Increasing drought favors nonnative fishes in a dryland river: evidence from a multispecies demographic model

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Abstract. Understanding how novel biological assemblages are structured in relation to dynamic environmental regimes remains a central challenge in ecology. Demographic approaches to modeling species assemblages show promise because they seek to represent fundamental relationships between population dynamics and environmental conditions. In dryland rivers, rapidly changing climate conditions have shifted drought and flooding regimes with implications for fish communities. Our goals were to (1) develop a mechanistic multispecies demographic model that links native and nonnative species with river flow regimes, and (2) evaluate demographic responses in population and community structure to changing flow regimes. Each fish species was represented by a stage-structured matrix, and species were coupled together into a multispecies framework through density-dependent relationships in reproduction. Then, community dynamics were simulated through time using annual flow events classified from gaged streamflow data. We parameterized the model with vital rates and flow–response relationships for a community of native and nonnative fishes using literature-derived values. We applied the simulation model to the Verde River (Arizona, USA), a major tributary within the Colorado River Basin, for the past half century (1964–2017). Model validation revealed a match between model projections and relative abundance trends observed in a long-term fish monitoring dataset (1994–2008). At the beginning of the validation period (1994), model and survey observations showed that native species comprised approximately 80% of total abundance. Model projections beyond the survey data (2008–2017) predicted a shift from a native dominant to a nonnative dominant assemblage, coinciding with increasing drought frequency. Trade-offs between native and nonnative species dominance emerged from differences in mortality in response to the changing sequence of major flow events including spring floods, summer high flows, and droughts. In conclusion, the demographic approach presented here provides a flexible modeling framework that is readily applied to other stream systems and species by adjusting or transferring, when appropriate, species vital rates and flow-event thresholds.

Key words: assemblage; climate change; community; demographic model; drought; freshwater fish; hydrology; invasive species; multispecies model; nonnative species; non-stationarity; stochasticity.

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INTRODUCTION

Novel biological assemblages, comprised of species combinations differing from the past, are increasingly widespread across the world (Hobbs et al. 2009). Mounting evidence suggests that novel assemblages do not arise from a random reshuffling of species, but from heterogeneous rates of species losses and gains over time and space (Zavaleta et al. 2009, Dirzo et al. 2014, Moore and Olden 2017). Species invasions and climate change are primary contributors to novel assemblages, causing changes in species composition and non-systematic reductions in species richness (Dornelas et al. 2014). Consequently, modeling changes in community structure in response to non-stationary climate regimes, particularly in light of on-going species invasions, is a primary research challenge.

Demography-based approaches to modeling communities show promise because they represent fundamental relationships between population dynamics and environmental conditions (Keith et al. 2008, van de Pol et al. 2010, Yen et al. 2013, Lytle et al. 2017). Demographic models utilize knowledge regarding the autecology of a species, such as rates of birth, growth, fecundity, and mortality of individuals (i.e., vital rates) to project population dynamics through time. Causal mechanisms can be readily incorporated by allowing vital rates to change as a function of environmental stochasticity or density dependence (Caswell 2001). Unlike modeling approaches that largely rely on combining single-species predictions to infer community change between two points in time, demographic approaches can model population responses to specific sequences of environmental events. This makes demographic approaches useful for understanding community responses to environmental change (Lytle and Merritt 2004, Yen et al. 2013, Wheeler et al. 2018). Representative communities can be modeled by linking individual species together via density dependence in space requirements, food, or some other limiting resource. The demographic community models are interaction neutral in the sense that pairwise species interactions arise from the model structure itself, rather than being specified a priori as parameters (Lytle et al. 2017, Tonkin et al. 2018). This community-wide approach has demonstrated an ability to recover realistic patterns of community dynamics in freshwater ecosystems and shows promise for revealing how species interact under novel environmental conditions.

A rapidly changing climate that includes more frequent and severe droughts and flooding is poised to reshape fish communities of temporary and perennial rivers in dryland regions (Datry et al. 2014, Kominoski et al. 2018). Climate models project that decreased snow accumulation and higher evapotranspiration rates in spring and summer months will lead to more frequent and severe droughts in the southwestern United States, especially when combined with growing human water demands (Christensen et al. 2004, Seager et al. 2013, Udall and Overpeck 2017). In fact, climate-driven changes to streamflow in the Colorado River Basin have already been observed (Solander et al. 2017), where increased low-flow anomalies and decreasing habitat connectivity threaten native fish persistence (Jaeger et al. 2014, Ruhí et al. 2016) and may favor non-native fishes in the future (Ruhí et al. 2016). Life-history traits have proven useful to understand past (Gido et al. 2013) and predict future responses of native and nonnative fishes to environmental change (Whitney et al. 2017). Thus, we expect that demographically based community models that account for species-specific relationships with hydrology will help to predict past and future changes in fish assemblages.

Dryland rivers of the southwestern United States are a flashpoint for the conservation challenges associated with changing river hydrology and a proliferation of introduced species. Widespread dam construction, flow diversions, and surface and groundwater abstraction for growing human populations have significantly altered environmental regimes in the region, creating conditions that threaten native species persistence and promote nonnative fishes (Minckley and Deacon 1968, Olden and Poff 2005, Strecker et al. 2011). As a result of nonnative species introduction and their establishment and proliferation from reservoirs, the number of nonnative species equals or exceeds native species in most watersheds throughout the southwestern United States (Pool et al. 2010, Walsworth and Budy 2015).
Here, we developed a mechanistic multi-species demographic model to evaluate how native and nonnative fish populations change in response to changing flow regimes. We modeled a fish community by examining species-specific vital rates that varied as a function of flow regimes describing patterns of drought and flooding. Density-dependent relationships coupled populations together into a multispecies framework, and community dynamics were simulated through a sequence of flow events defined by the streamflow record. We validated the model by comparing model projections to empirical data from a long-term fish monitoring program. We then examined how contemporary climate-driven change in the sequence of drought and flood events affects the composition of endemic native fishes in the Colorado River Basin, and broadly highlight the utility of multispecies demographic modeling approaches in ecology.

**METHODS**

**Study system and species**

The Verde River, a tributary within the Colorado River Basin, drains over 17,000 km² of central Arizona (Fig. 1). The perennial mainstem river runs approximately 270 km through private, state, tribal, and United States Forest Service lands, originating in Big Chino Wash (1325 m a.s.l.) and flowing to its confluence with the Salt River north of Phoenix, Arizona (402 m a.s.l.). We focused our study on the unregulated upper Verde River mainstem, where development is primarily limited to livestock grazing and reductions in baseflows are a result of groundwater withdrawals (Garner et al. 2013).

Highly valued for its natural beauty and management priorities as a Wild and Scenic River, the Verde River is a focal point for the conservation of endemic native fishes (Averitt et al. 1994, Turner and List 2007). At least 12 fish species were historically native to the system, but the fish assemblage is changing rapidly, and only five native species have been observed since 1997 (Rinne 2005). By contrast, numerous nonnative fishes are present in the Verde system, including several species of centrarchids, ictalurid catfishes, and minnows (Rinne 2005). The Verde River has been the focus of detailed monitoring efforts starting in the 1990s, where fish community composition in relation to flow and habitat requirements have been examined annually for the period 1994–2008 (Stefferd and Rinne 1995, Rinne and Miller 2006, Neary et al. 2012). These surveys included seven sites in the upper Verde River, encompassing a spatial extent of approximately 60 river kilometers and representing all valley types and habitats occurring within the Verde River (Fig. 1; Neary et al. 2012). River discharge representative of our study area has been measured in the mainstem upper Verde River continuously since 1963 (USGS gage 09503700).

