Multiple species drive flexible lake food webs with warming

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Abstract

Climate change is rewiring the food webs that determine the fate of diverse ecosystems. Mobile generalist consumers are responding to climate change by rapidly shifting their behaviour and foraging, driving food webs to flex. Although these responsive generalists form a key stabilizing module in food web structure, the extent to which they are present throughout whole food webs is largely unknown. Here, we show that multiple species comprising key trophic roles drive flexible lake food webs with warming. By examining lakes that span a 7°C air temperature gradient, we found significant reductions in nearshore derived carbon and nearshore habitat use with increased temperature in three of four fish species. We also found evidence that the response of lake trout to increased temperatures may reduce their biomass and cascade to release their preferred prey, the pelagic forage fish cisco. Our results suggest that climate warming will shift lake food webs toward increased reliance on offshore habitats and resources. We argue that species across trophic levels broadly couple lake macrohabitats, suggesting that potentially stabilizing responsive consumers are present throughout food webs. However, climate change appears to limit their ability to responsively forage, critically undermining a repeated stabilizing mechanism in food webs.

Keywords: behavioral response, boreal shield, climate change, fish ecology, generalist module, habitat coupling, responsive foraging, mobile consumer, stable isotopes, trophic level

Introduction

Ongoing rapid climate change is altering species interactions and reorganizing the food webs that determine the fate of diverse ecosystems (Barton *et al.* 2009; Tunney *et al.* 2014; Kortsch *et al.* 2015). As environmental conditions change, food web structure changes—or 'flexes'—with it (McMeans *et al.* 2016). Identifying these changes in food web structure is central to understanding the consequences of climate change on the major energy pathways of carbon and nutrient flow in ecosystems but is challenging given that food web structure is inherently complex (Polis 1991). Importantly, climate change is expected to 'rewire' food webs (Bartley *et al.* n.d.) by adding or removing whole trophic links (e.g., Kortsch *et al.* 2015) and by modifying the strengths of existing interactions (e.g., Barton & Schmitz 2009). These changes in the flow of energy among species are likely to have myriad consequences in ecosystems, from species to whole ecosystem function (Křivan & Schmitz 2003; Kondoh 2010; Guzzo *et al.* 2017). If so, future warming may alter food webs in ways that lead to ecosystems that differ substantially from past and present conditions (Kortsch *et al.* 2012). Despite the potential for these drastic changes, our understanding of food webs responses to climate change and the consequences of these response remain limited.

Food webs are flexible because of the foraging responses of mobile generalist consumers that can rapidly shift their foraging behaviour in response to environmental change. Many generalist consumers move across the landscape to forage in various spatially-distinct macrohabitats (Rooney *et al.* 2008) and can feed omnivorously across trophic levels (Thompson *et al.* 2007). As a result, these generalist consumers can respond in two key ways: by changing the degree to which they couple different macrohabitats (i.e., 'horizontal' shifts between energy pathways, Schindler & Scheuerell 2002; Vander Zanden & Vadeboncoeur 2002; Rooney *et al.* 2008) and

by shifting their degree of omnivory (i.e., 'vertical' foraging shifts along an energy pathway,

Barton & Schmitz 2009; Sentis *et al.* 2014; Ruiz-Cooley *et al.* 2017). Generalists that exhibit

both horizontal and vertical foraging flexibility form the generalist module (*sensu* McMeans *et*

al. 2016), a widely documented patterning of food web interactions (Polis *et al.* 1997; Nakano &

Murakami 2001; Rooney *et al.* 2008; Eloranta *et al.* 2015) that can expand and contract food

webs in response to changing environmental conditions (Tunney *et al.* 2012). This generalist

module may be a powerful stabilizing structure, allowing mobile generalists to respond to

changing densities in different habitats in a manner that buffers population dynamics (McCann &

Rooney 2009; McMeans *et al.* 2016). Despite this potential, few studies have empirically

documented the presence of fundamental module throughout food webs or how human impacts

like climate change may interrupt this module.

