress, whereas eosinophil levels can indicate infectious disease (Davis et al., 2008). Inflammation markers can provide evidence of infectious and noninfectious processes. Calprotectin, lipocalin, and lactoferrin are inflammation markers that limit bacterial growth (Mao et al., 2012) and are used to diagnose inflammatory bowel disease in humans (Van Rheenen et al., 2010). Such biomarkers, which are gaining traction in human clinical practice, have untapped potential for use in wildlife monitoring. Increased metabolism results in the production of chemically reactive metabolic by-products known as reactive oxygen species (ROS) (Sies, 1991). Typically, ROS are removed from the body by antioxidants, but if they are generated in excess, oxygen radicals build up and bind to a range of biological molecules. This oxidative stress results in cellular and DNA damage, reduced defense mechanisms, and accelerated aging (Finkel & Holbrook, 2000). Chronically elevated GC production is associated with oxidative stress across species (Costantini et al., 2011).

1.1.2 Chemical and physical stressors

In addition to natural stressors, organic compounds, trace elements, and pharmaceuticals have all been responsible for catastrophic species declines (Rowe, 2008). Chemicals that are persistent and can bioaccumulate in food webs are particularly dangerous as they can have destabilizing effects on ecosystems. Major environmental contaminants are pesticides, perfluorinated compounds, and pharmaceuticals. Bioaccumulation of persistent organochlorines, such as DDT and associated compounds, has been implicated as major environmental contaminants, which cause catastrophic bird declines and are implicated in endemic disruption in humans (Blus, 2011). Tributyltin (TBT) is an antifoulant that is well known for its endocrine disruptive effects. Although DDT and TBT are now banned globally, their persistence means that they still occur at appreciable levels in the environment. Perfluorooalkylated compounds are commonly used in various forms of manufacturing. They are persistent in the environment and are linked to endocrine disruption, fertility, and metabolism (Jensen & Leffers, 2008). These compounds also affect human health, for example, increasing cancer risk, and declines in reproductive health, and longevity. The widespread use of antibiotics for human and veterinary health is linked to environmental bioaccumulation that, in turn, is implicated in the spread of antimicrobial resistance (Singer et al., 2016). Anti-inflammatory drugs also bioaccumulate with detrimental effects. The unregulated veterinary use of the anti-inflammatory drug diclofenac resulted in catastrophic declines of Asian vulture populations (Green et al., 2004; Shultz et al., 2020). Seabirds are also subject to high rates of anthropogenic stress results in cellular and DNA damage, reduced defense mechanisms, and accelerated aging (Finkel & Holbrook, 2000). Chronically elevated GC production is associated with oxidative stress across species (Costantini et al., 2011).
injury (Dias et al., 2019). Critically, in addition to direct mortality, injury is associated with reproductive suppression and delayed mortality in birds (Fajardo et al., 2000; Parsons et al., 2018), fish (Mueller et al., 2020), and reptiles (Sack et al., 2017) and can lead to population decline associated with high levels of physical injury. Thus, although physical injury is not necessarily a widespread problem, in some species both the acute and chronic impacts of injury are significant conservation challenges.

1.2 | Interpreting and integrating indicators

Using functional markers at large scales to identify vulnerable or declining populations requires reference or benchmark values from well performing populations. Where this is not possible, for example, in a species undergoing widespread declines and range collapse, it may be possible to use benchmarks from historical records or use congeners as a reference population (Britnell et al., 2021; Bocherens et al., 2015; Kerley et al., 2012). The expectation is that individuals from populations in marginal habitats or under challenges will have either single or multiple functional indicators that diverge from an optimal benchmark. Negative indicators will increase and positive indicators will decrease with the distance from central or optimal habitats (either geographically or in terms of niche hypervolume).

