

# Quantitative hydrological preferences of benthic stream invertebrates in Germany

Kakouei, Karan<sup>1</sup>; Kiesel, Jens<sup>2</sup>; Kail, Jochem<sup>3</sup>; Pusch, Martin T.<sup>4</sup>; Jähnig, Sonja C.<sup>5</sup>

## DOI

[10.1016/j.ecolind.2017.04.029](https://doi.org/10.1016/j.ecolind.2017.04.029)

## Original publication date

15 May 2017 (Available online)

## Document version

Accepted manuscript

## Published in

Ecological Indicators

## Citation (Vancouver)

Kakouei K, Kiesel J, Kail J, Pusch MT, Jähnig SC. Quantitative hydrological preferences of benthic stream invertebrates in Germany. *Ecological Indicators*. 2017;79:163-72.

## Author affiliation

1: Department of Ecosystem Research, Leibniz-Institute of Freshwater Ecology and Inland Fisheries (IGB), Justus-Von-Liebig-Str. 7, 12489 Berlin, Germany.  <https://orcid.org/0000-0001-8665-6841>

2: Department of Ecosystem Research, Leibniz-Institute of Freshwater Ecology and Inland Fisheries (IGB), Justus-Von-Liebig-Str. 7, 12489 Berlin, Germany.  <https://orcid.org/0000-0002-4371-6434>

3: Department of Aquatic Ecology, University of Duisburg-Essen, Universitätstraße 5, 45141 Essen, Germany.

4: Department of Ecosystem Research, Leibniz-Institute of Freshwater Ecology and Inland Fisheries (IGB), Justus-Von-Liebig-Str. 7, 12489 Berlin, Germany.

5: Department of Ecosystem Research, Leibniz-Institute of Freshwater Ecology and Inland Fisheries (IGB), Justus-Von-Liebig-Str. 7, 12489 Berlin, Germany.  <https://orcid.org/0000-0002-6349-9561>



## 36 **1.1 Abstract**

37 Current knowledge regarding the flow preferences of benthic stream invertebrates is mostly based on  
38 qualitative data or expert knowledge and literature analysis. These established flow preferences are  
39 difficult to use in predictions of the effects of global change on aquatic biota. To complement the  
40 existing categories, we performed a large-scale analysis on the distribution of stream invertebrates at  
41 stream monitoring sites in order to determine their responses to various hydrological conditions.

42 We used 325 invertebrate surveys from environmental agencies at 238 sites paired to 217 gauges  
43 across Germany covering a broad range of hydrological conditions. Based on these data, we modelled  
44 the respective probabilities of occurrences for 120 benthic invertebrate taxa within this hydrological  
45 range using hierarchical logistic regression models.

46 Our analyses revealed that more than one-third of the taxa (18-40 %) can be considered as ubiquitous  
47 and having a broad hydrological tolerance. Furthermore, 22-41 % of the taxa responded to specific  
48 ranges of flow conditions with detectable optima. “Duration high flow event” represented the flow  
49 parameter that correlated best with the abundance of individual taxa, followed by “rate of change  
50 average event”, with 41 and 38 % of the taxa showing a peak in their probability of occurrence at  
51 specific ranges of these metrics, respectively. The habitat suitability for these taxa may be potentially  
52 affected by global change-induced hydrological changes.

53 Quantified hydrological traits of individual taxa might therefore support stream management and  
54 enable the prediction of taxa responses to flow alteration. The hydrological traits of stream benthic  
55 invertebrates may be used in forecasting studies in central Europe, and the methods used in this study  
56 are suitable for application in other regions with different flow regimes.

## 57 **1.2 Introduction**

58 Hydraulic conditions are key habitat variables for all biota living in running waters and result from the  
59 interaction between river morphology and discharge or flow. Benthic invertebrates show high  
60 biodiversity in streams and rivers, have been shown to include indicator species sensitive to flow  
61 conditions, occupy a central position in the functioning of river ecosystems, and display some  
62 fascinating adaptations to flowing waters, e.g., in terms of life history, nutrition, respiration, or  
63 behavioral and morphological characteristics (Bellard et al., 2012; Lytle and Poff, 2004; Poff et al.,  
64 2007; Statzner et al., 1988). However, quantitative empirical knowledge on the flow requirements or  
65 preferences of lotic benthic invertebrates is limited but is essential (i) to assess the effects of  
66 hydrological alterations, e.g., due to global change or water uses, and (ii) to identify environmental  
67 flow regimes that aim to preserve the ecological integrity of river ecosystems (Bunn and Arthington,  
68 2002; Poff and Zimmerman, 2010).



69 There are three main approaches to assessing flow preferences. First, they are usually assessed based  
70 on literature reviews and/or expert knowledge and described at nominal (e.g., “generalist”, “lentic” or  
71 “lotic”) or ordinal scales (e.g., “limnobiont” to “rheobiont”) (Schmidt-Kloiber and Hering, 2015) and  
72 have already been collated for many taxa and compiled in databases such as the *freshwaterecoloy.info*-  
73 database (Schmidt-Kloiber and Hering, 2015). Such descriptive classifications of invertebrate flow  
74 preferences are suitable and widely used to compare the flow trait composition of different sampling  
75 sites (Armanini et al., 2011). However, due to their qualitative nature, they are less suited to assess,  
76 model and predict the effects of flow changes that are described in quantitative terms (e.g., discharge  
77 changes due to global change). Second, the hydraulic preferences of invertebrates have already been  
78 described in semi-quantitative terms in several studies by recording species’ probability of occurrence  
79 and relating it to near-bed shear stress measured using FST-hemispheres (Schmedtje, 1995; Statzner et  
80 al., 1988). However, data requirements and computational time make it infeasible to map or model the  
81 hydraulic conditions at larger than reach scales (e.g., for whole river networks) to apply such hydraulic  
82 preferences, e.g., for their application in catchment or larger scale species distribution models.  
83 Moreover, the hydraulic shear stress recorded for a specific discharge only partly reflects the complex  
84 relationship between changing flow conditions over time, since it effects species throughout different  
85 life stages and finally determines reproductive success and hence, the presence or absence of  
86 individual invertebrate species. Third, flow preferences can be based on qualitative discharge  
87 measurements, which can be summarized into typical flow or hydrological regimes when analyzed  
88 over time. It has been shown that the flow regime strongly influences ecological processes and that  
89 changes in the abundance and distribution of aquatic invertebrates are caused, in part, by flow  
90 alterations (Brooks et al., 2011; Poff and Zimmerman, 2010). In contrast to shear-stress data, long-  
91 term discharge time series (gauging data) are readily available at large spatial scales. Additionally,  
92 these data are useful for statistical modelling and for its large-scale upscaling, e.g., to predict the  
93 effects of discharge changes due to global change. Despite this clear relationship between the  
94 hydrological conditions and biota, few studies have used hydrological data to quantify the flow  
95 preferences of benthic invertebrates in rivers. Among these, most studies represent specific case  
96 studies and reviews on flow alteration and associated ecological processes (Dunbar et al., 2010a;  
97 Monk et al., 2007; Monk et al., 2006; Poff and Zimmerman, 2010), with a prevailing focus on the  
98 community structure (Brooks et al., 2011; Death, 2008b; Konrad et al., 2008; Principe et al., 2007)  
99 preferentially on individual taxa (Armanini et al., 2011).