One well-documented example of a responsive generalist consumer that drives food web

flexibility is lake trout (*Salvelinus namaycush*), a common generalist top predator in lakes of

boreal North America. Lake trout forage on fishes and invertebrates in both of the two thermally

distinct lake macrohabitats: the nearshore (i.e., littoral) habitat and the offshore (i.e., pelagic)

habitat (Vander Zanden & Vadeboncoeur 2002). Because lake trout is cold-adapted and

thermally sensitive, the foraging habits of lake trout are mediated by thermal accessibility of the

relatively warm nearshore habitat (Plumb & Blanchfield 2009; Guzzo *et al.* 2017). Lake trout

show reduced nearshore habitat use with increasing nearshore temperature (Tunney *et al.* 2014;

Guzzo *et al.* 2017). When lake trout avoid entering the physiologically taxing nearshore habitat,

they forage less on nearshore resources and more on fish (Tunney *et al.* 2012, 2014). This

reduced access to abundant nearshore prey may undermine lake trout's ability to responsively

forage to resource densities in different macrohabitats (Tunney *et al.* 2014; Guzzo *et al.* 2017).

Importantly, this phenomenon may not be limited to lake trout; many fishes are generally highly

mobile and flexible foragers (Dill 1983), and boreal lakes in North America contain numerous

cold-adapted fish species that may exhibit similar behavioural and foraging responses due to

thermal limitations (Vander Zanden & Vadeboncoeur 2002; Hasnain *et al.* 2013). In contrast, the

species that prefer warmer waters may not be thermally limited and thus may not exhibit

behavioural or foraging responses to warming. However, the roles that species other than lake

trout play in determining how lake food webs flex are largely unexplored. To understand how

whole food webs are being altered by climate change requires that we examine how many

species across trophic levels and with various thermal preferences respond to warming.

Here, we use a natural climate gradient of approximately 7°C to examine how multiple

species behavioural and foraging responses drive flexes in food web structure. We investigate the

habitat use and diet of four widespread and abundant species of boreal shield fishes that

represent different key roles and trophic levels in the food web: the cold-adapted generalist top

predator lake trout, the cold-adapted pelagic planktivorous cisco (*Coregonus artedi*), the cool-

adapted piscivorous top predator walleye (*Sander vitreus*), and the cool-adapted mesopredator

yellow perch (*Perca flavescens*) (Coker *et al.* 2001; Hasnain *et al.* 2013). We use stable-isotope-

based food-web indices of nearshore feeding and trophic position along with catch-per-unit-

effort based metrics of behaviour to show that several species across trophic levels and from

multiple thermal guilds respond predictably to changes in temperature. We also use catch-per-

unit-effort data to show that lake trout show reduced biomass index under increased

temperatures, consistent with a reduction of the nearshore resource availability. Our results

suggestion that multiple species comprising key trophic roles drive flexible lake food webs with

warming. We argue that by studying the foraging and behavioral responses of this set of key

players, we can predict how food webs will flex with climate change. We end by discussing how

these changes suggest climate change will impact the function and stability of lake ecosystems,

and how many generalist consumers may be a fundamental feature of food web architecture.

Methods

Lake Selection

The Canadian boreal shield includes hundreds of thousands of lakes that span various

natural environmental gradients, including climate (Gunn *et al.* 2004; Keller 2007). We used data

for 66 lakes in the province of Ontario, all of which have been used previously to study food web

structure through stable isotope analysis (Dolson *et al.* 2009; Tunney *et al.* 2014, 2018). We use

lakes that have both nearshore and offshore intermediate consumers to ensure that changes in

food web structure or predator behaviour are not driven by differences in the presence or absence

of whole trophic groups. For 59 of our 66 lakes, we used catch-per-unit-effort (CUE) data from

the Ontario Ministry of Natural Resources and Forestry (OMNRF) Broad-scale Fisheries

Monitoring (BSM) Program (Sandstrom *et al.* 2013). In summary, for each lake, standardized

fish community surveys took place one time between May and September from 2008 to 2012

using overnight sets of two types of multipaneled mesh gillnets: North American standard

multipaneled gill nets (with 8 mesh sizes varying from 38 to 127 mm) and small mesh gillnets

(with 5 mesh sizes from 13 to 38 mm). Netting occurred randomly throughout the lake and

depending on the maximum depth of each lake, in each of up to 8 depth strata: 1-3m, 3-6m, 6-

12m, 12-20m, 20-35m, 35-50m, 50-75m, and >75m. Complete details can be found in Sandstrom

et al. 2013. Catch data were not available for the 7 lakes from Dolson *et al.* (2009).