We examined the seven most common fish species in the upper Verde River, collectively representing, on average, 87% of stream reach biomass (Gibson et al. 2015). Native fishes included desert sucker (*Catostomus clarki*), Sonora sucker (*Catostomus insignis*), and roundtail chub (*Gila robusta*); species that are endemic to the Colorado River Basin. Nonnative fishes included yellow bullhead (*Ameirus natalis*), green sunfish (*Lepomis cyanellus*), smallmouth bass (*Micropterus dolomieu*), and red shiner (*Cyprinella lutrensis*);
species with known ecological impacts (Ruppert et al. 1993, Dudley and Matter 2000, Propst et al. 2015). These seven species represent a range of body sizes and major life-history trade-offs between size and age at maturity (growth), juvenile survivorship (survival), and fecundity (reproduction; Olden et al. 2006a). This range of functional traits is represented in dryland streams throughout the western USA.

**Modeling framework**

All seven species populations were modeled simultaneously to represent the fish community of a 1-km river reach. The foundation was a stage-structured matrix population model for each species, modified to incorporate environmental variability and density-dependent relationships (Caswell 2001). The general model structure was adapted from a multispecies matrix population model originally designed to model riparian vegetation population dynamics as a function of river hydrology (Lytle and Merritt 2004, Lytle et al. 2017, Tonkin et al. 2018), but with a number of important modifications described below. Model implementation followed four major steps: parameterization, simulation, validation, and perturbation analysis. Parameter values were based on the flow regime, species biology, and biomass estimates. Model simulations projected fish community composition at annual time steps for each water year in the flow record (1964–2017). Model validation compared population and community model projections against the entire record of long-term fish surveys for the seven upper Verde sites from 1994 to 2008. Finally, perturbation analysis evaluated the effect of uncertainty in parameter estimates on the model output. These steps are described in detail below.

**Parameterization: streamflow, fish, and flow-response relationships.**—Hydrology in the upper Verde River mainstem is characterized by relatively steady, spring-fed baseflow, with high-flow events that vary in magnitude and timing among years in response to winter and summer storm runoff (Fig. 2A; Goetz and Schwarz 2018). Each time-step in the model represented one year. Using the historical flow record, each year was classified into a flow-event year type according to the timing, magnitude, and duration of flows for a water year (1 October–30 September) from USGS gage number 09503700 near Paulden, Arizona, USA. Flood events were defined using discharge thresholds based on recurrence intervals, and drought events were defined by the duration of baseflow (Fig. 2B). Spring high-flood events were years in which the maximum of late winter/early spring discharge (1 January–30 April) exceeded 19.8 m³/s, which has a 4-yr return interval during the spring time window (following Brouder 2001). Spring medium-flood events, corresponded roughly to bankfull flows, had a maximum discharge that exceeded 6.2 m³/s, with a 2.5-yr recurrence interval in the spring (Phillips and Ingersoll 1998, Neary et al. 2012). Summer (and monsoon season) high-flow events (1 May–30 September) exceeded a maximum discharge of 5.7 m³/s, representing a 4-yr recurrence interval in the summer. Drought events were categorized by the absence of floods, when low-flow conditions (i.e., 25th percentile of flows following Béche et al. 2009) persisted for a continuous duration of 40 or more days (i.e., exceeding the 75th percentile duration of low-flow events). Nonevents occurred by default if years were otherwise not defined by flood or drought. A year type with both spring flood and summer high-flow events was possible, but all other flow events were mutually exclusive. This resulted in six possible years: spring high flood, spring medium flood, summer high flow, spring flood and summer high flow, drought, and nonevent (Fig. 2B).

The life-history adaptations of fishes to the flow regime are directly related to their vital rates (Lytle and Poff 2004). Because of the relationship between growth, survival, reproduction, and the flow regime, the use of vital rates inherent to each species supports the transferability of our model to other riverine systems (Mims and Olden 2012, Chen and Olden 2018). We conducted an extensive literature search of peer-reviewed articles, graduate theses, and professional reports for each species, or closely related congeners, to determine parameter estimates for the vital rates used in the model (Table 1; Appendix S1).

Our integrated approach to demographic modeling requires knowledge of how these vital rates vary according to key components of the hydrologic regime. In another literature search, we reviewed studies about fish responses to flow...
components, specifically low-flow (drought) and high-flow (flooding) events in the Colorado River Basin as much as possible. Limiting our review to the Colorado River Basin minimized variability in vital rates that would be introduced by nonnative fish responses observed in other physiographic regions (Chen and Olden 2018). The breadth of information revealed in this literature review included observed relationships between abundance and discharge, effect sizes on changes in abundance in response to high- and low-flow events, and timing of reproductive
behavior (Table 2). This analysis allowed us to assign different vital rates according to types of flow-event years.

The riverine flow regime affects two key life stages that ultimately shape fish population structure and dynamics: juvenile survival and recruitment, and adult survival to reproduction (Schlosser 1985, Humphries et al. 1999). Thus, we allowed the vital rates for each species to vary according to literature-informed relationships between major flow events and fish abundance. To implement this, we used species-specific flow modifiers to adjust baseline vital rates for each year type in the flow record (Appendix S1: Table S1). For example, juvenile survival and recruitment of roundtail chub and smallmouth bass are influenced by the magnitude and timing of high-flow events. Spring flooding increases recruitment of juvenile roundtail chub, whereas elevated summer flows increase mortality of juvenile smallmouth bass (Brouder 2001, Smith et al. 2005). Adult survival is most affected by extended droughts. Low-flow events reduce survival and abundances of species (Stefferud and Stefferud 1998, Ruhi et al. 2015). Droughts typically create conditions where fish suffer because of limited resources and degraded water quality conditions as stream reaches are reduced to shallow isolated pools (Deacon and Minkley 1974, Lake 2003). The magnitude of these flow modifiers is set to reduce or increase mortality by a factor related to trends and effect sizes that were obtained from studies conducted in the region, and occasionally from other watersheds when data for nonnative species were otherwise unavailable (Table 2; Appendix S1: Table S1).

**Model structure and simulation.**—Each fish species was represented by a three-stage demographic matrix containing species-specific vital rates (Fig. 3, Table 1). The three life stages in the life-cycle model represented important ontogenetic shifts for each species. These stages corresponded to: year-1, juvenile recruitment into the population; year-2, sub-adults at first maturity; and year-3 or older adults of fully mature and reproductive individuals (Fig. 3). The parameters within the life-cycle model were adjusted by species to reflect real differences in population stage structure and traits such as lifespan and age at maturity (details follow). Individuals in each stage of the model occupied biomass calculated from length–weight relationships (Appendix S1: Table S2). Lengths at each stage corresponded to literature-reported values of young-of-year and/or immature fish (stage 1), average length at age of maturity (stage 2), and the average length of mature adult individuals from samples in the Upper Verde River (Appendix S1; unpublished data from Gibson et al. 2015).

Transition probabilities in the matrix differed according to the major flow events, which allowed recruitment and survival to vary

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**Table 1. Model parameter values for species vital rates.**

| Parameter          | Desert sucker (Catostomus clarki) | Sonora sucker (Catostomus insignis) | Roundtail chub (Gila robusta) | Yellow bullhead (Ameiurus natalis) | Red shiner (Cyprinella lutrensis) | Green sunfish (Lepomis cyanellus) | Smallmouth bass (Micropterus dolomieu) |
|--------------------|-----------------------------------|------------------------------------|------------------------------|------------------------------------|----------------------------------|----------------------------------|--------------------------------------|
| $G_{ij}$           | 0.08                              | 0.14                               | 0.06                         | 0.06                               | 0.11                             | 0.10                             | 0.07                                 |
| $D_{ij}$ (eggs/g)  | 894                               | 345                                | 1000                         | 533                                | 2123                             | 667                              | 484                                  |
| $D_{ij}$ (indiv/g) | 0.280                             | 0.030                              | 0.110                        | 0.070                              | 6.650                            | 0.860                            | 0.230                                |
| $D_{ij}$ (indiv/g) | 0.113                             | 0.004                              | 0.019                        | 0.009                              | 3.735                            | 0.249                            | 0.010                                |
| $D_{ij}$ (indiv/g) | 0.014                             | 0.002                              | 0.008                        | 0.007                              | 0.618                            | 0.094                            | 0.007                                |
| $M_{ij}$           | 0.999154                          | 0.999869                           | 0.999409                     | 0.998945                          | 0.98855                          | 0.99138                          | 0.999367                             |
| $M_{ij}$           | 0.290                             | 0.212                              | 0.310                        | 0.356                              | 0.320                            | 0.430                            | 0.188                                |
| $a_{ij}$           | 0.167                             | 0.167                              | 0.200                        | 0.200                              | 1.00                             | 0.333                            | 0.250                                |
| $B_{ij}$ ($\lambda, \kappa$) | 5284, 1.52                      | 34068, 1.33                        | 2376, 0.44                    | 1306, 0.36                         | 238, 1.78                        | 164, 0.34                        | 4202, 0.66                           |

