



Water storage decisions will determine the distribution and persistence of imperiled river fishes

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	<p>concerning water storage vary substantially throughout a river basin. In the absence of interspecific interactions, future Colorado River temperatures would appear to benefit both endemic and nonnative fish species. However, endemic species are already declining or extirpated in locations where their ranges overlap with warmwater nonnatives and changes in water storage may lead to warming in some of the coolest portions of the river basin, facilitating further nonnative expansion. Integrating environmental considerations into ongoing water storage negotiations may lead to better resource outcomes than mitigating nonnative species impacts after the fact.</p>

Title: Water storage decisions will determine the distribution and persistence of imperiled river fishes

Running Head: Storage determines fish distributions

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Abstract. Managing the world's freshwater supply to meet societal and environmental needs in a changing climate is one of the biggest challenges for the 21st century. Dams provide water security, however, the allocation of dwindling water supply among reservoirs could exacerbate or ameliorate the effects of climate change on aquatic communities. Here, we show that the relative sensitivity of river thermal regimes to direct impacts of climate change and societal decisions concerning water storage vary substantially throughout a river basin. In the absence of interspecific interactions, future Colorado River temperatures would appear to benefit both endemic and nonnative fish species. However, endemic species are already declining or extirpated in locations where their ranges overlap with warmwater nonnatives and changes in water storage may lead to warming in some of the coolest portions of the river basin, facilitating further nonnative expansion. Integrating environmental considerations into ongoing water storage negotiations may lead to better resource outcomes than mitigating nonnative species impacts after the fact.

Introduction. Reservoir operations that govern water storage and its release downstream are typically determined by large-scale water-supply policy that rarely considers ecological impacts to the river segments between reservoirs (Bair et al. 2019). River regulation has fragmented rivers on a global scale, altered natural flow regimes, disrupted sediment and organic material flux, severed long distance fish migrations, and created thermal discontinuities that profoundly influence aquatic communities (Ward and Stanford 1983, Schmidt and Wilcock 2008, Poff and Zimmerman 2010). Below large reservoirs, changes in storage affect river temperature and by extension aquatic communities. When full, most large reservoirs stratify and release cold water from the hypolimnion, and there is little seasonal variation. When relatively empty, water released from these same reservoirs is warmer, because withdrawals occur from the epilimnion

(Caissie 2006, Olden and Naiman 2010). As such, the extent of thermal modification can be strongly correlated with the depth of reservoir withdrawals (Appendix S1: Fig. S1).

Human consumption of water supply is expected to exacerbate climate-driven reductions in water availability in many areas (Haddeland et al. 2014), especially in arid regions already experiencing intensified drought and warming air temperatures (Udall and Overpeck 2017, Xiao et al. 2018). In some arid regions where water supply is already fully appropriated for human consumption, any additional decrease in water availability due to climate change will force reconsideration of longstanding water-supply allocation agreements, with inevitable changes in the rules governing reservoir storage and operation. We hypothesize that decisions arising from the renegotiation of water supply agreements have the potential to exacerbate or ameliorate the direct thermal effects caused by a warming climate on river segments below large reservoirs because the depth of water withdrawals can have a profound effect on riverine thermal regimes.

The Colorado River basin (hereafter, ‘basin’) is among the most highly regulated and overallocated river networks in the world, having the largest ratio of reservoir storage to mean annual flow in North America (Hirsch et al. 1990). The basin is also experiencing increasing air temperatures and declining watershed runoff, with subsequent reductions in water available for human consumption (Brekke et al. 2013, Udall and Overpeck 2017). Basin reservoirs, including the two largest in the United States, lakes Mead and Powell, store water to support agriculture, large human populations, and regional economies of the US and Mexico. The administrative framework that governs management of the river is known as *The Law of the River* and includes a bi-national treaty and its amendments, federal and state laws, a Supreme Court ruling, records of decisions arising from environmental impact statements, and other agreements. Starting in 2020, stakeholders will begin renegotiations concerning water supply allocation and reservoir

operations that will shape releases and storage volumes in response to future drought and water shortages. These negotiations have the potential to further alter thermal regimes through reservoir releases, and as such, native and nonnative fish assemblages across the basin.

The Colorado River basin was historically home to more than thirty mostly endemic native fish species, including the four ‘big river’ fishes that are federally endangered - Colorado pikeminnow (*Ptychocheilus lucius*), razorback sucker (*Xyrauchen texanus*), humpback chub (*Gila cypha*), and bonytail (*Gila elegans*)(Minckley and Deacon 1991, Mueller and Marsh 2002). Dams, diversions, and reservoirs have fundamentally changed the physical and biological template of the river and opened niche space for nonnative species through stabilization of flow regimes and thermal regime impairment (Olden et al. 2006, Bestgen and Hill 2016). In recent decades, rapidly spreading opportunistic nonnative fish have contributed to observed declines in native species (Martinez et al. 2014, Bestgen et al. 2018).

Here, we focus on the nexus among water-supply allocation policy, riverine thermal regimes controlled by reservoir releases, and fish community dynamics, and we examine how these relations might change in a warming climate where Colorado River watershed runoff declines. We describe the current thermal regime and its suitability for fish communities, predict water temperatures under climate change scenarios and water allocation strategies, and consider potential responses of native and nonnative fish populations to anticipated changes in water temperature. Further, we illustrate how societal decisions concerning water supply have the potential to trigger changes in fish communities across this highly engineered riverscape.

Methods. *Water temperature model.* We modified a heat-exchange model previously described by Wright et al. (2009) and empirically estimated parameters by fitting relationships to monthly average water temperature data collected from 1985-2015 at 44 gages along the Colorado,

Gunnison, Green, Yampa, Duchesne, White, San Juan, and Animas rivers. Solar radiation and air temperature represent the primary components in the simplified heat budget that determines river temperature (Appendix S2: Table S1). Our model also accounted for major tributaries (i.e., mean annual flow $\geq 10\%$ of the mainstem river). To estimate model suitability, we first fit the model to data from odd years and used data from even years to calculate the root mean square error (RMSE) and overall bias of out-of-sample prediction aggregated by river segment and month of year (Appendix S2: Table S2). We then fit the model using all data, producing estimates with similar means, but higher precision, and used these estimates to predict water temperatures for the current period (1985-2015) at a 1-river kilometer (rkm) resolution along 2,560 rkm of river.

Linking fish to temperature. We linked the distribution of native and nonnative species to river temperatures by first developing a thermal suitability metric for each of six important fish species based on literature review (Appendix S3; Table S1), and then fitting a generalized linear mixed model using this thermal suitability metric to predict the current status of each species in 16 river segments in the basin. The six species we modelled include three of the four endangered big-river fishes - Colorado pikeminnow, razorback sucker, and humpback chub. The fourth, bonytail, is extremely rare (Minckley and Deacon 1991). We also included three warmwater nonnative species in our analysis, smallmouth bass (*Micropterus dolomieu*), red shiner (*Cyprinella lutrensis*), and channel catfish (*Ictalurus punctatus*); these species have likely contributed to the decline of endangered big-river fishes via competition and predation (Johnson et al. 2008, Bestgen et al. 2018). We calculated the thermal suitability metric for each species in each segment by comparing monthly average water temperatures to the minimum, maximum, and optimal range of temperatures for growth of that species (Appendix S3: Table S1), and assigned a score from 0 (temperature too high or too low) to 1 (optimal temperature), with

fractions when temperatures were within minimum and maximum but not within the optimal range. These proportions were multiplied by the number of days in a month and summed across months. We categorized the status of each species as ‘never present’, ‘extirpated’, ‘rare’ or ‘common/abundant’ following Holden and Stalnaker (1975) using recent data from fish sampling efforts, government reports, and journal articles (Appendix S3: Table S2). For our analysis, we excluded river segments with a ‘never present’ status for a given species and classified extirpated/rare as 0 and common/abundant as 1. One exception was Colorado pikeminnow, which have not been common anywhere since the early 1900s when Laguna Dam was constructed (Mueller and Marsh 2002), so we classified extirpated as a 0 and rare a 1 (Appendix S3: Table S2). The generalized linear mixed model included random intercepts and slopes for species, a logit link, and a binomial error structure.

Case studies. While we hypothesized that thermal suitability would correlate with the current status of both native and nonnative fish species (see last section), we were also aware the long-term trend of the three native species was declining in segments that also support large populations of warmwater nonnatives. To highlight this important nuance, we summarized population-level data for humpback chub and Colorado pikeminnow as case studies relative to data on warmwater nonnatives. We compare Colorado pikeminnow abundance estimates to smallmouth bass removal efforts, which were modest prior to 2003 because the species had just invaded. Large smallmouth bass numbers in 2007 and 2012-2014 portray possible future responses where warm low flows may prevail.

Future water temperature scenarios. We used air temperature predictions from CMIP3 (SRES A.1B) and CMIP5 (RCP 4.5) models (n=109) to understand how climate change under moderate (somewhat reduced by mitigation) greenhouse gas emissions may change air

temperatures by mid-century (2040–2059) relative to 1950–1999. On average, air temperatures in the basin are predicted to increase by 1.85–3.01 °C per month by mid-century, averaging ~2.6 °C on an annual basis (Brekke et al. 2013). An annual air temperature increase of ~2.6 °C would reduce mean annual flow in the basin by ~17% mainly owing to greater evapotranspiration and sublimation (Udall and Overpeck 2017). We used predicted increases in monthly air temperature per Brekke et al. (2013), combined with predicted declines in Colorado River flow from Udall and Overpeck (2017), as inputs into our ‘climate change only’ model (Appendix S2: Table S3).

