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- Large-stream nitrate retention patterns shift during droughts: seasonal to
   sub-daily insights from high-frequency data-model fusion
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#### 18 Abstract

High-frequency nitrate-N  $(NO_3^- - N)$  data are increasingly available, while accurate assessments 19 of in-stream  $NO_3^- - N$  retention in large streams and rivers require a better capture of complex 20 21 river hydrodynamic conditions. This study demonstrates a fusion framework between high-22 frequency water quality data and hydrological transport models, that (1) captures river hydraulics 23 and their impacts on solute signal propagation through river hydrodynamic modeling, and (2) 24 infers in-stream retention as the differences between conservatively traced and reactively observed  $NO_3^- - N$  signals. Using this framework, continuous 15-min estimates of  $NO_3^- - N$ 25 26 retention were derived in a 6<sup>th</sup>-order reach of the lower Bode River (27.4 km, central Germany), 27 using long-term sensor monitoring data during a period of normal flow from 2015-2017 and a period of drought from 2018-2020. The unique  $NO_3^- - N$  retention estimates, together with 28 29 metabolic characteristics, revealed insightful seasonal patterns (from high net autotrophic uptake 30 in late-spring to lower uptake, to net heterotrophic release during autumn) and drought-induced 31 variations of those patterns (reduced levels of net uptake and autotrophic nitrate uptake largely

32 buffered by heterotrophic release processes, including organic matter mineralization). Four 33 clusters of diel uptake patterns were identified, potentially representing changes in dominant 34  $NO_3^- - N$  retention processes according to seasonal and hydrological conditions. For example, dominance of autotrophic  $NO_3^- - N$  retention extended more widely across seasons during the 35 drought years. Such cross-scale patterns and changes under droughts are likely co-determined by 36 37 catchment and river environments (e.g., river primary production, dissolved organic carbon 38 availability and its quality), which resulted in more complex responses to the sequential 39 droughts. Inferences derived from this novel data-model fusion provide new insights into  $NO_3^$ dynamics and ecosystem function of large streams, as well as their responses to climate 40 41 variability. Moreover, this framework can be flexibly transferred across sites and scales, thereby 42 complementing high-frequency monitoring to identify in-stream uptake processes and to inform 43 river management.

Key words: long-term high-frequency monitoring; river hydrodynamic model; conservative
tracer simulation; river metabolism; droughts; river ecosystem function

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### 47 **1 Introduction**

48 Anthropogenically induced high nitrate  $(NO_3^-)$  levels in rivers are a pervasive threat to

49 freshwater and costal ecosystems, especially under the changing climate (Costa et al., 2022;

50 Reusch et al., 2018; Rockström et al., 2009). In addition to efforts to control point- and diffuse-

51 sources, the "self-cleansing" capacity of river systems to retain and process  $NO_3^-$  has also been

52 recognized (Ensign and Doyle, 2006; Jäger et al., 2017). Restoring rivers and their floodplains

has been advocated as a part of "natural-based" solutions to reduce  $NO_3^-$  losses to water bodies,

54 with multiple co-benefits (EEA, 2021).

55 Large streams and rivers (e.g.,  $\geq 4^{\text{th}}$  order) receive considerable management attention, due to the

56 impacts of nutrient pollution on water-related ecosystem services, and because in-stream

57 biological activity likely contributes disproportionately to whole-river ecosystem function

58 (Wollheim et al., 2022). However, current understanding of in-stream processing is mostly

59 extrapolated from tracer-addition experiments or intensive process monitoring in small,

60 headwater streams (Ensign and Doyle, 2006; Mulholland et al., 2008). Large streams and rivers

are subject to a range of complex interactions between hydrodynamics, biological and

62 anthropogenic activities that may limit the applicability of extrapolation from headwater

63 measurements (Bernal et al., 2019). Moreover, owing to the practical challenges of making direct

64 process measurements in large, deep non-wadeable rivers, there is a shortage of data on whole-

65 stream  $NO_3^-$  biochemical cycling and its constituent pathways (e.g., assimilation, denitrification,

66 mineralization and nitrification).

67 Fostered by sensor technology development, high-frequency measurements are increasingly

being used to infer in-stream  $NO_3^-$  retention and release in large streams and rivers (Rode et al.,

69 2016b). Biochemical process-related information can be directly extracted from  $NO_3^-$  timeseries

70 (Burns et al., 2019; Heffernan and Cohen, 2010). However, inferences based purely on high-

71 frequency data are often conducted under well constrained conditions, e.g., a steady-state

vupstream boundary (Hensley and Cohen, 2016; Yang et al., 2019). While the more adaptive two-

station approach allows a dynamic upstream boundary, the inference accuracy is still largely

subject to influences of hydraulic transformation (e.g., dispersion Hensley and Cohen, (2016))

and dilution/enrichment from lateral inflow along the reach between upstream and downstream

stations (Zhang et al., 2022). Consequently, in-stream process quantifications remain uncertain

77 when extrapolating to various hydro-morphological river conditions outside the monitored

reaches. As a result, while continuous high-frequency monitoring has been deployed for multiple

79 years across river sizes and geomorphological conditions (Arndt et al., 2022; Bieroza et al.,

80 2023; Rode et al., 2016b), inferences of in-stream processes using such invaluable long-term

81 data, is often obscure due to dynamic flow conditions and their complex convolutions with

82 biological activity (Hensley and Cohen, 2016; Payn et al., 2017). This complexity has hindered

83 multi-year continuous investigations of in-stream  $NO_3^-$  retention, and the responses of in-stream

84 processes to the changing climate, such as the severe droughts in Europe over last five years.