100 We aimed to quantitatively determine the flow preferences of lotic invertebrates—thereby defining  
101 “hydrological traits” for central European rivers by analyzing existing hydrological and biomonitoring  
102 data. More specifically, we (i) investigated whether invertebrates show a clear response and have an  
103 optimum along the gradient of different hydrological variables and hence have specific hydrological

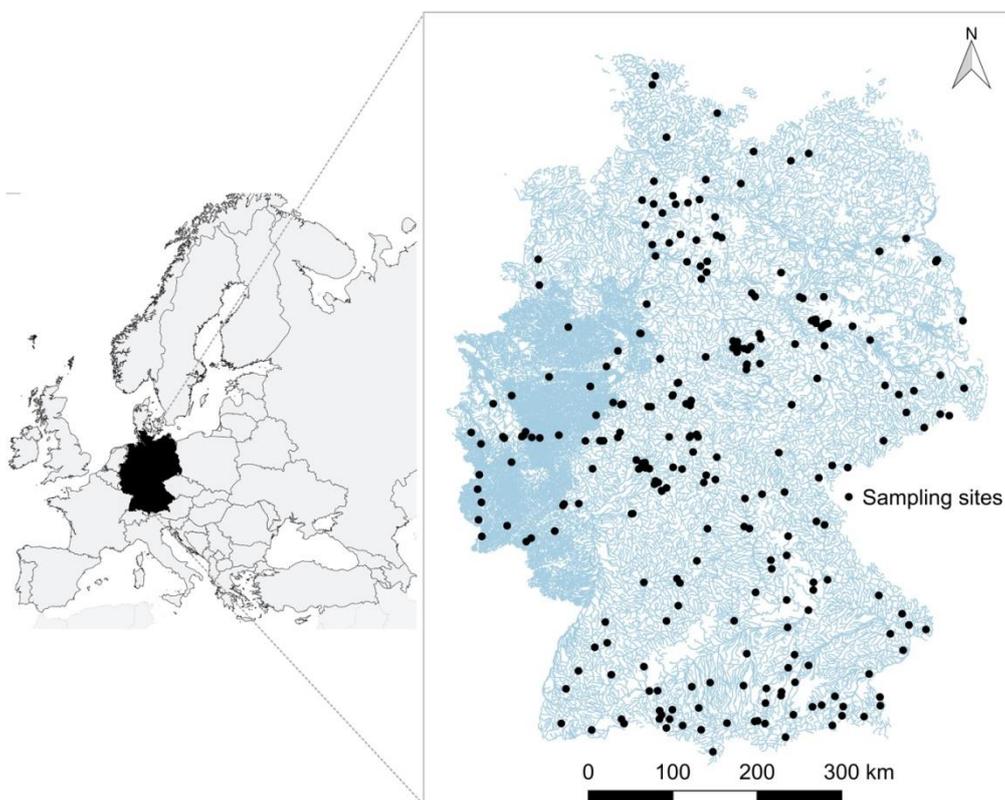


104 traits at all and (ii) aimed to quantify the hydrological thresholds at which species abundance and  
105 presence sharply change.

## 106 1.3 Methods

### 107 1.3.1 Datasets and pairing biomonitoring sites with gauging stations on the 108 river network

109 We gathered and analyzed two independent, already existing long-term datasets from Germany: (i)  
110 daily hydrological data (gauging data) and (ii) results from benthic invertebrate surveys conducted by  
111 regional water managers in German rivers. Our dataset covers a wide range of hydrological conditions  
112 in Germany, including streams and rivers in the northern lowlands, central lower-mountain areas, and  
113 Alpine region of southern Germany.



114

115 Figure 1 The locations and distribution of sampling sites in the German river network. The gauging  
116 stations are not shown as they are too close to sampling sites for being distinguishable on this scale

117 Using the German national flow gauge network and the geographical coordinates of the benthic  
118 invertebrate sampling sites, we searched for gauging stations located in the same river reach as at least  
119 one biomonitoring site. As the locations of biomonitoring sites did not usually match those of the  
120 gauging stations, they were assigned to the nearest station (DeWeber and Wagner, 2014) when the



following criteria were met: (i) having no tributaries in between and (ii) located within

122 a maximum distance of 12 km from the paired gauging station. This pairing resulted in 371  
 123 invertebrate surveys from 238 sites paired to 217 gauging stations (Figure 3). To consider the effect of  
 124 distance on discharge, the discharge data from the gauging station was recalculated for the sampling  
 125 sites according to the ratio between the catchment size at the biomonitoring site and at the paired  
 126 gauge.

127 The biological dataset included abundance data for benthic invertebrate taxa that had been sampled in  
 128 either spring or summer between 2004 and 2013 according to the currently used standard  
 129 biomonitoring protocols. All sites were in a good or high ecological status according to the EU Water  
 130 Framework Directive. We analyzed the hydrological preferences of 120 taxa that occurred in at least  
 131 eight sites for each season (spring and summer). Rare taxa with an abundance of fewer than three  
 132 occurring in fewer than eight sampling sites were excluded from the dataset because such sparse data  
 133 do not allow statistical analysis (Heino and Soinenen, 2010; Leigh and Datry, 2016). The taxonomic  
 134 resolution was the species level (111 taxa), while nine taxa were only identified to the genus level  
 135 (Supplementary Table ST1). The most frequent orders were Trichoptera (43 taxa), Ephemeroptera  
 136 (25), Coleoptera (12) and Diptera (12) (Table 2). Prior to all analyses, the abundance data were log  
 137 (x+1)-transformed.

**Table 1** Number of taxa per systematic unit analysed in this study

<b>Taxon</b>	<b>Number of species</b>	<b>Number of genera</b>
Trichoptera	41	2
Ephemeroptera	24	1
Coleoptera	11	1
Diptera	7	5
Crustacea	5	-
Plecoptera	3	1
Gastropoda	4	-
Turbellaria	3	-
Oligochaeta	3	-
Bivalvia	2	-
Megaloptera	2	-
Odonata	2	-
Hirudinea	2	-
Heteroptera	1	-
<b>Total number of taxa</b>		<b>120</b>



139 Since the addition of pseudo-absences is strongly recommended when modelling species preferences  
140 and distributions (Vaughan & Ormerod, 2005; Lobo & Tognelli, 2011) we added absence data for  
141 species at specific sites. Instead of randomly generated absence data (Lobo and Tognelli, 2011;  
142 VanDerWal et al., 2009), we preferentially generated absence data using a semi-random stratified  
143 approach, considering the stream type (Schmedtje et al., 2000) of the sampling sites according to their  
144 common environmental and hydromorphological characteristics. Sites with absences were selected  
145 based on two criteria: (i) having the same stream type as sites where the taxa were already recorded  
146 and (ii) being located in the same region/federal state as the present sites. These two criteria ensured  
147 the exclusion of sites representing inappropriate habitat conditions for the occurrence of taxa. All sites  
148 meeting these criteria were added as pseudo-absences to the analysis.

### 149 **1.3.2 Computation and pre-selection of hydrological metrics**

150 There are 171 hydrological metrics known as Indicators of Hydrologic Alteration (IHA) that are  
151 ecologically relevant and can be calculated based on daily discharge data, describing the duration,  
152 frequency, timing, magnitude, and rate of flow events (Olden and Poff, 2003). These metrics were  
153 calculated using discharge data from the 12-month period prior to the date of the biological sampling  
154 (e.g., for a macroinvertebrate sample from 12.06.2012, flow data between 13.06.2011 and 12.06.2012  
155 were considered). This period has been shown to best describe the effects of hydrological conditions  
156 on benthic invertebrates (Leigh and Datry, 2016). Twenty metrics were excluded due to the need for  
157 longer periods of discharge data, resulting in 151 metrics for further analysis. There was no significant  
158 gap (i.e., missing values for more than 5 days) in the discharge data for any of the sampling sites.  
159 Missing discharge data were filled in for individual gaps according to the trends before and after  
160 failures and by comparing trends with the data from nearby gauge(s) for which pairwise correlations  
161 exceeded the reliable threshold of  $|r| > 0.5$  (Kennard et al., 2010; Leigh, 2016). All flow metrics were  
162 computed using the R package EflowStats (Archfield et al., 2014; Henriksen et al., 2006).

163 We aimed to select at least one metric from each of the five flow regime categories (duration,  
164 frequency, timing, magnitude and rate) to minimize redundancies prior to the development of the  
165 hierarchical logistic regression models (see below). A pairwise collinearity test and a principal  
166 component analysis facilitated the selection among the 151 hydrological metrics using data from the  
167 217 paired gauges. When pairwise correlations exceeded the sensitive threshold of  $|r| > 0.7$  (Dormann  
168 et al., 2013), and hence redundancy occurred, the metric with the lower loading on the most significant  
169 principal component axis was removed from the list.

### 170 **1.3.3 Temporal and spatial pseudo-replication**

171 It was necessary to analyze temporal and spatial pseudo-replication because some sampling sites were  
172 paired with the same gauging station or may be flow-connected (Hale et al., 2016b).



To avoid temporal pseudo-replication, sampling sites paired to the same gauging

174 station and with overlapping 12-month periods were considered temporally dependent, and only the  
175 site located nearest to the gauging station was included in the analysis. This resulted in removing 46 of  
176 the 371 benthic invertebrate sampling sites.