**Notes:** Parameters included the Gonadal-Somatic Index ($GSI_j$), the conversion factor from individuals to biomass for eggs ($D_{ij}$), stage 1 juveniles ($D_{1j}$), stage 2 sub-adults ($D_{2j}$), stage 3 ($D_{3j}$), egg and larval mortality ($M_{ij}$), and mortality ($M_{ij}$) for each stage $i$ and species $j$. Transition probability of adult mortality in stage 3 ($e_{ij}$) for each species $j$, and starting adult population biomass ($B_{ij}$) for each species $j$ selected from a negative binomial distribution with mean ($\lambda$) and dispersion parameter ($\kappa$). References for vital rate parameter estimates can be found in Appendix S1.
according to the hydrologic conditions in a particular year. The seven single-species demographic matrices were coupled via density-dependent relationships in reproduction, limited by the total biomass carrying capacity of a representative river reach. In this way, the reproductive output of each species declined as the total aggregate fish biomass of the entire community approached a reach-wide carrying capacity (Eq. 2). This generalized density dependence is analogous to exploitative competition experienced by organisms competing for a single limiting resource (Hardin 1960). A similar approach has been implemented in multispecies models using spatial density dependence (Lytle et al. 2017).

The model had several assumptions regarding carrying capacity and vital rates. First, we assumed that the carrying capacity of the reach was limited by the amount of total fish biomass that could be sustained. Carrying capacity, \( K \), was set to the average total biomass found in a 1-km reach from surveys in nine replicate 100-m sampling sites located in the Upper Verde River and conducted in 2012 (Gibson et al. 2015). We chose to use average total biomass because some reaches will naturally be more or less productive and suitable for fishes than others. Second, population growth was limited by a density-dependent function in the reproductive term (fecundity, \( F \)) so no recruitment occurred if total biomass in the reach was greater than or equal to \( K \) (Eq. 2). Third, baseline mortality was the same for all life stages within each species, except for the egg and larval phase. The combination of egg and larval mortality was calculated as part of fecundity so that the starting population of adults produced sufficient offspring to equal their replacement after stage 1 and stage 2 baseline mortality was taken into account. Last, all species had a 1:1 sex ratio, and an individual’s life cycle could be completed in a 1-km reach for all species. This is a reasonable assumption given that fishes in the Verde River typically occupy

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**Table 2. Evidence for flow–response relationships in the study region.**

| Flow attribute | (+/-) | Species | Response variable | References |
|----------------|-------|---------|-------------------|------------|
| High flows (spring) | +     | Catostomus sp., Gila sp. | Abundance/density | Propst and Gido (2004), Propst et al. (2008), Stefferud et al. (2011), Gido et al. (2013), Ruhé et al. (2015) |
| High spring flows (summer, spring, or number of events) | +     | Gila robusta | Recruitment | Brouder (2001) |
| High summer flows | –     | Ameiurus sp., Cyprinella lutrensis, Leponis cyanellus, Micropterus sp. | Abundance/density | Minkley and Meffe (1987), Propst et al. (2008), Gido et al. (2013), Ruhé et al. (2015) |
| Low flows (constant baseflow) | –     | Micropterus dolomieu | Recruitment | Smith et al. (2005) |
| Low flows | +     | Ameiurus sp., C. lutrensis, L. cyanellus, Micropterus sp. | Abundance/density | Propst and Gido (2004), Propst et al. (2008), Gido et al. (2013), Ruhé et al. (2015) |
| Low flows | –     | Catostomus sp., Gila sp. | Abundance/density | Stefferud and Stefferud (1998), Propst et al. (2008), Stefferud et al. (2011), Gido et al. (2013), Ruhé et al. (2015) |

**Notes:** Information presented includes the directionality of response to generalized flow attributes reported in the literature (+/-), list of relevant fish species, the dependent response variable to the flow attribute, and supporting references. Flow modifier values assigned according to these relationships are provided in Appendix S1: Table S1.
small home ranges and most fish move <1 km (Jaeger et al. 2014, Comte and Olden 2018).

Species’ biomass, rather than species abundances, was the currency for the model framework (although abundance and biomass values were interchangeable using the stage-specific weight of individuals). For each species j at each stage i for a given year type k, the biomass change from time t to time t + 1 was given by:

\[
\begin{pmatrix}
B_{1j}^{t+1} \\
B_{2j}^{t+1} \\
B_{3j}^{t+1}
\end{pmatrix}
= \begin{pmatrix}
F_{1j} & F_{2j} & F_{3j} \\
G_{1jk} & 0 & 0 \\
0 & G_{2jk} & P_{3jk}
\end{pmatrix}
\begin{pmatrix}
B_{1j}^t \\
B_{2j}^t \\
B_{3j}^t
\end{pmatrix},
\]

where \(B_{ij}\) was the total biomass (g) of the species at the corresponding life stage. Stage-specific fecundity (\(F_{ij}\)) was linearly density dependent on the total fish biomass in the community:

\[
F_{ij} = \begin{cases} 
    f_i \times (K - \sum_{j=1}^{m} \sum_{i=1}^{n} B_{ij}) \times K^{-1} \times D_{ej} \times D_{ij}^{-1} & \text{if } \sum_{i=1}^{m} \sum_{j=1}^{n} B_{ij} \leq K \\
    0 & \text{otherwise}
\end{cases}
\]

\[f_i = 0.5 \times \text{GSI}_j \times (1 - M_{ij}) \quad (3)\]

where \(i = 1, 2, \ldots, m\) was an index representing stages for each of \(j = 1, 2, \ldots, n\) species in the community. Reach-wide carrying capacity \(K\) was the maximum aggregate biomass attainable for all species in all stages combined. Species began spawning at the stage that corresponded to the age of first maturity (Appendix S1: Fig. S3). Therefore, in each species’ matrix, the number of fecundity (\(F\)) terms corresponded to the number of years the species can reproduce and was zero otherwise. For example, red shiner may begin reproducing in its first year of life, so the matrix had three fecundity terms, whereas yellow bullhead begin reproducing in their third year of life, so they had one fecundity term at stage 3.

The proportion of egg biomass (\(f_j\)) produced by all females in the population after accounting for mortality during the first year depended on the Gonadal-Somatic Index (GSI), the proportion of gonad mass to total body mass, the proportion of female individuals (0.5), and the vital rate or combined egg and larval mortality (\(M_{ij}\)). The egg biomass of each species was converted to stage 1 biomass to account for fish development and growth. We converted egg biomass using egg density (\(D_{ej}\)) in units of number of eggs per gram and the average weights of stage 1 individuals (\(D_{ij}^{-1}\); Table 1).

Egg density of each species was the average value calculated from literature-reported values (Appendix S1) of the number of mature eggs in ripe females, divided by the total ovary mass. Ovary mass was reported or calculated using her length–weight relationship and GSI. If the relationship between total length and number of mature eggs was not published, egg density was estimated from the means of reported values.

