

Portfolio conservation of metapopulations under climate change

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Abstract

Climate change will likely lead to increasing population variability and extinction risk. Theoretically, greater population diversity should buffer against rising climate variability, and this theory is often invoked as a reason for greater conservation. However, this has rarely been quantified. Here we show how a portfolio approach to managing population diversity can inform metapopulation conservation priorities in a changing world. We develop a salmon metapopulation model where productivity is driven by spatially-distributed thermal tolerance and patterns of short- and long-term climate change. We then implement spatial conservation scenarios that control population carrying capacities and evaluate the metapopulation portfolios as a financial manager might – along axes of conservation risk and return. We show that preserving a diversity of thermal tolerances minimizes risk given environmental stochasticity and ensures persistence given long-term environmental change. When the thermal tolerances of populations are unknown, doubling the number of populations conserved may nearly halve metapopulation variability. However, this reduction in variability can come at the expense of long-term persistence if climate change increasingly restricts available habitat – forcing ecological managers to balance society’s desire for short-term stability and long-term viability. Our findings suggest the importance of conserving the processes that promote thermal-tolerance diversity, such as genetic diversity, habitat heterogeneity, and natural disturbance regimes, and demonstrate that diverse natural portfolios may be critical for metapopulation conservation in the face of increasing climate variability and change.

Keywords: biocomplexity, ecosystem based management, Pacific salmon, portfolio effect, prioritization, range contraction, response diversity, risk assessment, stability-diversity, stochastic simulation

Running head: Metapopulation portfolio conservation

Introduction

Untangling the mechanisms that underpin the stability of ecological systems is a critical focus of ecology (e.g. Ives and Carpenter 2007; de Mazancourt et al. 2013). Decades of research has focused on the role of species richness and functional diversity in driving stability; however, recent research has highlighted that the drivers of ecological stability are more complex and multidimensional than previously thought (e.g. Balvanera et al. 2006; Ives and Carpenter 2007; de Mazancourt et al. 2013). Two key drivers of population stability that have been comparatively understudied are response diversity (Winfree and Kremen 2009; Mori et al. 2013) – different responses to the environment by functionally similar species or populations (Elmqvist et al. 2003) – and the role of metapopulations (Schtickzelle and Quinn 2007). Here, we examine the role of response diversity conservation in stabilizing metapopulations given projected changes in climate. With unprecedented loss of biodiversity and levels of anthropogenic environmental change, it is more critical than ever to consider conservation approaches that maintain system stability in the face of environmental uncertainty (Lee and Jetz 2008; Ando and Mallory 2012).

Typically, conservation actions to maintain system stability and thereby reduce risk are driven by an *ad hoc* combination of scientific information, political influences, and feasibility (Margules and Pressey 2000); the management of financial portfolios provides another way of considering risk (e.g. Figge 2004; Koellner and Schmitz 2006; Ando and Mallory 2012; Haak and Williams 2012). Economists work to minimize risk and maximize returns by building a portfolio of individual investments (called assets) with different attributes. For example, different financial sectors can be expected to perform uniquely in some economic conditions; when one rises in value another may fall. Modern Portfolio Theory proposes that out of all possible portfolios, there is a small subset of portfolios that maximizes expected return for a level of risk or minimizes risk for a level of return (called the efficient frontier), and that only by considering risk and return in tandem can an investor achieve maximum benefit from a portfolio (Markowitz 1952).

Similarly, expected growth rate and variance of a metapopulation is a function of the variance, covariance, and size of the individual populations (Moore et al. 2010; Carlson and Satterthwaite 2011; Anderson et al. 2013). An ecological portfolio approach to managing risk for a metapopulation might therefore consider how conservation actions affect the weight of each population in a metapopulation portfolio. This investment weight could represent the conservation budget or the habitat conserved for each population. The population growth rate is then analogous to the financial rate of return and the variability of that growth rate a metric of risk.

Environmental conditions could represent the financial market conditions. Given this interpretation, ecological managers could consider how various conservation strategies affect the expected risk and return of their ecological portfolio. These risk and return elements are central to ecological management and conservation – management aims to ensure stability over environmental variability (risk), and increase population abundance (return). Different scenarios may suggest different desired trade-offs between the two. For example, a manager with a healthy population might prioritize short-term stability, while a manager with an endangered population might try to balance the two, or prioritize population growth initially.

Managing Pacific salmon under the uncertainty of climate change is an ideal scenario to consider through the lens of portfolio theory for four reasons. (1) The migration of Pacific salmon biomass profoundly influences aquatic and terrestrial coastal ecosystems throughout the North Pacific ocean from Korea to California (Quinn 2005). (2) Pacific salmon form metapopulations (e.g. Policansky and Magnuson 1998; Cooper and Mangel 1999; Schtickzelle and Quinn 2007) and we can consider, for example, the metapopulation in a river-catchment as a portfolio and the stream populations as assets (Schindler et al. 2010; Moore et al. 2010; Carlson and Satterthwaite 2011; Anderson et al. 2013; Yeakel et al. 2014). Fisheries often integrate across multiple populations, acting as investors in the salmon portfolio (Hilborn et al. 2003). Fisheries managers and conservation agencies can act as portfolio managers by choosing which salmon habitat to prioritize for protection or restoration. (3) Many Pacific salmon metapopulations are highly threatened (e.g. Gustafson et al. 2007) and will likely become more at risk as threats such as overfishing, damming, logging, and particularly changing climate, intensify (e.g. Lackey 2003). Indeed, recovery goals for Pacific salmon are often set at the metapopulation level (McElhany et al. 2000), and knowing what minimizes risk to the metapopulation can help choose efficient conservation actions (Policansky and Magnuson 1998; McElhany et al. 2000). (4) Given the scale and variety of the threats facing salmon, some prioritization will be required to recover these highly-valued, even iconic species (Allendorf et al. 1997; Ruckelshaus et al. 2002).

Two key mechanisms can generate the asynchrony in metapopulation dynamics that is critical to a diversified portfolio. First, localized habitat features can filter larger-scale environments, generating unique conditions for populations (Schindler et al. 2008) (*sensu* the Moran effect). Second, salmon populations may respond differently to environmental variability (i.e. response diversity (Elmqvist et al. 2003) and biocomplexity (Hilborn et al. 2003)) arising from unique local adaptations and traits (Fraser et al. 2011; Eliason et al. 2011; Thorson et al. 2014*b*). In reality, these

mechanisms can interact. For example, salmon response diversity in the marine environment can be driven by adaptation to localized freshwater environments (Johnson and Schindler 2013).

In addition to posing perhaps the greatest threat to global biodiversity in general (Thomas et al. 2004), climate warming poses a particular threat to riverine species whose ranges are largely confined to existing habitat (Thomas 2010). Among these species, salmon are strongly affected by climate warming (e.g. Patterson et al. 2007). Warmer water can lead to massive mortality of salmon populations (e.g. Patterson et al. 2007) and indirectly impact salmon productivity through alterations to snow-melt timing and extreme hydrological events (Crozier et al. 2008). Due to these effects, adverse stream temperatures are already impeding recovery of some Pacific salmon populations (McCullough 1999) and are expected to make recovery targets more difficult to achieve (Battin et al. 2007). However, despite the evidence that warming impacts salmon, salmon also show evidence of response diversity and local adaptation to temperature. For example, thermal tolerance of sockeye salmon in the Fraser River, British Columbia, Canada, varies within streams according to historical environmental conditions (Eliason et al. 2011).

Here we ask how portfolio theory can inform spatial approaches to prioritizing metapopulation conservation in a changing world. To answer this, we develop a salmon metapopulation simulation in which spatially-distributed thermal tolerance and patterns of short- and long-term climatic change drive population-specific productivity. We then implement scenarios that prioritize alternative sets of populations and evaluate the salmon portfolios along risk-return axes, as a financial portfolio manager might. We show that conserving a diversity of thermal tolerances buffers metapopulation risk given short-term climate forcing and ensures metapopulation persistence given long-term climate warming. We then show that dividing conservation among more populations buffers risk regardless of thermal-tolerance diversity or climate trend, but possibly at the expense of long-term growth rate and persistence when available habitat declines over time. We conclude that considering metapopulations through portfolio theory provides a useful additional dimension through which we can evaluate conservation strategies.

Methods

We developed a 100-year salmon metapopulation simulation model that includes both population dynamics and harvesting along with process, observation, and implementation uncertainty (Fig. 1). We tested different conservation scenarios under two kinds of environmental regimes (short-term climate variability and long-term climate change) and in cases where habitat capac-

ity remained constant or declined over time. We provide a package `metafolio` (Anderson 2014) for the statistical software R (R Core Team 2013), to carry out the simulations and analyses described in this paper (Supplement).

Defining the ecological portfolio

In our ecological portfolios, we defined assets as stream-level populations and portfolios as salmon metapopulations. The specific configuration of our model refers to salmon that spend extended time rearing in freshwaters (e.g. steelhead [*Oncorhynchus mykiss*], sockeye salmon [*O. nerka*], coho salmon [*O. kisutch*], and stream-type Chinook salmon [*O. tshawytscha*]), which will likely be more impacted by changes to stream temperature and flow (Mantua et al. 2010). We use the terms *stream* and *populations* interchangeably to represent the portfolio assets. We defined the portfolio investors as the stakeholders in the fishery and metapopulation performance. For example, the investors could be conservation agencies, First Nations groups, or civil society as a whole. The fisheries management agency then becomes the portfolio manager. We defined the asset value as the abundance of returning salmon in each stream and the value of the portfolio as the overall metapopulation abundance.

In this scenario, the equivalent to financial rate of return is the generation-to-generation metapopulation growth rate, calculated as the first difference of the log salmon returns. We defined the financial asset investment weights as the capacity of the stream populations — specifically the unfished equilibrium stock size — since maintaining or restoring habitat requires money, time, and resources and habitat size itself is a strong predictor of the occupancy of salmon (Isaak et al. 2007). Investment in a population therefore represents investing in salmon habitat conservation or restoration and the risk and return from investment strategies become emergent properties of our metapopulation model.

Salmon metapopulation dynamics

The salmon metapopulation dynamics in our simulation were governed by a spawner-return relationship with demographic stochasticity and straying between populations. We defined the spawner-return relationship with a Ricker model,

$$R_{i(t+1)} = S_{i(t)} e^{a_{i(t)}(1-S_{i(t)}/b_i)+w_{i(t)}}$$

where i represents a population, t a generation time, R the number of returns, S the number of spawners, a the productivity parameter (which can vary with the environment), and b the density-dependent term (which is used as the asset weights in the portfolios). The term $w_{i(t)}$ represents first-order autocorrelated error. Formally, $w_{i(t)} = w_{i(t-1)}\rho_w + r_{i(t)}$, where $r_{i(t)}$ represents independent and normally-distributed error with standard deviation of σ_r , mean of $-\sigma_r^2/2$ (bias corrected so the expected value after exponentiation is 1), and correlation between subsequent generation values of ρ_w . We set $\sigma_r = 0.7$ and $\rho_w = 0.4$ to match the mean values for salmonids in Thorson et al. (2014a).

We manipulated the capacity and productivity parameters b_i and $a_{i(t)}$ as part of the portfolio simulation. The capacity parameters b_i were controlled by the investment weights in the populations. For example, a large investment in a stream was represented by a larger unfished equilibrium stock size b for stream i . The productivity parameters $a_{i(t)}$ were controlled by the interaction between a temperature time series and the population thermal-tolerance performance curves. In a different context, investment could represent improving the productivity (a_i) parameters, say through culling, to offset mortality increases due to changing temperatures. However, such a scenario is unlikely in the case of an endangered species where population levels are often well below levels where culling would increase productivity.

We generated the thermal-tolerance curves according to

$$a_{i(t)} = \begin{cases} a_i^{\max} - W_i(e_t - e_i^{\text{opt}})^2, & \text{if } a_{i(t)} > 0 \\ 0, & \text{if } a_{i(t)} \leq 0 \end{cases}$$

where W_i controls the width of the curve for population i , e_t represents the environmental value at generation t , e_i^{opt} represents the optimal temperature for population i , and a_i^{\max} represents the maximum possible a value for population i . We set the W_i parameters (evenly spaced values increasing and decreasing between 0.08 and 0.04) to generate widths approximately as shown in Eliason et al. (2011). We set the area under each curve to 30 units to create a_i^{\max} values ranging roughly between 2.2 and 2.9 as in Dorner et al. (2008). These parameter values created some warm-tolerant populations, some cold-tolerant populations, and some populations with a wider range of thermal-tolerance but a lower maximum productivity (Fig. 2a). Although we refer to a thermal-tolerance curve because temperature is a dominant driver of salmon productivity (e.g. McCullough 1999; Patterson et al. 2007; Eliason et al. 2011), our model could apply to any environmental tolerance (e.g. tolerance to stream flow volume or changes in snow melt timing; Crozier et al. 2008).