Lake Attributes

We used two measures of temperature for our analysis: average recent air temperature in

°C growing degree days > 5 °C (Table S1). Our primary temperature variable was average recent

air temperature, which was calculated as the average hourly air temperature in °C for the 30 days

prior to field sampling because it corresponds to the time period reflected by the isotopic

signature of fish muscle tissue (Peterson & Fry 1987). This temperature variable was derived

from FetchClimate (Grechka *et al.* 2016) using the latitude and longitude of each lake. In

addition to the average recent air temperature, we corroborated our results with and growing

degree days above 5°C from 1981 to 2010 (provided by the Ontario Ministry of Natural

Resources and Forestry), which measures the accumulation of heat over time. We also account

for several physical and chemical factors that have been previously identified as important

drivers of habitat coupling or trophic position in boreal shield lakes (Table 1, all provided by the

Ontario Ministry of Natural Resources, Dolson *et al.* 2009; Tunney *et al.* 2014, 2018): lake

surface area in hectares, mean lake depth in metres, Secchi depth in metres, and total phosphorus

in µg•L^{-1}, and lake shape (calculated the shoreline development index):

$$SDI = \frac{Shoreline\ Distance}{2\sqrt{\pi \times Surface\ Area}}$$

Species Selection

We used four species that are typical of their respective trophic groups. Stable isotope

data for these species was available for 40 or more lakes and the trophic level, thermal

classifications, and habitat preferences for these species are known (Coker *et al.* 2001; Hasnain

et al. 2013). Lake trout (*Salvelinus namaycush*) is the most common top predator of offshore

habitats in boreal shield lakes, and much research on boreal shield lake food webs has focused on

this coldwater species (Dolson *et al.* 2009; Plumb & Blanchfield 2009; Tunney *et al.* 2014;

Guzzo *et al.* 2017). Cisco (or lake herring, *Coregonus artedi*) is one of the most common cold-

adapted planktivores and is a common prey item for lake trout and walleye. Walleye (*Sander*

vitreus) is a common cool-adapted piscivore and popular sport fish that is present in lakes across

the boreal shield. Yellow perch (*Perca flavescens*) is a widespread and abundant cool-adapted

mesopredator that is consumed by a wide variety of predatory fishes throughout its range,

including both lake trout and walleye. These four species comprise a large portion of the average

catch in the 59 lakes that we use here from the OMNRF's BsM surveys (see Supplementary

Information). The number of individual for each species sampled for stable isotope analysis in

each lake varied from 2 to 21 (mean 13.5) for lake trout, 3 to 32 (mean 17.0) for walleye, 1 to 20

(mean 10.8) for cisco, and 1 to 18 (mean 9.1) for yellow perch. For both lake trout and walleye,

only individuals greater than 250 mm were used for stable isotope analysis because these species

are known to show ontogenic diet shifts (Mittelbach & Persson 1998; Sherwood *et al.* 2002;

Galarowicz *et al.* 2006).

Food Web Metric Calculations using Stable Isotopes

As in many previous studies, we used stable isotopic signatures from our four fish species

and baseline invertebrates to calculate both the nearshore carbon index (based on the proportion

nearshore carbon, e.g., Tunney *et al.* 2014) and trophic position (Vander Zanden *et al.* 1999b, a;

Post 2002). Collection and processing methods for stable isotope data can be found in Dolson *et*

al. 2009; Tunney *et al.* 2014, 2018). We used baseline invertebrates from the nearshore and

offshore zones to account for variability in isotopic signatures across lakes. We incorporated data

for multiple trophic groups into both our nearshore and offshore baseline isotopic signatures to

reduce the number of estimated baseline values required for our analysis and to increase the

sample size for our baseline isotopic signatures. We corrected all δ^{13}C signatures using C:N

ratios as

$$\delta^{13}C_{corr} = \delta^{13}C_{raw} + (-3.32 + 0.99 * CN)$$

where $\delta^{13}C_{corr}$ is the corrected δ^{13}C signature, $\delta^{13}C_{raw}$ is the raw δ^{13}C signature, and CN is the

C:N ratio of that tissue sample (Post *et al.* 2007). For lakes that were missing either nearshore or

offshore baseline isotopic signatures, we estimated the δ^{13}C and δ^{15}N signatures of the missing

baseline using the available baseline and simple linear regression between the baselines across

lakes (see Supplementary Information).