The growth rate and survival probability (\(G\)) transitions from stage 1 to stage 2 and stage 2 to stage 3 were

\[
G_{ijk} = (1 - (M_{ij} \times Y_{ijk})) \times D_{ij} \times D_{ij}^{-1} (4)
\]

\[
P_{ijk} = (1 - (M_{ij} \times Y_{ijk})) \times (1 - a_{3j}) (5)
\]

Survival was one minus mortality \((M_{ij})\) multiplied by a flow modifier \(Y_{ijk}\) for stage \(i\) at species \(j\) for flow event \(k\) (Table 2; Appendix S1: Table S1).

The probability of surviving and remaining in the adult stage was related to the literature-reported lifespan of each species:

where the probability that adults stay in stage 3 \((a_{3j})\) is the reciprocal maximum age of species \(j\) after accounting for the first two life stages of the model. At the end of each model run, biomass output was converted to abundance for subsequent data analysis and interpretation.

We ran 1000 iterations of the model, simulating the community from 1964 through 2017. Each iteration began with different initial population biomass to account for spatial variation. Initial biomass for each species \(j\) was sampled from a negative binomial distribution with mean \(\lambda\) and dispersion parameter \(\kappa\) converted from the mean.
and variance in relative abundance of the long-term Verde dataset for all seven sites and 15 yr of data (n = 105). As with a Poisson distribution, the negative binomial is appropriate for counts of organisms that occur randomly over time or space, but the negative binomial allows the variance to exceed the mean.

Model validation.—Model validation was performed by comparing population and community projections to data from long-term fish surveys from seven sites in the Upper Verde River between 1994 and 2008 (Stefferud and Rinne 1995, Rinne et al. 1998, Neary et al. 2012). Data from 2002 were omitted from the analysis because only two of seven long-term sites were surveyed. Relative annual species abundances, averaged across the seven sites, were compared to model projected relative abundances according to the sum of non-juvenile (stages 2 and 3) individuals because juveniles (stage 1) are under-represented by electrofishing survey methods due to their more cryptic behavior and small body size (Bonar et al. 2009). Relative abundance was chosen over absolute abundance because survey effort was not reported and was not always consistent between years, despite the use of a standardized collecting protocol. The strength of association between observed and modeled relative abundances was assessed using Spearman rank correlations. Spearman rank correlations are reported for each species across years and each year across species between 1994 and 2008, again omitting 2002. We also reported root mean square error (RMSE) and coverage (C), the percentage of 95% confidence intervals that overlap true values, to evaluate model performance.

Perturbation analysis.—The influence of model uncertainty was evaluated with respect to mortality rate, GSI, flow event thresholds, and biomass carrying capacity. These values were chosen because they are the most likely to affect transitions between life stages (i.e., mortality rate), fecundity/reproduction (i.e., GSI), species responses to environmental changes (i.e., flow-event thresholds), and outcomes in species dominance (i.e., biomass carrying capacity). We conducted a direct perturbation analysis (sensu Regan et al. 2003, Bond et al. 2014) to estimate the effects of parameter uncertainty by adjusting each value by ±10% of the starting value. Flow events were evaluated by adjusting thresholds of spring and summer high-flow events and duration of low-flow drought events. Uncertainty was quantified by the proportional change in species’ relative abundances in response to parameter adjustments. Perturbation analysis was favored over analytical sensitivity analysis (via partial differentiation of a vital rate with respect to population growth rate) because it can be used to evaluate uncertainty in non-matrix elements, such as carrying capacity, and it does not require calculation of the long-term population growth rate (Akçakaya et al. 2003, Stott 2016). The model simulation and all analyses were performed with program R v.3.4.0 (R Core Team 2017).

RESULTS

Model population projections reflected compositional trends observed in the long-term fish surveys conducted from 1994 to 2008 (Fig. 4). During this validation period, trends in observed relative abundances of green sunfish, smallmouth bass, yellow bullhead, and roundtail chub were aligned with the simulated model populations (Fig. 4C, D, F, G). Model performance was best for green sunfish, with highest values for correlations (r), coverage (C), and lowest prediction errors (RMSE; Fig. 4F). The remaining species demonstrated varied correlations, coverage, and prediction errors. Predictions for relative abundance for roundtail chub and smallmouth bass were significantly correlated with observed data and had low to moderate prediction errors (Fig. 4C, G). Both smallmouth bass and yellow bullhead had high coverage (Fig. 4D, G). Desert sucker predictions correlated moderately well with observations and had moderate prediction errors and coverage (Fig. 4A). By contrast, the model tended to overestimate Sonora sucker and underestimate red shiner relative abundances (Fig. 4B, E). Relative abundances of red shiner, a species with small maximum body size that often occurs patchily in large schools of individuals, varied greatly among sites, evidenced by the large error bars (Fig. 4E). Sonora sucker had the lowest correlation with survey data and lowest coverage because the model did not capture an observed shift toward lower relative abundance in the middle of the survey period.
Community structure shifted from native dominant to nonnative dominant over the course of the model simulation period. Native species comprised approximately 80% of total abundance at the beginning of the validation period (1994) for both model and survey data (Fig. 5). Then, after a series of drought and nonevent flow years, there was a marked transition toward nonnative dominance. At the end of the observed survey data, after a spring high-flood event in 2005, native species rebounded to represent 50–60% of the assemblage abundance (Fig. 5).

Fig. 4. Modeled (lines) and observed (dots) relative abundances through time for native (A–C) and nonnative (D–G) fish species. Model and observed data include 95% confidence intervals, gray bands, and error bars, respectively. In the upper right corner of each panel, root mean square error (RMSE) is a measure of the difference, and Spearman rank correlations (r) is a measure of the association strength, and coverage is a measure of the percent overlap, between observed and model values.
Model simulations followed the general trend of decreasing, then increasing native abundances, but the magnitude of the model fluctuations was dampened compared to survey observations (Fig. 5). Despite some differences in magnitude, overall there was a strong correlation between rank order abundances of species between model projections and survey observations (mean $r \pm SE: 0.67 \pm 0.018$) with significant annual correlations for half (7 out of 15 yr) of the validation period (Appendix S1: Table S3). The model performed poorly between 1999 and 2005, coinciding with a period in the flow record from 1996 to 2004 that lacked spring flood events, which are important for juvenile (stage 1) survivorship of native Sonora sucker, desert sucker, and round-tail chub (Table 1; Appendix S1: Table S1). Beyond the period of the fish survey (post-2008), the modeled community projection, in response to observed streamflows, continued to gradually shift from native to nonnative dominant assemblage to the end of the flow record in 2017. Put in the perspective of flow events, the last ten years of the model had a drought frequency of 30%, compared to 9% drought frequency of the full flow record (Fig 2B).

Perturbation analysis on species vital rates had the largest effects on overall community composition, a larger effect than was seen by perturbing flow thresholds or biomass carrying capacity (Table 3). Perturbation of desert sucker and green sunfish mortality and red shiner reproduction (GSI) had the largest influence on community composition overall, but the most affected species was always the one whose parameter was being perturbed. For example, when desert sucker’s mortality rate experienced a 10% decrease, their relative abundance increased by 58% while other species compensated for this increase with a decline of 14–23% relative abundance. A decrease in red shiner’s GSI parameter resulted in a 44% decrease in their relative abundance with a compensatory increase of 6–30% for the other species. This species, with its short lifespan, small body size, and high reproductive rate, was the most influential among changes in GSI, but also the most influenced by increased carrying capacity. When carrying capacity for a reach was increased, red shiner relative abundance increased by 17%, due to its high reproductive rate. Other members of the community had smaller changes in relative abundance, with declines of 4–9%.