For our ‘storage + climate change’ scenario, we used nonlinear regression (least squares) to predict reservoir release temperatures as a function of storage elevation by month, which permitted an assessment of the degree to which changes in reservoir storage affect riverine thermal regimes relative to climate warming alone. This analysis included data spanning 1965–2015 from five large storage reservoirs in the basin (Fontenelle, Flaming Gorge, Navajo, Glen Canyon, Hoover; Appendix S2: Table S3). The predicted water temperature associated with the lowest recorded storage elevation for each reservoir was used to predict potential warming of releases if storage was deemphasized. As such, the low storage adjustment (Δ °C) represents predicted river temperature at the lowest storage after reservoirs initially filled relative to the base model (Appendix S2: Table S3). We calculated future thermal suitability and predicted status by fish species, in the absence of species interactions, using outputs from the climate change only and storage + climate change scenarios and compared outputs to current conditions.

Uncertainty. While our approach does not consider the full uncertainty in climate change forecasts, there is equal or greater uncertainty in how monthly flows out of or storage in reservoirs in the basin will be managed, because the rules for these decisions are currently being negotiated. There is also considerable uncertainty in nonflow management alternatives that may

be developed to mitigate threats posed by warmwater nonnatives. Our intent in comparing scenarios is to highlight the potential importance of reservoir storage decisions for water temperature and fish communities, acknowledging that more detailed modelling that fully incorporates uncertainties is an important next step. Data, models, and code generated during this study are available from the USGS ScienceBase-Catalog (Dibble et al. 2020).

Results. Water temperature model. The mean signed error (bias) of our water temperature model was -0.003°C and absolute error (RMSE) averaged 0.5°C . By month, RMSE ranged from 0.4 to 0.6°C . For individual river segments, signed error ranged from -0.2 to $+0.1^{\circ}\text{C}$ and RMSE ranged from 0.3 to 1.2°C , with most segments $<0.7^{\circ}\text{C}$ (Appendix S2: Table S2). We used parameters from the model that incorporated all data from 1985-2015 to generate temperature predictions throughout the basin at a 1-rkm resolution (Fig. 1). Different segments differ in the timing of their warmest six-month period (Appendix S2), the degree of discontinuity created by reservoirs (also see Appendix S1: Fig. S1), and the downstream rate of temperature recovery (Fig. 1). Artificially cold release temperatures persist over the entire length of some river segments, such as the Colorado River downstream from Lake Powell (476 rkm), never approximating ambient temperatures. In contrast, cool release temperatures in the Green River downstream from Flaming Gorge Reservoir persist for 90 rkm but then nearly equilibrate to unregulated tributary temperatures at the Yampa River confluence, 105 km downstream (Fig. 1).

Fish status and trends. There was a positive relationship between thermal suitability and current status for all six fish species (Appendix S3: Tables S3, S4), however closer examination of trends in humpback chub and Colorado pikeminnow illustrate the importance of interspecific interactions. In our statistical model describing current status, nonnative species all had higher slopes (and intercepts) than endangered species indicating greater nonnative species response to

179 increasing thermal suitability (Fig. 2; Appendix S3: Table S4). Throughout the basin, humpback
180 chub populations have declined where warmwater nonnative species are common (red arrows)
181 and are increasing where these species are rare (blue arrows, Fig. 3a). The greatest increases in
182 humpback chub populations have occurred in segments where thermal suitability has increased
183 AND nonnatives are rare, illustrating the interactive effects of water temperature and nonnative
184 abundance (Fig. 3a). Similarly, Colorado pikeminnow are not currently found anywhere in the
185 absence of nonnatives, and their rate of decline has increased concomitant with increases in
186 warmwater nonnative fish, likely due to reduced juvenile survival leading to low recruitment and
187 a smaller adult population (Fig. 3b). Notably, smallmouth bass populations have increased,
188 requiring intensive removal efforts, even though the habitat is currently less thermally suitable
189 for this species than Colorado pikeminnow (Fig. 3b). This is likely because nonnative fish have a
190 higher probability of being common in habitats with fewer thermally suitable days (Fig. 2).

191 *Future scenarios.* Climate change will increase water temperatures throughout the basin;
192 however, water storage decisions have greater potential to impact water temperature in some
193 river segments. Upstream from Lake Powell, release temperatures are less dependent on storage
194 levels, and warming occurs quickly downstream from reservoirs, so storage decisions are not as
195 significant for thermal regimes as the direct impacts of climate change on air temperature and
196 river flows. For example, the biggest change in thermal regime for the Green and San Juan rivers
197 occurs from the base model (interior color, Fig. 4a) to the climate model (middle color), whereas
198 additional heating from the storage model (exterior color) is minimal. In contrast, release
199 temperatures from lakes Powell and Mead are very sensitive to reservoir elevation and warming
200 is relatively slow. Warming temperatures increased thermal suitability for most species in most
201 river segments (Figs 4b-g; Appendix S3: Table S3). Consequently, models that ignore species

interactions suggest increasing probabilities that both native and nonnative species will become common in areas where they are not already common. Similar to raw temperature predictions, thermal suitability predictions are more sensitive to climate change in the upper portions of the basin, and more sensitive to water storage decisions below the large dams lower in the basin.

Discussion. Fish are ectotherms, and as such, the thermal regime of their environment is critically important in determining species distribution, abundance, and growth (Neuheimer and Taggart 2007, Isaak et al. 2017, Yackulic et al. 2018). There is substantial overlap in the thermal suitability of river segments for growth of warmwater native and nonnative fishes across the basin, and current evidence suggests nonnative species have a competitive or predatory advantage over native species in places where their ranges overlap (Olden et al. 2006, Johnson et al. 2008). Our analysis demonstrates that nonnative species in the basin have responded more strongly to recent river warming than native species. Thus, in the absence of effective management interventions, future warming is likely to disproportionately benefit nonnative species to the detriment of native species.

Resource managers are currently exploring flow management strategies to suppress smallmouth bass reproduction while also benefitting native endemic species through the timing of flood disturbance events (Bestgen and Hill 2016). These types of designer flood-flow regimes have been evaluated in other southwestern US rivers to maximize spawning and recruitment benefits to native fishes while disadvantaging nonnative predators (Tonkin et al. 2020). If flow management strategies are successful, then prioritizing storage rules that provide water for these flows will be important for the recovery of listed species. Upcoming renegotiations of storage rules are likely to consider how much water is consumptively used and where reservoir water is stored during periods of extended drought, thereby determining the proportional changes in mean

annual streamflow in different parts of the river network. Since less than 10% of the total reservoir storage capacity of the entire watershed occurs in reservoirs upstream from Lake Powell, modifying water storage rules to provide maximum flexibility in implementing designer flows will have a minimal impact on overall storage compared to the debate over the allocation of storage between the two largest reservoirs in the US – lakes Powell and Mead.

Recent declines in Lake Mead elevation have increased the extent of the Colorado River in western Grand Canyon by more than 100 rkm, and this segment is now dominated by native fish species. Lake Mead reduction below 346 meters above sea level (masl) has also led to the development of Pearce Ferry Rapid, where the river has cut through the reservoir delta outside the original channel, such that water now flows over a bedrock ledge. Fish biologists hypothesize this rapid is a barrier to movement of warmwater nonnative fishes from Lake Mead into upstream habitats that are currently suitable for nonnatives. Thus, if storage in Lake Powell were prioritized over Lake Mead in the future, Pearce Ferry Rapid would likely be maintained longer, but water temperatures in eastern Grand Canyon may return to unsuitably cold conditions for native fish species, similar to the 1980s and 1990s. Alternatively, prioritizing storage in Lake Mead over Lake Powell could lead to warmer water temperatures throughout Grand Canyon improving its thermal suitability for native and nonnative species alike. Two of the three native species we modelled (humpback chub, razorback sucker) are already present in this river segment and could benefit from improved thermal conditions, assuming of course that warmwater nonnatives can be effectively managed. Colorado pikeminnow, the third species we modeled, was last recorded in Grand Canyon in 1972 and it is thought that unfavorable thermal conditions played a major role in their extirpation; more favorable thermal conditions in Grand Canyon might aid potential reintroduction and recovery efforts for this species. The strategy of

prioritizing storage in Lake Mead over Powell would also open more than 100 rkm of riverine habitat upstream from Lake Powell in the lower San Juan and upper Colorado rivers. However, management of nonnatives in Grand Canyon under this scenario would be especially problematic because it could remove barriers to invasions by warmwater nonnatives from both Lake Mead via elimination of Pearce Ferry Rapid, and Lake Powell via elimination of the cold to cool-water barrier. An intermediate strategy would be to maintain the current, modified thermal regimes and physical barriers to invasion that exist in Grand Canyon by installation of selective water withdrawal on upstream Glen Canyon Dam or by maintaining intermediate storage levels in Lake Powell. This strategy would prevent water temperatures in Grand Canyon from recovering to natural patterns but may support persistence of the native-dominated fish community and aid continued growth and recovery for endangered humpback chub and razorback sucker by suppressing nonnative species.