Single marker studies can give an incomplete or even misleading picture of individual condition and population health as multiple stressors can act independently or in tandem causing additive, synergistic, or antagonistic effects (Beldomenico & Begon, 2010; Todgham & Stillman, 2013). Few studies employ multitool approaches to evaluate the impact of stress on multiple physiological pathways (Madliger et al., 2018) and studies, which investigate stressors, physiology, and demography together are even more scarce (Beehner & Bergman, 2017). Functional responses to multiple environmental challenges can be measured using the concept of allostatic load, which is the cumulative physiological impact of challenges, when the body can no longer buffer challenges this becomes allostatic overload (McEwen & Wingfield, 2003). Allostatic overload leads to loss of condition, immune, and reproductive suppression and disease.

Concurrently evaluating a suite of biomarkers can provide information about how the different pathways interconnect and impact fitness in relation to environmental stressors or challenges (Figure 1, Table 1). The relationship between multiple biomarkers and population performance can be evaluated with a multivariate model selection approach (Johnson & Omland, 2004), a growth curve model or similar structural equation modeling approaches (Schlotz et al., 2011), or multidimensional data analysis. Thus, a set of indicators can be used to set up alternative hypotheses to determine primary challenges causing poor performance (Figure 2). For example, acute stressors such as disturbance, predation, or persecution may be associated with space use or time budget changes (use of refuges or cover, increased vigilance, movement, and decreased feeding and/or resting) and increased HPA activation. Resource limitation should be associated with increased foraging effort, diet changes, and decreased metabolic rate. Diet changes can be manifest by either increased switching to low-quality "fall back" foods during times of

**FIGURE 1** A conceptual diagram showing the different biomarkers available that can be integrated into studies using the footprints and pathway approach.
| Class                  | Species                          | Challenge                | Functional indicators | Population/fitness proxies | Reference                                      |
|------------------------|----------------------------------|--------------------------|-----------------------|---------------------------|-----------------------------------------------|
| Physiological- Resource| Killer whales *Orcinus Orca*     | Fish abundance           | fGCs, fT3             | Pregnancy ↓                | Wasser et al. (2017)                          |
|                        | African elephants *Loxodonta africana* | Rainfall                  | fGCs, fPG            | Reproductive function ↓    | Foley et al. (2001)                           |
|                        | Shetland ponies *Equus caballus* | Seasonality              | Heart rate, Locomotion | Field metabolic rate ↑     | Brinkmann et al. (2016)                       |
|                        | Soay sheep *Ovis aries*          | Maternal effects, Genetic variation | Ig proteins, Fecal egg count | Survival ↓                | Sparks et al. (2018)                          |
|                        | Roe deer *Capreolus capreolus*   | Primary productivity     | fA, fPG, Estradiol fGCs, fN, IgA | Reproductive condition ↓   | Escribano-Avila et al. (2013)                |
|                        | Cape Mountain zebra *Equus zebra* | Season                   | GCs ↑, Diet shifts    | population growth rate ↓   | Lea et al. (2018)                             |
|                        | Barbary macaques                 | Seasonality-food availability | T3 ↑                  |                           | Cristóbal-Azkarate et al. (2016)              |
|                        | White-tailed deer *Odocoileus virginianus* | Seasonality-food availability | T3/T4               |                           | Bahnak et al. (1981)                          |
|                        | European badger *Taxidea taxus*  | Food availability       | T3                    |                           | Harlow and Seal (1981)                        |
|                        | Chimpanzee *Pan troglodytes*     | Habitat quality          | Creatinine, GCs      |                           | Wessling et al. (2018)                        |
|                        | Common vampire bats *Desmodus rotundus* | Habitat conversion     | Diet, behavior, microbiome | Immune function       | Ingala et al. (2019)                          |
|                        | Black howler monkey *Alouatta pigra* | Logging and deforestation | Diet and microbiome diversity |                           | Amato et al. (2013)                          |
|                        | Primates                         | Habitat quality          | Microbiome diversity |                           | Stumpf et al. (2016)                         |
|                        | Western fence lizard             | Central-peripheral populations | GCs, plasma protein, hematocrit | Body weight             | Dunlap (1995), Dunlap and Wingfield (1995)   |
|                        | Maned wolf *Chrysocyon brachyurus* | Transformed landscapes   | GCs ↑, T3↑, PG↓       | Suggested reduced reproduction | Vynne et al. (2014)                          |
|                        | Primates (Pan spp, Ateles spp, *Papio*) | Habitat quality         | Time budgets          | Range and occupancy dynamics | Bettridge et al. (2010), Dunbar et al. (2009), Korstjens and Dunbar (2007), Korstjens et al. (2010) |

