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1	Effects of fish predation on density and size spectra of prey fish
2	communities in lakes
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31 Abstract

32 Planktivorous and benthivorous fish have been documented to influence the density and size 33 structure of their prey communities in lakes. We hypothesized that piscivorous fish modify their prey 34 fish communities in the same way and sought to find evidence for such predation effects from a 35 comparison across 356 lakes located in nine European ecoregions. We categorized individual fish as 36 being either piscivore, non-piscivore or prey of piscivores, depending on species and individual size. 37 We calculated piscivore, non-piscivore and piscivore prey densities, respectively, and fit linear 38 abundance size spectra (SS) on lake-specific piscivore, non-piscivore and piscivore prey size 39 distributions. Multiple linear regressions were calculated to quantify the effect of piscivore density 40 and SS slopes on non-piscivore and piscivore prey densities and SS slopes, by accounting for 41 potentially confounding factors arising from lake morphometry, productivity and local air 42 temperature. Piscivore density correlated positively with piscivore prey density, but was 43 uncorrelated to density of non-piscivores. Across a subset of 76 lakes for which SS slopes of 44 piscivores were statistically significant, SS slopes of piscivores were uncorrelated with SS slopes of 45 either non-piscivores or piscivore prey. However, densities of piscivores, non-piscivores or piscivore 46 prey were a significant negative predictor of SS slopes of the respective groups. Our analyses suggest 47 that direct predation effects by piscivorous fish on density and size structure of prey fish 48 communities are weak in European lakes, likely caused by low predator-prey size ratios and the 49 resulting size refuges for prey fish. In contrast, competition may substantially contribute to between-50 lake variability in fish density and size.

51

52 Key words: abundance size spectra, predator-prey interaction, across-lake comparison

54 Introduction

55	Following two seminal papers on the effect of predation by planktivorous fish on zooplankton
56	communities (Hrbacek et al. 1961; Brooks and Dodson 1965), pelagic food webs have long been
57	considered classical examples for the structuring effects of predation in lakes. Numerous subsequent
58	studies have shown that planktivorous fish predators can have profound effects on the densities,
59	species composition and size structure of prey communities (e.g. Post and McQueen 1987; Carpenter
60	et al. 2001; Brucet et al. 2010). Similar predation effects have also been found for benthivorous
61	predatory fishes feeding upon macroinvertebrate prey (e.g. Brönmark et al. 1992; Diehl 1992;
62	Blumenshine et al. 2000). In large-scale comparisons between lakes, negative correlations between
63	predator and prey densities or size have supported the assumption that predation can explain a large
64	part of between-lake variability in zooplankton densities or size (Jeppesen et al. 2003; Matveev
65	2003).
66	There is a much smaller body of literature showing the same strong predation effects for
67	piscivorous fish feeding upon fish prey. In some lakes, massive disturbances of predator communities
68	after winter fish kills or by fish stocking or removal have induced correlated, often short-term,
69	changes in prey communities (Benndorf et al. 1984; Mittelbach et al. 1995; Potthoff et al. 2008). In
70	regional studies, the effect of predation by piscivores on prey fish densities has been compared
71	across several lakes (Nowlin et al. 2006; Mehner 2010; Friederichs et al. 2011). In one of these
72	analyses, predator and prey densities were positively correlated (Mehner 2010), whereas in others
73	the expected negative correlations of prey density with predator density were found only for a
74	limited number of small prey species (Nowlin et al. 2006; Friederichs et al. 2011). These studies also
75	revealed that lake productivity and morphometry may confound the potential effects of predators on
76	prey, and hence have to be taken into account when analysing correlations between abundances of
77	predator and prey across lakes (Mehner 2010). Shifts in the size structure of prey fish communities in
78	response to fish predation across lakes have not yet been explicitly demonstrated, although recent

79 publications have indicated that fish prey-size metrics may be sensitive to the strength of predation

80 by piscivorous fish (Mehner 2010; Emmrich et al. 2011).

81	Body size is one of the most important traits of animals, being linked to physical activities,
82	biological rates, the strength of ecological interactions and selective fisheries (Brown et al. 2004; Shin
83	et al. 2005; Brose et al. 2006). Therefore, characterizing the size distributions of animal communities
84	may capture much of the biologically meaningful variation in a relatively straightforward manner
85	(Woodward et al. 2005). In aquatic systems, indeterminate growth of organisms dominates and
86	ontogenetic changes in diet are common, hence the concept of species belonging to a single niche or
87	trophic level is less plausible and the size-based view has been more widely appreciated than in
88	terrestrial ecosystems (Trebilco et al. 2013). Furthermore, the main determinant of the trophic
89	position of an aquatic predator within a food web is often its size rather than its taxonomic identity
90	(Woodward et al. 2005), and strong patterns in food web structure related to body size in aquatic
91	systems become evident when analyzed using size-class subdivisions (Jennings et al. 2002).
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The aim of this study was to elucidate whether the strong predation effects on prey density and
 size demonstrated for planktivorous and benthivorous fish predators in aquatic food webs can

104 likewise be found for piscivorous fish predators in lakes. In contrast to earlier studies with a more 105 regional focus (Bertolo et al. 2005; Mehner 2010; Friederichs et al. 2011), we included information 106 on 356 lakes of the European continent, thus accounting for substantial variation of fish community 107 composition in response to geographical location and environmental temperature (see Brucet et al. 108 2013). Individual fish were classified as piscivores, non-piscivores or piscivore prey, depending on 109 species and size for facultative piscivores (Mittelbach and Persson 1998). We correlated piscivore 110 with non-piscivore or piscivore prey densities, and expected to find a negative relationship across the 111 lakes. Furthermore, we calculated linear abundance size spectra separately for piscivores non-112 piscivores and piscivore prey communities, thus explicitly searching for patterns in size structure in 113 response to predation that go beyond the analysis of shifts in mean size (Mehner 2010; Brucet et al. 114 2013). Although changes of size spectra of prey communities in response to predation have been 115 studied occasionally (Zimmer et al. 2001; Jonsson et al. 2005; Brucet et al. 2010; Murry and Farrell 116 2014), a simultaneous inspection and comparison of slopes of size spectra from both predator and 117 prey communities across several lakes is not yet available. We expected to find steeper slopes of size 118 spectra of non-piscivores and in particular of piscivore prey fish communities indicating a dominance 119 of smaller fish in lakes where their piscivorous predators were large due to size-selective predation 120 on larger prey (see Blumenshine et al. 2000; Blanchard et al. 2009). To account for confounding 121 effects of lake productivity, morphometry and temperature on prey density and size, we included 122 lake depth, lake area, total phosphorus concentration and maximum air temperature as covariates 123 into our analyses. Finally, we tested for geographical effects by splitting the dataset into regional 124 subsets from Northern and Central Europe.