85 Conventional two-station inferences of in-stream  $NO_3^-$  biogeochemical retention along the reach

86 were based on load  $(Q \times NO_3^- - N)$  differences between the upstream and downstream stations.

- 87 However, purely using the monitored high-frequency data is challenging to estimate lateral
- 88 inflows and transport time lags, and their inferences on the calculation require specific
- 39 justifications (Jarvie et al., 2018; Zhang et al., 2022). Alternatively, high-frequency discharge

90 (O) and solute dataset are increasingly used to calibrate river modeling approaches, which 91 normally include both hydraulic transport (advective and dispersive) and biogeochemical 92 retention terms in their master equation of mass balance (Hensley and Cohen, 2016; Huang et al., 93 2022). However, the simulations of biological processes are often challenged by complex 94 hydraulic-biological convolutions and model parameterization (Rode et al., 2007). To benefit 95 from the complementary data analysis and modelling, here we propose an innovative data-model 96 fusion framework for in-stream process inferences in large streams and rivers. The framework is 97 mainly based on the two-station high-frequency monitoring, but untangling impacts of hydraulic 98 transformations along the reach using the advanced features of the WASP8 water and tracer 99 simulations (Ambrose and Wool, 2017; Knightes et al., 2019). Such methodological 100 improvements are necessary to study the nitrate uptake dynamics under varying climatic and 101 flow conditions.

Here we illustrated the analysis using six years of data from a 27.4 km 6<sup>th</sup> order reach of the lower Bode River, central Germany. The objectives of this study were: (1) to estimate continuous in-stream  $NO_3^-$  overall retention (net uptake/release) over the normal (2015-2017) and drought (2018-2020) years based on the fusion of high-frequency data analysis and modeling; (2) to investigate responses of the in-stream processes to the droughts from sub-daily to inter-annual scales; and (3) to unravel potential environmental factors that control the in-stream processes and their pattern shifts under the drought disturbances.

## 109 2 Method and Materials

### 110 **2.1 Study site and long-term high-frequency monitoring**

111 The Bode River, ca. 169 km long with a watershed area of 3270 km<sup>2</sup>, originates in the Harz

112 Mountain area, central Germany. The studied 27.4 km reach of the lower Bode River (6<sup>th</sup> order)

is surrounded by intensive lowland arable land (Figure 1a). Between the study period 2015-2020,

114 multi-parameter high-frequency monitoring equipment was deployed at two stations: the

- 115 upstream station Groß Germersleben (GGL) and the downstream station Staßfurt (STF). The
- river reach mostly exhibits rectangular or trapezoidal cross-sections, due to artificial
- 117 modifications in the 1970s for the purpose of agricultural activities. There are also three
- 118 overflow weirs installed along the reach. The riverbed substrate consists of mostly sand and
- small gravel, with an average width and slope of 20 m and 0.4 ‰, respectively. Patchy bankside

- 120 deciduous trees partially shade the riparian margins during summer, while the relatively open
- 121 canopy allows high irradiance at the river surface and development of phytoplankton and benthic
- 122 algae (Huang et al., 2022).



Figure 1. (a) The Lower Bode river reach and the multi-parameter high-frequency monitoring scheme. (b) 15-min measurements of discharge (Q),  $NO_3^- - N$  concentrations, dissolved oxygen (DO) and water temperature (WT) at the upper Groß Germersleben (GGL) and lower Staßfurt (STF) stations.

- 128 At each station, YSI 610 sensors (Yellow Springs, US) measured dissolved oxygen (DO), pH,
- 129 water temperature (WT), and electric conductivity (EC) at 15-min intervals. ProPS-UV sensors
- 130 (TriOS GmbH, Germany) measured spectral absorbance at 254 nm (SAC, as a proxy of
- 131 dissolved organic carbon) and Nitrate-N ( $NO_3^- N$ , precision of 0.03 ±2%  $mgl^{-1}$ )
- 132 concentrations, also at 15-min intervals. More details about sensor accuracy and maintenance
- 133 were described in Rode et al. (2016a). Hourly data on photosynthetically active radiation (PAR)
- 134 were collected at an eddy-covariance flux tower at Wulferstedt station (15.6 km north-west of
- 135 GGL). Discharge (Q) data at 15-min intervals were obtained from the Saxony-Anhalt water
- 136 authority (Water Service data portal <u>https://gld.lhw-sachsen-anhalt.de/</u>, accessed 4/27/2023) at
- 137 station Hadmersleben (HAD, 2.7 km upstream of GGL without significant lateral inflows) and