**Physiological- Acute Stress**

| Class                  | Species                          | Challenge                | Functional indicators | Population/fitness proxies | Reference                                      |
|------------------------|----------------------------------|--------------------------|-----------------------|---------------------------|-----------------------------------------------|
|                        | Guadalupe fur seals              | Capture                  | Aldosterone ↑ return to baselines |                           | DeRango et al. (2019)                          |
|                        | Bottle-nosed dolphin             | Beaching                 | Aldosterone ↑         |                           | Champagne et al. (2018)                       |
|                        | Stingrays                        | Tourist activity         | ROS ↑                 |                           | Semeniuk et al. (2009)                        |
|                        | Damselfly *Enallagma cyathigerum* | Predation                | Stress proteins, O2 consumption, Enzyme activity, Oxidative stress | Growth rates ↓            | Slos and Stoks (2008)                         |
| Class                  | Species                          | Challenge                          | Functional indicators                  | Population/fitness proxies          | Reference                                                                 |
|-----------------------|----------------------------------|------------------------------------|----------------------------------------|-------------------------------------|---------------------------------------------------------------------------|
|                       | Black grouse *Tetrao tetrix*     | Human disturbance                  | Feeding times ↑                         | Energy expenditure ↑                | Arlettaz (2015) #1790                                                     |
|                       | Great tit *Parus major*          | Artificial light                   | Corticosterone ↑                       | Fledging ↓                          | Ouyang et al. (2015)                                                     |
|                       | Eastern black rhinos            | Captive environment               | PG ↓, androgens ↓                      | Reproduction ↓                      | Antwis et al. (2019), Edwards et al. (2015)                               |
|                       | Florida manatee *Trichechus manatus latirostris* | Release from rehabilitation; injury and disease | Serum and urinary creatinine ↑, creatine kinase ↑, urea nitrogen, GCs↑, lymphocyte proliferation ↓ |                                   | Manire et al. (2003), Tripp et al. (2011)                                 |
|                       | African elephants *Loxodonta africana* | Translocation                      | GCs ↑                                 |                                     | Jachowski et al. (2013), Vlijoen et al. (2008)                             |
|                       | Chimpanzee Pan troglodytes      | Human Disturbance                  | GCs ↑                                 |                                     | McLennan et al. (2019)                                                   |
| Disease               | Red grouse                       | Nematodes                          |                                       | Fecundity ↓                         | Hudson (1986), Hudson et al. (1998)                                      |
|                       | Seychelles warblers (*Acrocephalus sechellensis*) | Parasitism, habitat quality       | ROS ↑                                 | Survival and fecundity ↓           | van de Crommenacker et al. (2011), van de Crommenacker et al. (2012), van de Crommenacker et al. (2017) |
|                       | Soay sheep                       | Parasitism                         | IgA                                    | Survival ↓                         | Sparks et al. (2018), Watt et al. (2016)                                 |
| Chemical and Physical | Fathead minnow *Pimephales promelas* | Environmental estrogen EE2         |                                       | Survival and fecundity ↓           | Schwindt et al. (2014)                                                   |
|                       | Little auk                       | Mercury exposure                   |                                       | Body condition ↓, growth rate ↓     | Amélineau et al. (2019)                                                  |
|                       | Black legged kittiwakes          | Perfluorinated carboxylates        | GC ↓                                  | Body condition ↓, Hatching ↓        | Tartu et al. (2014)                                                      |
|                       | Monk seals                       | Lethal injury                      | GC, T3                                 | Body condition ↓                   | Gobush et al. (2014)                                                     |
|                       | African elephants *Loxodonta africana* | Foot injury                       | FGC↑                                  | Body condition ↓                   | Ganswindt et al. (2010)                                                  |