125

126 Methods

127 Sampling

128 Lake fish communities were sampled in about 1800 European lakes between 1990 and 2010, and 129 data have been accumulated into a database within the EU-project WISER (see for details Argillier et 130 al. 2013; Brucet et al. 2013). Complete information on fish size, lake productivity and lake 131 morphometry were available only for a subset of these 1800 lakes. We focused on those 356 lakes 132 located in nine European ecoregions (Illies 1978) which are dominated by percid (perch Perca 133 fluviatilis and/or pikeperch Sander lucioperca) or salmonid predators (mainly brown trout, Salmo 134 trutta) (Fig. 1). For these lakes, we acquired information on lake total phosphorus concentration (TP, mg m^{-3} , measured as the mean of a minimum of four samples taken in a single year), lake maximum 135 136 depth (m) and lake area (km²). Maximum air temperature at the geographic location of the lake was 137 calculated from the climate CRU model (New et al. 2002) and was used as an approximation to 138 maximum lake temperature. Earlier calculations have shown that size structure of fish communities 139 is more sensitive to maximum than to average local temperatures (Emmrich et al. 2014). An overview 140 about mean values and value ranges for all variables within the 356 lakes is given in the 141 Supplementary Material (Table S1). 142 Fish in these lakes were caught by stratified multimesh gill-net sampling according to the EU 143 standard for such sampling (CEN 2015). Each lake was divided into depth strata, and each stratum 144 was randomly sampled by a pre-defined number of benthic gill-nets (type NORDEN: length 30 m; 145 height 1.5 m; 12 panels of 2.5 m each with mesh sizes (knot to knot) of 5, 6.25, 8, 10, 12.5, 15.5, 19.5, 146 24, 29, 35, 43 and 55 mm), depending on lake area and maximum depth (Appelberg 2000; CEN 2015). 147 Deep lakes (maximum depth >6 m) were additionally sampled with a row of pelagic nets [similar type 148 as the benthic ones, but of 3 m height and 27.5 m length (5 mm mesh panel missing)] placed over the 149 deepest location in each lake. The number of pelagic nets was defined by the maximum lake depth 150 (i.e., number of nets = depth divided by the 3 m height of the pelagic nets) such that the pelagic net 151 row fished the entire water column in deep lakes. All sampling was conducted during late summer to 152 early autumn periods (CEN 2015). Catch was expressed as lake-specific number per unit effort (NPUE, fish net⁻¹ night⁻¹), averaged from all nets set per lake (but pelagic nets of 3 m height counted as two 153

nets because the net area was about twice as large as the one of benthic nets), and hence this NPUE
reflects primarily the catch in the benthic gill-nets (for an overview on numbers of nets set per lake,
see Supplementary Material Table S1).

157 In predator fish, ontogenetic diet shifts are common and piscivory is the dominant feeding strategy only beyond a certain fish size. Therefore, all individuals $>2^{5}=32$ g (roughly equivalent to 15 158 159 cm total length) of pike Esox lucius, pikeperch, catfish Silurus glanis, asp Aspius aspius, burbot Lota 160 lota, perch, brown trout, rainbow trout Oncorhynchus mykiss, Arctic charr Salvelinus alpinus, lake 161 trout Salvelinus namaycush and largemouth bass Micropterus salmoides were classified as piscivores. 162 All other 45 species were classified as non-piscivores, and were considered potential prey of the 163 piscivores. However, to account for the potentially underestimated proportion of small-sized fish in 164 gill-net catches and to prevent the inclusion of all non-piscivores that were definitely too large to be 165 fed upon by piscivores, we defined the narrow-sized group of piscivore prey (i.e., suitably sized prey 166 for the piscivores) which included the sum of all young piscivores in the size range >8 g to \leq 32 g and 167 all non-piscivores in the size range >8 g to ≤128 g. This group hence reflects all fish in the size range 168 between 8 g and 128 g except piscivores >32 g.

169

170 Evaluating the effect of piscivore density on non-piscivore and piscivore prey densities

We visualized the frequency distributions of NPUEpiscivores, NPUEnon-piscivores and
NPUEpiscivore-prey across the lakes by histograms, compared the NPUE between piscivores, nonpiscivores and piscivore prey across all lakes by Wilcoxon rank sum tests, and calculated Spearman's
rank correlation coefficient r_s between either log₁₀ NPUEpiscivores and log₁₀ NPUEnon-piscivores, or
between log₁₀ NPUEpiscivores and log₁₀ NPUEpiscivore-prey to document the raw correspondence
between predator and prey densities. Non-parametric tests were appropriate because of non-normal
distribution and heteroscedasticity in the total, untransformed dataset.

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178 Non-piscivore fish were found in 332 lakes only. The remaining 24 lakes contained only 179 piscivorous species. To evaluate the effects of piscivore density (NPUEpisivores) on non-piscivore 180 density (NPUEnon-piscivores) in these 332 lakes while accounting for the confounding effects of lake 181 morphometry, productivity and temperature, we calculated linear models with log10 NPUEnon-182 piscivores as the dependent variable and log_{10} NPUEpiscivores, log_{10} lake area, log_{10} lake depth, log_{10} 183 TP and maximum air temperature as independent variables. We did not consider interactions 184 between the continuous predictor variables. We sought for the most parsimonious linear model by 185 backwards elimination of the predictor with lowest significance from the initially full model, thus 186 stepwisely declining the Akaike Information Criterion (AIC) of the models. Accordingly, the final 187 model was the one with the lowest AIC (Faraway 2005). In addition, we compared the two models 188 with lowest AIC by ANOVA to decide whether to retain the respective predictor. We checked residual 189 plots of the final model for deviations from normality and homoskedasticity, and calculated variance 190 inflation factors (VIF) to detect potential collinearity between predictor variables. To compare the 191 relative strength of the significant predictors, we additionally calculated their standardized (beta) 192 coefficients. In a similar way, we calculated linear models with log₁₀ NPUEpiscivore-prey (n=354 193 lakes) as the dependent variable and log₁₀ NPUEpiscivores, log₁₀ lake area, log₁₀ lake depth, log₁₀ TP 194 and maximum air temperature as independent variables.