177 We further tested for spatial autocorrelation resulting in pseudo-replication among sites that are  
178 longitudinally flow-connected (Hale et al., 2016b). However, due to the large spatial scale of this  
179 study, less than five percent of the biomonitoring sites were flow-connected. To test the spatial  
180 autocorrelation and detect sites with pseudo-replication, we first grouped flow-connected sites and  
181 then divided them into several bins with different distance categories among sites. We then calculated  
182 the Moran's I autocorrelation coefficient (Gittleman and Kot, 1990) for each hydrological metric of all  
183 sites in a group and then for different distance categories. However, Moran's I was not significant for  
184 any of the sampling site pairs, and hence none of the sites had to be excluded from the analysis due to  
185 spatial autocorrelation.

### 186 **1.3.4 Hierarchical logistic regression modelling**

187 We selected extended Huisman-Olff-Fresco (eHOF) models to quantify the flow traits of benthic  
188 invertebrates because they offer a variety of ways to efficiently fit the response data of taxa (Jansen  
189 and Oksanen, 2013). Based on the complexity of the biological data, the models were ranked in the  
190 following order with increasing empirical evidence for a response of the taxa to the hydrological  
191 metrics and evidence for the existence of a hydrological threshold: (I) a flat response over the  
192 hydrological gradient, (II) monotone in-/decreasing model: a monotone increasing or decreasing trend  
193 with a data-driven optimum at the end or at the beginning, respectively, (III) interval optimum model:  
194 an increasing or decreasing trend with a plateau below the upper bound, (IV) symmetrical model: a  
195 symmetrical response curve with similar slopes on both sides, and (V) skewed model: a skewed  
196 response curve with a steeper slope toward the gradient end (Huisman et al., 1993a; Jansen and  
197 Oksanen, 2013) (Table 3).

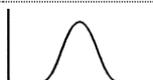
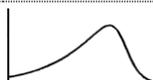
198 These models enable the determination and identification of taxa preferences for environmental  
199 conditions, e.g., min./max./optimum values for individual taxa. The set of five hierarchical models can  
200 be fitted to the observations and describe the response pattern over the environmental gradient with  
201 logistic and non-linear regression techniques (Huisman et al., 1993a; Jansen and Oksanen, 2013).

202 Covering a wide range of hydrological conditions across Germany, the probability of occurrence for  
203 individual taxa was determined and quantified by sorting the log-transformed abundance data along  
204 the gradient of each hydrological metric. The most adequate model type that best fitted the  
205 observations was selected according to its deviance from the log-likelihood of the predictions and an  
206 Akaike test (Akaike information criterion; AIC). The purity of the selected model type was quantified  
207 via bootstrapping with 100 re-sampling events (Supplementary material). The bootstrapping approach



208 changed the model selected for 15-25 % of the taxa (Figure SF1).

**Table 2** Description of eHOF models (according to Huisman et al., 1993; Jansen and Oksanen, 2013)

eHOF model type	Description	Model schemes
Flat response model (I)	A flat response over a hydrological gradient	
Monotone in-/decreasing model (II)	A monotone increasing or decreasing trend with an optimum at the end or at the beginning, respectively	
Interval optimum model (III)	An increasing or decreasing trend with a plateau below the upper boundary (the upper boundary is considered the optimum interval)	
Symmetrical model (IV)	A symmetrical response curve with similar slopes on both sides	
Skewed model (V)	A skewed response curve with a steeper slope toward one of the gradient ends	

209

### 210 1.3.5 Taxa responses to hydrological metrics and along the range of 211 hydrological gradients

212 Based on the individual model outcomes, the response shape and the highest probability of occurrence  
213 along the hydrological range, we evaluated the importance of each metric for the whole taxa pool.

214 To analyze the taxa responses along the range of hydrological gradients, the ranges of each metric  
215 were divided into quartiles. For each taxon, the quartile was recorded in which the optimum gradient  
216 value is reached (i.e., where the response is strongest). The optimum is an interval for taxa with an  
217 interval optimum model; therefore, their optimum gradient value might be affiliated with two or more  
218 quartiles. Prior to this analysis, we excluded taxa assigned to flat response and monotone in-  
219 /decreasing models, as they tolerate a wide range of hydrological conditions and an optimum value  
220 cannot be determined.

### 221 1.3.6 Hydrological thresholds

222 The eHOF models provide information on the shape of taxa responses along the hydrological  
223 gradients. Using this information, we identified hydrological thresholds (inflection points), where the  
224 maximum change occurs in taxa responses when moving along the hydrological gradient. These points  
225 can be detected on either the increasing or decreasing limb of the “interval optimum”,

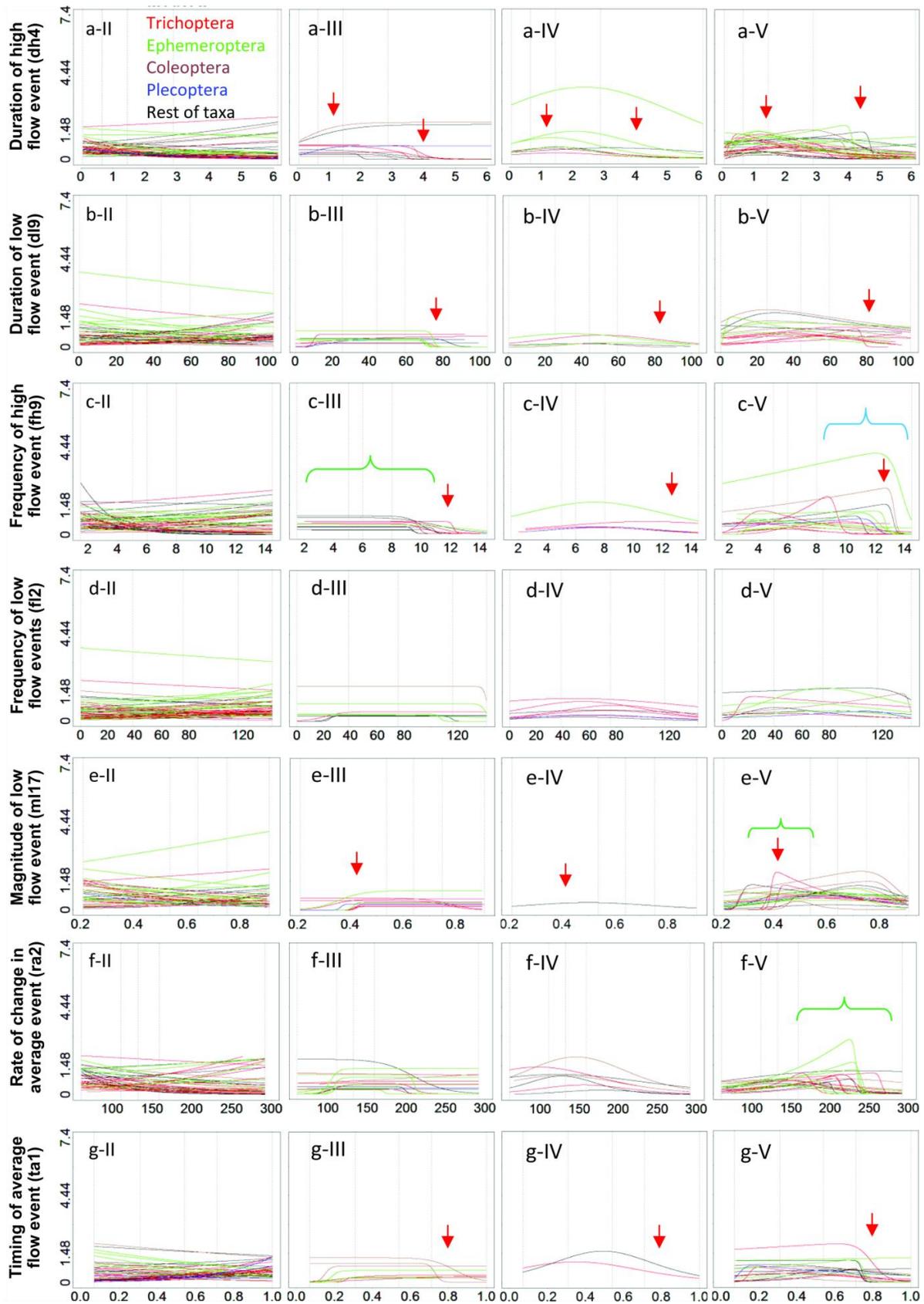


226 “symmetrical” and “skewed” models. All statistical analyses were carried out in R 3.2.3 (R  
 227 Development Core Team, 2016).