We implemented straying as in Cooper and Mangel (1999). We arranged the populations in a line and salmon were more likely to stray to streams near their natal stream (Appendix A). Two parameters controlled the straying: the fraction of fish f_{stray} (0.02) that stray from their natal stream in any generation and the rate m (0.1) at which this straying between streams decays with distance

$$\text{strays}_{ij(t)} = f_{\text{stray}} R_{j(t)} \frac{e^{-m|i-j|}}{\sum_{\substack{k=1 \\ k \neq j}}^n e^{-m|k-j|}}$$

where $R_{j(t)}$ is the number of returning salmon at generation t whose natal stream was stream j . The subscript k represents a stream ID and n the number of populations. The denominator is a normalizing constant to ensure the desired fraction of fish stray. Our simulation did not account for the homogenization of diversity due to straying. For example, all salmon in one population maintained the same thermal-tolerance curve regardless of how many salmon it received from another stream.

Fishing

Our simulation used a simple set of rules to establish the exploitation rate of fisheries and the remainder left to spawn (escapement target). First, to establish a range of spawner-return values and to mimic the start of an open-access fishery, for the first 30 years we drew the fraction of fish harvested randomly from a uniform distribution between 0.1 and 0.9. We discarded these initial 30 years as a burn-in period. Then, every five years for the remaining 100 years of our simulation, we fitted a spawner-return function to the cumulative data for individual populations. The target escapement rate E_{tar} (a proportion per year) was set based on Hilborn and Walters (1992) as

$$E_{\text{tar}} = \frac{R}{b(0.5 - 0.07a)}$$

where R represents the return abundance and a and b represent the Ricker model parameters. The target harvest rate is then a function of returns and the escapement target ($H_{\text{tar}} = R - E_{\text{tar}}$). We included implementation uncertainty in the actual harvest rate H_{act} as a function of the target harvest rate and a beta distribution with location parameter α_h , shape parameter β_h , and standard deviation of σ_h (set to 0.1 as observed for similar data in Pestes et al. (2008)).

$$\alpha_h = H_{\text{tar}}^2 \left(\frac{1 - H_{\text{tar}}}{\sigma_h^2} - \frac{1}{H_{\text{tar}}} \right)$$

$$\beta_h = \alpha_h \left(\frac{1}{H_{\text{tar}}} - 1 \right)$$

$$H_{\text{act}} = \text{beta}(\alpha_h, \beta_h).$$

Environmental dynamics

Environmental dynamics typically have both short- and long-term fluctuations, such as annual variability and directional climatic warming. We evaluated portfolio performance under these two components separately in our initial scenarios and combined in our final scenario. We did not explicitly model a cyclical climate trend, such as the Pacific Decadal Oscillation, but the effect of such a trend would largely be a product of the short-term variability and long-term trend. We represented short-term dynamics $e_{\text{short}(t)}$ as a stationary first-order autoregressive process, AR(1), with correlation ρ_e (0.1)

$$e_{\text{short}(t)} = e_{t-1}\rho_e + d_t, d_t \sim N(\mu_d, \sigma_d^2)$$

where d_t represents normally distributed deviations of some mean μ_d and standard deviation σ_d . We set μ_d to 16 °C and σ_d to 2 °C, to approximately match the stream temperature variation in Eliason et al. (2011). We represented long-term environmental dynamics $e_{\text{long}(t)}$ as a linear shift in the temperature through time

$$e_{\text{long}(t)} = e_0 + \beta_e t$$

where e_0 represents the starting temperature up until the burn-in period ends and β_e represents the annual increase in temperature. We set $e_0 = 15$ °C and $\beta_e = 0.04$ °C/generation to obtain an increase in stream temperature of 4 °C over the next century (assuming one generation equals one year) ending at or above the optimum thermal optimum of all populations. This increase approximately matches predicted increases in stream temperature – relative to the 1980s, stream temperatures in the Pacific Northwest have already increased by approximately 0.2 °C/decade (Isaak et al. 2012), and are predicted to increase 2 to 5 °C by 2080 (Mantua et al. 2010).

We summarize the chosen parameter values in Appendix B. Combining salmon population dynamics, fishing, and environmental dynamics, we illustrate the components of an example

simulation in Fig. 3 and the effect of varying population, fishing, and environmental parameters from their base values on metapopulation abundance in Appendix C.

Conservation scenarios

Spatial conservation scenarios: We evaluated four spatial conservation scenarios (Fig. 2b–e). We conserved four populations ($b_i = 1000$) and set the unfished equilibrium abundance of the six remaining populations to near elimination ($b_i = 5$) at the start of the simulation. These reduced populations could still receive straying salmon but were unlikely to rebuild on their own to a substantial abundance. The four spatial scenarios we considered were:

1. Conserve a full range of thermal tolerances (conserve some cool-, some intermediate-, and some warm-tolerant populations; Fig. 2b).
2. Conserve the middle section of the metapopulation (conserve the most thermal-tolerant populations with the widest response curves; Fig. 2c).
3. Conserve the lower half of the metapopulation (conserve cool-tolerant populations; Fig. 2d).
4. Conserve the upper half of the metapopulation (conserve warm-tolerant populations; Fig. 2e).

Unknown thermal tolerances: In reality we rarely know precise levels of thermal response diversity. We therefore also considered cases where conservation was randomly assigned with respect to thermal tolerance but where conservation effort ($\sum_{i=1}^n b_i = 2000$) could be distributed across different numbers of streams. We considered conserving from two to 16 streams with thermal tolerance distributed along the same range as in the spatial scenarios. As in the spatial strategies, we reduced the capacity of the remaining streams to the nominal level of $b_i = 5$.

Declining habitat availability: Habitat capacity in the Pacific Northwest is likely shrinking over time as salmon populations are squeezed between warming temperatures reducing habitat from below and declining stream flows reducing the habitat that remains from above. For example, temperature isotherms are shifting upstream at 1–10 km/decade in low gradient streams that Chinook use for spawning (Isaak and Rieman 2013). At the same time, summer-fall stream flow volumes have been decreasing 10–30% across the Pacific Northwest over the past 50 years (Luce and Holden 2009) and are likely to continue declining (Luce et al. 2013). We therefore considered a scenario where habitat capacity declined by a constant amount across all populations. We reduced the b parameters by 0.85 units per generation so that some of the smaller populations

would reach near extinction by the end of the simulation, as is likely for smaller isolated populations within this century (e.g. Gustafson et al. 2007). In this scenario, we considered cases where thermal tolerance was unknown but conservation effort could be distributed across between 16 and two streams. Climate followed a combination of the same long-term warming and short-term variability as before. For many Pacific salmon metapopulations, this scenario represents the most realistic scenario investigated.

Results

Spatial conservation scenarios

Given short-term environmental fluctuations (strong interannual variation), conserving a wide range of thermal tolerances is the safest choice because it reduces overall risk to an ecological portfolio (Fig. 4a; Appendix D Figs. D1, D2). The average variance of metapopulation growth rate was 1.6 times lower given balanced thermal tolerance conservation (conserving a full range of thermal tolerances or the middle section vs. the upper or lower half). Thermal tolerance diversity also led to more consistent stability — there was less spread in variance across simulated metapopulations (width of quantiles from left to right in Fig. 4a). These increases in stability occurred despite the portfolios being comprised of warm- and cool-thriving populations that individually showed greater variation in response to environmental variability than populations with wide thermal tolerance curves. We can see the mechanism behind these portfolio properties by inspecting example population time series (Fig. 4c, d). If only the upper or lower half of thermal tolerances is conserved, the portfolio tends to alternate between performing well and poorly, depending on the environmental conditions, resulting in a riskier portfolio (Fig. 4e). This risk is buffered when a diversity of thermal tolerances is conserved (Fig. 4c) and the resulting asynchrony in population abundance (Appendix E).

Given long-term environmental change, such as climate warming, an ecological manager is hedging his or her bets on the environmental trend and how the populations will respond by conserving a range of thermal tolerances. The choice of which populations to conserve affects the “rate of return” (metapopulation growth rate) properties of an ecological portfolio (Fig. 4b; Appendix D Figs. D3, D4). The typical metapopulation growth rate when thermal tolerances were balanced was near zero — the metapopulation neither increased nor decreased in abundance in the long run. The example metapopulation abundance time series (Fig. 4d, f) illustrate the

mechanism: by conserving a range of thermal tolerances, when one population is doing poorly, another is doing well and the metapopulation abundance remains stationary through time. If a manager had invested only in the populations that were doing well at the beginning they would have had the lowest metapopulation growth rate at the end (purple portfolios in Fig. 4f).

Unknown thermal tolerances

In a scenario where the distribution of population-level thermal tolerances are unknown, portfolio optimization informs us that investing in more populations buffers portfolio risk regardless of environmental trend (Fig. 5). *Given short-term environmental fluctuations*, conserving more populations buffers portfolio risk (Fig. 5a, c, d; Appendix D Figs. D5, D6). For example, a metapopulation with 16 conserved populations is on average 1.7 times less variable than a metapopulation with only eight. At the same time, the random conservation of thermal tolerances creates an increased spread of possible metapopulation risk given fewer populations conserved (increasing quantile width from left to right in Fig. 5a).

Given long-term environmental change, conserving more populations also buffers portfolio risk (Fig. 5b; Appendix D Figs. D7, D8). Furthermore, in comparison to the short-term environmental noise scenario, the long-term environmental change creates a greater spread of possible metapopulation growth rates. For example, the height of the 75% quantile of the mean metapopulation growth rate for the two-population systems (light grey polygons) is larger given long-term change than short-term change.

Declining habitat availability

Given a reduction in stream flow over time along with climate change and climate variability, a manager encounters a risk-return trade-off when deciding how many populations to distribute conservation efforts across (Fig. 6; Appendix D Figs. D9, D10). Conserving more populations buffers portfolio risk, but at the expense of expected metapopulation growth rate. For example, the mean metapopulation variance was 2.7 times lower when 12 populations were conserved instead of four, but the expected metapopulation growth rate was 2.0 times lower when 16 populations were conserved instead of eight. The conservation scenarios represent an efficient frontier where a manager must choose whether to hedge his or her bets on a smaller number of populations and take on greater expected variability or conserve more populations and accept a lower expected metapopulation growth rate.

Discussion

The importance of conserving populations with a diversity of responses to the environment is a key assumption of conservation ecology, but has rarely been tested quantitatively (Mori et al. 2013). We show how maintaining populations with a variety of thermal tolerances reduces risk caused by short-term environmental stochasticity and optimizes chances for long-term persistence given climate change. Further, conserving more populations reduces metapopulation variability but possibly at the expense of long-term metapopulation growth rate if available habitat is squeezed by climate change. In this discussion, we begin by linking our model with real-world conservation issues for Pacific Northwest salmon. We then consider broader implications for metapopulation conservation of any species and ecological stability in general.

Implications for salmon conservation

Our results emphasize the importance of promoting ecological conditions that promote diversity of environmental response to the environment if stability is to be maintained in the face of environmental uncertainty. This suggests three clear conservation actions. First, since habitat heterogeneity can lead to local adaptation (e.g. Fraser et al. 2011), our results emphasize the need to maintain a diversity of salmon habitat (Rogers and Schindler 2008). Second, if conservation actions must be prioritized, then our model suggests we should focus on populations that aren't spatially contiguous to maximize diversity of response to the environment. Third, our results demonstrate the advantages of avoiding structures that artificially remove diversity of environmental response. For salmon, dams are a prominent example (McClure et al. 2008a). Dams can have a double impact whereby their introduction selectively eliminates a large swath of contiguous habitat, perhaps analogous to our upper- or lower-half scenarios in Fig. 4, and then mitigation approaches such as hatcheries can further reduce response diversity if not carefully managed (McClure et al. 2008b). In fact, salmon habitat lost to dams in the western U.S. has been biased towards warmer, drier, higher habitats (McClure et al. 2008a) and our findings suggest the resulting loss of warm-tolerant species may compound the risk to current metapopulations in the face of global warming.

The goals of existing salmon management structures in the western US and Canada support a portfolio conservation perspective. In the US, salmon populations are divided into Evolutionarily Significant Units (ESUs), groups of populations that are reproductively isolated and share a common evolutionary heritage, and finer-scale Viable Salmonid Populations (VSPs), populations

that are demographically independent of other populations over a 100-year time frame (McElhany et al. 2000). In Canada, the rough equivalent to the ESU is a Conservation Unit (CU), which consists of a group of salmon that are reproductively isolated and that if lost would be unlikely to recolonize in a reasonable time frame (DFO 2005). A salmon portfolio in our model could represent an ESU or CU and the lessons learned from our models are thus directly applicable to management guidelines in the Pacific Northwest. In fact, a number of VSP guidelines agree with our findings. For example, VSP guidelines suggest maintaining diversity in a variety of forms, focusing conservation efforts not just where salmon are currently abundant, and maintaining metapopulations with some populations near each other and others further apart (McElhany et al. 2000).