Note: We highlight studies that link environmental challenges with multiple biomarkers and fitness proxies in terms of health, condition, or reproduction.
scarcity, or in extreme cases, the diet being completely comprised of low-quality items. Loss of condition and fitness costs, such as decreases in survival and fecundity, that are not associated with diet or behavioral change will likely be caused by disease when there are clear inflammatory responses, and by contaminants or chemical stressors when there are not. As each vulnerable population may face a unique set of challenges, a predefined set of testable hypotheses can be used to identify most likely candidates. A contingency table of expected responses can act as a starting point for formulating testable hypotheses based on the Functional Marginality Framework (Figure 2).

Multivariate modeling approaches can tease apart the relative importance of extrinsic factors. The role of environmental traits and functional indicators of health outcomes (i.e., reproductive failure, elevated mortality) can be evaluated using a model selection approach (Deelen et al., 2019). Multiple markers can also effectively evaluate the extent of “dysfunction” as a measure of deviation, such as Mahalanobis distance, from a multivariate central tendency (Milot et al., 2014). Evaluating the model weight for different factors (Johnson & Omland, 2004) can identify functional indicators that best predict fitness variance or population resilience, which can be used as key population health markers and focused on in future research. Clearly, a challenge with this macrophysiological approach is identifying markers that can be rapidly and noninvasively collected such as demography, behavior (association patterns, space use, and time budgets), and noninvasive biological samples (e.g., fecal and urine). Using model species, where noninvasive samples can be easily collected from a large number of individuals and tied to reproductive, survivorship, or population growth rate outcomes (Lea et al., 2018), is a key priority for developing a macrophysiological approach.

1.3 | Theoretical frameworks

Mechanistic distribution models use functional traits to link environmental variation with individual- and population-level performance (Buckley et al., 2010) as variation in physiological biomarkers of stress, health, and reproduction can act as these heuristic indicators of population viability (Chown & Gaston, 2008; Ellis et al., 2012; Gaston et al., 2009). Thus, they can predict likely population performance and range dynamics including the probability of colonization and extirpation under changing conditions (Figure 3a). For example, range-wide land transformation and climate changes can lead to an increased allostatic load (e.g., oxidative stress, inflammation, and disease) and a decline in positive biomarkers (reproductive function, metabolism, hematocrit levels) in adversely affected populations, with a net reduction in functional condition (Figure 3b). Functional traits can also evaluate patterns of range contraction, where the expanding edge will be associated with improved functional traits and

![Diagram](image-url)
**Positive Biomarkers**
- Reproductive hormones
- Metabolic rate
- Social/Reproductive behavior
- Body Condition/Fat reserves
- Diet quality
- Time budgets

**Negative Biomarkers**
- Glucocorticoids
- Parasite burden
- Microbiome dysbiosis
- Oxidative stress
- Vigilance behavior
- Chemical burden

**Shading**
- **Source**: population growth rate > 1
- **Sink**: population growth rate < 1

**FIGURE 3** Conceptual diagram of the Functional Marginality Framework. (a) Viable populations are determined by good functional condition leading to sustainable growth rates, range limits are determined by an increased burden of negative functional traits relative to positive ones. (b) Range shifts will be associated with improving functional condition on the expanding edge and declining condition on the retreating edge. (c) Habitat degradation leads to a net decline in functional condition (balance of positive indicators and negative allostatic load) across occupied habitat resulting in more sink populations and fewer source populations. (d) Functional condition can be tracked over time by repeatedly measuring positive and negative functional traits, and will exhibit characteristic profiles during periods of threat and recovery.

the retreating edge associated with declines in positive indicators and increasing burden of negative indicators (Figure 3c). Moreover, if range contraction is characterized by retreat into suboptimal refuges away from encroaching threats (Scheele et al., 2017), then in additional to poor functional condition in the contracting edge, remaining populations should have poor functional condition as compared to populations in optimal habitats, or where baselines are not available, relative to closely related species (Figure 3c). If environmental change causes species to retreat into optimal habitats (Channell & Lomolino, 2000), then functional traits in the remaining habitats should be consistent with those in optimal habitats or source populations. We would also expect a truncated distribution of markers toward less optimal states across the remaining range (Figure 3c).