195

196 Evaluating the effect of predator size on prey size

To evaluate the effect of piscivore size structure on non-piscivore and piscivore-prey size structure, we accumulated information on the size of fish. All fish caught were individually measured (rounded to cm total length, TL) and directly weighed (g wet mass, wm) in most cases. For several lakes, wm was calculated from TL by regional species-specific regressions. The multimesh gill-nets used underestimate the proportion of fish smaller than 4-6 cm TL (Prchalova et al. 2009), and hence fish of 1 to 4 g wm (i.e., fish in first year of life) were under-represented in our catches. To avoid a

203 bias in estimating the fish size structures, we excluded all fish ≤ 4 g from subsequent calculations, and 204 numbers of non-piscivores were accumulated into 13 \log_2 size classes (1st class: >2²=4 g to $\leq 2^3$ =8 g, 2^{nd} class: >2³=8 to ≤2⁴=16 g, etc., 13th class: >2¹⁴=16,384 to ≤2¹⁵=32,768 g). The size structure of 205 206 piscivores encompassed only 10 size classes (>32 to ≤32,768 g). The size structure of piscivore prey 207 encompassed only four size classes (>8 to \leq 128 g, see above). For overall inspection purposes, we 208 accumulated all piscivore, non-piscivore and piscivore prey fishes per size class across all 356 lakes 209 into a cumulative size spectrum. Cumulative size spectra represent the average decline of fish 210 numbers by size across all lakes, but weighted by the number of fish caught per lake. Accordingly, 211 these plots are biased towards the size structure of those lakes which contributed most fish to the 212 overall number. To check for regional differences in the cumulative size spectra, we split the total 213 lake dataset into two geographical subsets (Northern Europe with n=193 lakes from Norway and 214 Sweden; Central Europe with n=163 lakes primarily from Germany and France, accompanied by a few 215 lakes from U.K., Ireland, Northern Italy, Estonia and Slovenia), and compared the slopes of the 216 regional size spectra (see below) by ANCOVA. 217 Linear abundance size spectra (SS) as obtained by logarithmic binning (Kerr and Dickie 2001) were 218 calculated as linear least-square regressions between log₂ numbers accumulated per size class and 219 log₂ midpoint of size classes (g). The slopes of SS indicate the rate of decrease in numbers with 220 increasing body size, a parameter that may be sensitive to size-selective predation of piscivores. In 221 some of the lakes, a few size classes did not contain fish, and these empty size classes were excluded 222 from linear regressions (White et al. 2008). Furthermore, in some lakes, fish covered only one or two 223 size classes, and hence SS could not be calculated. Ultimately, separate slopes for piscivores, non-224 piscivores and piscivore prey fish communities were calculated for 353, 317 and 317 lakes, 225 respectively. We visualized the frequency distributions of piscivore, non-piscivore and piscivore-prey 226 SS slopes by histograms, compared the SS slopes between piscivores and non-piscivores or between 227 piscivores and piscivore prey across all lakes by Wilcoxon rank sum tests, and calculated the

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228 Spearman's rank correlation coefficient rs between piscivore SS slopes and non-piscivore or piscivore-

229 prey SS slopes to document the raw correspondence between predator and prey size spectra.

230 To evaluate more precisely the effect of predator size distributions on prey size distributions, we 231 calculated linear models with non-piscivore SS slopes as the dependent variable, and piscivore SS 232 slopes as the independent variable. To consider potential density effects on non-piscivore slopes, 233 log₁₀ NPUEnon-piscivores was added to the set of predictor variables. We further added log₁₀ lake 234 area, log₁₀ lake depth, log₁₀ TP and maximum air temperature as potentially confounding variables. 235 The most parsimonious model was found according to minimized AIC and ANOVA as described 236 above. Similarly, we calculated linear models with piscivore-prey SS slopes as the dependent variable, 237 and piscivore SS slopes as the independent variable, and added log₁₀ NPUEpiscivore-prey, log₁₀ lake 238 area, log_{10} lake depth, log_{10} TP and maximum air temperature as predictors.

239 The SS of fish communities were non-linear in some lakes (Emmrich et al. 2014). Therefore, we

240 created a lake subset by including only significant (P<0.05) and informative (R²>0.5) SS slopes for

piscivores (185 lakes) and non-piscivores (137 lakes). For piscivore prey (123 lakes), we applied a less

strict significance threshold (P<0.10) because these SS regressions were based on four value pairs

243 only. Significant and informative SS for both piscivores and non-piscivores in the same lake were

found in 76 lakes. Similarly, significant and informative SS for both piscivores and piscivore prey in

the same lake were found in 76 lakes. For these subsets, we repeated the visualization and all

246 calculations as described above, to evaluate whether the size structure of piscivores had a correlative

247 effect on the size structure of non-piscivores of piscivore-prey fish.

248 With the subsets with significant and informative SS slopes, we conducted two additional

analyses. First, we split the lakes into Northern or Central European origin (see above) and re-

250 calculated the linear models with SS slope of non-piscivores or piscivore prey as the dependent

- variable separately for both geographical subsets. Second, we reversed the analyses and tested
- 252 whether the SS slopes of non-piscivores or piscivore prey (independent variables) had an effect on SS

- slopes of piscivores (dependent variable). We included log₁₀ NPUEpiscivores as predictor, and log₁₀
- lake area, log₁₀ lake depth, log₁₀ TP and maximum air temperature as covariates.

256 Potential bias of gill-net catches for fish density and size estimates

257 All analyses were based on standardized catches by multi-mesh gill nets, and gill-net catches can 258 be biased with respect to abundance estimates and size distributions (Prchalova et al. 2009; 259 Prchalova et al. 2011; Clement et al. 2014). Saturation effects may lead to underestimation of fish 260 abundances if soaking time of gill-nets is long, and correction has been recommended for catches >3 261 kg per standard multi-mesh gill-net per night (Prchalova et al. 2011). However, the average biomass per net exceeded 3 kg in only 41 out of the 356 lakes, and was higher than 6 kg net⁻¹ in only four 262 263 lakes. Therefore, saturation may have only marginally biased our results. It has been shown that gill-264 nets consistently underestimate the relative proportions of fish smaller than about 6 cm (Prchalova 265 et al. 2009). However, there is a strong correspondence between catches by multi-mesh gill-nets and 266 fish densities calculated from hydroacoustic records (Emmrich et al. 2012), suggesting that the fish-267 catch index (NPUE) obtained from gill-nets is a relatively unbiased approximation to lake-specific fish 268 densities. Therefore, we assumed that the consistent underestimation of the smallest size classes of 269 fish by gill-net catches has no systematic effect on comparison of fish NPUE between the lakes. In 270 addition to the underestimation of small fish <4 g, the abundance of very large fish may also be 271 underestimated (Psuty and Borowski 1997; Prchalova et al. 2009; Smejkal et al. 2015). The 272 cumulative size spectra (Supplementary Material Fig. S1) suggest an underestimation of fish >4096 g. 273 However, this uncertainty for fish >4 kg may have had no substantial effect on the estimates of total 274 piscivore abundances or piscivore SS slopes. Overall, the SS slopes of piscivores and non-piscivores 275 were determined by the numbers of fish in up to 11 (usually 5-7) size classes in our dataset (compare 276 Fig. 3), and hence uncertainties for abundances in single size classes have not too strong effects on 277 the slope estimates. Furthermore, the biases introduced by gill-nets apply to all lakes in a

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278 comparative way, and hence we are convinced that the general trends found by our analyses are

279 robust and valid.