**Table 3** Descriptions, calculation procedures, units and temporal aspects of seven IHA metrics further used in this study (according to Olden & Poff (2003) and references therein).

IHA group (code)	IHA metric	Calculation procedure	Unit	Temporal aspect
Duration of high flow event (dh4)	Annual maximum 30-day moving average flows	Compute the max of 30-day moving average flows and take the max for each year; take the mean of these values. These values were log-transformed for use in the modelling approach.	Log(m <sup>3</sup> /s)	Daily
Duration of low flow event (dl9)	Variability in annual minimum 30-day moving average flows	Compute the standard deviation of the yearly min 30-day moving averages; multiply by 100 and divide by the mean of the yearly min 30-day moving averages	%	Daily
Frequency of high flow event (fh9)	Flood frequency	Compute the number of flow events with flows above the 75 % exceedance value for the full flow record; take the average number of events per year	1/year	Annual
Frequency of low flow event (fl2)	Variability in low pulse count	Compute the standard deviation of the average number of flow events per year below the 25th percentile for the full flow record; multiply by 100 and divide by the average number of flow events	%	Annual
Magnitude of low flow event (ml17)	Base flow	Compute mean annual flow, compute the min of a 7-day moving average annual flow and divide by the mean annual flow; calculate the mean of those ratios	Dimensionless	Annual
Rate of change in average event (ra2)	Variability in rise rate	Compute the standard deviation of positive flow changes for the full flow record, multiply by 100 and divide by the mean change in rising flows	%	Daily
Timing of average flow event (ta1)	Constancy	Compute constancy from the Colwell (1974) matrix	Dimensionless	Daily





228

229 **Figure 2** Results of models selected for all taxa among all metrics. Rows are separated according to  
 230 the seven IHA metrics (a to g), columns are separated according to eHOF model types.



The x-axis represents the gradient of the respective hydrological metrics, and the y-

232 axis is the probability of occurrence of the taxa, which is based on log-transformed abundance data.  
233 Quartiles are separated by gray dashed lines, and different colors represent orders. While taxa with the  
234 eHOF model types of “interval optimum”, “symmetrical” and “skewed” have preferences for specific  
235 ranges of hydrological values, taxa with a “flat response” model tolerate wide ranges of hydrological  
236 conditions and exhibit no response along the hydrological gradient. The plots of “flat response”  
237 models were excluded from this figure due to limited space and the simplicity of this model. The red  
238 arrows mark thresholds where the probability of taxa occurrence drastically decreases, and green and  
239 blue brackets mark gradient ranges that are preferred by taxa.

## 240 **1.4 Results**

### 241 **1.4.1 Taxa responses to hydrological metrics**

242 The following seven metrics (Table 4) remained after the pairwise collinearity test: “Duration of high  
243 flow event” (dh4), “duration of low flow event” (dl9), “frequency of high flow event” (fh9),  
244 “frequency of low flow event” (fl2), “magnitude of low flow event” (ml17), “rate of change in average  
245 event” (ra2) and “timing of average flow event” (ta1; Colwell , 1974). Figure 4 shows the results of  
246 the models selected for all taxa among all metrics.

247 The invertebrate taxa responded most strongly to dh4 (duration of high flow event) and ra2 (rate of  
248 change in average event), having the lowest share of flat response and monotone in-/decreasing  
249 models.

250 The flat response model was selected for 18-40 % of the taxa, depending on the hydrological metric  
251 (Figure 5). The share of taxa showing a monotone increase or decrease along the gradient of the seven  
252 hydrological variables was 35-53 %. The symmetrical model was the least often selected model  
253 among all metrics, selected, on average, for 4 % of the taxa (Figure 5).

### 254 **1.4.2 Taxa responses along the range of hydrological gradients**

255 23-41 % of all 120 taxa show clear preferences along the ranges of the seven hydrological metrics.  
256 The three eHOF model types “interval optimum”, “symmetrical” and “skewed” allow the  
257 determination of the positions of optimum values for taxa (Figure 6). The lowest proportion of taxa  
258 responding to specific ranges of a hydrological metric was to fl2 (frequency of low flow event), and  
259 the highest proportion was for dh4 (duration of high flow event).

260 Taxa occurrences according to the recorded quartile of optimum values varied among metrics (Figure  
261 6). The highest fraction of taxa with high occurrence probabilities in the first quartile were found for  
262 ml17 (magnitude of low flow event, 34 % of taxa) and ta1 (timing of average flow event, 42 % of  
263 taxa), i.e., taxa occurred more often when low flow events did not last long or when flow events were  
not highly constant, respectively. The last quartile ranked highest for fh9 (frequency of

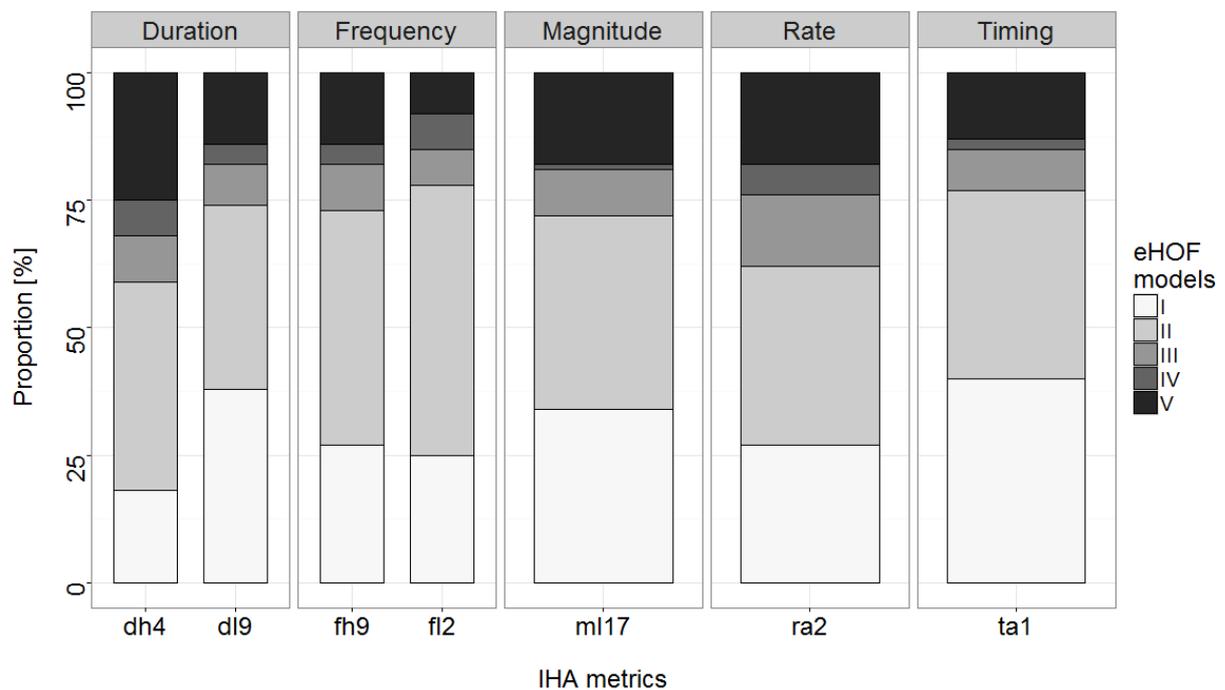


265 high flow event, 33 % of taxa) and ra2 (rate of change in average event, 38 % of taxa), i.e., taxa  
 266 occurred more often when a greater number of flow events were recorded with flows above the 75-  
 267 percent exceedance value for the entire 12-month period or taxa occurred more often with a higher  
 268 variability in the rise rate, respectively.

269 Taxa exhibit peak responses to values lower than the median for fl2 (frequency of low flow event, 61  
 270 %), ta1 (timing of average flow event, 58 %) and ml17 (magnitude of low flow event, 57 %) (Figure  
 271 6). Optimal responses were found for values higher than the median for ra2 (rate of change in average  
 272 event, 63 %), fh9 (frequency of high flow event, 59 %) and dh4 (duration of high flow event, 51 %).  
 273 Approximately equal proportions of taxa have peak preferences to either higher or lower median  
 274 values for both metrics belonging to the indicators of hydrologic alteration category of duration (dh4,  
 275 dl9); however, the preferences for quartiles differ significantly for all other categories (Figure 6).