Although variation in resource availability across landscapes is widely appreciated, metabolic costs also vary in terms of slopes, substrates, and thermal stresses (Shepard et al., 2013). Incorporating spatial patterns of energy availability and costs in “energetic landscapes” can provide a step change in our understanding of how environmental conditions impact on fitness at the individual and population level. Spatial variation in threats from predation, disturbance, and disease risk can be used to create predictive models of “landscapes of fear” and “landscapes of disgust” (Gallagher et al., 2017; Laundré et al., 2001; Weinstein et al., 2018). In particular, spatial variation in predation risk has consequences on physiology, reproduction, immune function, and behavior (Clinchy et al., 2011, 2013). These spatial models can be integrated to create “landscapes of stress,” where physiological or behavioral trade-offs can be directly incorporated into population or habitat use models (Koprivnikar & Penalva, 2015). For example, brown bears (Ursus arctos) near human settlements have lower heart rate variability, a cardiovascular indicator of stress, and they move further during increased human activity, which is expected to have an energetic cost (Støen et al., 2015). Similarly, landscape features, movement, and disturbance together predict physiological state in grizzly bears (Wilson et al., 2021).

We can also use indicators to test intervention success or the functional recovery of individuals or a population, which can provide insights into efficacy of restoration, colonization, and reintroduction. As humans have extensively changed and degraded habitats, conservation efforts often try to restore habitats or populations to reflect a historical state or ecological baseline (Britnell et al., 2021). A mechanistic approach can provide the evidence about how best to restore or manipulate degraded systems and how to establish whether an intervention has had the desired response (Hobbs et al., 2014). Successful interventions should increase population growth rates and nudge a population from being a sink to being self-sustaining or a source. Following an intervention, negative biomarkers should decrease, and positive biomarkers increase, relative to pre-intervention levels (Figure 3d; Cooke & Suski, 2008). The relationship between functional indicators and fitness can be assessed through changes in vital rates, for example, by monitoring changes in pregnancy rates of vertebrates before and after restoration or policy implementation (Pallin et al., 2018).

Planning for, and the short- and long-term effects of, active management such as handling, translocation, and reintroduction can
also be evaluated with functional indicators. Factors such as climate suitability are predictors of translocation failure (Bellis et al., 2020), and mechanistic distribution models can increase the robustness of habitat suitability predictions. Translocation success is also linked to stress responses and resilience, which occur during the translocation event and the establishment phase immediately after release (Dickens et al., 2010). Following an intervention, positive and negative biomarkers should return to pretranslocation levels after intervention and recovery. Conducting physiological monitoring before, during, and after release can improve our understanding of translocations, and the factors required for them to be a success. For example, a study using the Florida manatee (Trichechus manatus latirostris) in a simulated release process found that decreased food intake and changes in water salinity led to serum creatinine elevations and impaired immune function, indicated by lymphocyte proliferation assays (Manire et al., 2003). This is beneficial from a conservation perspective, as it increases the probability of future reintroduction success, and from a welfare perspective, as it allows methods to be refined to avoid stress and reduces the number of animals required (Tarszisz et al., 2014).