Policy makers will soon commence a formal renegotiation process of the 2007 Interim Shortage Guidelines that are part of the *Law of the River*. This renegotiation has the potential to shape water policy across the basin for the next few decades. Central to this discussion will be allocation of the basin's dwindling water supply and changing the rules of reservoir storage and operation. During past negotiations, the consideration of impacts of storage decisions on resources in the intervening river segments between dams have not been a priority. However, as our analysis demonstrates, these future negotiations will have important implications for the recovery of endemic fishes basin-wide. While we have focused on evaluating conditions in the Colorado River basin, global climate change is likely to present similar challenges in other large river networks, particularly those in arid regions associated with declining watershed runoff. We suggest that our approach to modeling water temperatures, which explicitly considers the twin

levers of climate change and storage decisions in driving river temperatures, could inform management of native and nonnative fish populations globally.

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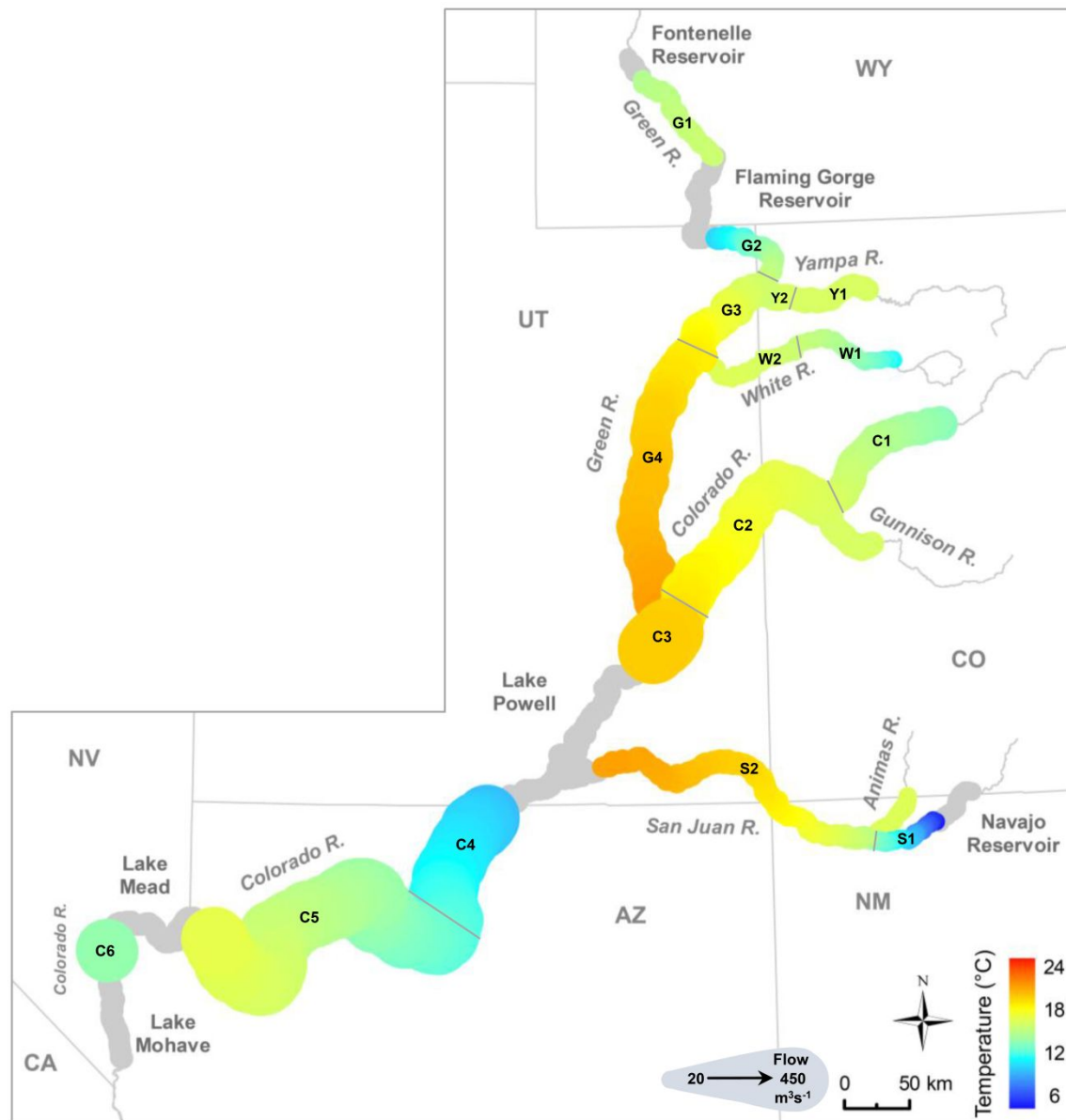
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Figure 1. Predicted river temperatures from the base model for the warmest growing season months by river segment (see Appendix S2). Segment width proportional to mean monthly flow.

Figure 2. Thermal suitability for: (a) Colorado pikeminnow, (b) razorback sucker, (c) humpback chub, (d) smallmouth bass, (e) red shiner, (f) channel catfish. Maps include a logistic regression of the probability a species is common relative to TSDs from the base model. Gray on LR plots= predictions beyond the data range; orange=zero TSDs; purple=increasing thermal suitability; black=species never present. Hash marks=current species status. Illustrations: Joseph Tomelleri.

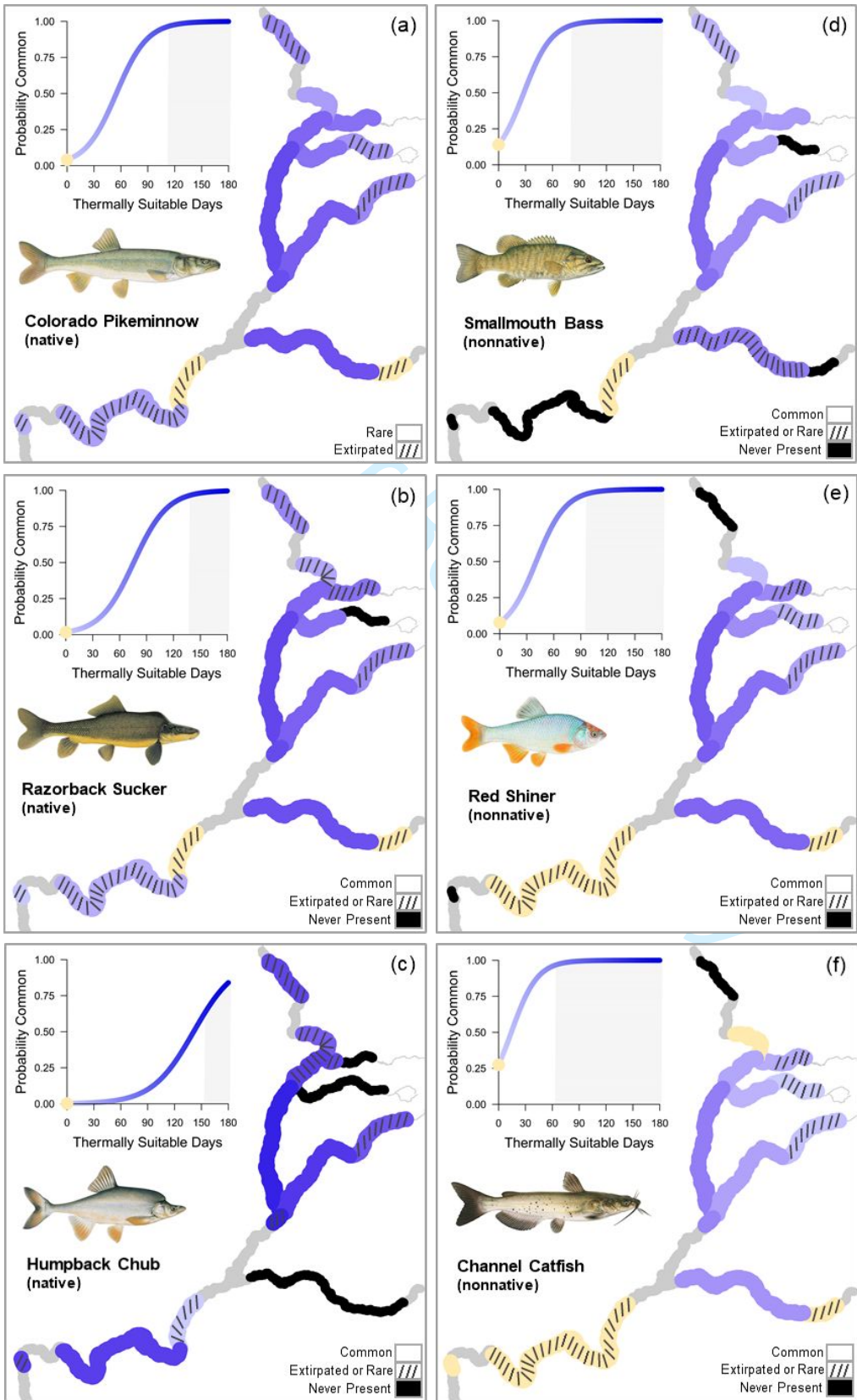
Figure 3. (a) Mean humpback chub estimates for Dinosaur National Monument (DNM), Desolation/Gray Canyons (DGC), Black Rocks (BR), Westwater Canyon (WC), Little Colorado River (LCR), and western Grand Canyon (WGC) relative to TSDs. An ‘*’ indicates estimate based on a portion of the population. Arrow bases represent 1990-1999 or 2000-2009, arrow tips represent 2010-2018. Red arrows represent overlap with warmwater (WW) nonnatives, blue lack overlap. (b) Adult Colorado pikeminnow estimates in the Green, Yampa, and White rivers (819 rkm; G3, G4, Y1, Y2, W2; blue), compared to smallmouth bass removed from a subset of that reach (247 rkm; G3, Y2; red). Smallmouth bass removals from 2012-2014 (15,750, 21,869, and 10187, respectively) capped at 10,000 to better show patterns. Point size proportional to TSDs.