We used two source mixing models to estimate the nearshore carbon index and the

trophic position of each species in each lake based on their relative isotopic signatures (Post

2002). We calculated the nearshore carbon index for each fish species as

$$NCI_{fish} = \frac{\delta^{13}C_{fish} - \delta^{13}C_{osb}}{\delta^{13}C_{nsb} - \delta^{13}C_{osb}}$$

where NCI_{fish} is the nearshore carbon index in the diet of a fish species, $\delta^{13}C_{fish}$ is the

average δ^{13}C signature for that fish species, $\delta^{13}C_{osb}$ is the average or estimated δ^{13}C signature

for all offshore baselines (i.e., mussels and/or zooplankton), and $\delta^{13}C_{nsb}$ is the average or

estimated δ^{13}C signature for all nearshore baselines (i.e., snails and/or aquatic insect larvae). The

nearshore carbon index is similar to the proportion nearshore carbon used by others (e.g., Tunney

et al. 2014) but is not constrained between 0 and 1. This approach minimizes data transformation

requirements of the stable isotopic signatures and increases the number of lakes in our analyses

while still allowing us to qualitatively understand changes in the relative contribution of

nearshore resources to fish diets.

Based on the nearshore feeding index, we estimated the trophic position of each species

as

$$TP_{fish} = 2 + \frac{\left(\delta^{15}N_{fish} - \left(\delta^{15}N_{nsb} \times NCI_{fish} + \delta^{15}N_{osb} \times (1 - NCI_{fish})\right)\right)}{3.4}$$

where TP_{fish} is the trophic position of a fish species, NFI_{fish} is the nearshore carbon index for that

fish species, $\delta^{15}N_{fish}$ is the average $\delta^{15}N$ signature for that fish species, $\delta^{15}N_{nsb}$ is the average or

estimated $\delta^{15}N$ signature for all nearshore baselines (i.e., snails and aquatic insect larvae),

$\delta^{15}N_{osb}$ is the average or estimated $\delta^{15}N$ signature for all offshore baselines (i.e., mussels and

zooplankton), 3.4 is the assumed increase in $\delta^{15}N$ due to fractionation (Post 2002), for each

trophic level, and 2 is the assumed trophic position of the baseline invertebrates.

Behaviour Metrics and Biomass Index Using Catch-per-unit-effort

We used catch-per-unit-effort data for each depth stratum (Sandstrom *et al.* 2013) to

calculate a weighted average mean depth of capture for each species as

$$SMD = \frac{\sum_{i=1}^{4} d_i \times CUE_i \times p_i}{\sum_{i=1}^{4} CUE_i}$$

where SMD is the mean depth of capture of a fish species, CUE_i is the CUE of that species

for depth stratum i, p_i is the proportion of the lake in depth stratum i, and d_i is the depth of at the

center of the depth range for stratum i (2m for stratum 1, 4.5m for stratum 2, 9m for stratum 3,

and 16m for stratum 4). We only used depth strata 1 through 4 to calculate mean depth of capture

because these are the only depth strata that are sampled with both small-mesh and large-mesh

gear (Sandstrom *et al.* 2013). The CUE data that we use here is the number of fish per 100 metre

nights. For CUE biomass index, we used the kg of fish per gang per 100 metre nights, calculated

by summing the area-weighted CUE (number of fish per 100 metre nights) for each depth

stratum and then multiplying by the mean weight of that fish species.