280

281 Testing for potential bias in the dataset caused by fisheries

282	Exploitation of fish communities by commercial or recreational fisheries may modify fish
283	abundance and size spectra (Jennings and Blanchard 2004). We tested for a potential bias of our
284	analyses induced by fisheries intensity within the total dataset of 356 lakes by comparing NPUE and
285	SS slopes of piscivores, non-piscivores and piscivore prey between 47 lakes with reported high
286	fisheries intensity (categorized according to local expert opinion), and 112 lakes with no or very low
287	fisheries intensity. For the remaining lakes, no information was available, or fisheries intensity was
288	considered intermediate. We ran linear models with fisheries (low or high) as the main categorical
289	factor, and lake area, lake depth, TP concentration and maximum air temperature as covariates. We
290	further included all two-way interactions between fisheries intensity and the four covariates.
291	Fisheries intensity had no effect on NPUEpiscivores (P=0.20). However, NPUEnon-piscivores
292	(P=0.042) and NPUEpiscivore-prey (P=0.012) were slightly higher in lakes with high fisheries intensity
293	indicating that more non-piscivore and prey fish are found in lakes with stronger fisheries. However,
294	the SS slopes of piscivores (P=0.89), non-piscivores (P=0.21) and piscivore prey (P=0.22) did not differ
295	between the low and high fisheries lakes. We conclude that fisheries intensity is likely to have had
296	only marginal effects on the validity of our analyses which focused on the effects of predator on prey
297	fish.
298	All statistical calculations were performed in R 3.1.2 (R Development Core Team 2014).

299

300 Results

301	In total, n=39,066 piscivore predators were caught in the 356 lakes, and the median standardized
302	NPUE of piscivores was 3.5 fish net ⁻¹ night ⁻¹ (Supplementary Material Table S1). In the majority of
303	lakes, NPUE of piscivores ranged between 1 and 10 fish net ⁻¹ night ⁻¹ (Fig. 2a), but the maximum
304	piscivore NPUE was 26 fish net ⁻¹ night ⁻¹ . A total of n=163,562 non-piscivore fish was caught in 332
305	lakes (the remaining lakes had fish communities exclusively composed of juvenile and adult
306	piscivores). The median NPUE of non-piscivores was 11.4 fish net ⁻¹ night ⁻¹ , and non-piscivore NPUE
307	was significantly higher than piscivore NPUE in these lakes (Wilcoxon rank sum test, W=94,828,
308	P<0.0001). In the majority of the lakes, NPUE of non-piscivores ranged between 1 and 50 fish net ⁻¹
309	night ⁻¹ , but higher NPUE up to the maximum of 244 fish net ⁻¹ night ⁻¹ were found in some of the
310	remaining lakes (Fig. 2b, Supplementary Material Table S1). There was a weak negative correlation
311	between the log_{10} -transformed piscivore and non-piscivore NPUEs across the 332 lakes (Fig. 2d;
312	Spearman's r_s =-0.113, P=0.039). A total of n=188,868 piscivore prey fish was caught in 354 lakes, with
313	a median NPUE of piscivore prey of 15.5 fish net ⁻¹ night ⁻¹ (Fig. 2c) which was higher than the median
314	piscivore NPUE in all lakes (Wilcoxon rank sum test, W=108,705, P<0.0001). There was no correlation
315	between the log_{10} -transformed NPUE of piscivore and NPUE of piscivore prey across the 354 lakes
316	(Fig. 2e; Spearman's r _s =0.023, P=0.67).
317	The best linear model for the 332 lakes with log_{10} NPUEnon-piscivores as the dependent variable
318	contained log_{10} TP, log_{10} maximum lake depth, log_{10} lake area, maximum air temperature and log_{10}

319 NPUEpiscivores as significant predictor variables (AIC=-504.9). The next best model (AIC=-503.4)

320 excluded log₁₀ NPUEpiscivores, but this model was not significantly different from the model

321 including NPUEpiscivores (ANOVA, P=0.49), and hence we excluded NPUEpiscivores as predictor

322 (Table 1, adj. R^2 =0.47, $F_{4,327}$ =73.7, P<0.0001). All variance inflation factors were <1.5 (Table 1). Positive

323 relationships to NPUEnon-piscivores were found for TP, temperature and lake area, whereas

324 NPUEnon-piscivores declined with lake depth (Table 1). According to beta coefficients, air

325 temperature was the strongest predictor, whereas the effect of lake area was the weakest (Table 1).

326 The most parsimonious linear model for the 354 lakes with log₁₀ NPUEpiscivore-prey as the

327	dependent variable variable contained \log_{10} TP, \log_{10} maximum lake depth, \log_{10} lake area, maximum
328	air temperature and log ₁₀ NPUEpiscivores as significant predictor variables (Table 1, adj.R ² =0.48,
329	F _{5,348} =65.1, P<0.0001, AIC=-723.4). The next best model (AIC=-720.4) excluded log ₁₀ TP, but was
330	significantly different from the model including TP (ANOVA, P=0.026), and hence we kept TP as
331	predictor variable. All variance inflation factors were <1.65 (Table 1). In contrast to our expectations,
332	the NPUEpiscivores was positively related to NPUEpiscivore-prey. Positive relationships to
333	NPUEpiscivore-prey were likewise found for TP, temperature and lake area, whereas NPUEpiscivore-
334	prey declined with lake depth (Table 1). According to beta coefficients, air temperature was the
335	strongest predictor, whereas the effects of TP and NPUEpiscivores were the weakest (Table 1).
226	
330	The piscivore sizes ranged between the lower threshold size of 32 g (several species) and the
337	maximum of 11,124 g (a pike). The size of non-piscivores ranged between 1 g (several species) and
338	20,000 g (a bighead carp, Hypophthalmichthys nobilis). The size of piscivore prey was defined
339	according to our thresholds between 9 g and 128 g. The cumulative size spectrum across all 356 lakes
340	had a slope of -1.68 for piscivores, -1.35 for non-piscivores, and -0.86 for piscivore prey
341	(Supplementary Material, Fig. S1a). The total size range of piscivores and non-piscivores was very
342	similar, except the missing smallest size classes of predators (<32 g) which were classified as piscivore
343	prey. If the lake dataset was split into subsets from two geographical regions, the resulting slopes of
344	cumulative size distributions of piscivores, non-piscivores and piscivore prey for the Northern and
345	Central European regions were similar (Supplementary Material Fig. S1b,c) and did not differ
346	between the geographical subsets (ANCOVA, piscivores: $F_{2,20}$ =0.47, P=0.63; non-piscivores: $F_{2,30}$ =2.1,
347	P=0.13; piscivore prey: F _{2,6} =0.65, P=0.55).
348	Among the 356 lakes, there were combinations of shallow piscivore but steep non-piscivore and