Perturbation analysis in flow event thresholds demonstrated that community composition was most influenced by increasing the threshold of spring medium-flood events and decreasing the threshold of summer high-flow events. By increasing the spring threshold, two fewer spring flow events and two more nonevent years occurred during the hydrologic record. This influenced community composition by increasing yellow bullhead and green sunfish relative abundance by 39% each, accompanied by smaller increases in relative abundance for the other nonnative species, and decreases in relative abundance for the native species (Table 3). By contrast, decreasing the threshold for summer high-flow events resulted in three fewer nonevent years and a decrease in yellow bullhead and green sunfish relative abundance by 35% and 52%, respectively, but with minor or no changes to the relative abundance of the other species (Table 3).

**DISCUSSION**

Demographic models provide new opportunities to better understand species responses to shifting environmental conditions such as
climate-induced changes to hydrologic regimes and human-caused flow alteration (Shenton et al. 2012, Bond et al. 2018). Approaches that enable temporally specific predictions of species responses to specific flow sequences are likely to be most useful in practice (Wheeler et al. 2018).

Here, we used demographic modeling to simulate how hydrologic drought and flood events interact with species’ vital rates to shape native and nonnative fish composition in a dryland river. Although biotic interactions, such as competition and predation, were not specified a priori in the model framework, pairwise species interactions arose from the model structure itself due to the assumption of aggregate density dependence. The approach captured community trends using trade-offs in flow-related mortality of different life stages, as evidenced by predicting ranked abundances of species over a decade-and-a-half time period. Mismatches between modeled and observed numbers were most noticeable for species with vital rates that are difficult to obtain, or arose from potential life histories that lead to observation error associated with field survey methodology. Mechanistically representing trends in community response to environmental drivers using independently published vital rates opens up the possibility for hypothesis testing and scenario analysis for exploring management options for multiple species at once.

In our simulated model of the Verde River, overall patterns in community structure, over a 54-yr period, demonstrated that more frequent drought events and fewer spring flood events created conditions where native fishes fared poorly compared to nonnative fishes. This finding is supported by empirical research in dryland rivers of the same region (Propst et al. 2008, Gido and Propst 2012, Gido et al. 2013). In the model, both nonevent years and droughts supported population growth by nonnative species. Non-events represent years of steady baseflows that favored the survival of all juvenile nonnative fishes and adult life stages of yellow bullhead, red shiner, and green sunfish. Similarly, during drought years, native suckers (Sonora sucker and desert sucker) and roundtail chub experienced high mortality rates in both juvenile and adult life stages. By contrast, nonnative species had lower mortality rates for juvenile life stages, and depending on the species, had higher or lower mortality rates for adult life stages during drought. Higher mortality rates of large native species (i.e., the suckers) during drought years facilitated a compensatory response by nonnative species. That is, newly available portions of carrying capacity in the form of biomass were taken up by small- and medium-bodied fishes with high reproductive and low mortality rates during drought years (i.e., red shiner and green sunfish).

Past research demonstrates that droughts modify the spatiotemporal connectivity of riverine habitats, ultimately driving patterns in the composition and trophic structure of fishes (Matthews and Marsh-Matthews 2003, Rolls et al. 2012). As streamflows decline, fishes move to seek deep refuge pools (Labbe and Fausch 2000, Magoulick and Kobza 2003, Marshall et al. 2016). As pool habitats contract, the density of organisms initially increases with several consequences. First, smaller volumes of water concentrate prey, providing a food subsidy and increasing survival probability for young-of-year fishes for species spawning in warm months, or for extended breeding seasons (Schlosser 1985, Craven et al. 2010). In the Verde River, all nonnative species included in this study spawn during the warm spring–summer months. Of these, red shiner has the longest breeding season being a serial spawner, thereby performing well in drought years. Second, crowding intensifies predation and competition for resources among fishes (Magoulick and Kobza 2003, Matthews and Marsh-Matthews 2006). In rivers of the southwestern United States, nonnative fishes such as smallmouth bass, green sunfish, and yellow bullhead outcompete or consume native fishes, especially juvenile life stages or small-bodied species, leading to local extirpations and community change when streamflow variability declines (Eby et al. 2003, Stefferud et al. 2011). Eventually, the combination of low flows and negative species interactions may lead to local native species extirpation. For instance, in the Verde River, the small-bodied native spikedace (Meda fulgida) has already been lost from the system but maintains viable populations where it has been repatriated in streams without nonnative fishes (Neary et al. 2012). As a corollary, high flows tend to favor native species,
sometimes at the cost of nonnative species. In particular, spring floods may displace nonnative species and delay their reproduction (Propst and Gido 2004). This suggests that low-flow conditions favor species invasions (Beche et al. 2009, Diez et al. 2012) and can have prolonged consequences even after the cessation of drought (Humphries and Baldwin 2003).

In the model results, low frequency or absence of spring flood events was detrimental to native fish populations. As observed in the perturbation analysis, minor increases in thresholds for medium spring flood events increased the number of nonevent years in the hydrologic period of record. This resulted in lower relative abundances of native species, including both sucker species and roundtail chub. The absence of spring flood events for nine years during the model simulation led to lower population growth rates for native species. The lack of spring flood events, which favored juvenile recruitment, resulted in native species experiencing higher mortality rates compared to nonnative species during these times. Therefore, given that the average lifespan of native species in this study range from 5 to 10 yr, it is likely that low recruitment potential for the nine-year period was detrimental to the persistence of these fish populations.

We found that native fish populations trended downward as nonnative fishes became dominant during a period of more droughts (three occurrences) and nonevents (two occurrences) starting in 2008. This decline was only punctuated by minor increases in native fish abundances following 2010 and 2015 spring floods. Because our model simulates a closed community with no rescue effects, once native fish population growth rates experienced a precipitous decline for several consecutive years after droughts, species within the model community were more likely to go extinct than in observational studies.
of riverine fish communities. Unlike our model community, fish communities observed under prolonged and supra-seasonal droughts can be resilient in spatially connected landscapes. For example, communities may eventually recover to pre-drought composition if deep refuge pools persisted during drought or other locations acted as sources for recolonization following the cessation of drought (Davey and Kelly 2007, Matthews et al. 2013, Rolls et al. 2016). However, persistent low-flow conditions have caused long-term changes in community structure where species adapted to low-flow conditions expanded their range and increased in abundance with complementary reductions in species favored by high flows (Lawson and Johnston 2016).

Demographic models that link species vital rates with hydrology have the advantage of addressing differential responses of native and nonnative fishes to streamflow variability because they leverage mechanistic associations between populations and specific flow events. This modeling approach implicitly includes species interactions via the density-dependent relationship between reproduction and biomass carrying capacity but does not explicitly include trophic interactions such as predation. For this reason, as with most modeling approaches, the strength of our conclusions depends on how much variability in population and community structure is explained by the unknown true contribution of flow events to demographic responses compared to other factors. Separating the influence of flow conditions independent of other extrinsic environmental factors, including species interactions, is inherently difficult (Chen and Olden 2018). However, a number of observational studies have demonstrated that the importance of biotic interactions in shaping fish communities is often overridden by environmental forcing (Grossman et al. 1998, Ruhí et al. 2015, Giam and Olden 2016, Bino et al. 2017). Our goal was to develop and present a transferable approach to modeling multiple species within a community, an approach that allows for the incorporation of non-stationary environmental change and demographic variability. Continued research will help establish the relative strength of different drivers to demographic and community responses and help improve model performance and interpretation of model results.

A persistent challenge of community modeling is to simultaneously represent all individual species (Olden et al. 2006b). Individual species correlations between observed and simulated model data were moderate, but it is important to note that our model predictions were derived from independently reported parameters. In other words, the demographic multispecies model we have presented is unlike a statistical model where empirical data are used to fit model parameters. Rather, model performance depended solely on independent information, including that from literature-based vital rates, initial population sizes taken from a probability distribution, and empirically informed flow modifiers. Discrepancies between observed and modeled results were at least partially due to uncertainty in parameter estimates. Despite long interest in flow-ecology relationships, the strength of our understanding of these associations for all species in a community remains limited (Davies et al. 2014, Rosenfeld 2017). Vital rate and flow-dependent transition parameters can be adjusted as new empirical knowledge is gained. Although no one method is perfect, natural observations and in-stream or mesocosm experiments remain critical for acquiring vital rate information as a function of streamflow and will help further improve model predictions of community structure (Shenton et al. 2012, Poff 2018, Wheeler et al. 2018).