276 Taxa responding according to the monotone in-/decreasing model show either a positive or a negative  
 277 trend in occurrence probabilities. The two metrics of dh4 and ra2 (duration of high flow event: 73 %  
 278 and rate of change in average event: 71 %) had the highest share of negative trends (Table 5), i.e., a  
 279 high proportion of taxa prefer low values along the gradient of maximum moving average flows or  
 280 variability in rise rate.

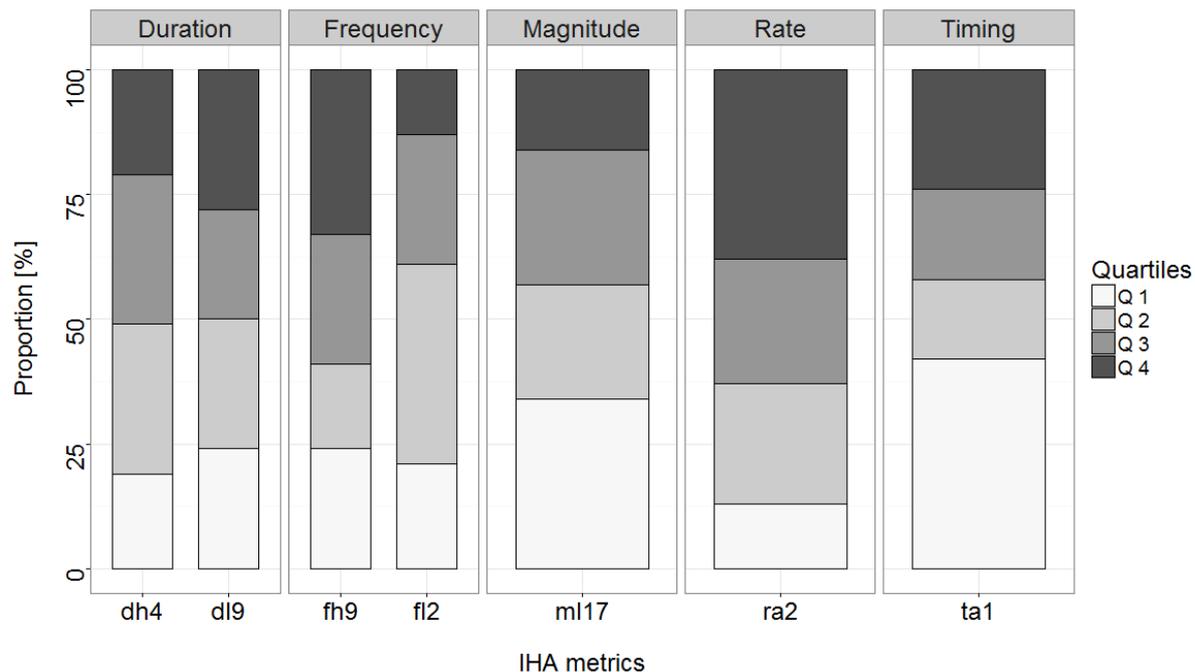
281



282

283 **Figure 3** Model frequencies and responses of taxa to selected IHA metrics (abbreviations in Table 4).





284

285 **Figure 4** Quartiles along the hydrological gradient where the taxa responses were strongest (i.e.,  
 286 location of optimum) according to the “interval optimum”, “symmetrical” and “skewed” eHOF  
 287 models (abbreviations in Table 4). Quartiles are shown as gray dashed lines in Figure 4.

288

289 In contrast, fl2 (frequency of low flow event, 70 %) and ta1 (timing of average flow event, 59 %)  
 290 include high proportions of positive trends in monotone in-/decreasing model outcomes, revealing  
 291 preferences for high values along the gradient of variability in low pulse count (Table 5).

292 For taxa with an interval optimum model, fh9 (frequency of high flow event, 100 %) and dh4  
 293 (duration of high flow event, 73 %) have the highest proportion of negative trends, while ml17  
 294 (magnitude of low flow event, 82 %) and ta1 (timing of average flow event, 78 %) include high  
 295 proportions of positive trends (Table 5).

296 Trichoptera had the highest proportion of taxa with the eHOF model types “interval optimum”,  
 297 “symmetrical” and “skewed” followed by Ephemeroptera, Coleoptera and Plecoptera taxa (Table 6),  
 298 which shows their preferences for specific ranges of hydrological conditions.

299 The highest response of Trichoptera taxa was to dh4 (duration of high flow event, 16 taxa) followed  
 300 by fh9 (frequency of high flow event, 14 taxa), while Ephemeroptera taxa responded mostly to ra2  
 301 (rate of change average event, 14 taxa) and dl9 (duration of low flow event, 8 taxa). Two Plecoptera  
 302 taxa respond to fh9 (frequency of high flow event), fl2 (frequency of low flow event) and ml17  
 303 (magnitude of low flow event), while only one taxon responds to dh4 (duration of high flow event),



304 dl9 (duration of low flow event), ra2 (rate of change in average event) and ta1 (timing of average flow  
 305 event).

**Table 4** Number and proportion of taxa with either a positive or negative trend in eHOF monotone in-/decrease or interval optimum model (abbreviations in Table 4).

	<b>dh4</b>	<b>dl9</b>	<b>fh9</b>	<b>fl2</b>	<b>ml17</b>	<b>ra2</b>	<b>ta1</b>
Number of monotone in-/decreasing model outcomes	49	43	55	64	46	42	44
Percentage of positive/negative trends for monotone in-/decreasing model	27 / 73	49 / 51	42 / 58	70 / 30	43 / 57	29 / 71	59 / 41
Number of interval optimum model outcomes	11	10	11	8	11	17	9
Percentage of positive/negative trends for interval optimum model	27 / 73	40 / 60	0 / 100	38 / 62	82 / 18	53 / 47	78 / 22

306

### 307 1.4.3 Hydrological thresholds

308 A strong decrease to nearly zero in taxa occurrence probabilities occurs if dh4 (duration of high flow  
 309 event) reaches values of either greater than four (i.e., 50 m<sup>3</sup>/s prior to log transformation) or less than  
 310 one (i.e., 1.7 m<sup>3</sup>/s prior to log transformation) as the maximum flow duration. Three taxa with interval  
 311 optimum models showed the highest occurrence probability for “duration of high flow event” > 4 (red  
 312 arrows mark these change points; Figure 4a).

313 Remarkable reductions in the probabilities of taxa occurrence were recorded for other hydrological  
 314 metrics, as well (red arrows for strong reductions; Figure 4b-g). A reduction was recorded at values  
 315 greater than 80 % coefficient of variation (CV) of dl9 (duration of low flow event), the magnitude of  
 316 minimum annual 30-day flows (Figure 4b). This threshold for fh9 (frequency of high flow event) was  
 317 more than 13, with the annual high flow events being above the threshold equal to the 75-percent  
 318 exceedance value. Taxa preferences for ranges of ml17 (magnitude of low flow event) were  
 319 substantially reduced by values less than 0.4, describing a very low base flow index (Figure 4e).  
 320 Constancy values of greater than 0.8 for ta1 (timing of average flow event) are also not preferred by  
 321 the taxa (Figure 4g).



322 Focusing on specific models, none of taxa with an interval optimum model tolerated a mean number  
 323 of annual high flow events larger than 12, and all preferred fewer than 9 according to dh4 (green curly  
 324 bracket; Figure 4c-III). Taxa with a skewed model highly prefer quartile four of fh9 (frequency of high  
 325 flow events) with gradient values between 8 and 13 (blue curly bracket; Figure 4c-V). The probability  
 326 of taxa occurrence increases at low flow magnitudes  $> 0.4$  (Figure 4e-III), but several taxa with a  
 327 skewed model preferred the values in quartile one based on this hydrological metric (green curly  
 328 bracket; Figure 4e-V). Almost all taxa with skewed models had an optimum in the fourth quartile for  
 329 ra2 (rate of change average event), which means they tolerate wide ranges of rise rates (green curly  
 330 bracket; Figure 4f-V). The gradient values of these thresholds are shown in Table ST1 for individual  
 331 taxa.

332

**Table 5** Proportion (%) of taxa from major orders and all others as “Rest” that respond to hydrological metrics with interval optimum, symmetrical or skewed model types. Taxa with a flat response model or monotone in-/decreasing model are not included due to their flat or nearly zero responses along the hydrological gradients (abbreviations in Table 4).