However, salmon populations in the Pacific Northwest are already heavily impacted (e.g. Gustafson et al. 2007) and VSP and CU recovery goals have not yet been achieved for most populations. Since European-Americans arrived, 29% of 1400 historical salmon populations in the Pacific Northwest and California have been lost (Gustafson et al. 2007). Furthermore, 44% of salmon habitat in the western US (in the lower 48 states) has been lost to dams and other freshwater blockages (McClure et al. 2008a). Changes to habitat, combined with increasing climate variability, has led to disturbance regimes that differ substantially in the frequency, magnitude, and duration from historical patterns, and threaten the resilience of salmon populations (Waples et al. 2009). Many remaining populations rely on hatcheries for long-term population viability — creating substantial evolutionary risks such as outbreeding depression, genetic homogenization, reduced effective population size, and domestication of fish (adaption to artificial environments and reduced fitness in wild environments) (McClure et al. 2008b). Reduction of long-term reliance on hatcheries, accompanied by habitat restoration through, for example, restoring connectivity of floodplains and stream flow regimes, remains a critical component of long-term salmon sustainability in the Pacific Northwest — particularly given predicted patterns of climate change (Beechie et al. 2013).

Our model complements other simulation-based salmon-habitat prioritization models. While these other models tend to focus on detailed assessment of individual fish stocks, our model is the first to consider the role of response diversity in buffering risk for metapopulations as a whole. The Shiraz model is one complementary prioritization scheme (Scheuerell et al. 2006). It focuses on detailed conditioning of the habitat-population-dynamics relationship at multiple life-history stages for a single salmon population. Whereas the Shiraz model can be applied to an entire watershed, it combines the populations together as a single unit thereby ignoring the role of

population-level environmental response diversity. A second salmon prioritization model proposes combining population viability measures with an assessment of the genetic consequences of losing particular populations (Allendorf et al. 1997). This model, however, also focuses on the assessment of individual stocks without considering their covariance and therefore the performance of the salmon portfolio as a whole. Our model does not replace these prioritization schemes. Rather, it proposes an additional focus on prioritization that optimizes metapopulation growth *and* risk and that considers diversity of tolerance to environmental conditions.

While our model captures many relevant aspects of salmon life history and environmental dynamics, it ignores others that could be investigated in future analyses and might improve our understanding of salmon portfolio conservation. First, some salmon populations, such as ocean-type Chinook, tend to spawn further downstream than stream-type salmon. Ocean-type Chinook may therefore be less affected by declining stream flow and be able to shift upstream to avoid shifting isotherms (Mantua et al. 2010). A model could consider evolutionary adaptation by having populations adopt more ocean-type-like characteristics. Second, our model ignores lost thermal-tolerance diversity from populations that reach low population sizes and are reestablished by straying from nearby streams. An individual-based model might more accurately penalize for this lost diversity and emphasize the need to define lower limits on the investment weights in a salmon conservation portfolio. Third, our model ignores fine-scale within-stream spatial and temporal environmental fluctuations. Fine-scale extremes in temperature and stream flow may be particularly important to population dynamics (Mantua et al. 2010) and could be incorporated into a future analysis. Such a model might show an increased benefit of portfolio optimization if the impact of increased magnitude and frequency of local climate extremes is important in addition to the mean trend (Jentsch et al. 2007).

Broad ecological implications and conservation priorities

To promote the stabilizing effect of a diversified ecological portfolio, there are two key components to identify: (1) the environmental drivers to which a varied response might occur, and (2) the conservation actions that can increase or decrease the diversity of response. A third component, identifying the traits and behaviours that mediate population responses to the environment may provide further insight into the mechanisms. Environmental drivers of response can include, for example, changes to temperature, habitat availability, air quality, water chemistry, or extreme weather (Elmqvist et al. 2003). Identifying conservation actions that promote environmental response diversity is critical to developing stable ecological systems (Mori et al. 2013).

However, merely measuring environmental response diversity in real ecological systems is challenging (albeit possible; Thibaut et al. 2012). Therefore, one realistic solution may be to create general guidelines from a small number of intensively-monitored systems in which we can associate changes in synchrony of populations with changes in conservation regimes (e.g. Moore et al. 2010; Carlson and Satterthwaite 2011). Another solution may be to monitor the diversity of environmental conditions themselves (e.g. temperature, stream flow, and gravel size in the case of salmon) since we know that traits affecting response to environmental conditions are heritable and are likely to adapt to local conditions (Carlson and Satterthwaite 2011) possibly producing diversity of response to subsequent disturbances.

We suggest a number of specific extensions to our simulation model. First, the environment-thermal-tolerance mechanism could be expanded — the distribution of environmental tolerance across a metapopulation does not necessarily follow a linear gradient, different forms of environmental tolerance could interact, and environmental conditions could affect populations through mechanisms other than productivity. Second, in addition to other taxa, our model could be extended to ecological communities or meta-communities after accounting for species interactions. Third, without any modifications, our model could consider the Moran or environmental-filter concept whereby populations experience increasingly different environmental forces at further distances (Schindler et al. 2008; Rogers and Schindler 2008). Fourth, a model could consider the contribution of contemporary evolution (Stockwell et al. 2003). These rapid adaptations to changes in the environment could strongly influence portfolio performance and emphasize the importance of maintaining genetic diversity and a variety of local habitat. Finally, our model could be conditioned on a system of interest — say a particular river basin in our example — and the metapopulation portfolio could be optimized across conservation and restoration options as part of a formal decision analysis.

Management decisions for exploited species often come with a trade-off between conservation and revenue generation. Our findings when habitat capacity declined over time illustrate another kind of trade-off more similar to the trade-off described by Markowitz (1952) in his seminal financial portfolio work. In this case, managers must navigate a trade-off between expected risk and return of the metapopulation/portfolio growth rate itself. No position along this trade-off is inherently better than another unless considered in the context of societal values. Does society value short-term stability or a greater assurance of long-term persistence? The optimal choice likely lies somewhere in the middle and parameterizing our model to a specific metapopulation could illustrate the nature of the trade-off and aid conservation decision making. However, if

environmental tolerance could be targeted for conservation as in Fig. 4, a manager could likely achieve portfolios closer to the efficient frontier in Fig. 6. In other words, a manager could achieve a lower expected variance for the same expected growth rate or a higher expected growth rate for the same expected variance — a better conservation outcome in either case.

Conservation planning is inherently a spatial activity (Pressey et al. 2007) and our results can inform how we approach spatial conservation planning. First, our results suggest focusing on conserving the processes and mechanisms underlying stability, not just biodiversity itself (Pressey et al. 2007; Beechie et al. 2013). In particular, our results suggest that response diversity should be a mainstream element of conservation, not just species and functional diversity (Mori et al. 2013). Our analysis also illustrates how conserving a portfolio of populations, ideally selected for a wide range of environmental tolerance, can help integrate across environmental uncertainty when spatial planning (Ando and Mallory 2012). This is particularly important given the uncertainty surrounding the future ecological responses to climate change (Walther et al. 2002). Finally, the increasing rapidness and variability of environmental change necessitates a dynamic approach in which spatial planning is reevaluated at regular intervals (Hannah et al. 2002) — perhaps testing for changes in population and species asynchrony in addition to changes in local productivity and variability. Combined, our results detail a pathway through which population diversity in environmental tolerance can underpin the stability of ecological systems. This pathway highlights that diverse natural portfolios may be critical for the conservation of metapopulations in the face of increasing climate variability and change.

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Supplemental Material

SUPPLEMENT Recreating the analyses with the `metafolio` R package.

APPENDIX A. An example straying matrix.

APPENDIX B. Simulation input parameters and default values.

APPENDIX C. Sensitivity illustration with alternative parameter values.

APPENDIX D. Example simulated time series from alternative conservation scenarios.

APPENDIX E. An illustration of the correlation between populations.

Figure legends

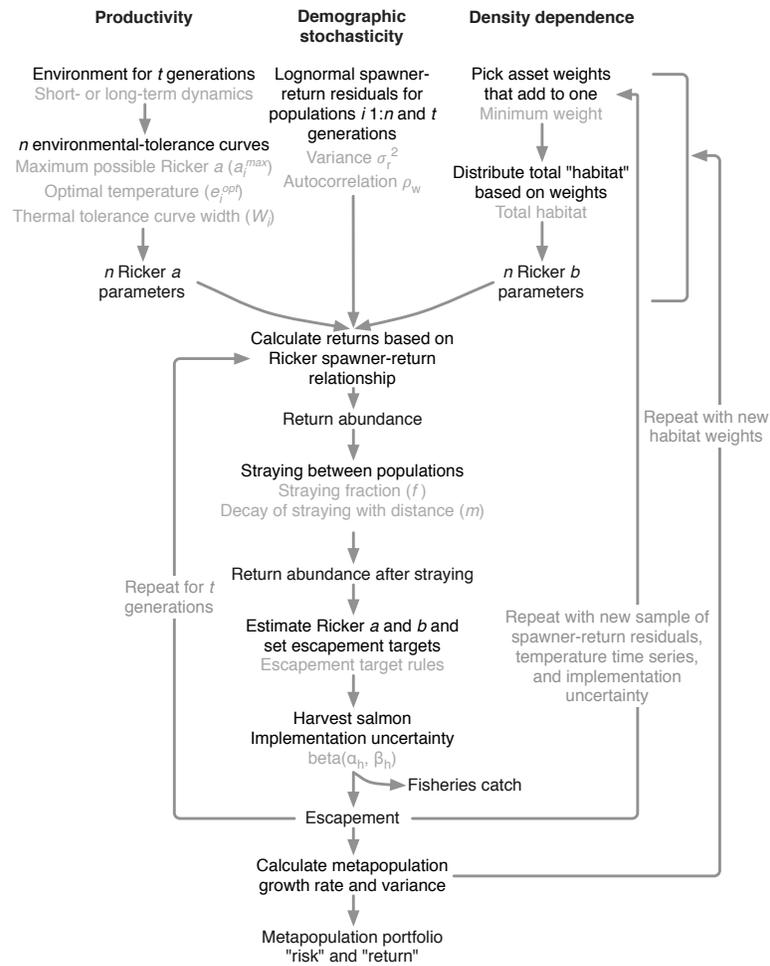


FIG. 1. Flow chart of the salmon-metapopulation simulation.

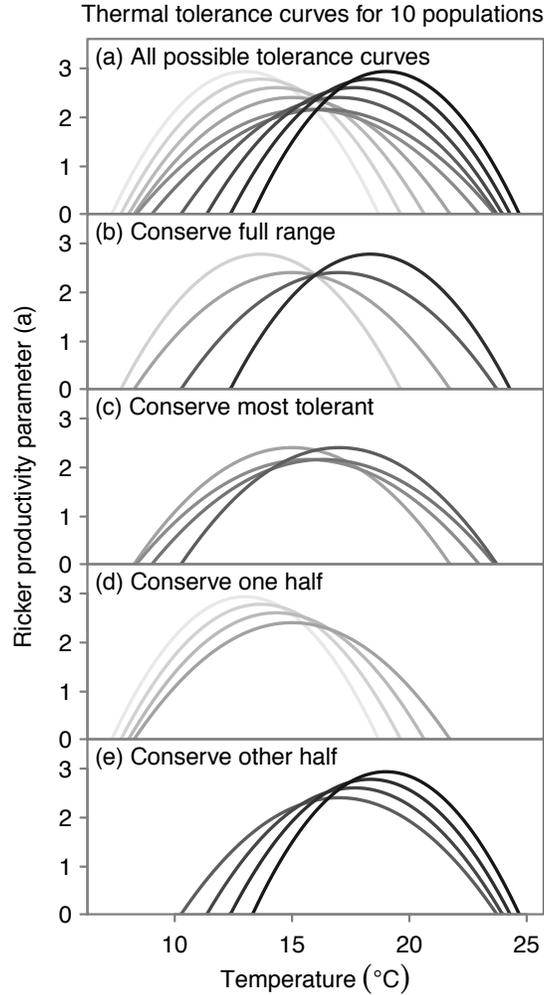


FIG. 2. Different ways of prioritizing thermal-tolerance conservation. Panel a shows thermal-tolerance curves for ten possible populations and panels b–e show different ways of prioritizing four of those populations. The curves describe how productivity varies with temperature for a given population. Some populations thrive at low temperatures (light greys) and some at warm temperatures (dark greys). Some are tolerant to a wider range of environmental conditions (mid greys) but with a lower maximum productivity. The total possible productivity (the area under the curves) is the same for each population.