349 piscivore-prey SS slopes (Mirower See, Fig. 3a), intermediate SS slopes for piscivores, non-piscivores

and piscivore prey (Wummsee, Fig. 3b), or steep piscivore but shallow non-piscivore and piscivore-

351 prey SS slopes (Fleesensee, Fig. 3c). Overall, the slopes of the SS of piscivores in the 353 lakes for

which SS could be calculated ranged between -2.93 and 0.50 (median= -0.828, Supplementary

353	Material Table S1), but the slopes were between -0.50 and -1.00 in about half of the lakes (Fig. 4a).
354	The slopes of the SS of non-piscivores in the 317 lakes for which SS could be calculated ranged
355	between -2.64 and 1.04 (median=-0.614, Fig. 4b; Supplementary Material Table S1), and non-
356	piscivore SS slopes were shallower than piscivore SS slopes (Wilcoxon rank sum test, W=65,055,
357	P<0.0001). There was a slightly negative correlation between piscivore and non-piscivore SS slopes
358	across the 315 lakes for which SS could be calculated for both fish groups (Fig. 4c; Spearman's
359	r_s =-0.118, P=0.036). The slopes of the SS of piscivore prey in the 317 lakes for which SS could be
360	calculated ranged between -3.33 and 1.36 (median=-0.883, Fig. 4e; Supplementary Material Table
361	S1), and piscivore-prey SS slopes did not differ from piscivore SS slopes (Wilcoxon rank sum test,
362	W=46,694, P=0.23). There was no correlation between piscivore SS slopes and piscivore-prey SS
363	slopes across the 314 lakes for which size spectra could be calculated for both fish groups (Fig. 4f;
364	Spearman's r _s =-0.016, P=0.77).

365 We excluded two lakes with outlier SS slopes of piscivores or non-piscivores (Fig. 4c) to reduce 366 heteroskedasticity and one lake for which no information on maximum depth was available .The best 367 linear model for the remaining 312 lakes with non-piscivore SS slope as the dependent variable 368 contained piscivore SS slopes, maximum temperature, log₁₀ TP, log₁₀ lake area and log₁₀ NPUEnon-369 piscivores as significant predictor variables (AIC=-615.7). However, the next best model (AIC=-613.8) 370 excluded maximum temperature, and this model was not significantly different to the one including 371 temperature (ANOVA, P=0.076). Therefore, we excluded maximum temperature from the final model 372 (Table 2, adj. R^2 =0.35, $F_{4,307}$ =41.4, P<0.0001). According to beta coefficients, the negative effect of the 373 piscivore SS slopes on non-piscivore SS slopes was the weakest among the significant predictors, 374 whereas NPUEnon-piscivores was the strongest predictor (Table 2).

We excluded two lakes with outlier SS slopes of piscivores or piscivore prey (Fig. 4f) to reduce heteroskedasticity.The best linear model for the remaining 312 lakes with SS slopes of piscivore prey as the dependent variable contained log₁₀ maximum depth, log₁₀ TP and log₁₀ NPUEpiscivore-prey as significant predictor variables (AIC=-284.0). However, the next best model (AIC=-283.5) excluded

379 maximum depth, and this model was not significantly different to the one including depth (ANOVA,

380 P=0.15). Therefore, we excluded maximum depth from the final model (Table 2, adj.R²=0.10,

381 F_{2,309}=19.0, P<0.0001). According to beta coefficients, NPUEpiscivore-prey was the strongest

382 predictor of SS slopes of piscivore prey (Table 2).

383 By considering only significant (P<0.05) and informative (R²>0.5) SS slopes, the range of SS slopes

384 for piscivores in the remaining 185 lakes became narrower than that calculated from all lakes

385 (median =-1.01), but the majority of slopes was still found between -0.50 and -1.0 (Fig. 4g). The range

386 of significant and informative non-piscivore SS slopes in 137 lakes became likewise narrower (median

387 =-0.843, Fig. 4h), but non-piscivore SS slopes still were significantly shallower than piscivore SS slopes

388 (Wilcoxon rank sum test, W=16,457, P<0.0001). There was no significant correlation between

piscivore and non-piscivore SS slopes in the 76 lakes for which significant and informative SS could be

390 calculated (Fig. 4i; Spearman's r_s =-0.102, P=0.39). The median of significant (P<0.1) and informative

391 (R²>0.5) SS slopes of piscivore prey in 123 lakes was steeper than the one in all lakes (median

392 =-1.184, Fig. 4k), and SS slopes of piscivore prey were significantly steeper than piscivore SS slopes

393 (Wilcoxon rank sum test, W=9,072, P=0.003). There was no significant correlation between SS slopes

394 of piscivores and SS slopes of piscivore prey in the 76 lakes for which significant and informative SS

could be calculated (Fig. 4l; Spearman's r_s=0.063, P=0.56).

396 The best linear model for the 76 lakes with significant SS slopes of non-piscivores as the

397 dependent variable contained SS slopes of piscivores, maximum temperature and log₁₀ NPUEnon-

398 piscivores as significant predictor variables (AIC=-247.5). However, the next best model (AIC=-247.0)

399 excluded SS slopes of piscivores, and this model was not significantly different to the one including

400 piscivore slopes (ANOVA, P=0.23). Therefore, we excluded SS slopes of piscivores from the final

401 model (Table 2, adj.R²=0.36, F_{2,73}=21.9, P<0.0001). The negative effect of NPUEnon-piscivores on SS

402 slopes of non-piscivores was stronger than the positive effect of temperature on SS slopes (Table 2).