Models	Category	Metric	Trichoptera (43 taxa)	Ephemeroptera (25 taxa)	Plecoptera (4 taxa)	Coleoptera (12 taxa)	Rest (36 taxa)
Interval optimum, symmetrical or skewed model (III, IV and V)	Duration	dh4	37.2 %	8.0 %	25.0 %	50.0 %	50.0 %
	Duration	dl9	25.6 %	36.0 %	25.0 %	41.7 %	16.7 %
	Frequency	fh9	32.6 %	32.0 %	50.0 %	41.7 %	25.0 %
	Frequency	fl2	18.6 %	24.0 %	50.0 %	16.7 %	25.0 %
	Magnitude	ml17	16.3 %	28.0 %	50.0 %	50.0 %	13.9 %
	Rate	ra2	30.2 %	56.0 %	25.0 %	58.3 %	36.1 %
	Timing	ta1	14.0 %	24.0 %	25.0 %	58.3 %	22.2 %

333

## 334 1.5 Discussion

### 335 1.5.1 Hydrological metrics

336 We used 12 months of continuous daily discharge data antecedent to each individual benthic  
 337 invertebrate sampling date to compute IHA metrics describing the hydrological



338 conditions at each sampling site. The short-term hydrological conditions prior to biological sampling  
339 are important in the occurrences and diversity of individual taxa and describe any changes according  
340 to recent hydrological conditions (Stewart-Koster et al., 2011). The calculation of IHA metrics based  
341 on very long periods of discharge and continuous biological data are more of interest if focusing on  
342 historical adaptations and long-term changes in assemblages of benthic invertebrates (Leigh and Datry,  
343 2016).

## 344 **1.5.2 Taxa responses to metrics and along the range of hydrological gradients**

345 A variety of taxa responses to recent hydrological conditions were revealed. Our analysis determined  
346 the analytical optimum of taxa to seven hydrological metrics according to individual modelling  
347 responses per taxon. Taxa for which the flat response model was selected can be considered  
348 ubiquitous, as these taxa have no clear optima or preferences along the hydrological gradients. Taxa  
349 for which the symmetrical or skewed model (and potentially the interval optimum model) was chosen,  
350 showed clear responses to specific ranges of the hydrological gradients. They might cope with changes  
351 in that range compared to changes at the gradient ends, where the probability of occurrence is lower.  
352 Our results of the model frequency analysis (Figure 5) show that taxa responded more often to metrics  
353 describing high flow magnitudes and frequencies rather than low flows and to other metrics such as  
354 ra2 (rate of change in average event). These results are in concordance with other studies that reported  
355 strong influences of high flow conditions on species of benthic invertebrates (Clausen and Biggs,  
356 1997; Death and Winterbourn, 1995; Suren and Jowett, 2006).

357 We covered wide ranges of quantified hydrological conditions, while previous semi-quantitative  
358 studies have covered limited ranges of flow conditions at the reach scale due to their methodology and  
359 data availability, e.g., the FST-hemispheres that were introduced by Statzner et al. (1988). The existing  
360 qualitative or semi-quantitative data described at the nominal or ordinal scales (Schmedtje, 1995;  
361 Schmidt-Kloiber and Hering, 2015) are barely comparable with the quantitative hydrological traits of  
362 benthic invertebrates and responses of taxa and thresholds along the hydrological gradients evaluated  
363 here. However, there are ecologically meaningful links between the ecological and hydrological  
364 preferences of taxa. For example, the taxa showing a clear negative response to high flows (e.g. dh4,  
365 fh9 and ra2) are prone of drifting by high flows as taxa respond to hydrological and hydraulic stress  
366 (Statzner and Holm, 1982). *Anabolia nervosa* and *Pisidium subtruncatum* show negative responses to  
367 metrics describing high flows, and preferably occur in standing waters and avoid current (Schmidt-  
368 Kloiber and Hering, 2015 and references therein); therefore, may be prone of being affected by higher  
369 high flows. Besides, taxa showing a clear negative response to low flows (e.g. dl9, fl2 or ml17) have a  
370 high oxygen demand, and hence are vulnerable to extreme low flow conditions in summer related to  
371 high water temperature and low oxygen content (Brooks et al., 2011). *Habroleptoides confusa*,

372 *Hydropsyche pellucidula*, *Baetis rhodani* and *Heptagenia sulphurea* show negative



373 responses to metrics describing low flow conditions, and preferably occur in streams with moderate to  
374 high current (Schmidt-Kloiber and Hering, 2015 and references therein), thus may be prone of being  
375 affected by low flow conditions.

376 The “duration of high flow event” (dh4) describes the amount of discharge a taxon might tolerate over  
377 a period of maximum 30-day moving average flows. This metric therefore describes the river size at  
378 which discharge is larger in rivers compared to streams and within the same river size the point at  
379 which it is larger for more dynamic flow regimes compared to those that are more monotonous. This  
380 reflects the river continuum concept, i.e., some taxa prefer to inhabit upstream areas of small streams,  
381 while others prefer larger streams or rivers (Vannote et al., 1980). The ecological trait of “stream  
382 zonation preference” (freshwaterecology.info, Schmidt-Kloiber and Hering, 2015) may describe taxa  
383 hydrological preferences to dh4 (duration of high flow event) best. However, the information is  
384 available for only 88 of 120 taxa. The quantitative responses of more than 90 % of 88 taxa make sense  
385 ecologically and fit to the expert judgment in the freshwater ecology database. For example, all taxa  
386 with a “flat response model” are marked as having preferences to almost all categories of “stream  
387 zonation preference”. *Baetis buceratus*, *Baetis muticus* and *Glossiphonia complanata* are indicator  
388 taxa that occur in almost all categories and show a flat response along the range of “duration of high  
389 flow event”. The inconsistencies for less than 10 % of the taxa might be due to data deficiencies or  
390 methodological constraints.

391 The information on the ecological traits of stream benthic invertebrates is lacking for dozens of taxa,  
392 which hinders the description of their ecological and hydrological requirements; however, we  
393 successfully determined quantitative hydrological requirements for all studied taxa.

### 394 **1.5.3 Hydrological thresholds**

395 Both very high and very low flow conditions influence the abundances of benthic invertebrates in river  
396 ecosystems (Dewson et al., 2007b; Suren and Jowett, 2006). Although the gradient ends of  
397 hydrological metrics are not well suited for taxa, all taxa show strong responses to the first or fourth  
398 quartile of hydrological metrics (e.g., taxa with a skewed model; Figure 5). High values of maximum  
399 flow duration render taxa unable to resist against flow and drift downstream as a result (Lake, 1990).

400 Moreover, high gradient values of dl9 (duration of low flow event) and low gradient values of ml17  
401 (magnitude of low flow event) are not suitable for taxa, as critical thresholds of low flows might be  
402 reached (Acuna et al., 2005), which are associated with high water temperatures and linked with low  
403 dissolved oxygen concentrations (Brooks et al., 2011).

404 Global change might potentially affect taxa by leading to changes in flow regime and discharge  
405 conditions in similar ranges of the gradient affecting some taxa. A vulnerability analysis of taxa  
406 according to their hydrological thresholds requires high-resolution hydrological data



407 from climate change hydrological models. Quantitative hydrological traits are therefore suitable  
408 information for modelling and predicting the effects of flow changes due to global change.

#### 409 **1.5.4 Methodological constraints**

410 The hydrological metrics are inherently co-correlated (Olden and Poff, 2003). We aimed to analyze  
411 taxa responses to each IHA category; therefore, at least one metric per category was selected as being  
412 representative for that specific hydrological feature, resulting in seven metrics from five IHA  
413 categories. The seven selected metrics are representative of many other metrics of the same group.  
414 Even in this highly reduced set of metrics, some correlation occurs, for example, dh4 (the annual  
415 maximum 30-day moving average flows, duration of high flow event) is highly correlated with dl9  
416 (annual minimum 30-day moving average flows) with a pairwise correlation value of negative one ( $|r|$   
417 = -1). Therefore, taxa responses cannot be judged as unique with certainty, and a currently unknown  
418 proportion of taxa could respond to either metric.

419 Taxa with monotone in-/decreasing model show a preference—through an increasing or decreasing  
420 trend—to either low or high values along the range of hydrological metrics with the analytical  
421 optimum at the gradient end. Taxa with an interval optimum model also have a threshold at which the  
422 occurrence probabilities increase/decrease drastically with a plateau at the upper level. The decreasing  
423 limb of taxa with this model is missing. However, these models are extremely sensitive at the gradient  
424 ends and can be affected even by a single data point (Jansen and Oksanen, 2013). This implies that the  
425 hydrological range of the respective taxa is probably not fully covered in the data.