- 138 station STF (Figure 1a). There are eight small tributaries along the study reach; however, total
- tributary inputs contributed less than 2% of total Q over 2015-2020 as simulated by the
- 140 catchment mHM-Nitrate model (Zhou, et. al., 2022).
- 141 The study region has experienced dramatically changing climatic conditions, especially severe
- summer droughts since 2018 (Bakke et al., 2020). Annual average Q at STF decreased from 9.50
- 143  $m^3 s^{-1}$  during the "normal" years (2015-2017) to 7.21  $m^3 s^{-1}$  during the "drought" years (2018-
- 144 2020) (Figure 1b). Comparing normal- and drought-year summers (Jun-August), average Q
- 145 decreased from 6.58 to 2.75  $m^3 s^{-1}$ , average  $NO_3^- N$  concentration decreased from 1.79 to
- 146 1.39  $mgl^{-1}$ , and average WT increased from 20.1 to 21.6 °C.

# 147 2.2 Two-station inferences of NO<sub>3</sub><sup>-</sup> retention based on the fusion of high frequency data 148 monitoring and the WASP water and tracer simulations

149 The WASP model (version 8.32) has been developed by the United States Environmental 150 Protection Agency and has been applied to different surface water systems (Wool et al., 2020). 151 The model allows users to flexibly define and arrange control volumes (i.e., fully mixed finite 152 segments), and resolves the mass balance over each control volume, considering material 153 exchanges via advective and dispersive transports. WASP further distinguishes free-flowing, 154 ponded and backwater/tidal influenced segments in a stream network, and solves them using 155 equations of kinematic wave, weir overflow and dynamic flow, respectively (Ambrose and 156 Wool, 2017). The tracer module implemented in WASP8 is a dummy sub-model for substances 157 with no kinetic interactions.

158 In this study, we conducted 15-min interval, one-dimensional WASP8 simulations of water and tracer (i.e., conservative  $NO_3^- - N$ ) transport in the study reach. We upgraded a prior WASP 159 160 model setup in the study reach by Huang et al. (2022), with 31 segments defined as free-flowing 161 reaches and 3 segments as ponded reaches at the locations of the weirs (Figure 1a). Discharge at station HAD was used as the upstream flow inputs, and the measured  $NO_3^- - N$  concentrations at 162 163 GGL were taken as the tracer inputs. Inputs of flow and tracer loads from the tributaries were 164 simulated from a catchment mechanistic model mHM-Nitrate (Yang et al., 2018; Yang and 165 Rode, 2020) (see detailed model simulations in the Bode catchment in Zhou et al. (2022)). Using the WASP hydrodynamic simulations, the upstream tracing signals of conservative  $NO_3^- - N$ 166 concentration  $(N_{trck})$  were transported to the downstream station STF. Thus, differences 167

between the tracked loading ( $L_{trck}$ ) predicted by the conservative WASP modeling and the observed loading ( $L_{obs}$ ) at STF were taken as retention along the river reach. Areal retention rates ( $U_T$ ,  $mgNm^{-2}$  per time step) and efficiencies ( $E_{UT}$ , %) can be further obtained based on this data-model fusion inferences:

172 
$$U_T = \frac{L_{trck} - L_{obs}}{A} = \frac{(N_{trck} - N_{obs}) \times Q_{STF}}{A}$$
(1)

$$E_{UT} = \frac{L_{trck} - L_{obs}}{L_{trck}} \times 100 \tag{2}$$

where *A* denotes the total river benthic area (27400×20 m<sup>2</sup>). Note that we used observed discharge at STF ( $Q_{STF}$ ) for  $L_{trck}$  to further reduce model uncertainty leveraged from hydraulic simulations (see Section 3). Therefore,  $U_T$  represents overall  $NO_3^-$  net uptake or net release (with

177 positive and negative values, respectively), and  $E_{UT}$  was only analyzed in periods exhibiting net 178 uptake ( $U_T$ >0).

## 179 **2.3 Calculations of stream metabolism and in-stream autotrophic** *NO*<sup>-</sup><sub>3</sub> assimilation

180 Estimates of whole-stream metabolisms (i.e., gross primary production - GPP and ecosystem

181 respiration - ER) were calculated based on the 15-min DO measurements. The hydraulic

182 characteristics at STF were heavily impacted by impoundment effects due to bridge piers and

submerged macrophytes, and this hindered the calculations of stream metabolism at STF.

184 Therefore, we applied the single-station method (Odum, 1956) using the data at the upper station

- 185 GGL, which is representative of the metabolic conditions of the Lower Bode reaches (see
- 186 detailed justifications in **Text S1** and **Table S1**). Please also refer to Yang et al. (2019) for

187 detailed calculation equations of the single-station approach.