403 Similarly, the most parsimonious model for significant SS slopes of piscivore prey in 76 lakes

404 contained only maximum temperature and NPUEpiscivore-prey as significant predictors (Table 2,

405	adj.R ² =0.16, F _{2,73} =7.89, P=0.0008, AIC=-127.9). The better model (AIC=-129.3) included also log ₁₀ TP,
406	but this model was not significantly better than the one without TP (ANOVA, P=0.53).
407	When the lake dataset with significant and informative SS slopes was split into regional subsets,
408	the linear models revealed that the NPUE of non-piscivores or piscivore prey were the strongest
409	predictors for SS slopes on non-piscivores or piscivore prey, respectively, in particular for the region
410	Central Europe (Table S2, Supplementary Material). A negative effect of piscivore SS slopes on non-
411	piscivore or piscivore-prey SS slopes could not be found in these geographical subsets.
412	In the reversed analyses, we calculated the most parsimonious linear model with SS slopes of
413	piscivores as the dependent variable, and TP, lake area, lake depth, maximum temperature, SS slope
414	of non-piscivores or piscivore prey, and NPUEpiscivores as predictors. In both linear models, all lake
415	descriptors and SS slopes of non-piscivores or piscivore prey were excluded, and NPUEpiscivores was
416	the single significant independent variable for SS slopes of piscivores (for piscivore SS slopes
417	including non-piscivores as predictor: final model adj. R^2 =0.38, $F_{1,74}$ =47.3, P<0.0001, log_{10}
418	NPUEpiscivores t=-6.88, P<0.0001; for piscivore SS slopes including piscivore prey as predictor: final
419	model adj.R ² =0.33, F _{1,74} =37.9, P<0.0001, log ₁₀ NPUEpiscivores t=-6.19, P<0.0001).

421 Discussion

422 The correspondence between the density and size of piscivorous fish and the density and size of 423 non-piscivores or piscivore prey fish produced by our calculations was in part unexpected. Piscivore 424 density correlated positively with piscivore prey density. Piscivore SS slopes did not predict the non-425 piscivore and piscivore-prey SS slopes in the linear models in any of the subsets including only 426 significant slopes. In contrast, the SS slopes of piscivores, non-piscivores and piscivore prey were 427 strongly negatively correlated with NPUE of the respective groups in all datasets, suggesting that the 428 size structure of fish in lakes is primarily density-dependent. A substantial decline of prey fish 429 densities or a consistent removal of maximum-sized piscivore prey fish from predation by piscivorous

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fish, similar to the patterns found for example for the interactions between planktivorous fish and
zooplankton prey in the pelagic area of lakes (Brooks and Dodson 1965; Jeppesen et al. 2003), could
not be detected by the ataxonomic approach as used here.

433 The strong positive correlations between non-pisicivore or piscivore-prey densities and 434 temperature, lake productivity or morphometry confirm earlier studies which have used various 435 subsets of European lakes (Jeppesen et al. 2000; Mehner et al. 2005; Brucet et al. 2013). Usually, 436 higher TP concentrations directly boost the biomass of primary producers and indirectly increase the 437 biomass of primary consumers (zooplankton and macroinvertebrates), and hence the resource 438 availability becomes similarly higher for fish as secondary consumers. The positive correlation 439 between piscivore densities and piscivore prey densities found in our data suggests that the 440 enhanced resource availability translates even into tertiary consumers. A similar positive correlation 441 between predator and prey fish densities has already been found in a subset of about 60 German 442 lakes (Mehner 2010). The correspondence between resource availability and fish density is further modified by lake morphometry. Polymictic shallow, large and warm lakes facilitate a stronger 443 444 response of fish density to productivity than observed in stratified lakes with a cool hypolimnetic 445 area (Brucet et al. 2013). In polymictic lakes, TP concentrations are translated into higher primary 446 productivity than in stratified lakes at comparable TP concentrations, attributable to the interaction 447 between frequent mixing and hence continuous nutrient availability for phytoplankton, and the 448 higher average temperatures (Jeppesen et al. 1997).

The overall positive correlation between piscivore density and piscivore-prey density contrasts with the findings of lake-specific studies in which massive interventions for piscivorous fish communities changed the equilibrium between predator and prey. For example, experimental stocking or removal of piscivores conducted in small lakes induced correlated short-term changes in prey fish communities (Benndorf et al. 1984; Mittelbach et al. 1995; Potthoff et al. 2008). However, in most of these cases, after a few years the predator-to-prey density or species ratios returned to the original values observed before disturbance, hence suggesting that strong negative density

456 effects of piscivores on prey communities may reflect transient states which cannot be upheld 457 without permanent disturbance (Mittelbach et al. 1995; Donald and Anderson 2003). In studies 458 comparing fish communities across several lakes, densities and size of single, usually small, prey fish 459 species have been found to correlate negatively to predator fish densities (Nowlin et al. 2006; 460 Friederichs et al. 2011). One may argue that our sampling approach was not suitable to find a 461 negative effect of predation on densities of small-sized fish because piscivore prey individuals smaller 462 than 2 g (about 6 cm length) representing the newly hatched cohorts in the year of catch were 463 underrepresented in the gill-nets (Prchalova et al. 2009). Hence, we cannot totally exclude the 464 possibility that fish predation reduces in particular the densities of newly hatched fish. However, this 465 effect obviously does not translate into lower recruitment and likewise lower densities of fish at 466 higher age because our analyses demonstrate that the total densities of piscivores and piscivore prey 467 in lakes tend to be positively correlated. This positive correlation suggests that productivity and 468 competition strength keep predator and prey densities in equilibrium and override potential local 469 predation effects. It has to be noted that there was no positive correlation between densities of 470 piscivores and non-piscivores, but only between piscivores and piscivore prey. The group of piscivore 471 prey also included the smallest size groups (<32 g or about <15 cm) of predators (primarily perch and 472 zander in European lakes) which can form a substantial proportion of total fish abundance in the 8 g 473 to 32 g size classes. Therefore, piscivores are more abundant in lakes where young piscivores are 474 likewise more abundant, and hence intra-guild predation and cannibalism within and among 475 piscivorous fish populations (e.g., Schulze et al. 2006) contribute to the energy transfer from primary 476 production to production of piscivores.