426 Taxa responses vary across life stages (Lancaster and Downes, 2010a). Biological sampling at only  
427 one specific time of the year results in invertebrate species at a specific stage in their life cycle being  
428 represented. This might affect the integrity of the determination of taxa responses to hydrological  
429 conditions (Lancaster and Downes, 2010a; Lancaster et al., 2009). Although we used benthic  
430 invertebrate sample data from two seasons over a 10-year period, our dataset cannot fully overcome  
431 this difficulty, as data on all life stages of the life cycle are not available through the standard  
432 biomonitoring procedure.

433 The *in-situ* probability of taxa occurrences depends on many environmental variables. In particular,  
434 land use, habitat availability and water quality are known to be influential even over long periods of  
435 time (Allan, 2004; Harding et al., 1998). Although the range of hydrological conditions was well-  
436 covered by the rich biological data from sampling sites with good or high ecological status, other  
437 environmental variables might still influence taxa occurrences (Stoll et al., 2016; Tonkin et al., 2016).  
438 Furthermore, suitable data for ecological processes such as competition are lacking, and these  
439 processes were not considered in this study.



440 Evolved traits enable benthic invertebrates to survive flow conditions within the context of natural  
441 flow regimes (Lytle and Poff, 2004), and the abundance and structure of their communities are  
442 believed to be significantly affected by changing hydrological conditions (Sousa, 1984). Global  
443 change is influencing all aspects of the flow regime in space and over time, causing, e.g., an increase  
444 in extremely low or high flow conditions (IPCC, 2007, 2014). Germany is also facing the impacts of  
445 global change-induced flow alteration, with low and high flow conditions projected to occur more  
446 often (Nilson, 2014), which affect the distribution and probability of occurrence of several taxa. As  
447 ecological processes and the abundance and distribution of aquatic invertebrates are strongly  
448 influenced by the actual type of flow regime (Poff et al., 1997), the benthic invertebrate community  
449 will respond to flow alteration by changes in their diversity and abundance (Arthington et al., 2006;  
450 Brooks et al., 2011; Poff and Zimmerman, 2010) as well as by plasticity and adaptations (Stoks et al.,  
451 2014).

### 452 **1.5.5 Summary and outlook**

453 Our study represents a shift from existing studies on ecological traits, which are based on largely  
454 qualitative data and often grounded in expert knowledge and literature analysis, to describing  
455 hydrological traits, which are quantitative and data-based. However, these quantitative hydrological  
456 traits do not replace the categories of ecological traits that are linked to hydrology, e.g.,  
457 resistance/resilience to droughts (Schmidt-Kloiber and Hering, 2015), but preferentially append new  
458 categories that might be useful for forecasting changes.

459 The quantified hydrological traits of individual taxa might therefore support stream management and  
460 enable the prediction of taxa responses to flow alteration. Such large-scale studies of flow preferences  
461 for modelling individual taxa responses to hydrological gradients can be implemented to optimize  
462 taxon-specific hydrological models.

463 The hydrological traits of stream benthic invertebrates may be used in forecasting studies in central  
464 Europe, and the methods used in this study are suitable for application in other regions, where a  
465 different flow regime might suggest the need to analyze other flow metrics. Other hydrological traits,  
466 e.g., those regarding extreme events, could also be modelled depending on research questions and  
467 interests.

### 468 **Acknowledgements**

469 This study was funded through the “GLANCE” project (Global Change Effects in River Ecosystems;  
470 01 LN1320A) supported by the German Federal Ministry of Education and Research (BMBF). We  
471 thank the German Working Group on Water Issues of the Federal States and the Federal Government  
472 (LAWA) for providing the biomonitoring data and German federal state environmental agencies for



providing high quality discharge data.

474 **References**

- 475 Acuna, V., Munoz, I., Giorgi, A., Omella, M., Sabater, F., Sabater, S., 2005. Drought and  
476 postdrought recovery cycles in an intermittent Mediterranean stream: structural and  
477 functional aspects. *Journal of the North American Benthological Society* 24, 919-933.
- 478 Allan, J.D., 2004. Landscapes and riverscapes: the influence of land use on stream  
479 ecosystems. *Annual review of ecology, evolution, and systematics*, 257-284.
- 480 Archfield, S.A., Kennen, J.G., Carlisle, D.M., Wolock, D.M., 2014. An objective and  
481 parsimonious approach for classifying natural flow regimes at a continental scale. *River  
482 Research and Applications* 30, 1166-1183.
- 483 Armanini, D.G., Horrigan, N., Monk, W.A., Peters, D.L., Baird, D.J., 2011. Development of a  
484 Benthic Macroinvertebrate Flow Sensitivity Index for Canadian Rivers. *River Research and  
485 Applications* 27, 723-737.
- 486 Arthington, A.H., Bunn, S.E., Poff, N.L., Naiman, R.J., 2006. The challenge of providing  
487 environmental flow rules to sustain river ecosystems. *Ecological Applications* 16, 1311-1318.
- 488 Bellard, C., Bertelsmeier, C., Leadley, P., Thuiller, W., Courchamp, F., 2012. Impacts of  
489 climate change on the future of biodiversity. *Ecology Letters* 15, 365-377.
- 490 Brooks, A.J., Chessman, B.C., Haeusler, T., 2011. Macroinvertebrate traits distinguish  
491 unregulated rivers subject to water abstraction. *Journal of the North American Benthological  
492 Society* 30, 419-435.
- 493 Bunn, E.S., Arthington, H.A., 2002. Basic Principles and Ecological Consequences of Altered  
494 Flow Regimes for Aquatic Biodiversity. *Environmental Management* 30, 492-507.
- 495 Clausen, B., Biggs, B., 1997. Relationships between benthic biota and hydrological indices in  
496 New Zealand streams. *Freshwater biology* 38, 327-342.
- 497 Death, R.G., 2008. The effect of floods on aquatic invertebrate communities. *Aquatic insects:  
498 challenges to populations*, 103-121.
- 499 Death, R.G., Winterbourn, M.J., 1995. Diversity patterns in stream benthic invertebrate  
500 communities: the influence of habitat stability. *Ecology* 76, 1446-1460.
- 501 DeWeber, J.T., Wagner, T., 2014. A regional neural network ensemble for predicting mean  
502 daily river water temperature. *Journal of Hydrology* 517, 187-200.
- 503 Dewson, Z.S., James, A.B.W., Death, R.G., 2007. A review of the consequences of decreased  
504 flow for instream habitat and macroinvertebrates. *Journal of the North American  
505 Benthological Society* 26, 401-415.
- 506 Dormann, C.F., Elith, J., Bacher, S., Buchmann, C., Carl, G., Carre, G., Marquez, J.R.G., Gruber,  
507 B., Lafourcade, B., Leitao, P.J., Munkemuller, T., McClean, C., Osborne, P.E., Reineking, B.,  
508 Schroder, B., Skidmore, A.K., Zurell, D., Lautenbach, S., 2013. Collinearity: a review of  
509 methods to deal with it and a simulation study evaluating their performance. *Ecography* 36,  
510 27-46.
- 511 Dunbar, M.J., Pedersen, M.L., Cadman, D., Extence, C., Waddingham, J., Chadd, R.P., Larsen,  
512 S.E., 2010a. River discharge and local-scale physical habitat influence macroinvertebrate LIFE  
513 scores. *Freshwater Biology* 55, 226 - 242.
- 514 Gittleman, J.L., Kot, M., 1990. Adaptation: Statistics and a Null Model for Estimating  
515 Phylogenetic Effects. *Systematic Zoology* 39, 227-241.
- 516 Hale, A.N., Noble, G., Piper, K., Garmire, K., Tonsor, S.J., 2016. Controlling for hydrologic  
517 connectivity to assess the importance of catchment-and reach-scale factors on  
518 macroinvertebrate community structure. *Hydrobiologia* 763, 285-299.



519 Harding, J., Benfield, E., Bolstad, P., Helfman, G., Jones, E., 1998. Stream biodiversity: the  
520 ghost of land use past. *Proceedings of the national academy of sciences* 95, 14843-14847.

521 Heino, J., Soininen, J., 2010. Are common species sufficient in describing turnover in aquatic  
522 metacommunities along environmental and spatial gradients? *Limnology and Oceanography*  
523 55, 2397–2402.