188 We further estimated  $NO_3^- - N$  uptake via autotrophic assimilation based on the whole-stream 189 metabolism at daily time step. Gross assimilations by autotrophs  $(U_A, mgNm^{-2}d^{-1})$  can be 190 calculated based on the measured stream metabolism and the stoichiometric C:N ratio:

191 
$$U_A = \frac{r_a \times GPP}{2.286 \times C:N} \times 1000$$
(3)

192 where autotrophic respiratory rate  $r_a$  was set as 0.5, assuming one mol of C fixed per mol of  $O_2$ 

193 produced, and half of GPP  $(gO_2m^{-2}d^{-1})$  becomes net production (Rode et al., 2016a). The

- 194 molar C:N ratio was 9.4, taken from local biofilm measurements (Kamjunke et al., 2015), and
- the constant 2.286 (32/14) converts the molar ratios to a mass basis.

## 196 **2.4 Diel pattern detection and statistical analysis**

- 197 The 15-min continuous  $NO_3^- N$  retention timeseries allowed assessment of in-stream processes
- 198 from sub-daily (i.e., the diel patterns) to inter-annual scales. At hourly scale, we specifically
- 199 investigated diel patterns of net uptake ( $U_T > 0$ ,  $mgNm^{-2}h^{-1}$ ). Dates for this diel pattern analysis
- 200 were detected sequentially according to (1)  $U_T > 0$  for all 24 hours, (2) minimal diel variations in
- 201 Q (i.e., <50<sup>th</sup> quantile of coefficient of variations), and (3) generally low-flow conditions (i.e.,
- 202 daily  $Q < 25^{th}$  quantile for each year). The constraints of relative steady Q were set to avoid
- 203 impacts of flow variations on  $NO_3^- N$  signals.
- 204 The *k*-mean clustering method was performed to further classify them into different clusters,
- using the "stat" R package (R core team, 2022). Please refer to **Text S2** for details of the k-mean
- 206 clustering analysis. Moreover, statistical analyses were all performed in R software, including
- 207 the one-way ANOVA test (*"stats"* package, with assumptions fulfilled when reporting *p* values)
- and the curve fitting (the GAM-based smooth function "geom\_smooth" in "ggplot2" package).
- 209 **3 Results**

## 210 **3.1 WASP simulations of discharge and traced** $NO_3^-$ dynamics

- 211 The WASP discharge simulations performed very well at the downstream station STF (i.e.,
- 212 Kling-Gupta Efficiency of 0.98 and percentage bias of 0.7%, **Figure 2**). This indicates that
- 213 WASP captured water dynamics well in the lower Bode River, in line with Huang et al. (2022).
- 214 We noted that discharge discrepancies were slightly greater during the recession and driest
- 215 periods of the drought years 2018-2020, though they were negligible in terms of the general river
- 216 water dynamics (Figure 2a).



Figure 2. The WASP simulations of discharge (a) and  $NO_3^- - N$  tracer concentrations (b) in the reach of the lower Bode River plotted at a daily timestep. Discharge performance criteria KGE and PBIAS represent Kling-Gupta Efficiency and Percentage bias, respectively.

# 3.2 Daily overall $U_T$ and stream metabolisms and their seasonal variations in the normal and drought years

Continuous daily retention rates  $U_T$  were estimated from the data-model fusion from 2015 to 223 2020 (Figure 3a).  $U_T$  mostly ranged between -337.8 and 419.4  $mgNm^{-2}d^{-1}$  (i.e., 5% and 95% 224 225 quantiles, respectively), while the values varied with season and across years. Apart from 226 disturbances of annual high-flows and short-term flow events,  $U_T$  exhibited a consistent seasonal 227 pattern in the normal period 2015-2017 (Figure 3a, upper panels): (1) winter and early spring seasons exhibited extensive net  $NO_3^-$  release ( $U_T < 0$ ), though largely affected by the annual high 228 229 flows; (2) general net uptake  $(U_T>0)$  occurred after the annual high-flow season, with 230 significantly higher retention in May-June than July-September (236.0±224.9 vs 104.7±65.2  $mgNm^{-2}d^{-1}$ , among 123 and 230 days, respectively, with Q below the normal-year median 231 value 7.10  $m^3 s^{-1}$ ; ANOVA test, p < 0.001); (3) after October, net  $NO_3^-$  release occurred until the 232 233 next early spring (except for the low winter Q in 2016). For the drought-impacted period 2018-234 2020 (Figure 3a, lower panels),  $U_T$  seemed to be still higher during May-June than the later 235 months, while the pattern was heavily disturbed by the annual-high flow recessions and small- to

- median-size flow events. Moreover,  $U_T$  was generally reduced during the annual low-flow seasons (e.g.,  $47.9\pm57.2 \ mgNm^{-2}d^{-1}$  in July-September, significantly lower than that of the normal years; ANOVA test, p<0.001), and the late-autumn net release during the drought years was not as consistently present as it was during the normal years. The uptake efficiency  $E_{UT}$  was higher during the drought than in normal years (10.8±8.1% vs 8.6±9.5%, respectively), and the
- drought years exhibited prolonged periods with higher  $E_{UT}$  (e.g., around 25%, Figure S1a).