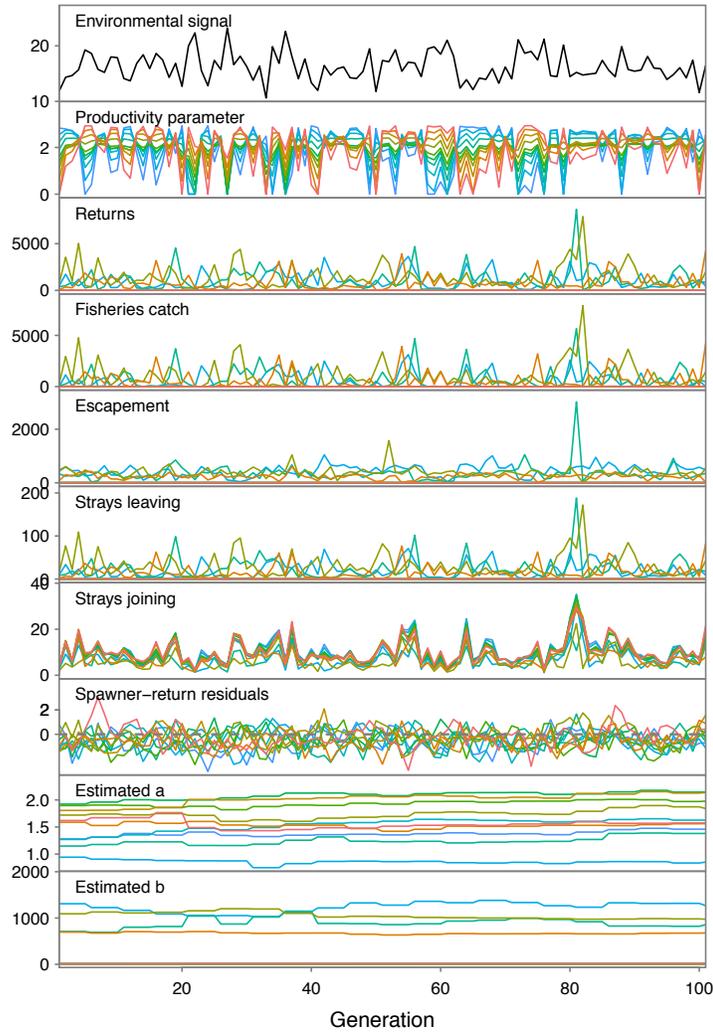


FIG. 3. The components of an example metapopulation simulation. We show, from top to bottom, the temperature signal, the resulting productivity parameter (Ricker a), the salmon returns, fisheries catch, salmon escapement, salmon straying from their natal streams, salmon joining from other streams, spawner-return residuals on a log scale, and the estimated a and b parameters in the fitted Ricker curve. The coloured lines indicate populations that thrive at low (blue) to high (red) temperatures.

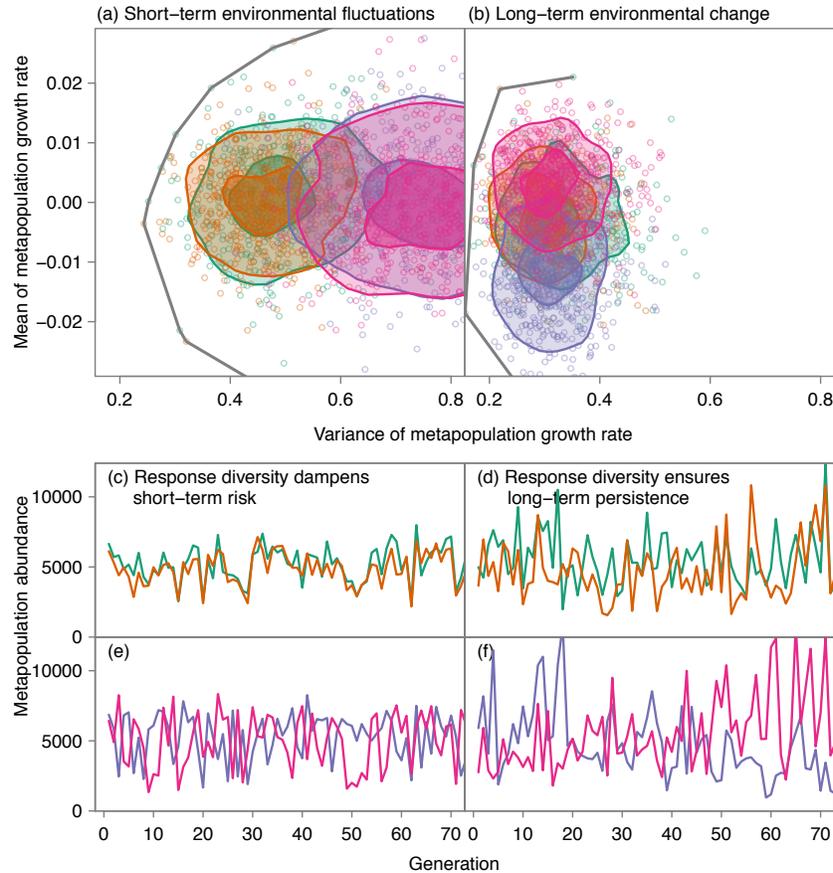


FIG. 4. The importance of preserving thermal-tolerance diversity through spatial conservation strategies. The conservation strategies correspond to figure 2 and represent conserving a range of responses (green), the most stable populations only (orange), or one type of environmental response (purple and pink). In risk-return space we show environmental scenarios that are comprised primarily of (a) short-term and (b) long-term environmental fluctuations. The dots show simulated metapopulations and the contours show 25% and 75% quantiles across 500 simulations per strategy. We also show example metapopulation abundance time series for the (c, e) short-term and (d, f) long-term environmental-fluctuation scenarios. The thick grey line (a, b) indicates the efficient frontier across all simulated metapopulations – metapopulations with the minimum variability for a given level of growth rate.

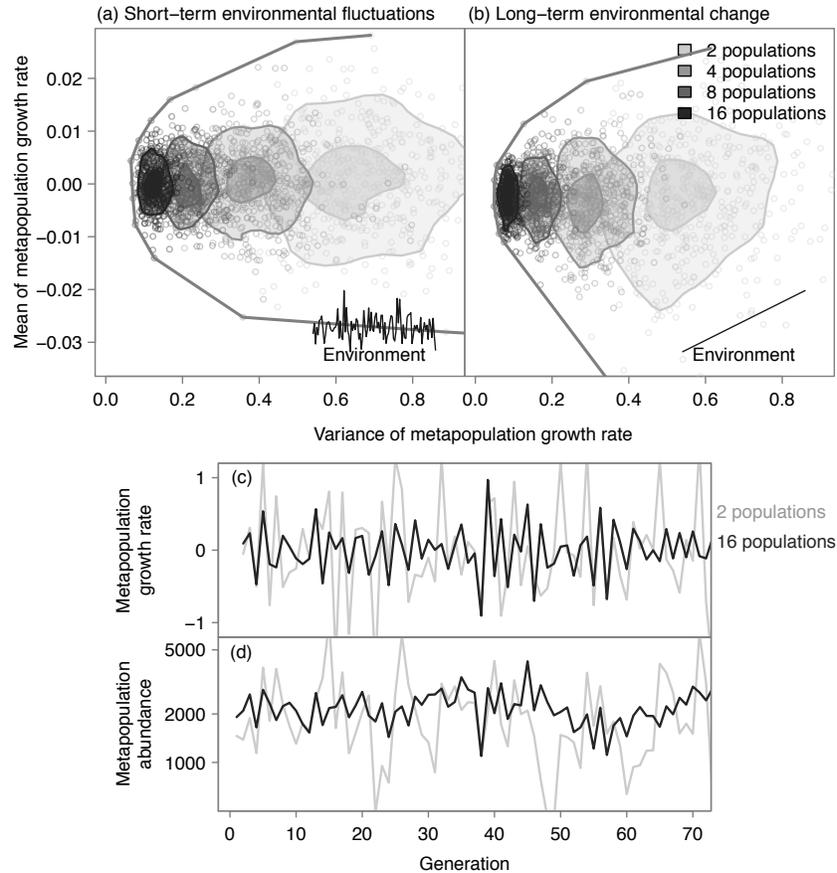


FIG. 5. The importance of preserving as many populations as possible when we do not know how thermal-tolerance is distributed. In risk-return space we show environmental scenarios that are comprised primarily of (a) short-term and (b) long-term environmental fluctuations. We show metapopulations in which two through 16 populations are conserved. The dots show simulated metapopulations and the contours show 25% and 75% quantiles across 500 simulations per strategy. We also show example metapopulation (c) rate-of-change and (d) abundance time series for the short-term environmental-fluctuation scenario. The thick grey line (a, b) indicates the efficient frontier across all simulated metapopulations – metapopulations with the minimum variability for a given level of growth rate.

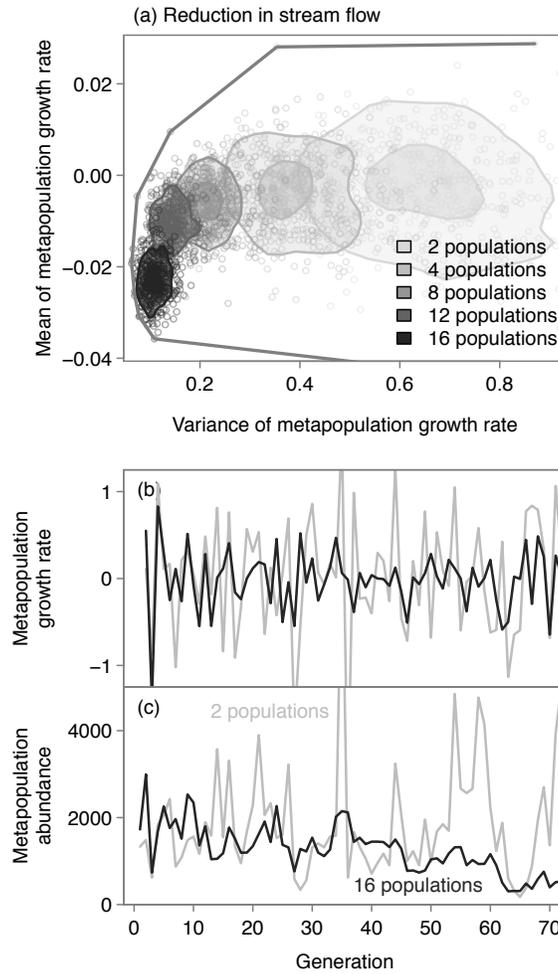


FIG. 6. Risk-return trade-off in the case where habitat is lost over time through stream flow reduction. The temperature follows both short-term fluctuations and a long-term increase. Thermal tolerance is randomly conserved. Shading indicates conservation plans where two through 16 populations are conserved. (a) Conserving more populations decreases expected variance but also decreases expected growth rate. Dots show simulated metapopulations and contours show 25% and 75% quantiles across 500 simulations per strategy. The thick grey line indicates the efficient frontier across all simulated metapopulations — metapopulations with the minimum variability for a given level of growth rate. Also shown are (b) example metapopulation growth rate and (c) abundance time series from the 2 and 16 population scenarios.

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APPENDIX A. An example straying matrix.

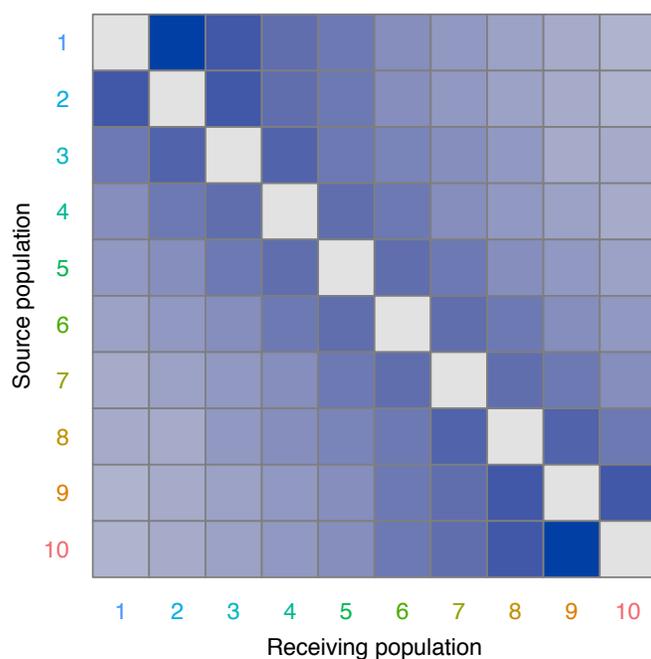


FIG. A1. An example straying matrix. The rows and columns represent different populations (indicated by population number). Dark blue indicates a high rate of straying and light blue indicates a low rate of straying.

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APPENDIX B. Simulation input parameters and default values.

Table B1: Input parameters to the salmon metapopulation simulation with default values.