Increasing fish density might be compensated for by reduced average individual growth
(Holmgren and Appelberg 2001; Rose et al. 2001). Therefore, fish achieve a relatively smaller size at
higher densities, as indicated by the dominant negative correlation between SS slopes of piscivores,
non-pisicivores or piscivore prey, and the NPUEof the respective groups. This strong negative density
dependence has been found by several previous analyses on the fish communities of European lakes

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(Emmrich et al. 2011; Arranz et al. 2015). Lowered growth rates leading to 'stunted' fish populations
are well described in response to strong competition for resources at high fish densities (Ylikarjula et
al. 1999; Amundsen et al. 2007). Stunting can be seen as an adaptive response of life history by
which age and size at maturity and investment into reproduction are adjusted to the local conditions.
It can be expected for fish that live in environments characterized by steep size spectra (i.e., low
average size of food resources) and low overall resource abundance such that growth is limited by
high foraging activity (Giacomini et al. 2013).

489 The median SS slopes of piscivores, non-piscivores and piscivore prey were relatively similar, and 490 the majority of the slopes ranged between -0.5 and -1.0, in particular in the subsets with significant 491 SS slopes. However, SS slopes of piscivores were unrelated to SS slopes of non-piscivores or piscivore 492 prey. Obviously, the correspondence between size distributions of piscivores and piscivore prey is 493 much weaker than the strong negative effect of planktivorous fish predators on zooplankton size 494 (Brooks and Dodson 1965). The decisive difference to the fish-zooplankton interaction is the very 495 similar and largely overlapping size range of piscivores and non-piscivores. We found both piscivores 496 and non-piscivores between 1 g and >10 kg in the catches. This feature characterizes size-structured 497 populations in which adult fish have to recruit through much smaller juvenile stages during which 498 juvenile piscivores can be competitors of piscivore prey fish (Werner and Gilliam 1984). In contrast, 499 there is almost no size overlap between the largest zooplankton (about 5 mm) and the smallest 500 zooplanktivorous fish (fish larvae of about 5-6 mm) and predator-to-prey length ratio (PPLR) in 501 interactions between adult fish and zooplankton is in most cases >100 which translates into a 502 predator-to-prey mass ratio (PPMR) of at least 100³=1,000,000. In contrast, the PPLR of piscivore fish 503 in temperate European lakes is only around 4 to 5 (Mittelbach and Persson 1998; Wysujack and 504 Mehner 2005; Dörner et al. 2007), and hence their average PPMR is about 4^3 =64 to 5^3 =125 (Brose et 505 al. 2006).

Based on this low PPMR, the range of piscivore-prey sizes still available to the piscivorous
predators is surprisingly small. An upper threshold for vulnerable size classes of prey fish can be

508 estimated at about 100 g because piscivores capable of feeding on 100 g prey have to be 64 to 125 509 times heavier, i.e. their mass would exceed 6400 g. Piscivores of this size are rare in the lake fish 510 communities analysed here (only 14 of the 39,066 piscivores were in the size classes >4096 g). In 511 turn, prey fish bigger than about 100 g may reach a size refuge (Hambright 1994; Wysujack and 512 Mehner 2005). According to the cumulative data from all 356 lakes in our dataset, size classes >128 g 513 represent about 4.9% of all non-piscivore individuals, a proportion big enough to facilitate continued 514 reproduction (Meijer et al. 1994). In addition, the largest female fish also have the highest absolute 515 fecundity. Therefore, the size refuge may explain why there was no negative effect of piscivore 516 densities on non-piscivore densities. However, we have also tested for a stronger negative effect of 517 predation on structure of the most vulnerable prey sizes by limiting the size range of piscivore prey 518 to between 8 and 128 g. This narrow size range of prey prevented inclusion of potentially 519 underestimated small sizes and also prevented the inclusion of all non-piscivores that were definitely 520 too large to be fed upon by piscivores. Accordingly, a negative correlation between SS slopes of 521 piscivores and these four size classes of piscivore prey would have been expected if gape-size 522 limitation is the single explanation for low predation effects on the size structure of the prey fish 523 community. However, even by having these few prey size classes included, there was no response of 524 piscivore-prey size structure to the piscivore size structure. These results suggest that the SS slopes 525 of non-piscivore fish communities are relatively stable and independent of predation, and may 526 reflect an emergent characteristics of the ecosystem (Sprules 2008; Yurista et al. 2014). This is 527 confirmed by the fact that we found systematically steeper slopes at the higher fish densities which 528 occur in more productive lakes, suggesting that there is a correspondence between fish density, 529 ecosystem productivity and fish size structure.

The lack of significant slopes in many lakes indicates that linear SS may not be an ideal approach for characterizing the size structure of fish communities. Only in about half of all lakes we obtained significant and informative SS slopes by logarithmic binning and least-square estimation. Continuous approaches such as the fit by maximum likelihood of a Pareto I cumulative distribution function are

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534 recommended (White et al. 2008), but are likewise inappropriate if the log-log relationship between 535 numbers and size is not linear (Emmrich et al. 2011; Emmrich et al. 2014). The cumulative size 536 spectra shown here likewise indicated a tendency towards a non-linear log-log size-density 537 relationship. Our own tests showed that Bayesian minimum mean square error estimations of Pareto 538 II functions result in somehow improved fits of the size distributions, probably because Pareto II 539 models approximate to non-linear log-log relationships between numbers and size (results not 540 shown). However, the resulting Pareto II function has two or even three parameters which are 541 strongly correlated to each other. Accordingly, the ecological interpretation of the form of Pareto II 542 curves is not straightforward (Vidondo et al. 1997), and a meaningful conclusion with respect to the 543 effect of predator size on prey size could not be drawn. Accordingly, we had to base our conclusions 544 primarily on the relationship between SS of piscivores and non-piscivores or piscivore prey for the 545 subsets of lakes with significant SS slopes. 546 In summary, our analyses did not find evidence for predation effects of piscivorous fish on prey 547 fish communities in European lakes which mimic the strong effects of planktivorous and 548 benthivorous fish predators on size and density of their invertebrate prev groups. Size ranges were 549 remarkably similar between piscivores and non-piscivore fish across all lakes included in the analyses, 550 and hence there was no negative effect of piscivore density or size on non-piscivore or piscivore prey 551 numbers and size structure. As a consequence of this lack of density control by predation, the density 552 and size structure of non-piscivore fish communities were substantially predicted by variables related 553 directly to temperature, resource availability and the correlated system size (see Chase 2003), 554 indicating that the intensity of intra- and interspecific competition strongly modifies composition and 555 size variability of fish communities in lakes.