24.

Figure 3. (a) Daily total retention rate  $U_T$  in each year based on the model-data fusion framework, and monthly counts of the four clusters of diel patterns. (b) Daily ratios of wholestream gross primary production and respiration (i.e., P/R ratio) in the normal (2015-2017) and the drought (2018-2020) years. Note that, for subplot (b), P/R ratios on dates exhibiting diel patterns are highlighted with the same cluster colors as in subplot (a).

248 Such varying seasonal patterns between the normal and drought years were also observed in the 249 whole-stream metabolic characteristics (Figures 3b and S2). In spring seasons, the ratio of GPP 250 to ER (P/R) was >1, indicating autotroph-dominated system. This occurred in both the normal 251 and drought years (Figure 3b), though in the former it was likely driven by relatively high levels 252 of GPP, while in the latter by relatively low ER (Figure S2). During the summer seasons, the 253 P/R ratios varied largely among the normal years, often with similar levels of GPP but more 254 variable ER (Figure S2). In contrast, the ratios were more consistent and relatively high in the 255 drought years (0.77±0.20). In autumn and winter, P/R ratios were extensively less than 1,

- indicating a heterotroph-dominated system for both normal and drought years. Comparatively,
- the drought years exhibited higher P/R ratios (Figure 3b) primarily induced by the significantly
- lower ERs (i.e., 2.85±1.04 vs 3.65±2.27  $gO_2m^{-2}d^{-1}$  over October-February of the drought and
- normal years, respectively; ANOVA test, p < 0.001, Figure S2). The GPP-informed autotrophic
- 260  $NO_3^-$  uptake ( $U_A$ ) maintained similar in the normal and drought years (i.e., 47.92±42.76 vs
- 261 51.04±42.38  $mgNm^{-2}d^{-1}$ , respectively, **Figure S1a**), while its proportions to  $U_T$  were more
- 262 consistently >1 during July-October of the drought years (calculated among dates with  $U_T$ >0 and
- 263 below-median Q, Figure S1b).

# **3.3 Diel patterns of net** $NO_3^- - N$ uptake and their predominance under the normal and drought years

266 The high-frequency data-model fusion inference provided the unique opportunity to investigate sub-daily patterns of  $NO_3^- - N$  uptake and their variations under different hydrological 267 268 conditions. Quantifiable diurnal variation in  $U_T$  was detected in a total of 178 days (90 and 88 269 days in the normal and drought years, respectively), and could be further grouped into four 270 distinct clusters (Figures 4 and 3, and detailed results of the *k*-mean clustering in Figure S3). 271 Clusters C1 and C2 exhibited increased uptake during the daytime hours, with the diel maxima 272 occur after and before 12:00, respectively (Figure 4a and b); Clusters C3 and C4 exhibited 273 decreased uptake during the daytime hours, with the diel minima occur around 12:00 for the 274 former (Figure 4c) while the latter exhibited more noteworthy decreases before noon and 275 delayed diel minima (Figure 4d). Meanwhile, the seasonal occurrence of the four clusters likely 276 changed between the normal and drought years (the upper and lower panels of **Figure 3a**, 277 respectively). Specifically, in the normal years, C1 and C2 mostly occurred before August, while 278 C3 and C4 dominated the late summer and autumn seasons; in contrast, during drought years C1 279 and C2 patterns persisted into the late summer and autumn (accounting for 67% of total days 280 detected in the drought years, compared to that of 38% in the normal years), and also more 281 evenly distributed across summer-autumn seasons. The four diel patterns did not show 282 substantial differences between normal and drought years, with the exception of the C1 cluster 283 that exhibited more delayed diel maxima in the normal years than the drought years, (Figure 4, 284 upper panels). However, the absolute uptake rates were significantly lower under the drought

conditions, except for the C3 cluster even though there existed extraordinarily high uptake days
in 2017 (Figure 4, lower panels).



Figure 4. The four types of  $U_T$  diel patterns clustered among the detected 178 days and the corresponding hourly variations of retention rates (upper and lower panels, respectively). Note that the *k*-mean clustering was performed based on the max-min normalized  $U_T$  values.