Description	Symbol	Value	Reference
<i>Population dynamics parameters</i>			
Stock-recruit residual standard deviation (on log scale)	σ_r	0.7	Thorson et al. 2014
AR(1) serial correlation of stock-recruit residuals	ρ_w	0.4	Thorson et al. 2014
Fraction of fish that stray from natal streams	f_{stray}	0.02	Quinn 2005 and references therein
Exponential rate of decay of straying with distance	m	0.1	Cooper and Mangel 1999
Range of maximum productivities	a_i^{max}	2.2–2.9	Dorner et al. 2008
<i>Environmental parameters</i>			
Width parameter for thermal-tolerance curves for populations i 1 to n (values generate widths in line with listed references)	W_i	0.08–0.04–0.08	Brett 1952; Eliason et al. 2011
Optimum environmental value for populations i 1 to n	e_i^{opt}	13–19	Eliason et al. 2011
Standard deviation of annual temperature fluctuations	σ_d	2	Eliason et al. 2011
AR(1) autocorrelation of annual temperature fluctuations	ρ_e	0.1	
Annual increase in stream temperature in degrees Celcius	β_e	0.04	Mantua et al. 2010
<i>Fishery parameters</i>			
Standard deviation of beta distribution for implementation error	σ_h	0.1	Pestes et al. 2008
Frequency of assessment (years)	f_{assess}	5	

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APPENDIX C. Sensitivity illustration with alternative parameter values.

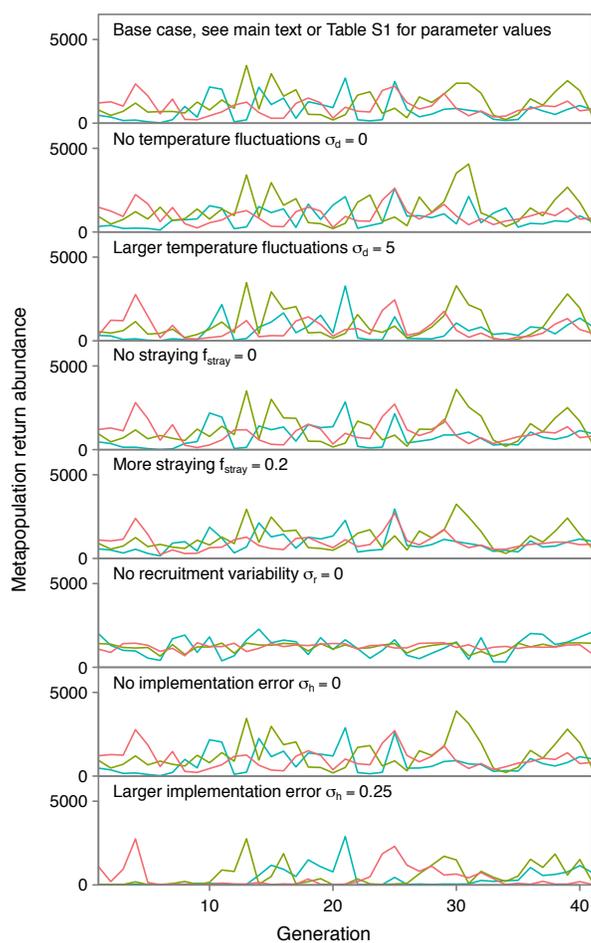


FIG. C1. The impact of increasing or decreasing various parameter values on metapopulation return abundance. The different coloured lines represent three example salmon populations. The base case represents the base-case values for the short-term environmental fluctuation scenario.

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APPENDIX D. Example simulated time series from alternative conservation scenarios.

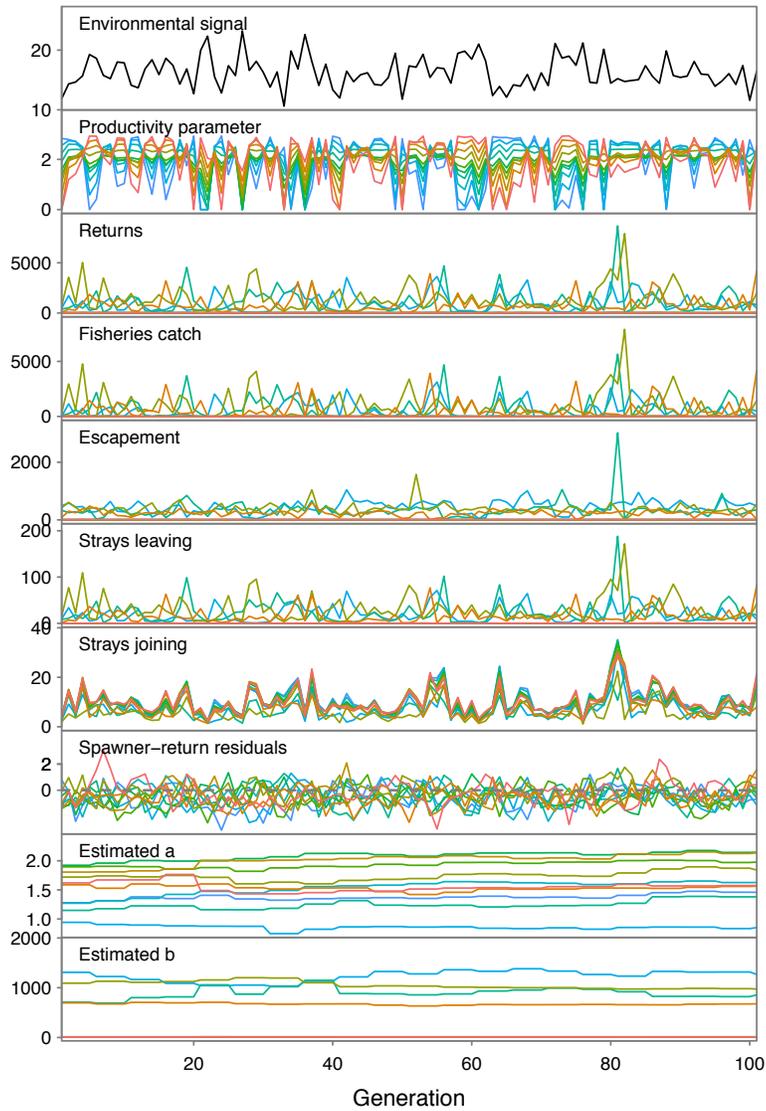


FIG. D1. Conserving a **full range** of response diversity (spatial conservation strategy) with **short-term** environmental fluctuations. This is the same as Fig. 3 but in colour.

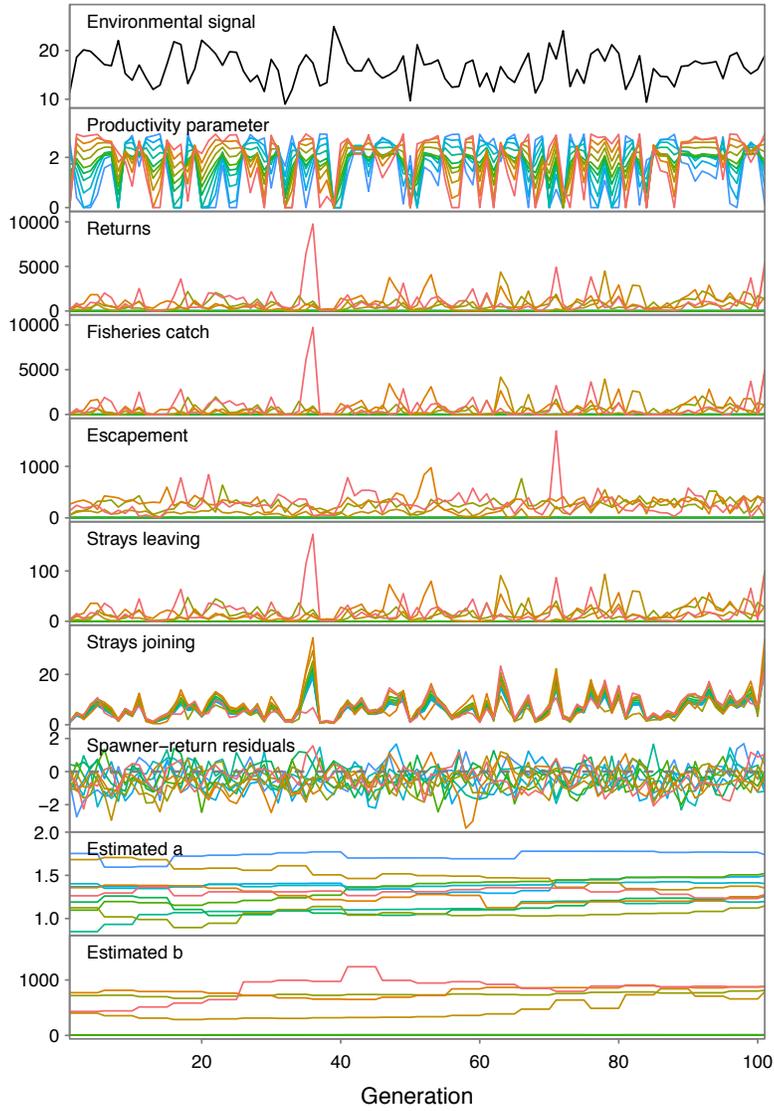


FIG. D2. Conserving **one half** of response diversity (spatial conservation strategy) with **short-term** environmental fluctuations.

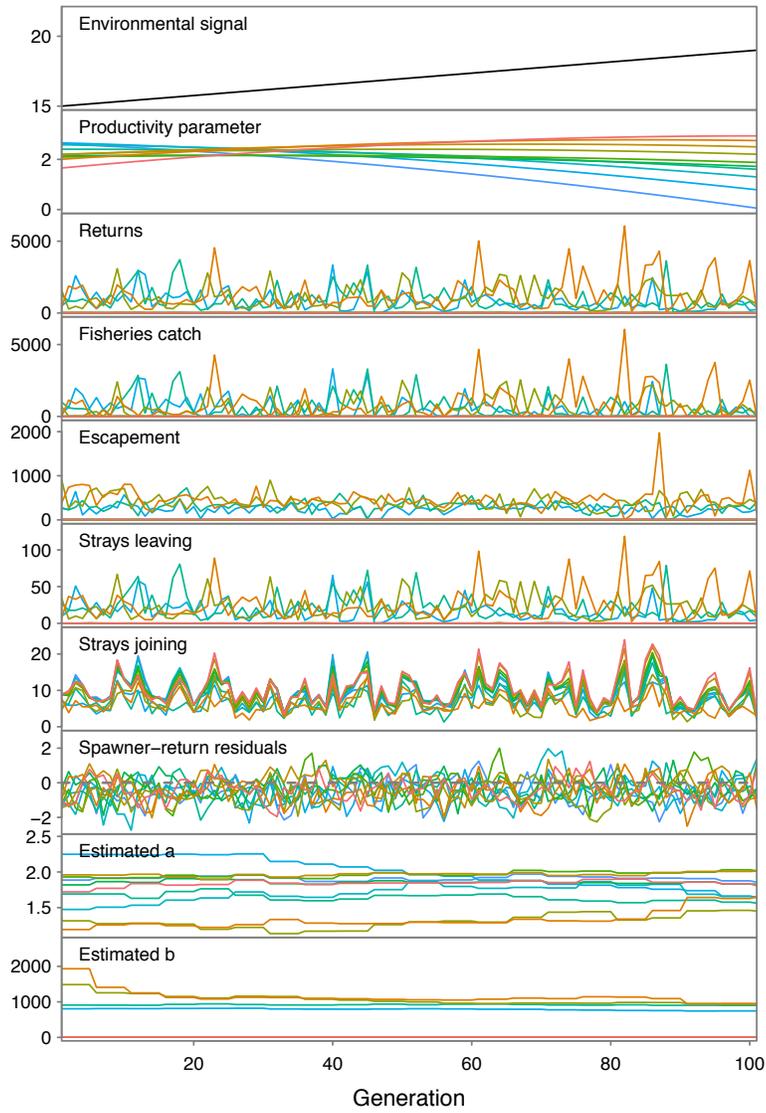


FIG. D3. Conserving a **full range** of response diversity (spatial conservation strategy) with **long-term** environmental change.