Statistical Analyses

To test for the effect of temperature on each response variable (nearshore carbon index,

trophic position, mean depth of capture, nearshore/offshore presence, and the log-transformed

CUE biomass index of each species), we used a two-step approach. Firstly, to account for

morphometric or chemical factors known or suspected to influence the foraging and behaviour of

these fish species, we first constructed multiple regression models for each response variable and

lake size in hectares, lake shape (SDI), Secchi depth in m, mean lake depth in m, and total

phosphorous in $\mu g \cdot L^{-1}$. We then compared all subsets of each full multivariate regression model

using AIC. We selected the best model as the model with the smallest number of explanatory

variables from the set of models within 2 AIC points of the lowest AIC value (Aho *et al.* 2014,

see Supplementary Information for best models). Secondly, to test the effect of each temperature

variable, we took the best model for each response variable and ran a multiple regression that

included the temperature variable (Figures 2, 3, and 4, Tables 3 and 4). To meet statistical

requirements, we used $\log_{10}$ transformed lake surface area and natural log transformed mean lake

depth and total phosphorous. We removed all data points with a Cook's distance greater than 1

from their respective regression models, and we used variance inflation factor to check for

inflated coefficient estimates in multiple linear regressions. All regression analyses were

performed in the R statistical language (v3.2.3), and model selection was conducted using the R

package "glmulti" (Calcagno 2013).

Results

Food Web Metric Calculations using Stable Isotopes

We found strong evidence that lake food web structure varied across an approximately 7° C

climate gradient. Three of the four species showed evidence of changes in nearshore feeding

with increased temperature. Both top predators (lake trout and walleye) showed a significant

decrease in the nearshore carbon index with increasing average recent air temperature, and the

cold-adapted cisco showed a similar marginally significant decrease (Figure 2a, c, and e, Table

3). The cool-adapted yellow perch showed no significant relationship between average recent air

temperature and the nearshore carbon index (Figure 2g, Table 3). Lake trout showed a significant

increase in trophic position with increased average recent air temperature (Figure 2b, Table 3),

but cisco, walleye and yellow perch showed no significant relationship between trophic position

with increasing average recent air temperature (Figure 2d, Table 3) (see Supplementary

Information and Table S3).

Behaviour Metrics and Biomass Index Using Catch-per-unit-effort

In agreement with changes in nearshore carbon index, the same three species (lake trout,

walleye, and cisco) showed strong evidence of reduced nearshore habitat use. All three species

showed a significant increase in mean depth of capture with increasing average recent air

temperature (Figure 3a, c, and e, Table 3). Consistent with these results, lake trout and cisco

showed a significant decrease in probability of nearshore presence and walleye showed a

marginally significant increase in probability of offshore presence with increasing average recent

air temperature (see Supplementary Information). In contrast, yellow perch showed no

relationship between either mean depth of capture or probability of offshore presence and

average recent air temperature (Figure 3g and inset, Table 3).

We found evidence of a reduction in biomass with increasing temperature for one

species, lake trout, which showed a significant decrease in CUE biomass index with increasing

average recent air temperature (Figure 3b, Table 3). In contrast, cisco, walleye, and yellow perch

showed no significant relationship between average recent air temperature CUE biomass index

(Figure 3d, f and h, Table 3).

Discussion

Understanding how climate change is altering whole food webs is a challenging but critical

endeavor for understanding the responses of whole ecosystems. Previous research showed that

lake trout (*Salvelinus namaycush*), a generalist top predator in lake ecosystems, responds to

increased temperature by reducing its nearshore habitat and resource use, and its degree of

omnivory (Tunney *et al.* 2014; Guzzo *et al.* 2017), thus forming a flexible 'generalist module'

(McMeans *et al.* 2016). Here, we extend that research to examine the impacts of climate

warming on four key species that represent different key roles and trophic levels in boreal lake

food webs: lake trout, walleye (*Sander vitreus*), cisco (*Coregonus artedi*), and yellow perch

(*Perca flavescens*). Our stable-isotope-based food-web metrics and our catch-per-unit-effort-

based behavioural metrics show a consistent response of species across multiple trophic levels to

warming to nearshore-derived carbon and offshore habitat use. In fact, temperature was the most

significant variable for predicting food web structure in this study (see Supplementary

Information). By harnessing the variation in food web structure across natural temperature

gradients, we have detected consistent food web responses to warming in boreal shield lake

ecosystems.