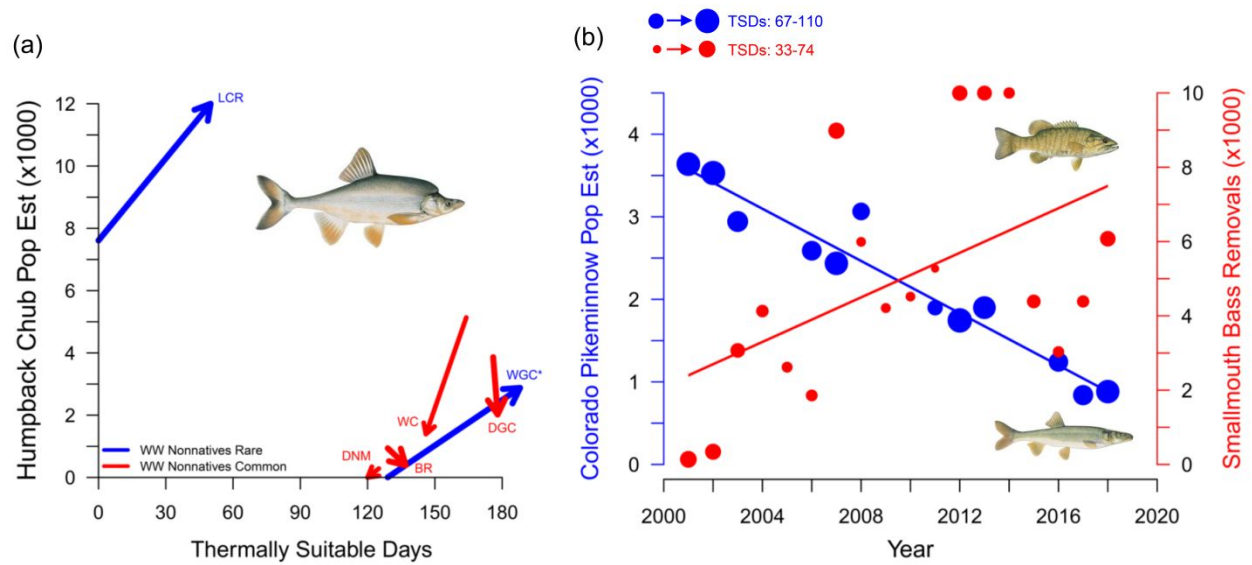
Figure 4. (a) Predicted river temperatures in the warmest months for the base (interior), climate (middle), and storage models (exterior). Water temperatures not substantially influenced by reservoir storage decisions shaded in gray (exterior). Probability (b) Colorado pikeminnow, (c) razorback sucker, (d) humpback chub, (e) smallmouth bass, (f) red shiner, and (g) channel catfish will become common based on predicted TSDs. Interior color represents current species status, middle and exterior colors indicate probability the species will become common based on predicted temperatures from the climate and storage models, respectively. Projections in reaches where fish species were historically absent can be interpreted as future thermal suitability.

365 **Figure 1**

366

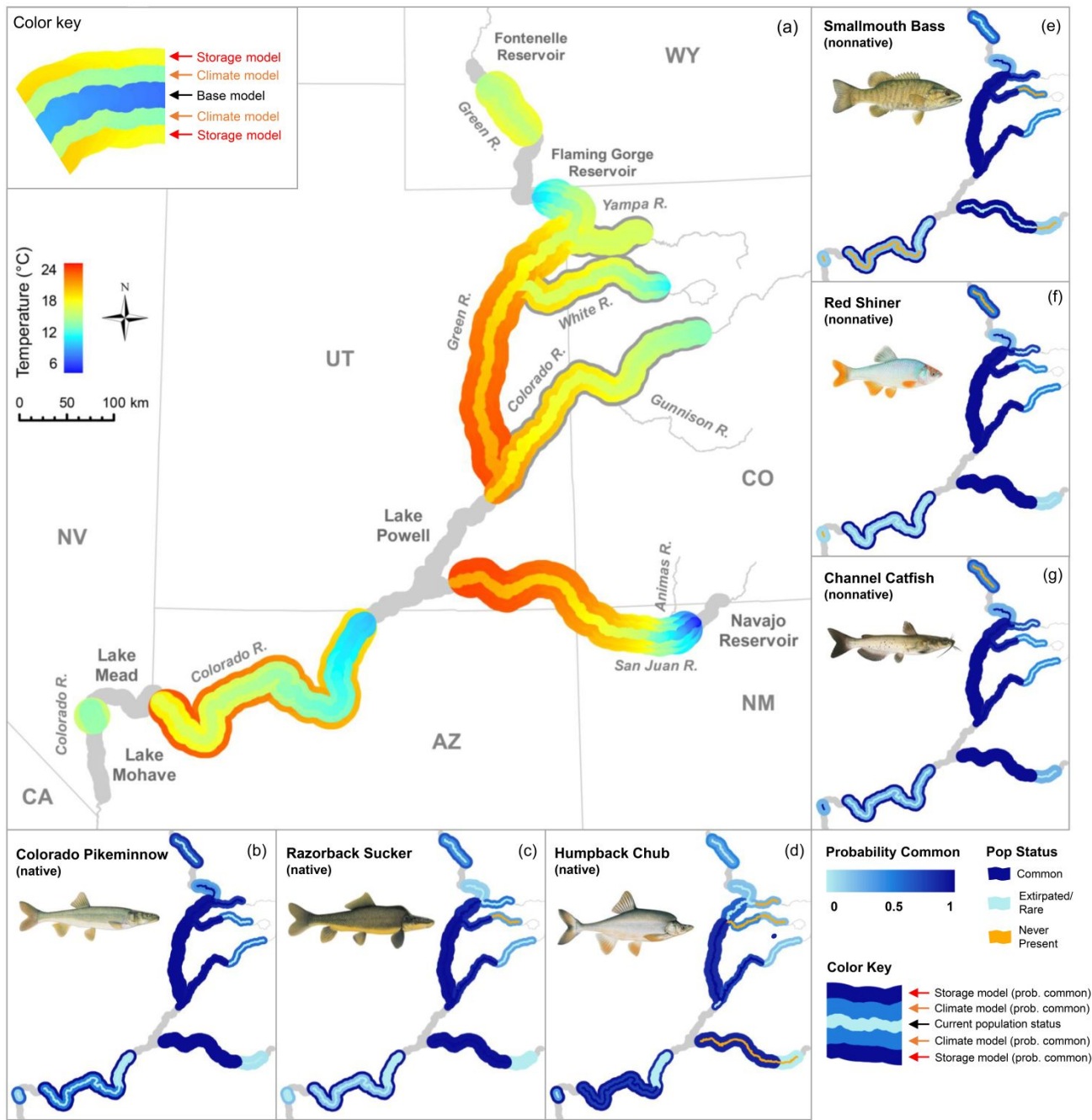
368 **Figure 2**



371 **Figure 3**

372

Figure 4



Supporting Information: Appendix S1

Journal: Ecological Applications

Title: Water storage decisions will determine the distribution and persistence of imperiled river fishes

Running Head: Storage determines fish distributions

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Appendix S1. Thermal discontinuity created by dams in western North America

We quantified the impact of reservoir withdrawal depth on the magnitude of thermal discontinuity downstream of 20 dams in the western US that spanned the Pacific Northwest, Missouri, Upper Colorado, Lower Colorado, and Arkansas-White-Red river basins. Temperature discontinuities were defined as the difference between mean monthly air temperature and river temperature in July. We calculated mean monthly river temperature (°C) at stations located within the first 25 kilometers downstream from each dam using data from the U.S. Geological Survey, National Water Information System (USGS NWIS) and the U.S. Fish and Wildlife Service, San Juan River Basin Recovery Implementation Program (USFWS SJRRIP). We located air temperature stations that were closest to each dam using the Haversine method ('geosphere' package; R Core Team 2018) and calculated mean monthly air temperature using data from the National Oceanic and Atmospheric Administration, National Climate Data Center (NOAA NCDC).

We estimated reservoir withdrawal depth by obtaining daily or monthly surface water elevation data (in meters above mean sea level, masl) from the USGS NWIS, U.S. Bureau of Reclamation (USBR) Hydromet System, U.S. Army Corps of Engineers (USACE), Montana Fish, Wildlife, and Parks, Natural Resources Conservation Service National Water and Climate Center, Colorado Division of Water Resources, and the Northern Colorado Water Conservancy District. We obtained fixed and variable penstock elevation data from descriptions and dam diagrams provided by the USBR, USACE, and from privately owned hydropower companies. We defined reservoir withdrawal depth in terms of the average withdrawal depth of water from fixed and variable penstocks relative to the reservoir surface. We estimated the degree to which reservoir withdrawal depth influences temperature discontinuity using simple linear regression.