Discrepancies between model predictions and observed values may also occur because of observation error. For example, we found that model predictions underestimated the abundance of red shiner and overestimated the abundance of Sonora sucker. Together, these discrepancies led to a period where native dominance in the community was overestimated. Models for red shiner demonstrated the poorest performance, which may be due to highly variable sampling efficiencies due to the schooling behavior of this species. Sampling bias may also affect the observation of species with large maximum body sizes such as Sonora sucker and smallmouth bass. Older age groups of these species may be observed less than expected by model predictions because they occupy habitats too deep to sample with standard seining and backpack electrofishing techniques. Although fish surveys document low-recruitment periods during dry years with
smaller sample sizes of Sonora sucker in wadeable habitats, model projections of adult Sonora sucker were relatively stable. This apparent stability may reflect mechanisms related to the temporal storage effect, whereby species are able to store up gains during favorable periods to persist during non-favorable periods, enabling multiple species to coexist in variable environments (Chesson and Warner 1981, Warner and Chesson 1985). Here, high-recruitment periods (i.e., spring flood years) allow Sonora sucker populations to remain stable because long-lived adults survive through periods more favorable to other species.

In our model, the sequence of flow events appeared to support long-term multispecies coexistence, by favoring different species in years with different environmental conditions, at least over the 54-yr time frame examined in this analysis. In explorations over longer timescales (centuries or longer), the model is expected to eventually predict the complete dominance of a single species—the species with traits that resulted in the highest stochastic population growth rate for that particular hydrologic regime—due to the fact that we are modeling finite populations in a finite reach under stochastic conditions. Body size, age at maturity, and fecundity represent strong trade-offs in the life-history strategies of fishes (Olden et al. 2006a). Therefore, one might expect that the largest or earliest maturing species with the highest fecundity or reproductive rate, respectively, would always outcompete all other species in the community (Cushing 1992). However, because modeled vital rates are tightly coupled to flow conditions, patterns reflected observed species coexistence dynamics. In the model, the sequence and frequency of particular flow events played an important role in determining which species persisted over longer timescales.

Demographic models can be applied to any component of river ecosystems to explore species responses to changing hydrology, or other drivers of persistence and mortality. Where vital rates are known, they have been applied to individual species or functional guilds of invertebrates (McMullen et al. 2017), riparian plants (Lytle et al. 2017, Tonkin et al. 2018), and fish (Yen et al. 2013). Demographic models are also flexible because different parameters may be applied to forecast the effects of environmental change in other ecosystems (McMullen et al. 2017). For example, one could apply our model to another stream system by adjusting or transferring, when appropriate, species vital rates, and flow event thresholds specific to the hydrograph and ecology of the system of interest. The flexibility of these models, and ability to predict non-stationary temporal dynamics, makes them useful for exploring scenarios of environmental change and the outcomes of various flow management interventions into the future. In-stream flow management plans to benefit species and ecosystems are often challenged by other competing demands on river flows, and population models that forecast the outcomes of various flow futures are a valuable and much-needed tool for decision makers.

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LITERATURE CITED

Akcakaya, H. R., J. L. Atwood, D. Breining, C. T. Collins, and B. Duncan. 2003. Metapopulation dynamics of the California Least Tern. Journal of Wildlife Management 67:829–842.

Averitt, E., F. Steiner, R. A. Yabes, and D. Patten. 1994. An assessment of the Verde River corridor project in Arizona. Landscape and Urban Planning 28:161–178.

Beche, L. A., P. G. Connors, V. H. Resh, and A. M. Merenlender. 2009. Resilience of fishes and invertebrates to prolonged drought in two California streams. Ecography 32:778–788.

Bino, G., S. Wassens, R. T. Kingsford, R. F. Thomas, and J. Spencer. 2017. Floodplain ecosystem dynamics under extreme dry and wet phases in semi-arid Australia. Freshwater Biology 63:224–241.

Bonar, S. A., W. A. Hubert, and D. W. Willis. 2009. Standard methods for sampling North American freshwater fishes. American Fisheries Society, Bethesda, Maryland, USA.
Bond, N. R., S. R. Balcombe, D. A. Crook, J. C. Marshall, N. Menke, and J. S.lobegeiger. 2014. Fish population persistence in hydrologically variable landscapes. Ecological Applications 25:901–913.

Bond, N. R., N. Grigg, J. Roberts, H. McGinnness, D. Nielsen, M. O’Brien, I. Overton, C. Pollino, J. R. W. Reid, and D. Stratford. 2018. Assessment of environmental flow scenarios using state-and-transition models. Freshwater Biology 63:804–816.

Brouder, M. J. 2001. Effects of flooding on recruitment of roundtail chub, Gila robusta, in a southwestern river. Southwestern Naturalist 46:302–310.

Caswell, H. 2001. Matrix population models: construction, analysis, and interpretation. Second edition. Oxford University Press, Oxford, New York, USA.

Chen, W., and J. D. Olden. 2018. Evaluating transferability of flow–ecology relationships across space, time and taxonomy. Freshwater Biology 63:817–830.

Chesson, P. L., and R. R. Warner. 1981. Environmental variability promotes coexistence in lottery competitive systems. American Naturalist 117:923–943.

Christensen, N. S., A. W. Wood, N. Voisin, D. P. Lettemaier, and R. N. Palmer. 2004. The effects of climate change on the hydrology and water resources of the Colorado River Basin. Climatic Change 62:337–363.

Comte, L., and J. D. Olden. 2018. Fish dispersal in flowing waters: a synthesis of movement– and genetic-based studies. Fish and Fisheries 19:1063–1077.

Craven, S. W., J. T. Peterson, M. C. Freeman, T. J. Kwak, and E. Irwin. 2010. Modeling the relations between flow regime components, species traits, and spawning success of fishes in warmwater streams. Environmental Management 46:181–194.

Cushing, J. M. 1992. A discrete model for competing stage-structured species. Theoretical Population Biology 41:372–387.

Datry, T., S. T. Larned, and K. Tockner. 2014. Intermittent rivers: a challenge for freshwater ecology. BioScience 64:229–235.

Davey, A. J. H., and D. J. Kelly. 2007. Fish community responses to drying disturbances in an intermittent stream: a landscape perspective. Freshwater Biology 52:1719–1733.

Davies, P. M., R. J. Naiman, D. M. Warfe, N. E. Pettit, A. H. Artthington, and S. E. Bunn. 2014. Flow-ecology relationships: closing the loop on effective environmental flows. Marine and Freshwater Research 65:133–141.

Deacon, J. E., and W. L. Minckley. 1974. Desert fishes. Pages 385–488 in G. W. Brown Jr, editor. Desert biology. Volume II. Academic Press, New York, New York, USA.

Diez, J. M., et al. 2012. Will extreme climatic events facilitate biological invasions? Frontiers in Ecology and the Environment 10:249–257.

Dirzo, R., H. S. Young, M. Galetti, G. Ceballos, N. J. B. Isaac, and B. Collen. 2014. Defaunation in the Anthropocene. Science 345:401–406.

Dornelas, M., N. J. Gotelli, B. McGill, H. Shimadzu, F. Moyes, C. Sievers, and A. E. Magurran. 2014. Assemblage time series reveal biodiversity change but not systematic loss. Science 344:296–299.