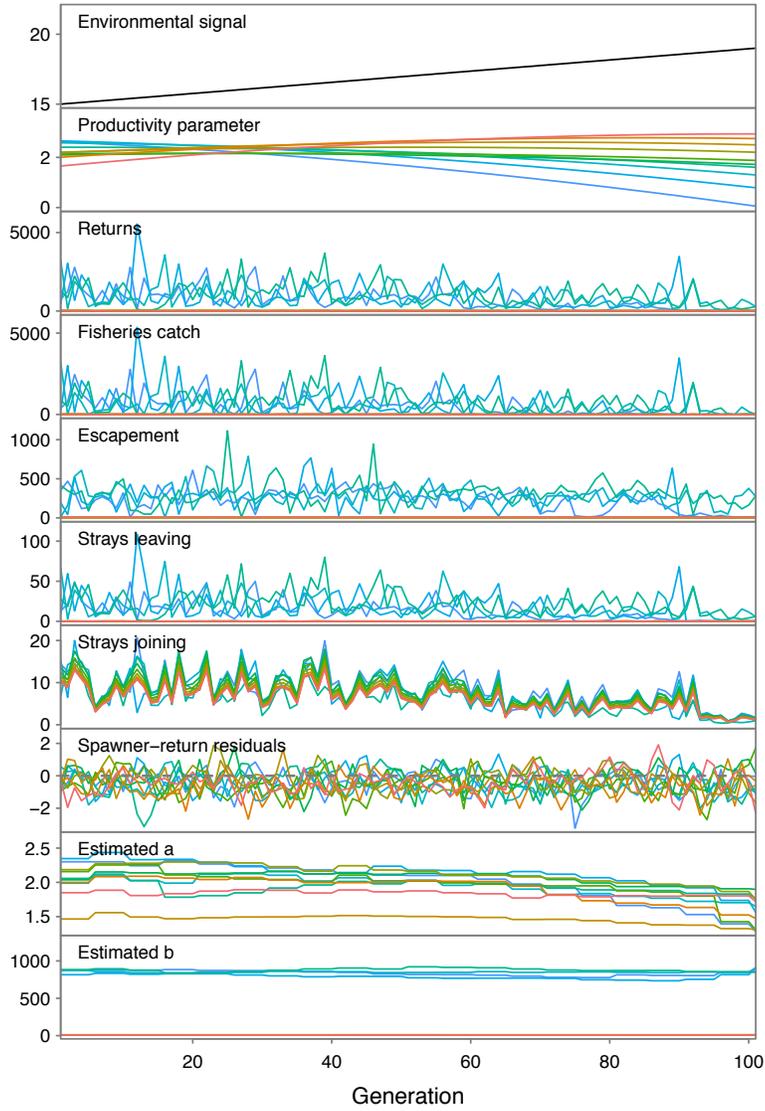


FIG. D4. Conserving **one half** of response diversity (spatial conservation strategy) with **long-term** environmental change.

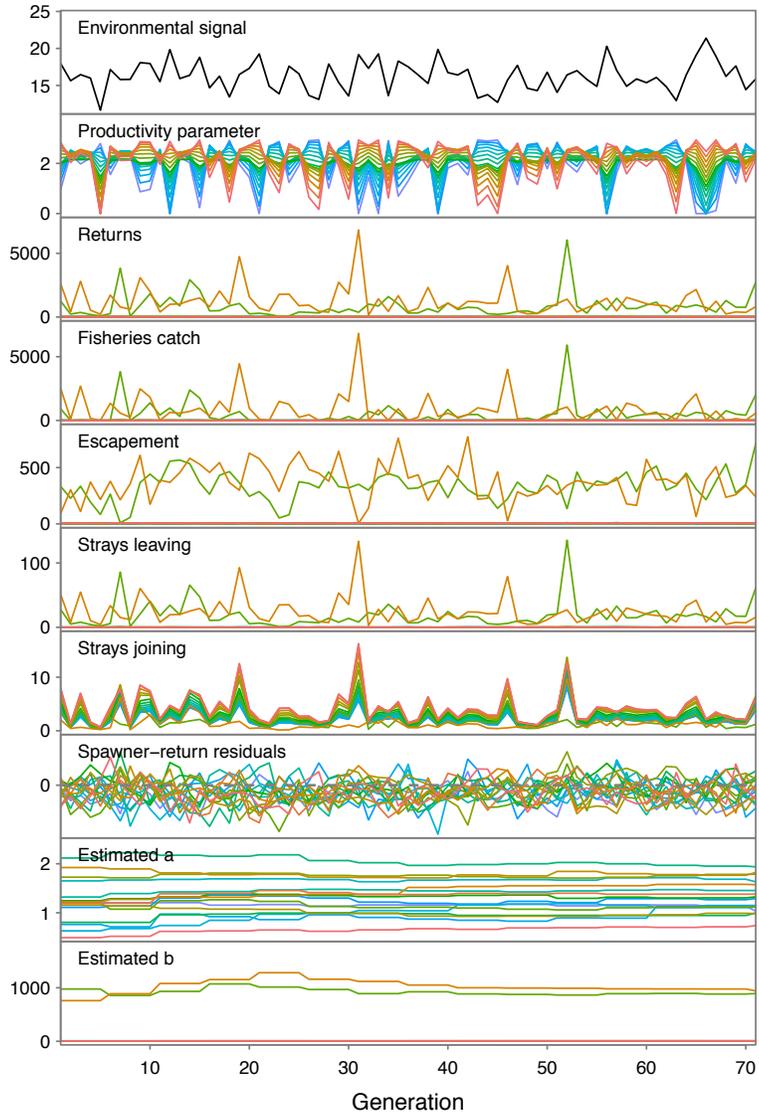


FIG. D5. **Two populations** conserved with random response diversity and **short-term** environmental fluctuations.

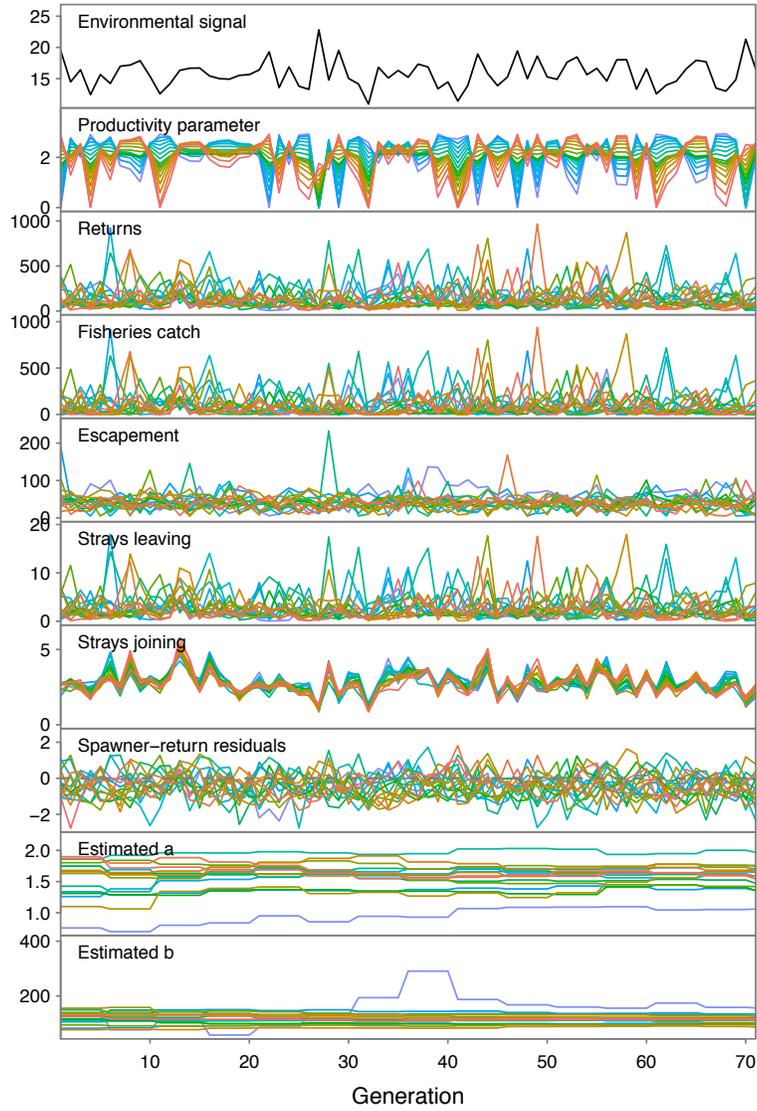


FIG. D6. **Sixteen populations** conserved with random response diversity and **short-term** environmental fluctuations.

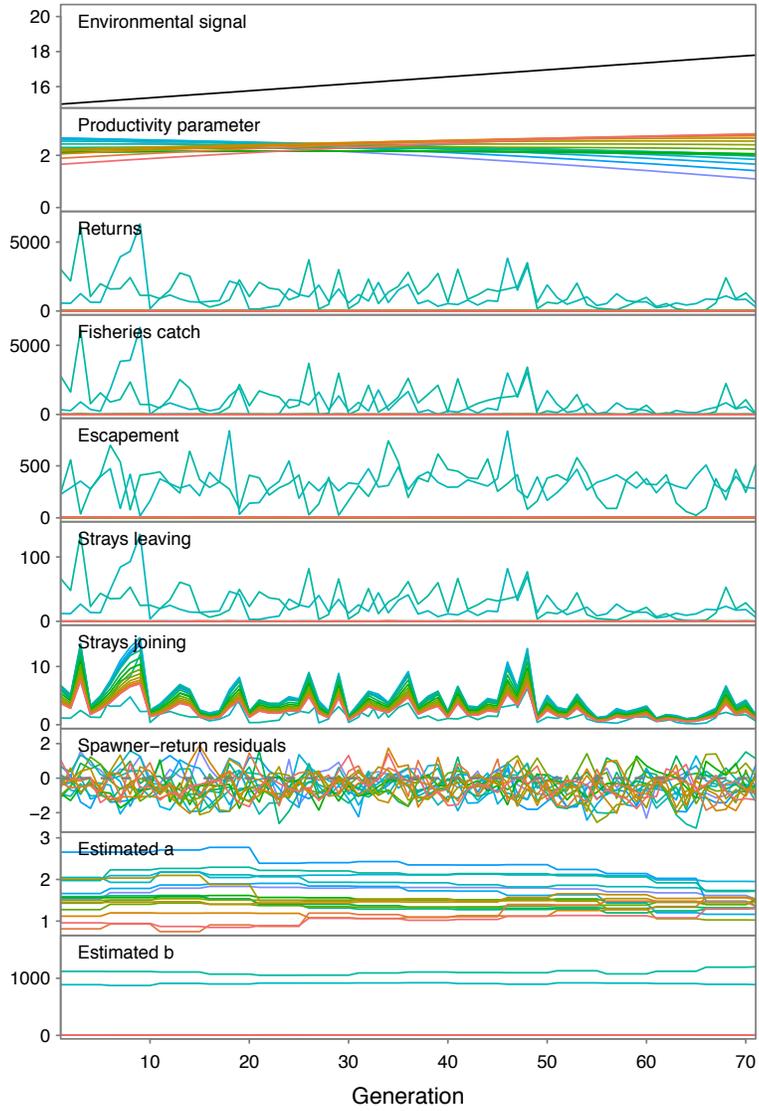


FIG. D7. **Two populations** conserved with random response diversity and long-term environmental change.

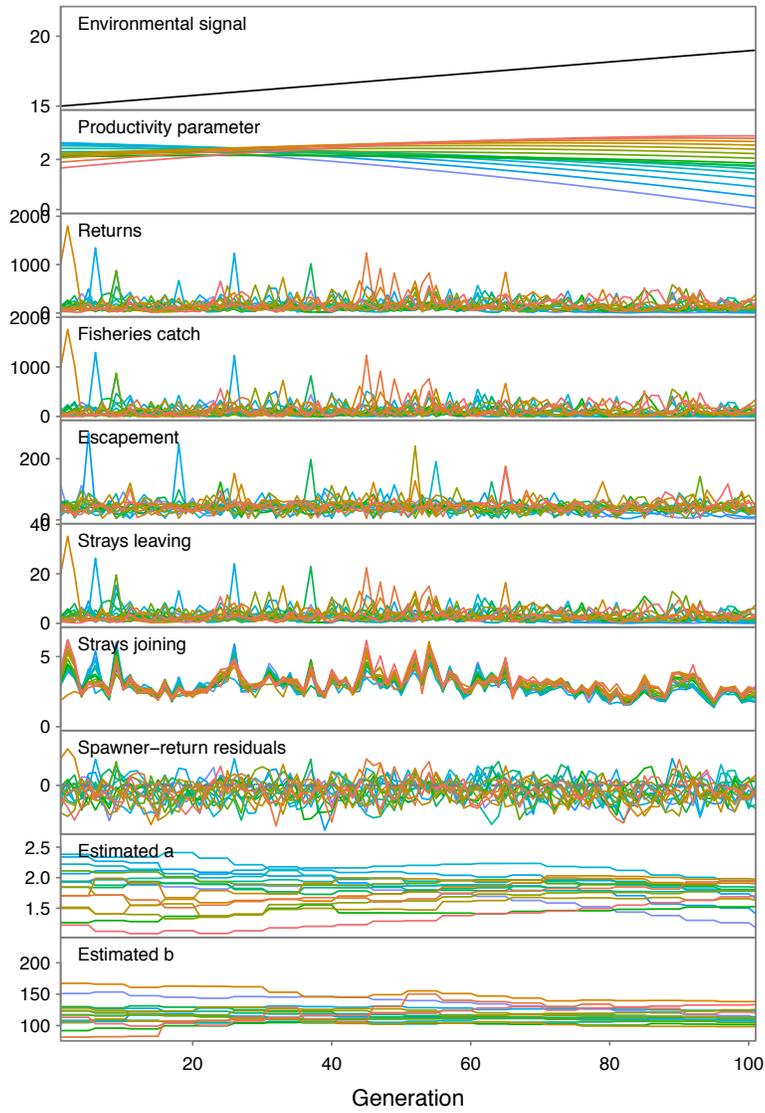


FIG. D8. **Sixteen populations** conserved with random response diversity and **long-term** environmental change.