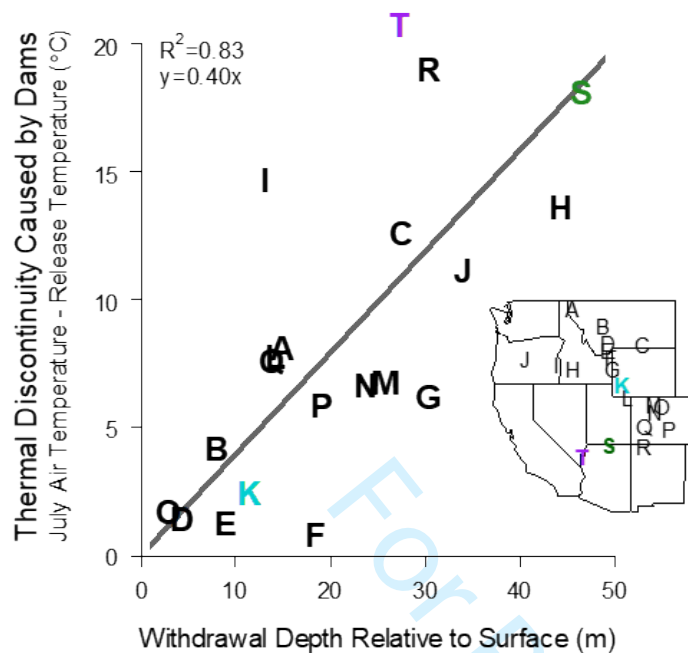


Figure S1. Thermal discontinuity is presented as the difference between mean air temperature and mean reservoir release temperature relative to withdrawal depth. Thermal discontinuities are largest downstream of large, high storage dams (e.g., ‘T’, Hoover Dam; ‘S’, Glen Canyon Dam) and lowest downstream from small run-of-river or low storage dams (e.g., ‘K’, Fontenelle Dam).

Map Key: A: Libby Dam, Kootenai River, MT; B: Holter Dam, Missouri River, MT; C: Yellowtail Dam, Bighorn River, MT; D: Madison Dam, Madison River, MT; E: Hebgen Dam, Madison River, MT; F: Island Park Dam, Henry’s Fork of the Snake River, ID; G: Palisades Dam, South Fork of the Snake River, ID; H: Anderson Ranch Dam, South Fork of the Boise River, ID; I: Owyhee Dam, Owyhee River, OR; J: Bowman Dam, Crooked River, OR; K: Fontenelle Dam, Green River, WY; L: Flaming Gorge Dam, Green River, UT; M: Ritschard Dam, Muddy Creek, CO; N: Green Mountain Dam, Blue River, CO; O: Windy Gap Dam, Colorado River, CO; P: Pueblo Dam, Arkansas River, CO; Q: Crystal Dam, Gunnison River, CO; R: Navajo Dam, San Juan River, NM; S: Glen Canyon Dam, Colorado River, AZ; and T: Hoover Dam, Colorado River, NV/AZ.

56 **Literature Cited**

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58 Statistical Computing, Vienna, Austria.

59

For Review Only

Supporting Information: Appendix S2

Journal: Ecological Applications

Title: Water storage decisions will determine the distribution and persistence of imperiled river fishes

Running Head: Storage determines fish distributions

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Appendix S2. Water temperature model

Data

Water Temperature and Discharge: We acquired subdaily or daily river temperature data from the U.S. Geological Survey, National Water Information System or Glen Canyon Dam Monitoring and Research Center (USGS), and the U.S. Fish and Wildlife Service, San Juan River Basin Recovery Implementation Program (USFWS) and calculated mean monthly water temperature. If continuous daily river temperature data were not available, we used point river temperature data from USGS NWIS water quality samples since point data were highly correlated with mean monthly data (Dibble et al. 2018). Due to the lack of river temperature data in the 24-km segment between the Colorado/Green River confluence and Lake Powell, we estimated initial river temperature for that segment by linear averaging of the Colorado and Green Rivers by flow volume, and then modeled temperatures by applying parameters β_0 , β_A , β_S , κ , b (Appendix S2: Table S1) from the adjacent upstream Colorado River segment (USGS gage 09085150 to Green River confluence). The 16-rkm segment from Hoover Dam to the inflow to Lake Mohave reservoir lacked multiple river temperature stations so we used data from one station and applied parameters from the adjacent upstream Colorado River segment (Appendix S2: Table S1; Glen Canyon Dam to Pearce Ferry). The White River lacked water temperature data to model temperatures from November to April.

The warmest growing season months of the year differed by river segments and included April-September for the Green River downstream from Flaming Gorge Dam, the San Juan River downstream from Navajo Dam, and the mainstem Colorado River. May-October was the warmest time period for the Green River downstream from Fontenelle Dam, the Yampa and White rivers, and the Colorado River downstream from Glen Canyon Dam, whereas June-

November was the warmest time period for the Colorado River downstream from Hoover Dam.

These time periods were used for visual purposes only in Figures 1 and 4.

We acquired subdaily discharge data from USGS NWIS gages closest to each dam or at the upstream boundary of each segment and calculated mean monthly values by year. See below for gages and data sources from the USGS and USFWS.

River	Gage ID	Source	River	Gage ID	Source
Animas	9364500	USGS	Duchesne	9302000	USGS
Animas	AR	USFWS	Green	9211200	USGS
Colorado	9085150	USGS	Green	9217000	USGS
Colorado	9095500	USGS	Green	9234500	USGS
Colorado	9163500	USGS	Green	9261000	USGS
Colorado	9180500	USGS	Green	9315000	USGS
Colorado	9380000	USGS	Green	404417108524900	USGS
Colorado	9383050	USGS	Gunnison	9152500	USGS
Colorado	9383100	USGS	San Juan	9379500	USGS
Colorado	9402352	USGS	San Juan	4C (or 09371010)	USFWS/USGS
Colorado	9402430	USGS	San Juan	AR (or 09355500)	USFWS/USGS
Colorado	9402500	USGS	San Juan	FM (or 09365000)	USFWS/USGS
Colorado	9403270	USGS	San Juan	LA	USFWS
Colorado	9404120	USGS	San Juan	ND	USFWS
Colorado	9404200	USGS	San Juan	SR (or 09368000)	USFWS/USGS
Colorado	9404220	USGS	White	9304200	USGS
Colorado	9421500	USGS	White	9304800	USGS
Colorado	CR033	USGS	White	9306224	USGS
Colorado	CR132	USGS	White	9306290	USGS
Colorado	CR149	USGS	White	9306500	USGS
Colorado	CR194	USGS	Yampa	9251000	USGS
Colorado	Glen Canyon Dam (release)	USGS	Yampa	9260050	USGS

Air Temperature: We acquired air temperature data from the National Oceanic and Atmospheric Administration, National Climate Data Center using stations located closest to each dam or tributary segment using the Haversine method ('geosphere' package; R Core Team 2018), with the closest stations listed below. We used these data to calculate mean monthly air temperature by year.

River	Dam	NCDC Location ID
Colorado	Colorado River mainstem	GHCND.USC00053307
Colorado	Colorado River mainstem	GHCND.USC00421240
Colorado	Colorado River mainstem	GHCND.USC00421241
Colorado	Colorado River mainstem	GHCND.USC00421243
Colorado	Colorado River mainstem	GHCND.USR0000CLIT
Colorado	Colorado River mainstem	GHCND.USR0000CPNR
Colorado	Glen Canyon	GHCND.USC00026180
Colorado	Glen Canyon	GHCND.USW00003162
Colorado	Hoover	GHCND.USC00028516
Colorado	Hoover	GHCND.USC00029376
Colorado	Hoover	GHCND.USC00260125
Colorado	Hoover	GHCND.USC00261371
Colorado	Hoover	GHCND.USW00023169
Green	Flaming Gorge	GHCND.USC00422864
Green	Fontenelle	GHCND.USC00484065
Green	Fontenelle	GHCND.USC00487847
San Juan	Navajo	GHCND.USC00420788
White	White River mainstem	GHCND.USW00094050
Yampa	Yampa River mainstem	GHCND.USC00055446

Solar Radiation: We assessed the effects of direct and reflected light on river temperatures using stations located midway along each segment by acquiring subdaily GHI data from the National Solar Radiation Database, Physical Solar Model (PSM, v. 3; data available 1998-2015) and Meteorological Statistical Models (MTS, v.1-2; data available 1985-2005). Since the PSM and MTS models overlapped from 1998-2005, we applied a slight correction to MTS data using the slope of the zero-intercept model for each river segment. We calculated mean GHI by month and year.

River	Dam	NSRDB Location ID	Latitude	Longitude	Model
Colorado	Colorado River mainstem	752	-108.56	39.13	MTS
Colorado	Colorado River mainstem	138108	-108.58	39.09	PSM
Colorado	Glen Canyon	553	-112.13	35.94	MTS
Colorado	Glen Canyon	0	-111.69	35.13	MTS
Colorado	Glen Canyon	97653	-112.18	35.45	PSM
Green	Flaming Gorge	1018	-109.50	40.44	MTS
Green	Flaming Gorge	153980	-109.50	40.41	PSM
Green	Fontenelle	149	-109.06	41.59	MTS
Green	Fontenelle	1024	-109.06	41.59	MTS
Green	Fontenelle	168092	-109.06	41.57	PSM

San Juan	Navajo	541	-108.25	36.75	MTS
San Juan	Navajo	110876	-108.26	36.69	PSM
White	White River mainstem	149614	-107.86	40.05	PSM
Yampa	Yampa River mainstem	1017	-107.56	40.50	MTS
Yampa	Yampa River mainstem	154518	-107.58	40.45	PSM

Reservoir Storage: We estimated the influence of mean monthly reservoir elevation on mean monthly reservoir release temperatures for five large dams in the basin using nonlinear regression (least squares) and historic data from Lake Powell (1965–1987, 1990–2002, 2004–2015), Lake Mead (1980, 1985–1990, 1996, 1998, 2011, 2013–2014), Fontenelle Reservoir (1966–1992, 2007–2015), Flaming Gorge Reservoir (1978–2000, 2002–2015; post-Selective Withdrawal Structure), and Navajo Reservoir (2000–2011), encompassing data from both low and high storage conditions. Results are presented in Appendix S2: Table S3.