556

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Table 1: Coefficients (estimates ± standard error, S.E.), beta (standardized) coefficients and variance
inflation factors (VIF) of the most parsimonious linear model of log₁₀ NPUEnon-piscivore fish (in fish
net⁻¹ night⁻¹, upper part) or log₁₀ NPUEpiscivore-prey fish (in fish net⁻¹ night⁻¹, lower part) as the
dependent variable in 332 and 354 European lakes, respectively. Significant predictors were log₁₀
piscivore fish density (NPUE_Pisc, in fish net⁻¹ night⁻¹), lake area (in ha), lake maximum depth
(max_depth, in m), total phosphorus concentration (TotalP, in mg m⁻³) and maximum air
temperature (max_temperature, in °C).

	Estimate	S.E.	Beta coeff.	VIF	t-value	Pr(> t)		
Non-Piscivores (n=332 lakes)								
(Intercept)	-0.892	0.219			-4.067	<0.0001		
log10(lake_area)	0.155	0.049	0.14	1.20	3.171	0.0017		
log10(max_depth)	-0.495	0.073	-0.31	1.30	-6.806	<0.0001		
log10(TotalP)	0.258	0.064	0.20	1.47	4.018	<0.0001		
max_temperature	0.135	0.014	0.45	1.36	9.600	<0.0001		
Piscivore prey (n=354 lakes)								
(Intercept)	-0.246	0.171			-1.432	0.1531		
log10(NPUE_Pisc)	0.189	0.053	0.14	1.11	3.546	0.0004		
log10(lake_area)	0.142	0.037	0.17	1.23	3.867	0.0001		
log10(max_depth)	-0.460	0.056	-0.37	1.34	-8.217	<0.0001		
log10(TotalP)	0.109	0.049	0.11	1.61	2.225	0.0267		
max_temperature	0.107	0.010	0.49	1.45	10.535	<0.0001		

769

- 771 **Table 2:** Coefficients (estimates ± standard error, S.E.), beta (standardized) coefficients and variance
- inflation factors (VIF) of the most parsimonious linear models of the slope of abundance size spectra
- (SS) of non-piscivore or piscivore prey fish communities as the dependent variables. Linear models
- vere calculated for the entire dataset (all lakes with SS slopes) and for subsets with significant
- 775 (P<0.05 for non-piscivores, P<0.10 for piscivore prey) and informative (R²>0.5) SS slopes only.
- Significant predictors were density of non-piscivores or piscivore prey (NPUE_Non-Pisc;
- 777 NPUE_PiscPrey; both in fish net⁻¹ night⁻¹), SS slope of piscivores (SlopeSS_Pisc), lake area (in km²),
- total phosphorus concentration (TotalP, in mg m⁻³) and maximum air temperature
- 779 (max_temperature, in °C).
- 780

	Estimate	S.E. E	Beta coeff.	VIF	t-value	Pr(> t)
Non-Piscivores						
All lakes (n=312)						
(Intercept)	-0.198	0.083			-2.392	0.0174
log10(lake_area)	-0.116	0.038	-0.15	1.05	-3.075	0.0023
log10(TotalP)	0.117	0.051	0.12	1.39	2.267	0.0241
SlopeSS_Pisc	-0.119	0.048	-0.12	1.15	-2.485	0.0135
log10(NPUE_Non-Pisc)	-0.510	0.046	-0.58	1.34	-10.993	<0.0001
Lakes with significant slop	es only (n=76))				
(Intercept)	-1.204	0.221			-5.441	<0.0001
max_temperature	0.052	0.015	0.38	1.30	3.569	0.0006
log10(NPUE_Non-Pisc)	-0.422	0.064	-0.70	1.30	-6.606	<0.0001
Piscivore prey						
<u>All lakes (n=314)</u>						
(Intercept)	-0.558	0.122			-4.557	<0.0001

log10(TotalP)	0.291	0.080	0.21	1.16	3.628	0.0003
log10(NPUE_PiscPrey)	-0.537	0.090	-0.34	1.16	-5.967	<0.0001
Lakes with significant slopes only (n=76)						
(Intercept)	-1.754	0.484			-3.625	0.0005
max_temperature	0.076	0.032	0.28	1.24	2.392	0.0194
log10(NPUE_PiscPrey)	-0.518	0.133	-0.46	1.24	-3.902	0.0002

782 Captions for figures

- Fig. 1: Geographical map of Europe, indicating location of the 329 lakes dominated by percid
 predators (dots) and 27 lakes dominated by salmonid predators or without predators at all
 (triangles).
- 786 Fig. 2: Frequency distribution of the fish abundance index (NPUE, number net⁻¹ night⁻¹) of fish caught
- by multi-mesh gill nets in 356 European lakes, split into NPUE of piscivores (a), non-piscivores (b) and
- piscivore prey fish (c), and scatter plot of log₁₀ NPUEpiscivores versus log₁₀ NPUEnon-piscivores (d)
- and log₁₀ NPUEpiscivores versus log₁₀ NPUEpiscivore-prey (e) in all lakes. The Spearman rank
- 790 correlation coefficients r_s are indicated in addition (d,e; *=P<0.05).
- 791 Fig. 3: Examples of linear abundance size spectra (SS) with opposing slopes of piscivores and non-

792 piscivores or piscivore prey fishes in three German lakes. The midpoint of log₂ size classes of fish (g

wm) is plotted against log₂ numbers of fish in this size class, and the slope of the regression is

794 determined by least-squares estimation. Examples show Lake Mirower See with shallow piscivore

slope and steep non-piscivore and piscivore-prey slopes (a), Lake Wummsee with intermediate

piscivore, non-piscivore and piscivore-prey slopes (b), and Lake Fleesensee with steep piscivore slope

and shallow non-piscivore and piscivore-prey slopes (c). Number of fish included (n), SS slopes and

variance predicted by linear regression (R²) are shown for each plot. All SS slopes are significant

799 (P<0.05).

800 Fig. 4: Frequency distribution of the slopes of abundance size spectra (SS) of fish caught by multi-

801 mesh gill nets in European lakes, split into predators (a,d, g, j), non-piscivores (b,h) and piscivore prey

fish (e, k). Note that plots (a) and (d), and (g) and (j), show the same dataset. Scatter plots show SS

- slopes of piscivores versus SS slopes of non-piscivores (c, f) or vs. piscivore prey (i, l). The Spearman
- rank correlation coefficient r_s is indicated in addition (c, f i, l; *=P<0.05). In the left part (a-f), we
- 805 included all lakes for which SS slopes could be calculated (number of lakes indicated in graph titles).

- 806 In the right part (g-l), only lakes are included for which SS slopes were significant (P<0.05 for
- 807 piscivores and non-piscivores, P<0.10 for piscivore prey) and informative ($R^2 \ge 0.5$).



813 Fig. 1



817 Fig. 2



820 Fig. 3



822

823 Fig. 4