### 291 4 Discussion

# 4.1 Continuous inferences of $NO_3^- - N$ retention in complex rivers enabled by a novel highfrequency data-model fusion framework

294 Using a novel data-model fusion framework, we derived continuous estimates of high-frequency  $NO_3^- - N$  retention over six years in the 6<sup>th</sup> order reach of the lower Bode River. To our 295 296 knowledge, this is among the first time that such in-stream inferences have been made in a large 297 stream or river with complex flow dynamics as well as across different hydrological conditions. 298 Importantly, the model-data fusion framework provided reasonable estimates of in-stream  $NO_3^- - N$  retention in the well-studied lower Bode region, as cross validated by previous studies 299 300 using different methodologies. For example, previous work (Zhang et al., 2022) using the 301 conventional two-station method reported a similar range and seasonal variations in net  $NO_3^-$  –

302 *N* retention and release (-205 and 381  $mgNm^{-2}d^{-1}$  as for 5<sup>th</sup> and 95<sup>th</sup> quantiles, respectively,

303 with consistently higher values in campaigns during May-June than July-September). Huang et

al. (2022) applied the WASP biogeochemical modeling in the same lower Bode River reach and

305 calibrated model parameters using the high-frequency data (including only the first drought year

306 2018). They also obtained generally similar levels of net  $NO_3^- - N$  uptake, and similar seasonal

307 patterns (Huang et al., 2022).

308 Our estimates of in-stream retention rate and retention efficiency were also well in line with

309 wider literature ranges (Alexander et al., 2009; Ensign and Doyle, 2006; Miller et al., 2016;

310 Mulholland et al., 2008; Seitzinger et al., 2002). Ensign and Doyle (2006) reported the

311 interquartile range of 9.1 and 376.7  $mgNm^{-2}d^{-1}$  based on 14 nutrient-addition/isotope-tracer

312 experiments conducted in 4<sup>th</sup> order streams. This fits well with our estimated range of the  $NO_3^-$ 

313 uptake (**Figure 3a**), which is also similar to the headwater measurements from the pioneering

LINX II project (Hall Jr. et al., 2009). Heffernan and Cohen (2010) deployed a high-frequency

315 monitoring in a Florida spring-fed river and revealed that  $U_T$  is two times higher in spring than in

fall seasons, though the reported magnitude was relatively high due to high denitrification rates

in subtropical rivers (Heffernan et al., 2010). Using  $E_{UT}$  as a more cross-comparable uptake

318 metric, Seitzinger et al. (2002) found that individual reaches generally retain <20% of N input,

and specifically for  $5^{\text{th}}$ - and higher-order reaches, the proportions are mostly <10%. This is well

320 in line with our estimated ranges (e.g., mostly < 20.8% as of the 90% quantile value over the six

321 years).

322 The high-frequency data-model fusion framework allows assessment of in-stream processes to

be extended across temporal and spatial scales, thereby advancing understanding of  $NO_3^-$ 

324 processes. First, the framework can make use of the increasingly available long-term time series

325 of high-frequency monitoring, being collected as part of routine monitoring by water authorities.

326 Given the logistic convenience and advances of cross-parameter analysis (Rode et al., 2016b;

327 van Geer et al., 2016), these continuous sensor deployments are often co-located with flow

328 gauging stations, allowing assessment of varying and complex flow regimes and their impacts on

329 biogeochemical processes (Bieroza et al., 2023; Oldham et al., 2013). Integrating the robust in-

330 stream hydraulic and conservative tracer modeling facilitates the extraction of biogeochemically

induced nutrient signals, even with shifts between advective and dispersive flow dominance and

variations in water residence/travel times (Hensley and Cohen, 2016). This, in our view, serves
as a step forward in maximizing the multi-benefits of these unique datasets, especially compared
to the uses of constraining synthesized in-stream water quality models at a finer temporal
resolution (though still informative, see Huang et al., (2022); Khorashadi Zadeh et al., (2019)).

336 Second, the framework quantifies in-stream  $NO_3^-$  retention from sub-daily to inter-annual scales,

providing new process understanding (see discussion in Section 4.2) and reference values for

338 specific processes. For instance, the WASP modeling by Huang et al., (2022) also revealed the

339 consistent pattern of net release after late-autumn (i.e.,  $U_T < 0$ ), while with more conservative

340 estimates (17.4  $mgNm^{-2}d^{-1}$  compared to our inferences of >100  $mgNm^{-2}d^{-1}$ , Figure 3a).

341 Given the well-acknowledged poor process understanding (von Schiller et al., 2015) and the

342 simplified model conceptualization, our inferences derived more directly from high-frequency

343 data, can be further used to validate model simulations and improve in-stream process

344 representation in water quality models.

345 There are several considerations when transferring the framework to a wider context. Firstly, the 346 estimation uncertainty may be large during high-flow seasons and short-term runoff events, 347 primarily induced by contributions from tributaries and lateral subsurface inflows. These lateral 348 inputs are often ungauged and elevated during hydrologically active periods. Particularly for the 349 lower Bode region, the high groundwater concentrations (up to > 10  $mgNl^{-1}$ , the authority data 350 portal https://gld.lhw-sachsen-anhalt.de/) in the agricultural landscape might strongly affect 351 riverine concentrations even under small runoff events (as illustrated by the discharge-352 concentration hystereses in Zhang et al. (2020)). However, we cautiously estimated tributary 353 inputs using the grid-based catchment modeling (Zhou et al., 2022), and found that the lateral 354 loading inputs were marginal, as this was generally a loosing channel (Kunz et al., 2017; Zhang 355 et al., 2022). Moreover, hyporheic exchange might need to be explicitly included in the hydraulic 356 and tracer simulations, where the process is known to be significant and sufficient data are 357 available to constrain it (Gomez-Velez et al., 2015).