1.4 | Examples

There is now a small but growing number of studies that have used functional traits to understand range dynamics and differences between central and peripheral populations. For examples, GCs, blood parameters, and body condition vary between central and marginal populations of Western fence lizards (Sceloporus occidentalis) (Dunlap, 1995; Dunlap & Wingfield, 1995). Combining variation in time budgets with species distribution models in primates is an excellent example of using behavior patterns to understand drivers of population and range dynamics (Bettridge et al., 2010; Dunbar et al., 2009; Korstjens & Dunbar, 2007; Korstjens et al., 2010). There is also extensive evidence for how pollution and contaminants affect fitness proxies and functional indicators in birds (Rattner et al., 1984), although this approach has not been widely used to evaluate spatial range dynamics in a macroecology context. Despite this, there are limited examples of studies that evidence links between environmental stressors, physiology, behavior, and fitness measures to predict both individual- and population-level responses to challenges (Beehner & Bergman, 2017; Cooke et al., 2013). We have summarized a range of studies that have used functional markers to assess the impact of challenges and population or fitness consequences (Table 1). There are, however, a few key studies that have evaluated links between environmental challenges, functional markers, and population-level variation in resilience and viability. Physiological and behavioral biomarkers have been used to identify marginal or “refuge” populations in Cape mountain zebra (Equus zebra zebra) (Lea et al., 2018) where poor reproduction performance is associated with elevated androgens and glucocorticoids as a consequence of resource limitation and skewed population sex ratios. Functional markers including elevated creatinine, C-peptide, and glucocorticoids were used to evidence thermal stress in chimpanzees (Pan troglodytes) inhabiting a savannah-mosaic habitat at the margins of their range (Wessling et al., 2018). Hawaiian monk seals (Neomonachus schauinslandi) have experienced declines since the 1950s that have been variously attributed to poor juvenile survival due to resource limitations, injury, and disease (Harting et al., 2021). Declining subpopulations were associated with chronic elevation of fGCMs and low fT3, especially in immature individuals and had, on average, poorer survival rates and lower intrinsic population growth rates (Gobush et al., 2014). In better performing populations, multiple markers still highlighted how anthropogenic injury and disease relative to malnutrition affected intrinsic growth rates (Harting et al., 2021). This modeling approach that identifies how local stressors affect growth rates could be applied to most conservation scenarios.

2 | CONCLUSION

In recent decades, much research has been carried out to develop biomarkers, which provide an indication of how the environment affects the physiological and behavioral state of an organism and ultimately on fitness. This is a difficult task as physiology is extremely complex. Physiological responses are the result of multiple interconnecting pathways, which can respond to the same stressors and interact with each other, making the change in a single biomarker difficult to relate to fitness. We propose that the establishment of complementary and integrated biomarkers to indicate population health, properly validated and applied to testable hypotheses, would be a major advance for large-scale ecology and conservation. Validation, the discovery of relevant biomarkers or combination thereof, is a key part of this approach. The approaches we describe can be used to show which biomarkers are useful at predicting future changes in fitness measures associated with population changes. Once established, these biomarkers can be the basis for investigating the causes of poor individual health and changes in survival and reproduction and testing ecological and conservation hypotheses. This information can help to uncover the causes of distributional limits and predict future changes, estimate resilience of populations to novel threats, assess the efficacy of conservation efforts, and reveal macroecological trends and processes. This approach provides conservation biologists and practitioners the ability to produce evidence for the causal mechanisms underlying conservation problems and macro- or evolutionary ecologists the ability to investigate the physiological mechanisms underlying long-term and large-scale processes. Advances in these fields can contribute toward the calls for evidence-based conservation and help to alleviate the threat of species extinctions and ecological collapse.

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CONFLICT OF INTEREST
We declare there are no conflicts of interest associated with this article.

AUTHOR CONTRIBUTIONS
Susanne Shultz: Conceptualization (equal); Project administration (lead); Supervision (lead); Writing-original draft (equal); Writing-review & editing (supporting).
Nicholas Harvey: Conceptualization (equal); Project administration (lead); Supervision (lead); Writing-original draft (equal); Writing-review & editing (supporting).

DATA AVAILABILITY STATEMENT
There are no primary data presented in this manuscript. All studies discussed are referenced in the manuscript.

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