Technical Details of the Model

Data: We used available data for water temperature, discharge, air temperature, and solar insolation from 1985–2015 (above). Mean monthly metrics were calculated by taking the mean daily measurement (from subdaily values, if available) and averaging daily values across each month by water year (WY; October 1–September 30).

Model description: Following Wright et al. (2009) and Walters et al. (2000), we begin by assuming a simplified heat balance equation for an advection dominated system can adequately describe changes in water temperature, $T(x)$, over longitudinal distance, x , below dams and between major tributary inputs by the following equation:

$$U \frac{dT(x)}{dx} = \kappa [T_e - T(x)] \quad (\text{Equation 1})$$

Where U is mean velocity, κ is a bulk surface-heat exchange coefficient and T_e is the equilibrium temperature (see Thomann and Mueller 1987). If we further assume U , κ , and T_e are

constant with respect to x , and define the upstream temperature conditions as T_o , we can integrate equation 1 to yield:

$$T(x) = T_e + (T_o - T_e)e^{-\frac{\kappa}{U}x} \quad (\text{Equation 2})$$

Wright et al. (2009) further simplified equation 2, while Walters et al. (2000) fit a version of equation 2. Past temperature modelling based on equation 2 or similar equations, modelled T_e by adding a constant to nearby observations of air temperature. While this approach is adequate for some uses it is problematic when making climate change predictions as solar insolation, not air temperature, is the primary driver of heat balance in most rivers, yet is not expected to change under climate change scenarios (unless cloud cover changes considerably) to the same extent as air temperature. Therefore, we estimated T_e as a function of standardized solar insolation (S) and air temperature (T_a) according to:

$$T_e = \beta_0 + \beta_A T_a + \beta_S S \quad (\text{Equation 3})$$

Where standardization consists of centering each covariate on its mean and dividing by its standard deviation to assist model fitting and β_0 , β_A , and β_S are estimated coefficients. Wright et al. (2009) and Walters et al. (2000) both relied on power law relationships between discharge, Q , and velocity, U , developed for the Grand Canyon river segment of the Colorado River basin from empirical methods and estimated κ from data. We took a different approach for two reasons: 1) such relationships have not been developed for other river segments in the basin, and 2) it is reasonable to assume that κ might also exhibit a power law relationship with discharge. If both κ and U exhibit power law relationships (i.e., $\kappa = a_1 Q^{b_1}$ and $U = a_2 Q^{b_2}$) it is impossible to estimate their individual powers or intercepts (i.e., a_1 , a_2 , b_1 , b_2), but straightforward to estimate

parameters representing the ratio of the intercepts (i.e., $\kappa = a_1/a_2$) and the difference of the powers (i.e., $b = b_1 - b_2$) allowing us to replace $\frac{\kappa}{U}$ in equation 2, with κQ^b .

The final difference between our approach and prior work by Walters et al. (2000) and Wright et al. (2009) was the inclusion of an additional equation to integrate data from major tributaries. Our approach was to set T_o equal to the discharge weighted average of the mainstem and tributary water temperatures for all points below a tributary and to replace the distance below the dam, x , in equation 2 with the distance below the nearest tributary. In other words, in a river segment with a tributary at distance, x_t , discharge, Q_t , and water temperature, T_t , we would replace T_o in equation 2 for all $x > x_t$ with T'_o given by:

$$T'_o = \frac{Q * T(x_t) + Q_t * T_t}{Q + Q_t} \quad (\text{Equation 4})$$

And x with x' defined as $x' = x - x_t$.

With enough replication under different values of S , T_a , Q and x , all five parameters (β_0 , β_A , β_S , κ , b) described above can be estimated for any river segment below a dam with a thermal discontinuity. Within the river segments we modelled in the Colorado River basin we found that gages (i.e., different values of x) were limiting in some segments, leading to imprecise estimates of κ . At the same time, we also found through preliminary analyses of river segments with multiple gages that κ was relatively conserved among river segments. Therefore, we decided to share κ across all river segments and independently estimate the other parameters for each river segment. To estimate parameters, we assumed observation errors were normally distributed. We fit models in the statistical package ‘Stan’ using the ‘rstan’ package (R Core Team 2018) and provide code in the data release.

Table S1. Water temperature model parameters

Parameters from the most parsimonious models used to predict river temperatures in parts of the Colorado River basin. “Segments” are identified in Figure 1. “WY” indicates the water years (October 1–September 30) included in models, with the number of individual years in parentheses. “rkm” is the number of river kilometers modeled for each river segment. Parameters β_0 , β_A , β_S , and b are specific to each model and river segment, while κ is a constant (see Appendix S2). Since the difference in parameters for a model using all data (odd and even years) vs. just odd years (for model validation step in Appendix S2: Table S2) was minor, we used parameters from the model that included all available data (below) to generate water temperature predictions.

River	Segment Description	Segments	WY	rkm	β_0	β_A	β_S	b	κ
Green	Fontenelle Dam to Flaming Gorge Dam	G1	1985-2015 (17)	117	12.2	6.3	2.6	0.5	0.1
Green	Flaming Gorge Dam to Colorado River confluence	G2, G3, G4	1999-2015 (17)	663	14.4	7.4	3.7	0.5	0.1
Yampa	USGS gage 09251000 to Green River confluence	Y1, Y2	2002-2015 (14)	142	11.7	5.3	3.6	1.0	0.1
White	USGS gage 09304200 to Green River confluence	W1, W2	2008-2013 (6)	283	17.9	8.2	0.8	0.5	0.1
San Juan	Navajo Dam to Lake Powell	S1, S2	1995-2015 (21)	410	14.7	8.8	0.1	0.5	0.1
Colorado	USGS gage 09085150 to Green River confluence	C1, C2, C3	2006-2015 (10)	428	15.4	8.1	4.2	0.7	0.1
Colorado	Glen Canyon Dam to Pearce Ferry	C4, C5	1995-2015 (21)	476	19.5	7.0	1.7	0.6	0.1
Colorado	Hoover Dam to Willow Beach (Lake Mohave)	C6	1985-2014 (30)	16	19.5	7.0	1.7	0.6	0.1

Table S2. Water temperature model fit

Model calibration and validation occurred using mean monthly data. Model calibration was conducted using half the water temperature data focused on odd years from 1985-2015, while we used data from even years and model parameters from the calibration step for validation of the model. Residuals (i.e., difference between model and observations) were used to analyze model fit. Signed and absolute errors (root mean square error) for the full model and by river segment are reported below. Signed errors provide information on systemic positive or negative biases in model results, whereas absolute errors indicate model fit.

River	Segment Description	Signed Errors	Absolute Errors (Root Mean Square Error)												
			All months	Jan	Feb	Mar	Apr	May	June	Jul	Aug	Sept	Oct	Nov	Dec
All	Full model; across river segments	-0.003	0.5	0.6	0.5	0.5	0.4	0.4	0.5	0.5	0.5	0.6	0.6	0.5	0.6
Green	Fontenelle Dam to Flaming Gorge Dam	0.02	1.2	0.7	1.8	1.1	1.2	1.8	0.7	1.2	0.8	1.8	1.5	1.2	0.4
Green	Flaming Gorge Dam to Colorado River confluence	-0.2	0.9	1.3	0.9	1.0	0.6	0.6	0.7	0.6	0.7	1.1	1.1	1.2	1.0
Yampa	USGS gage 09251000 to Green River confluence	0.1	0.7	0.7	0.8	0.7	0.5	0.8	0.8	1.0	0.7	0.7	0.2	0.4	0.7
White	USGS gage 09304200 to Green River confluence	-0.2	0.4	NA	NA	NA	NA	0.1	0.5	0.4	0.4	0.7	0.9	NA	NA
San Juan	Navajo Dam to Lake Powell	0.1	0.7	0.7	0.5	0.6	0.6	0.6	0.7	0.8	0.6	0.7	0.8	0.5	0.8
Colorado	USGS gage 09085150 to Green River confluence	-0.02	0.6	0.5	0.8	0.7	0.5	0.4	0.6	0.6	0.7	0.6	0.5	0.5	1.0
Colorado	Glen Canyon Dam to Pearce Ferry	-0.03	0.3	0.3	0.3	0.2	0.2	0.2	0.2	0.2	0.3	0.2	0.3	0.3	0.3
Colorado	Hoover Dam to Willow Beach (Lake Mohave)	-0.03	0.3	0.3	0.3	0.2	0.2	0.2	0.2	0.2	0.3	0.2	0.3	0.3	0.3

141 **Table S3. Climate change and low storage inputs to water temperature model**

142 We used air temperature predictions from CMIP3 (SRES A.1B) and CMIP5 (RCP 4.5) models (n=109) to understand how
143 climate change under a moderate emissions, mid-century (2040-2059) scenario may change relative to 1950-1999 in the Colorado
144 River basin. We took the expected increases in mean monthly air temperature per Brekke et al. (2013) combined with anticipated
145 declines in Colorado River flow (-17%) from Udall and Overpeck (2017) and used those as inputs into our climate model. We used
146 nonlinear regression (least squares) to predict reservoir release temperatures under low storage scenarios by month. The low storage
147 adjustment (Δ °C) in this table represents predicted river temperature at the lowest reservoir storage recorded by month, relative to
148 release temperatures predicted by the base model by river segment and month.