We found strong evidence of foraging and behavioural responses to increased temperature in

three of the four key food web members that we examined. As expected, our cold-adapted

species, the predatory lake trout and the planktivorous cisco, showed decreased nearshore

feeding and decreased near-shore habitat use with increasing temperatures. Our results are

consistent with the thermally accessibility hypothesis that has been previously shown for lake

trout (Tunney *et al.* 2014; Guzzo *et al.* 2017). Because offshore fish communities are largely

comprised of species that prefer cold water (Coker *et al.* 2001; Hasnain *et al.* 2013), these

thermal constraints likely operate on whole offshore fish communities, causing them to respond

uniformly to warming. We found that the cool-adapted piscivorous walleye showed behavioural

and foraging responses through decreased coupling and decreased nearshore habitat use with

higher temperature, but in contrast the cool-adapted mesopredator yellow perch showed no such

response. Since walleye and yellow perch have similar thermal tolerances (Hasnain *et al*. 2013),

the reason for the different responses of these species is unclear. Despite the lack of response in

yellow perch, the foraging responses of these fish species strongly indicate that boreal shield lake

food webs will collectively flex towards offshore resource and habitat use in response to

warming.

We found evidence that a reduction in nearshore coupling with warming may impact the

biomass of lake trout. The biomass of lake trout decreased significantly with increased

temperature, but none of the other species showed this pattern (Figure 3). Such reductions in

biomass with increasing temperature may be a consequence of a reduction in overall energy

availability due to decreased nearshore coupling. Mobile top predators that garner energy from

multiple spatially distinct macrohabitats are expected to have elevated biomass as a result of a

higher overall resource availability, producing relatively top-heavy food webs (i.e., non-Eltonian,

Mccauley *et al.* 2018; Woodson *et al.* 2018). Therefore, reduced access to the nearshore energy

pathway in warm lakes could therefore reduce overall energy availability for lake trout and so

lower lake trout biomass and population density (Guzzo *et al.* 2017; Tunney *et al.* 2018). A

reduction in lake trout biomass across could subsequently reduce the top-down control of lake

trout on cisco, their preferred prey. Reduced top-down control by lake trout in warmer lakes in

conjunction with a bottom-up reduction in their biomass from reduced access to nearshore

resources may result in no discernable trend in standing biomass of cisco with temperature,

consistent with our results here. Changes in trophic biomass structure and concomitant changes

in top-down control are one possible consequence of the flexes in lake food webs towards the

offshore with warming. If climate change broadly decouples predators from multiple resource

pools, changes in biomass structure may be a widespread signatures of climate change, possibly

altering top-down effects within ecosystems (McCann *et al.* 2005; McCauley *et al.* 2018).

Our results demonstrate that many mobile generalist consumers at trophic levels and

throughout compartments drive flexibility throughout food webs in boreal shield lakes. Our work

builds on previous studies that have started to expose the role that single generalist top predators

species play in determining food web responses to climate change in many ecosystems (Barton

& Schmitz 2009; Woodward *et al.* 2010; Eloranta *et al.* 2016). The presence of flexible foragers

at various trophic levels is consistent with both recent ecological theory that predicts large,

mobile generalists ought to rapidly respond to environmental change (Rooney & McCann 2012),

and is consistent with empirical evidence of flexible foraging from various ecosystems (Schmitz

et al. 1997; Held *et al.* 2002; Vander Zanden & Vadeboncoeur 2002; Shochat *et al.* 2004;

Rooney *et al.* 2008; Taipale *et al.* 2008; Gonçalves *et al.* 2012; Edmunds *et al.* 2016). The

presence of responsive foraging throughout diverse food webs rather than just at the top of food

chains would make the generalist module the type of remarkably scale-invariant feature of food

web architecture that has long interested ecologists (Sugihara *et al.* 1989).

The presence of responsive consumers across trophic levels and throughout food web

compartments has strong implications for food web stability. Responsive foraging by generalists

can be potently stabilizing in food webs (Post *et al.* 2000; Rooney *et al.* 2006; Rip *et al.* 2010;

McMeans *et al.* 2016). Generalist consumers can average variability in their resources (akin to

the portfolio effect of Tilman *et al.* (1998)) by rapidly responding to resource variability in space

and time by shifting away from low-density resources and towards high density resources,

impeding that variability from emanating throughout food webs (Valdovinos *et al.* 2010). This

switching ability also increases trophic redundancy, which can prevent secondary extinctions