#### 4.2 Insights into seasonal patterns of $NO_3^-$ retention and release under the changing climate

Both net  $NO_3^-$  uptake ( $U_T > 0$ ) and net  $NO_3^-$  release ( $U_T < 0$ ) exhibited substantial seasonality,

360 which also differed between the normal (2015-2017) and drought (2018-2020) years. Moreover,

this variation was well supported by the independent calculations of stream metabolic
characteristics, which also exhibited strong seasonality (Figure 3 and S2).

363 Despite the flow-induced uncertainty during the high-flow periods, winter to early-spring 364 seasons showed extensive net  $NO_3^-$  release, which was more pronounced in the normal years 365 than the drought years. Meanwhile, the normal-year  $P/R \ll 1$  and ER remained as high as that in 366 the summer seasons, indicating higher heterotrophic microbiological activity during the normal 367 year. Such net heterotrophy likely also reflects reduced rates of autotrophic activity (due to 368 seasonally lower light availability and water temperature) and promoted remineralization 369 processes (given that riverine dissolved organic carbon (DOC) is abundantly flushed from the 370 catchment and riparian areas). The DOC in anthropogenically affected rivers is likely more labile 371 to the microbial community than in unaffected rivers (Graeber et al., 2012; Stutter et al., 2018). 372 Meanwhile the relatively high water temperature (i.e., interquartile range of 3.4 and 7.9 °C) may 373 have contributed to higher rates of microbiological transformation of DOC (Lu et al., 2013). 374 However, under sequential droughts, the recalcitrance of DOC could have been increased due to 375 the longer exposure times of terrestrial organic matter to catchment microbiological 376 transformations before being transported (Catalán et al., 2016) and that the enzymatic hydrolysis 377 of the recalcitrant components is limited with increasing temperature (Evans et al., 2005; 378 Freeman et al., 2001). This agrees well with the observed low ERs (Figure S2) but with largely 379 maintained DOC concentrations during the drought years (as indicated by the surrogate SAC 380 measurements, although direct DOC quality information was missing). Also, the reduced levels 381 of net  $NO_3^-$  release during winter and early-spring seasons may be ascribed to the drought-382 induced reduction of heterotrophic activity.

383 The extensive net uptake patterns during the mid-spring to early-autumn warmer seasons are 384 well in line with general literature findings that highlight the significant nutrient "cleansing" 385 function" of lotic ecosystems, while the dominant mechanisms could vary largely under different 386 stream and climatic conditions (Heffernan and Cohen, 2010; Jarvie et al., 2018; Jones et al., 387 2015). The net uptake reached the highest level between mid-spring and early-summer, 388 associated with the promoted and autotrophy-dominated ecosystem activities (the high P/R ratios 389 and increased GPP and ER, Figures 3 and S2). The normal-year overall net uptake was still 390 greater than the autotrophic assimilation, although with increasing amount and proportions for

the latter pathway. This indicates that (1) autotrophic assimilation played a dominant role in

- $NO_3^- N$  uptake, primarily due to the sufficient light availability during the season before leaf-
- 393 out (Rode et al., 2016a; Yang et al., 2019) and (2) other uptake pathways via heterotrophic direct
- 394 assimilation and denitrification were likely substantial in the normal years, given the abundant
- nutrient and organic matter availability (Kamjunke et al., 2013). In contrast, the drought years
- 396 exhibited relatively reduced overall net uptake and ER, but a similar level of GPP, together
- resulting in earlier, more consistent  $U_A/U_T > 1$  (Figure S1b). This indicates an earlier
- 398 predominance of autotrophic assimilation, which is also buffered partly by N-release processes
- 399 (Jarvie et al., 2018). The reduction in other heterotrophic pathways is likely ascribed to the
- 400 limited allochthonous labile DOC during the droughts, given the synchronous changes between
- 401 ER and  $NO_3^- N$  uptake (Stutter et al., 2018; Sunjidmaa et al., 2022).