Month	Air Temperature Δ °C	River Flow Δ %Q	Low Storage Fontenelle Δ °C	Low Storage Flaming Gorge Δ °C	Low Storage Navajo Δ °C	Low Storage Glen Canyon Δ °C	Low Storage Hoover Δ °C
Jan	+2.39	-0.17	-2.94	+0.01	+0.10	-1.33	-0.18
Feb	+2.38	-0.17	-2.95	+0.98	-0.18	-0.67	-0.30
Mar	+2.31	-0.17	-1.00	+0.76	-0.45	+0.49	+0.28
April	+2.31	-0.17	+2.91	-0.22	-0.25	+1.00	+0.19
May	+2.79	-0.17	+2.67	-0.52	-0.48	+2.92	+1.68
June	+3.01	-0.17	+2.79	+0.40	-0.62	+6.11	+1.60
July	+2.81	-0.17	+3.51	-0.10	-0.35	+9.56	+2.46
Aug	+2.74	-0.17	+1.69	-1.23	-0.04	+9.38	+2.55
Sept	+2.74	-0.17	-1.54	-0.81	+0.14	+9.20	+1.98
Oct	+2.40	-0.17	-1.87	-1.01	+0.17	+11.55	+1.20
Nov	+1.85	-0.17	-3.65	-0.71	+0.01	+5.36	+0.66
Dec	+1.97	-0.17	-2.43	-0.68	-0.15	+0.62	+0.92
Climate Change Only Scenario							
Storage + Climate Change Scenario (includes Δ °C in air temperature and Δ %Q in discharge from climate model)							

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Supporting Information: Appendix S3

Journal: Ecological Applications

Title: Water storage decisions will determine the distribution and persistence of imperiled river fishes

Running Head: Storage determines fish distributions

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Appendix S3. Thermally suitable days

We estimated an index of thermal suitability, thermally suitable days (TSDs), from predicted monthly water temperature using the following equation:

$$TSD = \left(\sum_{i=1}^{12} f(T_i) \right) * 30 \quad (\text{Equation 1})$$

$$\text{where } f(T_i) = 0, \quad \text{if } T_i < T_{min} \mid T_i > T_{max}$$

$$f(T_i) = \frac{T_i - T_{min}}{T_{min.opt} - T_{min}}, \quad \text{if } T_{min} < T_i < T_{min.opt}$$

$$f(T_i) = 1, \quad \text{if } T_{min.opt} < T_i < T_{max.opt}$$

$$f(T_i) = \frac{T_{max} - T_i}{T_{max} - T_{max.opt}}, \quad \text{if } T_{max.opt} < T_i < T_{max}$$

Where T_i is mean monthly water temperature in month i , and T_{min} , $T_{min.opt}$, $T_{max.opt}$, and T_{max} are the minimum, minimum-optimal, maximum-optimal, and maximum temperatures needed for growth by a given species (Appendix S2: Table S1).

Table S1. Water temperatures used to calculate thermally suitable days for fish species

This table includes water temperatures needed by species for fish growth. This includes minimum, minimum-optimal, maximum-optimal, and maximum temperatures for fish growth.

Species	Minimum	Minimum Optimal	Maximum Optimal	Maximum	References
Colorado Pikeminnow	13	22	31	37	(Bulkley et al. 1981, Black and Bulkley 1985, Osmundson 1987, Petersen and Paukert 2005, Bestgen et al. 2006, Lamarra 2007, Osmundson 2011, Valdez et al. 2013)
Razorback Sucker	14	21	31	37	(Bulkley and Pimentel 1983, Carveth et al. 2006, Lamarra 2007, Bestgen 2008, Valdez et al. 2013, USFWS 2018)
Humpback Chub	12	16	30	37	(Hamman 1982, Gorman and VanHoosen 2000, Voichick and Wright 2007 [pre-dam water temperatures], Dzul et al. 2017, USFWS 2017, Yackulic et al. 2018, USGS WaterWatch Data: Station 09383100)
Smallmouth Bass	16	25	29	37	(Coble 1967, Shuter et al. 1980, Jobling 1981, Smale and Rabeni 1995, Bestgen and Hill 2016b)
Red Shiner	16	22	30	38	(Gale 1986, Smale and Rabeni 1995, Lentsch et al. 1996, Carveth et al. 2006, Peterson 2017)
Channel Catfish	18	26	30	35	(Shrable et al. 1969, Andrews and Stickney 1972, Cheetham et al. 1976, Jobling 1981, McMahon and Terrell 1982, Wellborn 1990, Buentello et al. 2000, Lamarra 2007, Valdez et al. 2013)

Table S2. Native and nonnative fish species status across the Colorado River basin

Classifications are adapted from Holden and Stalnaker (1975), but we combine the ‘occasional’ and ‘rare’ categories and the ‘common’/‘abundant’ categories. NP=never present, Ext=extirpated (historically present but now absent), R=rare, C=common/abundant. Species population status was assigned using information from the following sources: (Bestgen et al. 2006, Johnson et al. 2008, Osmundson and White 2009, Albrecht et al. 2010, Breen et al. 2011, Gido and Propst 2012, Breton et al. 2014, Osmundson and White 2014, Breton et al. 2015, Marsh et al. 2015, Bestgen and Hill 2016a, Bestgen and Hill 2016b, Farrington et al. 2016, Rubenson and Olden 2016, Zelasko et al. 2016, Kegerries et al. 2017, Osmundson and White 2017, USFWS 2017, Bestgen et al. 2018, Miller 2018, Stone et al. 2018, USFWS 2018, Valdez 2018, Van Haverbeke et al. 2018, Yackulic and Hull 2019, Bestgen et al. 2020, Hedden and Gido 2020, USFWS 2020, Van Haverbeke et al. 2020). In the bottom half of the table we convert categories to numerals for use in the generalized linear mixed model. We excluded river segments with a ‘never present’ status and classified extirpated/rare as 0 and common/abundant as 1. One exception was Colorado pikeminnow, which have not been common anywhere since the early 1900s when Laguna Dam was constructed (Mueller and Marsh 2002), so we classified extirpated as a 0 and rare a 1.

River	Segment	Segment Description	rkm	Colorado Pikeminnow	Razorback Sucker	Humpback Chub	Smallmouth Bass	Red Shiner	Channel Catfish
Green	G1	Fontenelle Dam to Flaming Gorge Reservoir	1-117	Ext	Ext	Ext	R	NP	NP
Green	G2	Flaming Gorge Dam to Yampa River	1-103	R	Ext	Ext	C	C	C
Green	G3	Yampa River to White River	104-256	R	C	Ext	C	C	C
Green	G4	White River to Colorado River	257-663	R	C	C	C	C	C
Yampa	Y1	Yampa River, starts at USGS gage 09251000	1-97	R	Ext	NP	C	R	R
Yampa	Y2	Yampa River, lower 45 km to Green River	98-142	R	R	Ext	C	C	C
White	W1	USGS gage 09304200 to Kenney Reservoir	1-111	Ext	NP	NP	NP	R	R
White	W2	Kenney Reservoir to Green River	112-283	R	C	NP	C	C	C
San Juan	S1	Navajo Dam to Animas River	1-74	Ext	Ext	NP	NP	R	R

San Juan	S2	Animas River confluence to Lake Powell	75-410	R	C	NP	R	C	C
Colorado	C1	USGS gage 09085150 to Gunnison River	1-148	Ext	Ext	Ext	R	R	R
Colorado	C2	Gunnison River to Green River	149-428	R	C	C	C	C	C
Colorado	C3	Green River confluence to Lake Powell	429-452	R	C	R	C	C	C
Colorado	C4	Glen Canyon Dam to Little Colorado River	1-124	Ext	Ext	R	R	R	R
Colorado	C5	Little Colorado River to Lake Mead	125-476	Ext	R	C	NP	R	R
Colorado	C6	Hoover Dam to Lake Mohave	1-16	Ext	Ext	Ext	NP	NP	C
Green	G1	Fontenelle Dam to Flaming Gorge Reservoir	1-117	0	0	0	0	NP	NP
Green	G2	Flaming Gorge Dam to Yampa River	1-103	1	0	0	1	1	1
Green	G3	Yampa River to White River	104-256	1	1	0	1	1	1
Green	G4	White River to Colorado River	257-663	1	1	1	1	1	1
Yampa	Y1	Yampa River, starts at USGS gage 09251000	1-97	1	0	NP	1	0	0
Yampa	Y2	Yampa River, lower 45 km to Green River	98-142	1	0	0	1	1	1
White	W1	USGS gage 09304200 to Kenney Reservoir	1-111	0	NP	NP	NP	0	0
White	W2	Kenney Reservoir to Green River	112-283	1	1	NP	1	1	1
San Juan	S1	Navajo Dam to Animas River	1-74	0	0	NP	NP	0	0
San Juan	S2	Animas River confluence to Lake Powell	75-410	1	1	NP	0	1	1
Colorado	C1	USGS gage 09085150 to Gunnison River	1-148	0	0	0	0	0	0
Colorado	C2	Gunnison River to Green River	149-428	1	1	1	1	1	1
Colorado	C3	Green River confluence to Lake Powell	429-452	1	1	0	1	1	1
Colorado	C4	Glen Canyon Dam to Little Colorado River	1-124	0	0	0	0	0	0
Colorado	C5	Little Colorado River to Lake Mead	125-476	0	0	1	NP	0	0
Colorado	C6	Hoover Dam to Lake Mohave	1-16	0	0	0	NP	NP	1

Table S3. Thermally suitable days and probability fishes are common by river segment

This table includes the thermally suitable days calculated by segment for native and nonnative fish species in the Colorado River basin using mean modeled river temperatures from the base, climate, and storage models. Using this thermal suitability metric, we also calculated the probability that the species is common (for those species currently in each segment), and the probability the species will be common in the future given anticipated warming in the climate and storage models. NP=fish species ‘never present’ in the segment, per Appendix S3: Table S2. “--” indicates temperatures from the storage model were not available to compute thermally suitable days because the segment was not affected by a dam.