Dudley, R. K., and W. J. Matter. 2000. Effects of small green sunfish (Lepomis cyanellus) on recruitment of Gila chub (Gila intermedia) in Sabino Creek, Arizona. Southwestern Naturalist 45:24–29.

Eby, L. A., W. F. Fagan, and W. L. Minckley. 2003. Variability and dynamics of a desert stream community. Ecological Applications 13:1566–1579.

Garner, B. D., D. R. Pool, F. D. Tillman, and B. T. Forbes. 2013. Human effects on the hydrologic system of the Verde Valley, central Arizona, 1910–2005 and 2005–2010, using a regional groundwater flow model: U.S. Geological Survey Scientific Investigations Report 2013–5029. U.S. Geological Survey, Reston, Virginia, USA.

Giam, X., and J. D. Olden. 2016. Environment and predation govern fish community assembly in temperate streams. Global Ecology and Biogeography 25:1194–1205.

Gibson, P. P., J. D. Olden, and M. W. O’Neill. 2015. Beaver dams shift desert fish assemblages toward dominance by non-native species (Verde River, Arizona, USA). Ecology of Freshwater Fish 24:355–372.

Gido, K. B., and D. L. Propst. 2012. Long-term dynamics of native and nonnative fishes in the San Juan River, New Mexico and Utah, under a partially managed flow regime. Transactions of the American Fisheries Society 141:645–659.

Gido, K. B., D. L. Propst, J. D. Olden, K. R. Bestgen, and J. Rosenfeld. 2013. Multidecadal responses of native and introduced fishes to natural and altered flow regimes in the American Southwest. Canadian Journal of Fisheries and Aquatic Sciences 70:554–564.

Goetz, J., and C. J. Schwarz. 2018. fasstr: Tools for streamflow data tidying, analyses, and visualization. R package version 0.2.5.1. Province of British Columbia, Canada. https://github.com/bcgov/fasstr

Grossman, G. D., R. E. Ratajczak, M. Crawford, and M. C. Freeman. 1998. Assemblage organization in stream fishes: effects of environmental variation and interspecific interactions. Ecological Monographs 68:395–420.

Hardin, G. 1960. The competitive exclusion principle. Science 131:1292–1297.
Humphries, P., A. J. King, and J. D. Koehn. 1999. Fish, Hobbs, R. J., E. Higgs, and J. A. Harris. 2009. Novel Humphries, P., and D. S. Baldwin. 2003. Drought and aquatic ecosystems: implications for conservation and restoration. Trends in Ecology & Evolution 24:599–605. Humphries, P., A. J. King, and J. D. Koehn. 1999. Fish, and their environment in the Murray-Darling river system, Australia. Environmental Biology of Fishes 56:129–151. Jaeger, K. L., J. D. Olden, and N. A. Pelland. 2014. Climate change poised to threaten hydrologic connectivity and endemic fishes in dryland streams. Proceedings of the National Academy of Sciences USA 111:13894–13899. Keith, D. A., H. R. Akcakaya, W. Thuiller, G. F. Midgley, R. G. Pearson, S. J. Phillips, H. M. Regan, M. B. Araújo, and T. G. Rebelo. 2008. Predicting extinction risks under climate change: coupling stochastic population models with dynamic bioclimatic habitat models. Biology Letters 4:560–563. Kominoski, J. S., A. Ruhí, M. M. Hagler, K. Petersen, J. L. Sabo, T. Sinha, A. Sankarasubramanian, and J. D. Olden. 2018. Patterns and drivers of fish extirpations in rivers of the American Southwest and Southeast. Global Change Biology 24:1175–1185. Labbe, T. R., and K. D. Fausch. 2000. Dynamics of intermittent stream habitat regulate persistence of a threatened fish at multiple scales. Ecological Applications 10:1774–1791. Lake, P. S. 2003. Ecological effects of perturbation by drought in flowing waters. Freshwater Biology 48:1161–1172. Lawson, K. M., and C. E. Johnston. 2016. The role of flow dependency and water availability in fish assemblage homogenization in tributaries of the Chattahoochee River, Alabama, USA. Ecology of Freshwater Fish 25:631–641. Lytle, D. A., and D. M. Merritt. 2004. Hydrologic regimes and riparian forests: a structured population model for cottonwood. Ecology 85:2493–2503. Lytle, D. A., and N. L. Poff. 2004. Adaptation to natural flow regimes. Trends in Ecology & Evolution 19:94–100. Lytle, D. A., D. M. Merritt, J. D. Tonkin, J. D. Olden, and L. V. Reynolds. 2017. Linking river flow regimes to riparian plant guilds: a community-wide modeling approach. Ecological Applications 27:1338–1350. Magoulick, D. D., and R. M. Kobza. 2003. The role of refugia for fishes during drought: a review and synthesis. Freshwater Biology 48:1186–1198. Marshall, J. C., et al. 2016. Go with the flow: the movement behaviour of fish from isolated waterhole refugia during connecting flow events in an intermittent dryland river. Freshwater Biology 61:1242–1258. Matthews, W. J., and E. Marsh-Matthews. 2003. Effects of drought on fish across axes of space, time and ecological complexity. Freshwater Biology 48:1232–1253. Matthews, W. J., and E. Marsh-Matthews. 2006. Persistence of fish species associations in pools of a small stream of the southern Great Plains. Copeia 2006:696–710. Matthews, W. J., E. Marsh-Matthews, R. C. Cashner, and F. Gelwick. 2013. Disturbance and trajectory of change in a stream fish community over four decades. Oecologia 173:955–969. McMullen, L. E., P. De Leenheer, J. D. Tonkin, D. A. Lytle, and T. Wootton. 2017. High mortality and enhanced recovery: modelling the countervailing effects of disturbance on population dynamics. Ecology Letters 20:1566–1575. Mims, M. C., and J. D. Olden. 2012. Life history theory predicts fish assemblage response to hydrologic regimes. Ecology 93:35–45. Minckley, W. L., and J. E. Deacon. 1968. Southwestern fishes and the enigma of “endangered species”. Science 159:1424–1432. Minckley, W. L., and G. K. Meffe. 1987. Differential selection for native fishes by flooding in stream communities of the American southwest. Pages 93–104 in W. J. Matthews and D. E. Heins, editors. Evolutionary and community ecology of North American stream fishes. University of Oklahoma Press, Norman, Oklahoma, USA. Moore, J. W., and J. D. Olden. 2017. Response diversity, nonnative species, and disassembly rules buffer freshwater ecosystem processes from anthropogenic change. Global Change Biology 23:1871–1880. Neary, D. G., A. L. Medina, and J. N. Rinne. 2012. Synthesis of Upper Verde River research and monitoring 1993–2008. General Technical Report. RMRS-GTR-291. USDA Forest Service, Rocky Mountain Research Station, Fort Collins, Colorado, USA. Olden, J. D., N. L. Poff, and K. R. Bestgen. 2006a. Life-history strategies predict fish invasions and extirpations in the Colorado River Basin. Ecological Monographs 76:25–40. Olden, J. D., M. K. Joy, and R. G. Death. 2006b. Rediscovering the species in community-wide predictive modeling. Ecological Applications 16:1449–1460. Olden, J. D., and N. L. Poff. 2005. Long-term trends of native and non-native fish faunas in the American Southwest. Animal Biodiversity and Conservation 28.1:75–89. Phillips, J. V., and T. D. Ingersoll. 1998. Verification of roughness coefficients for selected natural and
constructed stream channels in Arizona. U.S. Geological Survey professional paper 1584. U.S. Geological Survey, Reston, Virginia, USA.

Poff, N. L. 2018. Beyond the natural flow regime? Broadening the hydro-ecological foundation to meet environmental flows challenges in a non-stationary world. Freshwater Biology 63:1011–1021.

Pool, T. K., J. D. Olden, J. B. Whittier, and C. P. Paukert. 2010. Environmental drivers of fish functional diversity and composition in the Lower Colorado River Basin. Canadian Journal of Fisheries and Aquatic Sciences 67:1791–1807.