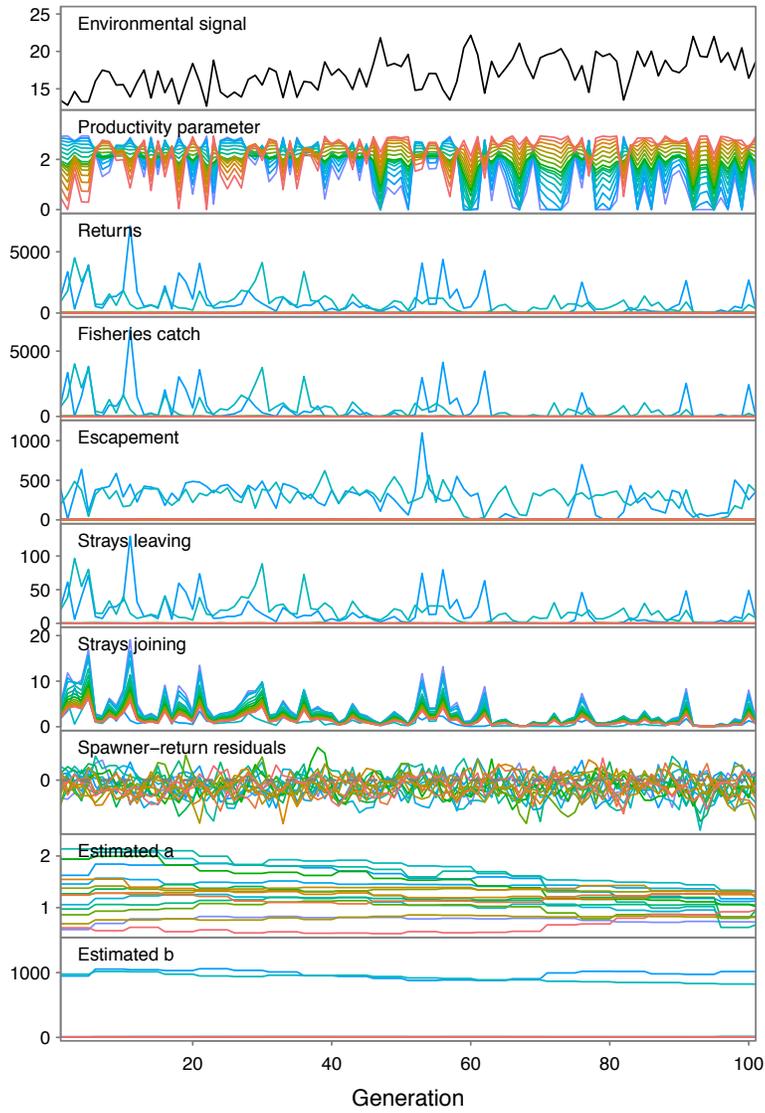


FIG. D9. **Two populations** conserved with random response diversity and long-term declining stream flow.

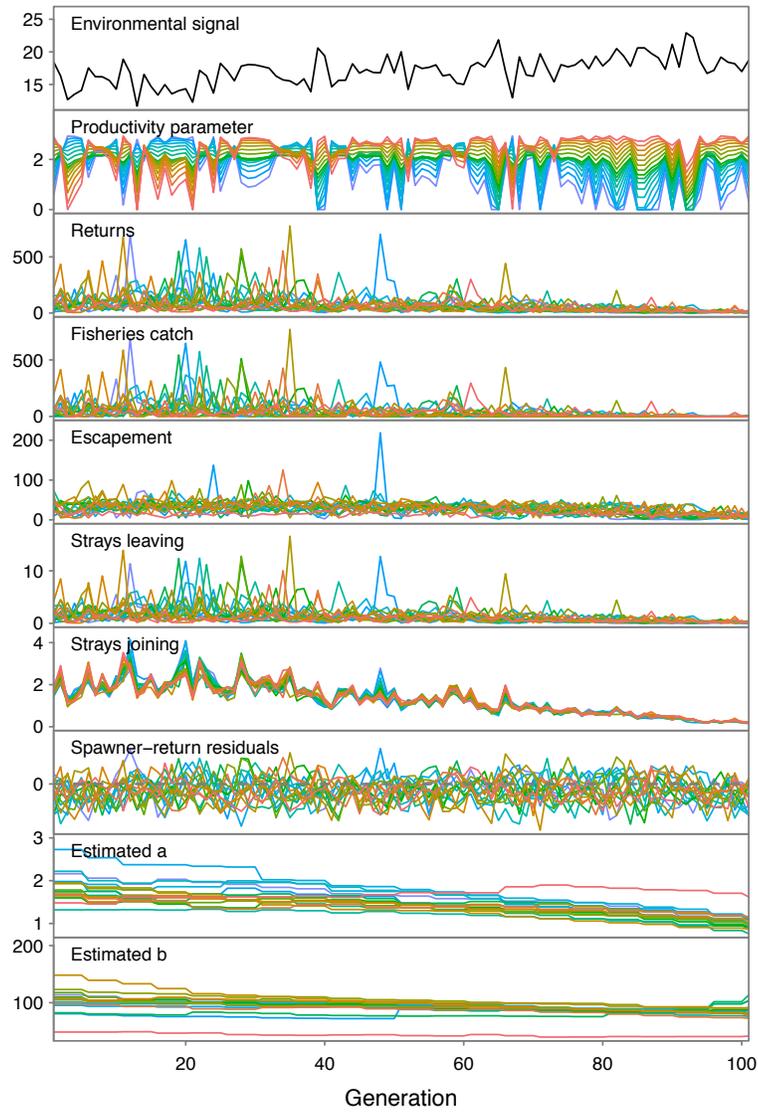


FIG. D10. **Sixteen populations** conserved with random response diversity and **long-term declining stream flow**.

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APPENDIX E. An illustration of the correlation between populations.

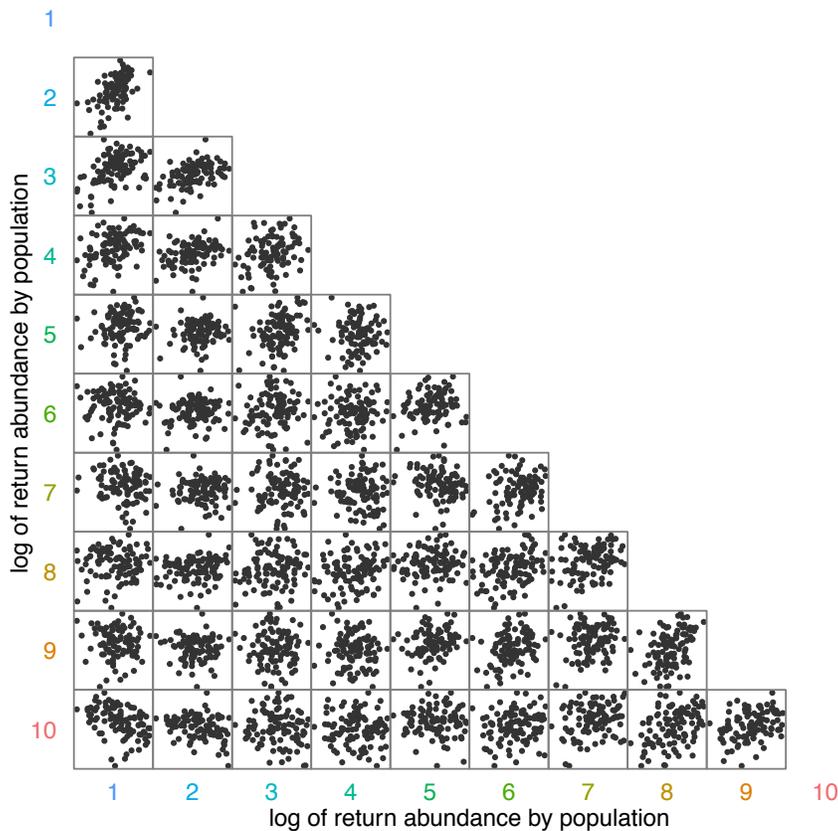


FIG. E1. A comparison of the $\log(\text{returns})$ between populations. The sub-population IDs are coloured from warm tolerant (warm colours) to cool tolerant (cool colours). Note how populations 1 and 10 have asynchronous returns whereas populations with more similar thermal-tolerance curves (say populations 9 and 10) have more synchronous dynamics. Populations with thermal tolerance curves in the middle (e.g. population 6) are less correlated with other populations. Their population dynamics end up primarily driven by demographic stochasticity and less so by temperature-induced systematic changes in productivity.

Recreating the analyses with the `metafolio` R package

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1 Loading the package

The `metafolio` R package is a tool to simulate metapopulations and apply financial portfolio optimization concepts to those metapopulations. The package is primarily intended for salmon simulations, but could be adapted for other taxonomic groups. In this document, we describe how to install and load the package, and run the analyses in our paper.

`metafolio` is available on CRAN: <http://cran.r-project.org/package=metafolio> and can be installed with:

```
install.packages("metafolio")
```

The analyses in this paper used `metafolio` version 0.1.0.

Alternatively, you can view the code and install the package from <http://github.com/seananderson/metafolio>. Installing the package from GitHub will require that you have a C++ compiler installed.

Load the package with:

```
library("metafolio")
```

The included vignette describes the package and illustrates some example simulations. You can view the vignette with:

```
vignette("metafolio")
```

You can view the help for the package with:

```
?metafolio  
help(package = "metafolio")
```

The figures from this paper can be re-created by downloading the source code from CRAN or GitHub and sourcing the file `README.R` in the `inst/examples` folder:

```
setwd("metafolio/inst/examples")  
source("README.R")
```

We have copied some examples from that code below. See the source code on GitHub for the full analysis. To save time compiling this document, we have run fewer iterations in these examples than in the paper.

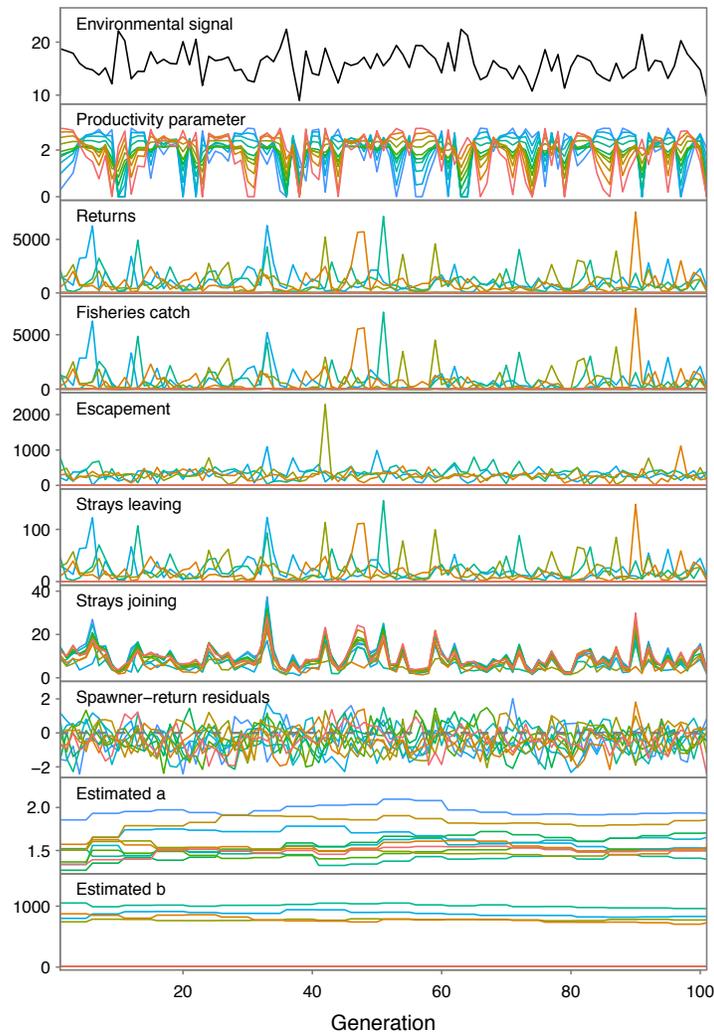
2 Prioritizing spatial response diversity

Setup the scenarios:

```
set.seed(1)  
w_plans <- list()  
w_plans[[1]] <- c(5, 1000, 5, 1000, 5, 5, 1000, 5, 1000, 5)  
w_plans[[2]] <- c(5, 5, 5, 1000, 1000, 1000, 1000, 5, 5, 5)  
w_plans[[3]] <- c(rep(1000, 4), rep(5, 6))  
w_plans[[4]] <- rev(w_plans[[3]])  
plans_name_sp <- c("Full response range", "Most stable only",  
  "Lower half", "Upper half")  
n_trials <- 200 # number of trials at each conservation plan  
num_pops <- c(10, 10, 10, 10)  
n_plans <- length(num_pops) # number of plans  
w <- list()  
for(i in 1:n_plans) { # loop over plans  
  w[[i]] <- list()  
  for(j in 1:n_trials) { # loop over trials  
    w[[i]][[j]] <- matrix(w_plans[[i]], nrow = 1)  
  }  
}
```