(Borrvall *et al.* 2000; Sanders *et al.* 2018). If a broad suite of organisms can make rapid, smart

foraging responses to environmental change, they may accentuate the stabilizing effect of a

single generalist module. The presence of many responsive consumers increases stability in

complex food web models (Kondoh 2010; Valdovinos *et al.* 2010). This reveals a potential

powerful, repeated stabilizing mechanism throughout food webs that Levin (1998) argues is

fundamental to persistence of ecosystems. However, when generalist consumers decouple

macrohabitats to avoid entering physiologically taxing habitats, their ability to respond to

consumer densities is limited, undermining their stabilizing potential (Schmitz & Barton 2014;

Tunney *et al.* 2014; Guzzo *et al.* 2017). Thus, the decoupling of habitats with warming may be

widespread across ecosystems, forcing the food webs of many ecosystems to flex with climate

change in ways that alter ecosystem stability worldwide.

Our work here adds a key piece to the mounting evidence that generalist consumers likely to

play a vital role in determining how climate change rewires food webs (Bartley *et al.* n.d.).

Generalist consumers positioned throughout food webs may ultimately reorganize whole

ecosystems in response to climate change (Blanchard 2015; McMeans *et al.* 2016). Determining

both the prevalence and positions of these responsive generalist consumers and the factors that

constrain these species responses are critical tasks for predicting the myriad ways that food webs

will adapt with a changing climate. Importantly, our results suggest that these species throughout

food webs contribute to adaptive capacity—the ability of an ecosystem to change in response to

changing conditions—a key part of the resilience of ecosystems in the face of environmental

variation (McMeans *et al.* 2016). Although we have showed that lake food webs flex at many

axes, it is unclear when this adaptive capacity will be exceeded and entire ecosystems collapse.

This uncertainty highlights the clear need to forecast major changes in food web structure.

Monitoring the behaviour and foraging of responsive consumers is a promising way to track

changes in food web structure and give prescient warnings of future drastic changes in ecosystem

dynamics (Bartley *et al.* n.d.; Velarde *et al.* 2013). This makes responsive generalist consumers

critical to understanding of how ecosystems will fare through climate change.

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Author Contributions

T.J.B. and K.S.M. conceived and designed this study. T.J.B. and T.D.T. collected field data and

conducted lab work. T.J.B. analyzed data and wrote the initial draft of the manuscript. T.J.B.,

T.D.T., N.P.L., B.J.S., R.H.H., and K.S.M. all contributed substantially to writing and revisions.

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552   **Figure 1.** Nearshore foraging index and the mean log(depth) of capture of lake trout (a and b),
553   cisco (c and d), walleye (e and f), and yellow perch (g and h) in lakes across a gradient of
554   average recent air temperature. A solid line indicates a p-value less than 0.05, and a dashed line
555   indicates a p-value less than 0.1. Regression model details can be found in Table S3. Fish
556   silhouettes were taken from PhyloPic (http://www.phylopic.org).
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558
559 **Figure 2.** Trophic position of lake trout (a), walleye (b), cisco (c), and yellow perch (d) in lakes
560 across a gradient of average recent air temperature. A solid line indicates a p-value less than
561 0.05, and a dashed line indicates a p-value less than 0.1. Regression model details can be found
562 in Table S3. Fish silhouettes were taken from PhyloPic (http://www.phylopic.org).
563
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566 **Figure 3.** The log-transformed CUE biomass index for lake trout (a), walleye (b), cisco (c), and
567 yellow perch (d) in lakes across a gradient of average recent air temperature. A solid line
568 indicates a p-value less than 0.05, and a dashed line indicates a p-value less than 0.1. Regression
569 model details can be found in Table S3. Fish silhouettes were taken from PhyloPic
570 (http://www.phylopic.org).
571

a) typical cold lake

i) thermal conditions

ii) species' responses

b) warmed lake after climate change

iii) thermal conditions

iv) species' responses

Figure 4: Schematic of the effects of warming on four widespread and abundant fish species of
boreal shield lake food web: lake trout, cisco, walleye, and yellow perch. The x axis indicates
shifts in the degree of nearshore foraging based on $\delta^{13}C$ signatures. The y axis indicates shifts in
habitat use in terms of the mean depth of capture. Lake trout, cisco, walleye all show shifts
towards offshore foraging and offshore habitat use with increased temperature. Fish silhouettes
were taken from PhyloPic (http://www.phylopic.org). Figure is adapted from Tunney *et al.* 2014.