524 Henriksen, J.A., Heasley, J., Kennen, J.G., Niewswand, S., 2006. Users' manual for the  
525 Hydroecological Integrity Assessment Process software (including the New Jersey  
526 Assessment Tools): U.S. Geological Survey Open-File Report. 1093 - 2006.

527 Huisman, J., Olf, H., Fresco, L., 1993. A hierarchical set of models for species response  
528 analysis. *Journal of Vegetation Science* 4, 37-46.

529 IPCC, 2007. *Climate Change 2007: The Physical Science Basis. Contribution of Working Group*  
530 *I to the Fourth Assessment Report of the Intergovernmental Panel on Climate Change*, in: S.  
531 Solomon, D.Q., M. Manning, Z. Chen, M. Marquis, K.B. Averyt, M. Tignor, H.L. Miller (Ed.).  
532 Cambridge University Press., Cambridge, United Kingdom and New York, NY, USA.

533 IPCC, 2014. *Climate Change 2014: Synthesis Report. Contribution of Working Groups I, II and*  
534 *III to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change*, in: R.K.  
535 Pachauri, Meyer, L.A. (Eds.). IPCC, Geneva, Switzerland, p. 151pp.

536 Jansen, F., Oksanen, J., 2013. How to model species responses along ecological gradients –  
537 Huisman–Olf–Fresco models revisited. *Journal of Vegetation Science* 24, 1108-1117.

538 Konrad, C.P., Brasher, A.M.D., May, J.T., 2008. Assessing streamflow characteristics as  
539 limiting factors on benthic invertebrate assemblages in streams across the western United  
540 States. *Freshwater Biology* 53, 1983-1998.

541 Lake, P.S., 1990. effects of successive flow perturbations on stream invertebrates. *Australian*  
542 *Journal of Ecology* 15, 477 - 488.

543 Lancaster, J., Downes, B.J., 2010. Ecohydraulics needs to embrace ecology and sound  
544 science, and to avoid mathematical artefacts. *River Research and Applications* 26, 921-929.

545 Lancaster, J., Downes, B.J., Glaister, A., 2009. Interacting environmental gradients, trade-offs  
546 and reversals in the abundance-environment relationships of stream insects: when flow is  
547 unimportant. *Marine and Freshwater Research* 60, 259-270.

548 Leigh, C., Datry, T., 2016. Drying as a primary hydrological determinant of biodiversity in river  
549 systems: a broad-scale analysis. *Ecography*, n/a-n/a.

550 Lobo, J.M., Tognelli, M.F., 2011. Exploring the effects of quantity and location of pseudo-  
551 absences and sampling biases on the performance of distribution models with limited point  
552 occurrence data. *Journal for Nature Conservation* 19, 1-7.

553 Lytle, D.A., Poff, N.L., 2004. Adaptation to natural flow regimes. *Trends in Ecology &*  
554 *Evolution* 19, 94-100.

555 Monk, W.A., Wood, P.J., Hannah, D.M., Wilson, D.A., 2007. Selection of river flow indices for  
556 the assessment of hydroecological change. *River Research and Applications* 23, 113-122.

557 Monk, W.A., Wood, P.J., Hannah, D.M., Wilson, D.A., Extence, C.A., Chadd, R.P., 2006. Flow  
558 variability and macroinvertebrate community response within riverine systems. *River*  
559 *Research and Applications* 22, 595-615.

560 Nilson, E., 2014. Auswirkungen des Klimawandels auf das Abflussgeschehen und die  
561 Binnenschifffahrt in Deutschland: Schlussbericht KLIWAS-Projekt 4.01. Bundesanstalt für  
562 Gewässerkunde.

563 Olden, J.D., Poff, N.L., 2003. Redundancy and the choice of hydrologic indices for  
564 characterizing streamflow regimes. *River Research and Applications* 19, 101-121.



565 Poff, N.L., Allan, J.D., Bain, M.B., Karr, J.R., Prestegard, K.L., Richter, B.D., Sparks, R.E.,  
566 Stromberg, J.C., 1997. The natural flow regime. *Bioscience* 47, 769-784.

567 Poff, N.L., Julie Zimmerman, K.H., 2010. Ecological responses to altered flow regimes: a  
568 literature review to inform the science and management of environmental flows. *Freshwater*  
569 *Biology* 55, 194–205.

570 Poff, N.L., Olden, J.D., Merritt, D.M., Pepin, D.M., 2007. Homogenization of regional river  
571 dynamics by dams and global biodiversity implications. *Proceedings of the National Academy*  
572 *of Sciences* 104, 5732–5737.

573 Principe, R.E., Raffaini, G.B., Gualdoni, C.M., Oberto, A.M., Corigliano, M.C., 2007. Do  
574 hydraulic units define macroinvertebrate assemblages in mountain streams of central  
575 Argentina? *Limnologia* 37 323–336.

576 Schmedtje, U., 1995. Ökologische Grundlagen für die Beurteilung von Ausleitungsstrecken.  
577 Beziehung zwieschen der sohnlnahen Strömung, dem Gewässerbett und dem  
578 Makrozoobenthos in Fließgewässern. *Informationsberichte des Bayerischen Landesamtes für*  
579 *Wasserwirtschaft*, Heft 25.

580 Schmedtje, U., Sommerhäuser, M., Braukmann, U., Briem, E., Haase, P., Hering, D., 2000.  
581 Top-down. Konzept einer biozönotisch begründeten Fließgewässertypologie Deutschlands.  
582 In: *Deutsche Gesellschaft für Limnologie (Hrsg.): Tagungsbericht*, 147-151.

583 Schmidt-Kloiber, A., Hering, D., 2015. [www.freshwaterecology.info](http://www.freshwaterecology.info) – An online tool that  
584 unifies, standardises and codifies more than 20,000 European freshwater organisms and their  
585 ecological preferences. *Ecological Indicators* 53, 271–282.

586 Sousa, W.P., 1984. The role of disturbance in natural communities. *Annual review of ecology*  
587 *and systematics* 15, 353-391.

588 Statzner, B., Gore, J.A., Resh, V.H., 1988. Hydraulic Stream Ecology - Observed Patterns and  
589 Potential Applications. *Journal of the North American Benthological Society* 7, 307-360.

590 Statzner, B., Holm, T.F., 1982. Morphological adaptations of benthic invertebrates to stream  
591 flow—an old question studied by means of a new technique (Laser Doppler Anemometry).  
592 *Oecologia* 53, 290-292.

593 Stewart-Koster, B., Olden, J., Kennard, M.J., Pusey, B.J., Boone, E., Douglas, M., Jackson, S.,  
594 2011. Fish response to the temporal hierarchy of the natural flow regime in the Daly River,  
595 northern Australia. *Journal of Fish Biology* 79, 1525-1544.

596 Stoks, R., Geerts, A.N., De Meester, L., 2014. Evolutionary and plastic responses of  
597 freshwater invertebrates to climate change: realized patterns and future potential.  
598 *Evolutionary applications* 7, 42-55.

599 Stoll, S., Breyer, P., Tonkin, J.D., Früh, D., Haase, P., 2016. Scale-dependent effects of river  
600 habitat quality on benthic invertebrate communities—Implications for stream restoration  
601 practice. *Science of the Total Environment* 553, 495-503.

602 Suren, A.M., Jowett, I.G., 2006. Effects of floods versus low flows on invertebrates in a New  
603 Zealand gravel-bed river. *Freshwater Biology* 51, 2207-2227.

604 Team, R.C., 2016. R: A language and environment for statistical computing. R Foundation for  
605 Statistical Computing, Vienna, Austria. 2015. URL h ttp. [www.R-project.org](http://www.R-project.org).

606 Tonkin, J.D., Stoll, S., Jähnig, S.C., Haase, P., 2016. Anthropogenic land-use stress alters  
607 community concordance at the river-riparian interface. *Ecological Indicators* 65, 133-141.

608 VanDerWal, J., Shoo, L.P., Graham, C., Williams, S.E., 2009. Selecting pseudo-absence data  
609 for presence-only distribution modeling: How far should you stray from what you know?  
610 *Ecological Modelling* 220, 589-594.



611 Vannote, R.L., Minshall, G.W., Cummins, K.W., Sedell, J.R., Cushing, C.E., 1980. The river  
612 continuum concept. Canadian journal of fisheries and aquatic sciences 37, 130-137.  
613