Here's one example of generating a time series plot:

```
set.seed(123)
arma_env_params <- list(mean_value = 16, ar = 0.1, sigma_env = 3, ma = 0)
eg_arma <- meta_sim(b = w[[1]][[1]], n_pop = 10, env_params = arma_env_params,
  env_type = "arma", assess_freq = 5)
plot_sim_ts(eg_arma, years_to_show = 100, burn = 30)
```



Now, we'll run the simulations:

```
x_arma_sp <- run_cons_plans(w, env_type = "arma", env_params =
  arma_env_params, show_progress = FALSE)
x_arma_sp$plans_port <- NULL # save space
```

```

linear_env_params <- list(min_value = 15, max_value = 19, sigma_env = 0.001,
  start_t = 30)
x_linear_sp <- run_cons_plans(w, env_type = "linear", env_params =
  linear_env_params, max_a = thermal_integration(10),
  show_progress = FALSE)
x_linear_sp$plans_port <- NULL # save space

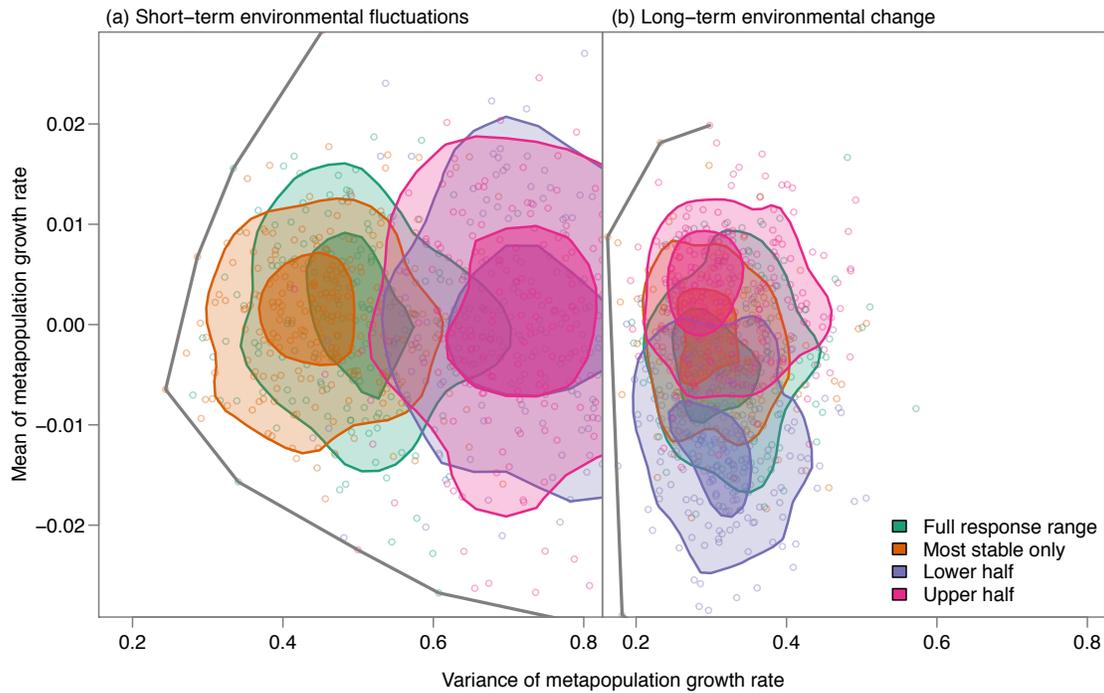
```

And plot the output:

```

cols <- RColorBrewer::brewer.pal(5, "Dark2")
xlim <- c(0.18, 0.80)
ylim <- c(-0.027, 0.027)
par(mfrow = c(1, 2))
par(las = 1, cex = 0.8, mar = c(0, 0, 0, 0), oma = c(4, 5.2, 1.5, .5),
  tck = -0.02, mgp = c(2, .6, 0))
plot_cons_plans(x_arma_sp$plans_mv, plans_name = plans_name_sp, cols = cols,
  add_all_efs = FALSE, xlim = xlim, ylim = ylim, add_legend = FALSE)
mtext("(a) Short-term environmental fluctuations", side = 3, line = 0.2, cex =
  0.8, adj = 0.05)
par(las = 0)
mtext("Mean of metapopulation growth rate", side = 2, line = 3, outer = FALSE,
  cex = 0.8)
par(las = 1)
plot_cons_plans(x_linear_sp$plans_mv, plans_name = plans_name_sp, cols = cols,
  add_all_efs = FALSE, xlim = xlim, ylim = ylim, y_axis = FALSE, add_legend =
  TRUE, legend_pos = "bottomright")
mtext("(b) Long-term environmental change", side = 3, line = 0.2, cex = 0.8,
  adj = 0.05)
mtext("Variance of metapopulation growth rate", side = 1, line = 2.25,
  outer = FALSE, cex = 0.8, adj = -1)

```



3 Numbers of populations conserved

Here, we'll show the effect of increasing the number of streams that conservation effort is distributed across. We'll conserve the populations without knowing how response diversity is distributed.

We'll set up the scenarios:

```
set.seed(1)
# In this version we start with a set amount of b and can split it up among many
# or invest it in a few
n_trials <- 200 # number of trials at each n conservation plan
num_pops <- c(2, 4, 8, 16) # n pops to conserve
b_conserve <- 2000 / num_pops
n_plans <- length(num_pops) # number of plans
w <- list()
for(i in 1:n_plans) { # loop over number conserved
  w[[i]] <- list()
  for(j in 1:n_trials) { # loop over trials
    w[[i]][[j]] <- matrix(rep(b_conserve[i], 16), nrow = 1)
    # conserve num_pops[i] populations; wipe out rest:
    w[[i]][[j]][-sample(1:16, num_pops[i])] <- 5
  }
}
plans_name_n <- paste(num_pops, "populations")
```

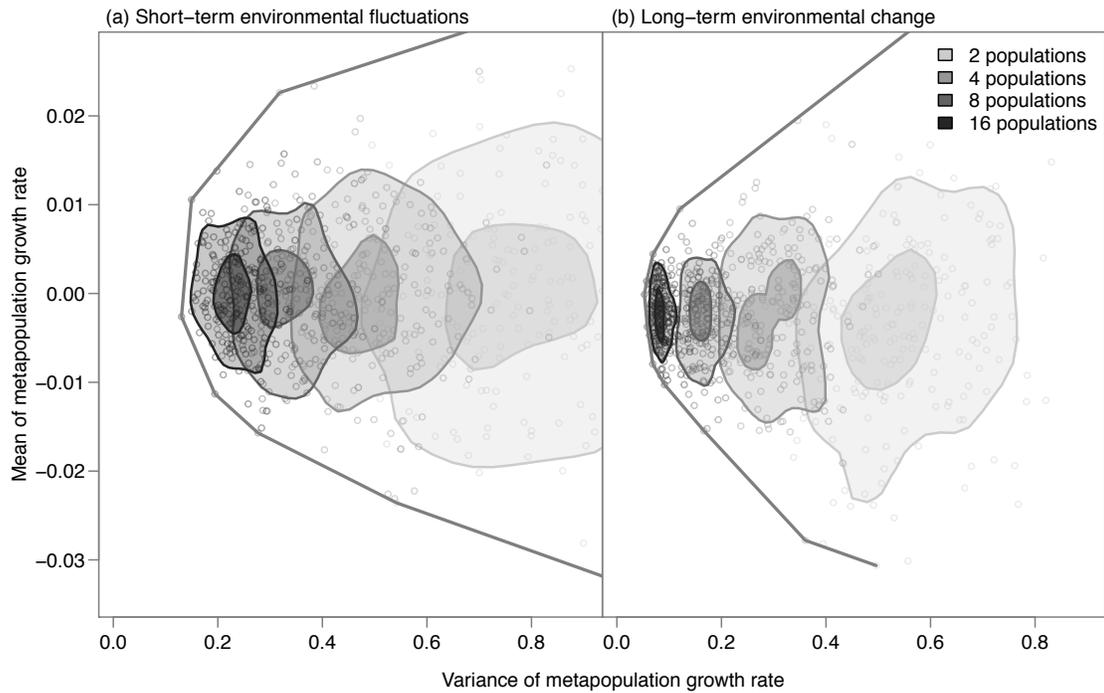
Now we'll run the simulations for the stationary climate variability and the non-stationary scenarios:

```
x_arma_n <- run_cons_plans(w, env_type = "arma", env_params =
  arma_env_params, max_a = thermal_integration(16),
  show_progress = FALSE)
x_arma_n$plans_port <- NULL # to save space
```

```
linear_env_params <- list(min_value = 15, max_value = 19, sigma_env = 0.001,
  start_t = 30)
x_linear_n <- run_cons_plans(w, env_type = "linear",
  env_params = linear_env_params, max_a = thermal_integration(16),
  show_progress = FALSE)
x_linear_n$plans_port <- NULL # save space
```

And plot the output:

```
cols <- RColorBrewer::brewer.pal(5, "Greys")[c(2:5)]
xlim <- c(0.008, 0.90)
ylim <- c(-0.034, 0.027)
par(mfrow = c(1, 2))
par(las = 1, cex = 0.8, mar = c(0, 0, 0, 0), oma = c(4, 5.2, 1.5, .5),
  tck = -0.02, mgp = c(2, .6, 0))
plot_cons_plans(x_arma_n$plans_mv, plans_name = plans_name_n, cols = cols,
  add_all_efs = FALSE, xlim = xlim, ylim = ylim, add_legend = FALSE)
mtext("(a) Short-term environmental fluctuations", side = 3, line = 0.2,
  cex = 0.8, adj = 0.05)
par(las = 0)
mtext("Mean of metapopulation growth rate", side = 2, line = 3,
  outer = FALSE, cex = 0.8)
par(las = 1)
plot_cons_plans(x_linear_n$plans_mv, plans_name = plans_name_n, cols = cols,
  add_all_efs = FALSE, xlim = xlim, ylim = ylim, y_axis = FALSE, add_legend = TRUE)
mtext("(b) Long-term environmental change", side = 3, line = 0.2, cex = 0.8,
  adj = 0.05)
par(xpd = NA)
mtext("Variance of metapopulation growth rate", side = 1, line = 2.25,
  outer = FALSE, cex = 0.8, adj = -1)
```



4 Reduction of habitat over time

Here we'll reduce the available habitat by a fixed quantity at every generation. The climate will follow a combination of short-term variability and long-term change.

Set up the scenarios:

```
set.seed(1)
n_trials <- 200 # number of trials at each n conservation plan
num_pops <- c(2, 4, 8, 12, 16) # n pops to conserve
b_conserve <- 2000 / num_pops
n_plans <- length(num_pops)
w <- list()
for(i in 1:n_plans) { # loop over number conserved
  w[[i]] <- list()
  for(j in 1:n_trials) { # loop over trials
    w[[i]][[j]] <- matrix(rep(b_conserve[i], 16), nrow = 1)
    # conserve num_pops[i] populations; wipe out rest:
    w[[i]][[j]][-sample(1:16, num_pops[i])] <- 5
  }
}
plans_name_n <- paste(num_pops, "populations")
cols <- RColorBrewer::brewer.pal(6, "Greys")[c(2:6)]
```

Run the simulations:

```
linear_arma_env_params <- list(min_value = 15, max_value = 19,  
  start_t = 30, mean_value = 16, ar = 0.1, sigma_env = 2, ma = 0)  
x_linear_arma_n <- run_cons_plans(w, env_type = "linear_arma", env_params =  
  linear_arma_env_params, max_a = thermal_integration(16), decrease_b = 0.85,  
  show_progress = FALSE)  
x_linear_arma_n$plans_port <- NULL # save space
```

Plot the output:

```
xlim <- c(0.08, 0.9)  
ylim <- c(-0.038, 0.028)  
par(las = 1, cex = 0.8, mar = c(0, 0, 0, 0), oma = c(4, 5.2, 1.8, .5),  
  tck = -0.02, mgp = c(2, .5, 0))  
plot_cons_plans(x_linear_arma_n$plans_mv, plans_name = plans_name_n, cols = cols,  
  add_all_efs = FALSE, xlim = xlim, ylim = ylim, add_legend = TRUE,  
  add_poly = TRUE, legend_pos = "bottomright")  
mtext("Reduction in stream flow", side = 3, line = .4,  
  cex = 0.8, adj = 0.05)  
mtext("Variance of metapopulation growth rate", side = 1, line = 2.25,  
  outer = FALSE, cex = 0.8)  
par(las = 0)  
mtext("Mean of metapopulation growth rate", side = 2, line = 3,  
  outer = FALSE, cex = 0.8)
```