Species	Segment	Thermally Suitable Days			Probability Species is Common		
		Base Model	Climate Model	Storage Model	Base Model	Climate Model	Storage Model
Colorado Pikeminnow	G1	59	68	89	0.56	0.68	0.88
Colorado Pikeminnow	G2	37	48	44	0.26	0.40	0.35
Colorado Pikeminnow	G3	88	112	111	0.87	0.96	0.96
Colorado Pikeminnow	G4	115	144	144	0.97	0.99	0.99
Colorado Pikeminnow	Y1	78	81	--	0.79	0.82	--
Colorado Pikeminnow	Y2	77	84	--	0.78	0.84	--
Colorado Pikeminnow	W1	44	62	--	0.35	0.60	--
Colorado Pikeminnow	W2	75	104	--	0.76	0.94	--
Colorado Pikeminnow	S1	0	5	4	0.04	0.05	0.05
Colorado Pikeminnow	S2	108	146	146	0.95	0.99	0.99
Colorado Pikeminnow	C1	56	63	--	0.52	0.61	--
Colorado Pikeminnow	C2	95	116	--	0.91	0.97	--
Colorado Pikeminnow	C3	111	143	--	0.96	0.99	--
Colorado Pikeminnow	C4	0	0	132	0.04	0.04	0.99
Colorado Pikeminnow	C5	36	58	169	0.25	0.54	1.00
Colorado Pikeminnow	C6	22	23	66	0.13	0.14	0.65
Razorback Sucker	G1	59	70	90	0.29	0.43	0.68
Razorback Sucker	G2	31	44	39	0.09	0.16	0.12
Razorback Sucker	G3	89	113	112	0.67	0.88	0.87
Razorback Sucker	G4	118	142	142	0.90	0.97	0.97
Razorback Sucker	Y1	79	81	--	0.54	0.57	--
Razorback Sucker	Y2	78	84	--	0.53	0.61	--
Razorback Sucker	W1	40	63	--	NP	0.34	--
Razorback Sucker	W2	79	104	--	0.54	0.82	--
Razorback Sucker	S1	0	1	0	0.02	0.02	0.02
Razorback Sucker	S2	109	145	145	0.85	0.98	0.98
Razorback Sucker	C1	55	64	--	0.25	0.35	--

Razorback Sucker	C2	93	115	--	0.72	0.89	--
Razorback Sucker	C3	113	140	--	0.88	0.97	--
Razorback Sucker	C4	0	0	140	0.02	0.02	0.97
Razorback Sucker	C5	24	49	167	0.06	0.19	0.99
Razorback Sucker	C6	3	4	50	0.02	0.02	0.20
Humpback Chub	G1	122	126	140	0.30	0.34	0.49
Humpback Chub	G2	97	110	107	0.13	0.21	0.19
Humpback Chub	G3	125	156	154	0.33	0.65	0.63
Humpback Chub	G4	156	185	185	0.65	0.87	0.87
Humpback Chub	Y1	118	118	--	NP	0.27	--
Humpback Chub	Y2	119	120	--	0.28	0.29	--
Humpback Chub	W1	97	112	--	NP	0.22	--
Humpback Chub	W2	123	139	--	NP	0.48	--
Humpback Chub	S1	6	26	24	NP	0.01	0.01
Humpback Chub	S2	159	193	193	NP	0.90	0.90
Humpback Chub	C1	105	108	--	0.17	0.19	--
Humpback Chub	C2	133	168	--	0.41	0.76	--
Humpback Chub	C3	156	188	--	0.65	0.88	--
Humpback Chub	C4	1	5	187	0.00	0.00	0.88
Humpback Chub	C5	127	170	222	0.35	0.77	0.97
Humpback Chub	C6	126	129	216	0.34	0.37	0.96
Smallmouth Bass	G1	19	28	44	0.36	0.50	0.74
Smallmouth Bass	G2	7	13	10	0.21	0.28	0.24
Smallmouth Bass	G3	50	79	77	0.80	0.96	0.96
Smallmouth Bass	G4	83	110	110	0.97	0.99	0.99
Smallmouth Bass	Y1	39	42	--	0.67	0.71	--
Smallmouth Bass	Y2	38	45	--	0.65	0.75	--
Smallmouth Bass	W1	15	26	--	NP	0.47	--
Smallmouth Bass	W2	35	64	--	0.61	0.91	--
Smallmouth Bass	S1	0	0	0	NP	0.14	0.14
Smallmouth Bass	S2	68	104	104	0.93	0.99	0.99
Smallmouth Bass	C1	22	28	--	0.40	0.50	--
Smallmouth Bass	C2	53	78	--	0.83	0.96	--
Smallmouth Bass	C3	78	102	--	0.96	0.99	--
Smallmouth Bass	C4	0	0	74	0.14	0.14	0.95
Smallmouth Bass	C5	0	7	108	NP	0.21	0.99
Smallmouth Bass	C6	0	0	4	NP	0.14	0.18
Red Shiner	G1	29	41	65	NP	0.51	0.82
Red Shiner	G2	10	19	15	0.14	0.21	0.17
Red Shiner	G3	72	96	95	0.87	0.97	0.97
Red Shiner	G4	98	128	127	0.97	1.00	0.99
Red Shiner	Y1	58	63	--	0.74	0.80	--
Red Shiner	Y2	57	67	--	0.73	0.83	--
Red Shiner	W1	22	39	--	0.25	0.48	--
Red Shiner	W2	52	89	--	0.67	0.95	--
Red Shiner	S1	0	0	0	0.08	0.08	0.08
Red Shiner	S2	89	126	125	0.95	0.99	0.99
Red Shiner	C1	34	42	--	0.40	0.52	--
Red Shiner	C2	78	97	--	0.91	0.97	--

Red Shiner	C3	95	124	--	0.97	0.99	--
Red Shiner	C4	0	0	108	0.08	0.08	0.98
Red Shiner	C5	0	10	146	0.08	0.14	1.00
Red Shiner	C6	0	0	6	NP	0.08	0.11
Channel Catfish	G1	6	11	26	NP	0.45	0.69
Channel Catfish	G2	0	0	0	0.27	0.27	0.27
Channel Catfish	G3	31	59	57	0.76	0.96	0.95
Channel Catfish	G4	66	93	93	0.97	1.00	1.00
Channel Catfish	Y1	26	28	--	0.69	0.72	--
Channel Catfish	Y2	25	30	--	0.68	0.75	--
Channel Catfish	W1	1	12	--	0.29	0.46	--
Channel Catfish	W2	19	42	--	0.58	0.87	--
Channel Catfish	S1	0	0	0	0.27	0.27	0.27
Channel Catfish	S2	47	84	84	0.90	0.99	0.99
Channel Catfish	C1	9	13	--	0.41	0.48	--
Channel Catfish	C2	33	58	--	0.78	0.95	--
Channel Catfish	C3	58	87	--	0.95	0.99	--
Channel Catfish	C4	0	0	48	0.27	0.27	0.91
Channel Catfish	C5	0	0	81	0.27	0.27	0.99
Channel Catfish	C6	0	0	0	0.27	0.27	0.27

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Table S4. Generalized linear mixed model of species population status relative to thermally suitable days

Results from generalized linear mixed model to determine the probability a fish species is common relative to thermally suitable days in each river segment using mean predicted temperatures by month from the base model. The generalized linear mixed model included both random intercepts and slopes for species, a logit link, and a binomial error structure.

Model	Estimate	Std. Error	Z Value	Pr
Fixed Effects				
Intercept	-3.06	1.27	-2.41	0.02
Thermally Suitable Days	0.06	0.02	3.72	0.0002
Random Effects				
Species	Variance	Std. Dev		
	3.03	1.74		
	Intercept	Slope	Equation	
<i>Colorado Pikeminnow</i>	-0.10	-0.0005	$y = -3.16 + 0.06x$	
<i>Razorback Sucker</i>	-0.96	-0.005	$y = -4.02 + 0.05x$	
<i>Humpback Chub</i>	-3.00	-0.015	$y = -6.06 + 0.04x$	
<i>Smallmouth Bass</i>	1.25	0.006	$y = -1.80 + 0.06x$	
<i>Red Shiner</i>	0.59	0.003	$y = -2.47 + 0.06x$	
<i>Channel Catfish</i>	2.09	0.010	$y = -0.97 + 0.07x$	

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