Propst, D. L., and K. B. Gido. 2004. Responses of native and nonnative fishes to natural flow regime mimics in the San Juan River. Transactions of the American Fisheries Society 133:922–931.

Propst, D. L., K. B. Gido, and J. A. Stefferud. 2008. Natural flow regimes, nonnative fishes, and native fish persistence in arid–land river systems. Ecological Applications 18:1236–1252.

Propst, D. L., K. B. Gido, J. E. Whitney, E. I. Gilbert, T. J. Pilger, A. M. Monie, Y. M. Paroz, J. M. Wick, J. A. Monzingo, and D. M. Myers. 2015. Efficacy of mechanically removing nonnative predators from a desert stream. River Research and Applications 31:692–703.

R Core Team. 2017. R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria.

Regan, H. M., H. R. Akçakaya, S. Ferson, K. V. Root, S. Carroll, and L. R. Ginzburg. 2003. Treatments of uncertainty and variability in ecological risk assessment of single-species populations. Human and Ecological Risk Assessment 9:889–906.

Rinne, J. N. 2005. Changes in fish assemblages in the Verde River, Arizona. Pages 115–126 in J. N. Rinne, R. M. Hughes, and B. Calamusso, editors. Historical changes in fish assemblages of large rivers in the Americas. Symposium 45, American Fisheries Society, Bethesda, Maryland, USA.

Rinne, J. N., and D. Miller. 2006. Hydrology, geomorphology and management: implications for sustainability of native Southwestern fishes. Reviews in Fisheries Science 14:91–110.

Rinne, J. N., J. A. Stefferud, A. Clark, and P. Sponholtz. 1998. Fish community structure in the Verde River, Arizona, 1974–1997. Hydrology and Water Resources in Arizona and the Southwest 28:75–80.

Rolls, R. J., J. Heino, and B. C. Chessman. 2016. Unravelling the joint effects of flow regime, climatic variability and dispersal mode on beta diversity of riverine communities. Freshwater Biology 61:1350–1364.

Rolls, R. J., C. Leigh, and F. Sheldon. 2012. Mechanistic effects of low-flow hydrology on riverine ecosystems: ecological principles and consequences of alteration. Freshwater Science 31:1163–1186.

Rosenfeld, J. S. 2017. Developing flow-ecology relationships: implications of nonlinear biological responses for water management. Freshwater Biology 62:1305–1324.

Ruhí, A., E. E. Holmes, J. N. Rinne, and J. L. Sabo. 2015. Anomalous droughts, not invasion, decrease persistence of native fishes in a desert river. Global Change Biology 21:1482–1496.

Ruhí, A., J. D. Olden, and J. L. Sabo. 2016. Declining streamflow induces collapse and replacement of native fish in the American Southwest. Frontiers in Ecology and the Environment 14:465–472.

Ruppert, J. B., R. T. Muth, and T. P. Nesler. 1993. Predation on fish larvae by adult red shiner, Yampa and Green Rivers, Colorado. Southwestern Naturalist 38:397–399.

Scollo, I. J. 1985. Flow regime, juvenile abundance, and the assemblage structure of stream fishes. Ecology 66:1484–1490.

Seager, R., M. Ting, C. Li, N. Naik, B. Cook, J. Nakamura, and H. Liu. 2013. Projections of declining surface-water availability for the southwestern United States. Nature Climate Change 3:482–486.

Shenton, W., N. R. Bond, J. D. L. Yen, and R. Mac Nally. 2012. Putting the “ecology” into environmental flows: ecological dynamics and demographic modelling. Environmental Management 50:1–10.

Smith, S. M., J. S. Odenkirk, and S. J. Reeser. 2005. Smallmouth bass recruitment variability and its relation to stream discharge in three Virginia Rivers. North American Journal of Fisheries Management 25:1112–1121.

Solander, K. C., K. E. Bennett, and R. S. Middleton. 2017. Shifts in historical streamflow extremes in the Colorado River Basin. Journal of Hydrology: Regional Studies 12:363–377.

Steffeurud, J. A., K. B. Gido, and D. L. Propst. 2011. Spatially variable response of native fish assemblages to discharge, predators and habitat characteristics in an arid–land river. Freshwater Biology 56:1403–1416.

Steffeurud, J. A., and J. N. Rinne. 1995. Sustainability of fishes in desert river: preliminary observations on the roles of streamflow and introduced fishes. Hydrology and Water Resources in Arizona and the Southwest 22:25–32.

Steffeurud, J. A., and S. E. Stefferud. 1998. Influence of low flows on abundance of fish in the upper San Pedro River, Arizona. Pages 167–181 in Cross Border Waters: Fragile Treasures for the 21st Century, 9th US/Mexico Border States Conference on Recreation, Parks, and Wildlife. Tucson, Arizona, June 3–6, 1998. Proceedings RMRS-P-5. USDA Forest
Service, Rocky Mountain Research Station, Fort Collins, Colorado, USA.

Stott, I. 2016. Perturbation analysis of transient population dynamics using matrix projection models. Methods in Ecology and Evolution 7:666–678.

Strecker, A. L., J. D. Olden, J. B. Whittier, and C. P. Paukert. 2011. Defining conservation priorities for freshwater fishes according to taxonomic, functional, and phylogenetic diversity. Ecological Applications 21:3002–3013.

Tonkin, J. D., D. M. Merritt, J. D. Olden, L. V. Reynolds, and D. A. Lyle. 2018. Flow regime alteration degrades ecological networks in riparian ecosystems. Nature Ecology & Evolution 2:86–93.

Turner, D. S., and M. D. List. 2007. Habitat mapping and conservation analysis to identify critical streams for Arizona’s native fish. Aquatic Conservation: Marine and Freshwater Ecosystems 17:737–748.

Udall, B., and J. Overpeck. 2017. The twenty-first century Colorado River hot drought and implications for the future. Water Resources Research 53:2404–2418.

van de Pol, M., Y. Vindenes, B.-E. Saether, S. Engen, B. J. Ens, K. Oosterbeek, and J. M. Tinbergen. 2010. Effects of climate change and variability on population dynamics in a long-lived shorebird. Ecology 91:1192–1204.

Walsworth, T. E., and P. Budy. 2015. Integrating nonnative species in niche models to prioritize native fish restoration activity locations along a desert river corridor. Transactions of the American Fisheries Society 144:667–681.

Warner, R. R., and P. L. Chesson. 1985. Coexistence mediated by recruitment fluctuations: a field guide to the storage effect. American Naturalist 125:769–787.

Wheeler, K., S. J. Wenger, and M. C. Freeman. 2018. States and rates: complementary approaches to developing flow-ecology relationships. Freshwater Biology 63:906–916.

Whitney, J. E., J. B. Whittier, C. P. Paukert, J. D. Olden, and A. L. Strecker. 2017. Forecasted range shifts of arid-land fishes in response to climate change. Reviews in Fish Biology and Fisheries 27:463–479.

Yen, J. D. L., N. R. Bond, W. Shenton, D. A. Spring, and R. Mac Nally. 2013. Identifying effective water-management strategies in variable climates using population dynamics models. Journal of Applied Ecology 50:691–701.

Zavaleta, E., J. Pasari, J. Moore, D. Hernandez, K. B. Suttle, and C. C. Wilmers. 2009. Ecosystem responses to community disassembly. Pages 311–333 in R. S. Ostfeld and W. H. Schlesinger, editors. The year in ecology and conservation biology 2009. Annals of the New York Academy of Sciences, Blackwell Publishing, Oxford, New York, USA.

DATA AVAILABILITY

Code and data are available at https://doi.org/10.5281/zenodo.1309024

SUPPORTING INFORMATION

Additional Supporting Information may be found online at: http://onlinelibrary.wiley.com/doi/10.1002/ecs2.2681/full