402 Nevertheless, autochthonous DOC from antecedent primary production may have provided an 403 important energy source for heterotrophs in the lower Bode River during the driest summer and 404 autumn seasons, in accordance with reduced catchment hydrological connectivity. This is 405 evidenced by the increasingly buffered autotrophic assimilation  $(U_A/U_T \gg 1)$  and generally decreased net  $NO_3^-$  uptake, both of which exhibited higher degrees of changes in the drought 406 407 years. This is supported by DOC measurements across the Bode River network by Kamjunke et 408 al. (2013), that the September samples from the lower Bode River exhibited better DOC quality 409 (i.e., lower molecular weights and lower humic content) than the headwater samples. Also, 410 Dupas et al. (2017) demonstrated that the Bode lowland aquatic systems likely acted as a DOC 411 source. There is a pressing and strategic need for better mechanistic understanding of the 412 responses of river ecosystems to climate variability. Our results suggest that reduced inputs of 413 allochthonous labile DOC during prolonged droughts may have resulted in shifts to greater and 414 more sustained utilization of autochthonous DOC by microbes, corresponding with reductions in 415 both net  $NO_3^-$  retention and net release.

## 416 **4.3 Insights into diel patterns of** *NO*<sup>-</sup><sub>3</sub> **retention under the changing climate**

- 417 The high-frequency data-model fusion framework also allowed in-depth investigations of diel
- 418 patterns of  $NO_3^- N$  uptake. Here we identified four distinct clusters of diel patterns, as well as
- 419 their seasonal distributions, revealing shifts in dominant  $NO_3^-$  retention processes under normal

and drought conditions. We can speculate as to some of the underlying mechanisms potentiallyresponsible for generating these patterns.

422 Cluster C1 represented a typical autotrophic assimilation induced diel pattern, which is 423 correlated with the diurnal variations of GPP and related environmental factors (e.g., radiation, 424 temperature and DO; Figures 4a, S4a and S5). Interestingly, the diel uptake maxima were 425 delayed relative to those of GPP, indicating physiological time lags between  $NO_3^- - N$  uptake 426 and photosynthesis. This is potentially due to photosynthesis providing additional energy to 427 further reduce  $NO_3^-$  for assimilation and biosynthesis (Mulholland et al., 2006). The observed 428 time-lag differences between normal and drought years are in line with the spring-fall differences 429 reported by Heffernan and Cohen (2010), which may be related to seasonal differences of such

430 energetic costs.

431 Cluster C2 also largely represented the dominance of autotrophic assimilation, given the

432 extensive positive correlations with GPP-related factors (Figure S4) and their well-fitted falling

433 limbs after 12:00 (Figure S5b). Besides, as a concatenation of Cluster 1 and Clusters 3/4, NO<sub>3</sub><sup>-</sup>

434 autotrophic assimilation in the afternoon might be buffered to a greater extent by release

435 processes, resulting in apparently earlier net uptake maxima. These diurnally earlier  $NO_3^-$  uptake

436 peaks are also reported in high-frequency  $NO_3^-$  data analysis, with complex, often unclear

437 mechanisms (Aubert and Breuer, 2016; Greiwe et al., 2021; Heffernan and Cohen, 2010). Here

438 we observed that (1) this pattern occurred mostly after high-flow recessions in the normal years

439 (Figure 3a), which is coincidently presented by Aubert and Breuer (2016) via data mining of

440  $NO_3^-$  timeseries; (2) the high correlations with SAC in drought years at the diurnally normalized

scale (**Figures S4 and S5b**). This indicates a crucial role of DOC availability, particularly its

442 quality as aforementioned, in shaping (either promoting or constraining) the diel pattern of  $NO_3^-$ 443 uptake.

444 Clusters C3 and C4 revealed less-explored diel patterns with diel minima uptake during the

445 daytime. For C3, diel minima occurred nearly at mid-day hours, the timing of GPP and

446 temperature maxima (**Figure S5c**); meanwhile, the  $NO_3^-$  uptake ( $U_T$  and the  $U_A$  pathway) and

447 metabolisms (GPP, ER and their ratios) were similar to Clusters C1 and C2, indicating high

448 daytime photosynthesis. The exact mechanisms would need further specific investigations, while

449 possible explanations could be: (1) this pattern occurred mostly during late summer and autumn,

- 450 when diel variations of denitrification rates govern the net  $NO_3^-$  uptake patterns; (2) redox
- 451 controls (linked to high daytime O<sub>2</sub> production by photosynthesis) will likely suppress
- 452 denitrification, resulting in diel minima during daytime hours. The C4 pattern was highly
- 453 comparable with C3 (see **Figure 4c-d**, and the clustering results in **Figure S3**), while likely
- 454 exhibited higher rates of heterotrophic-related uptake like denitrification (linked to lower P/R
- 455 ratios Figure 3b, and significantly positive correlations with SAC Figure S4).
- 456 In addition to specific mechanistic understanding of the four cluster types, their seasonal 457 distributions and changes between normal and drought years (Figure 3a, bar-plots), also reveal 458 shifts in river ecosystem function. In the normal years, the autotrophy-characterized C1 and C2 459 patters mostly occur during earlier seasons, which are then followed by a shift to heterotrophy-460 dominated C3 and C4 patterns during later summer-autumn low-flow periods. However, the 461 sequential droughts have shifted this seasonal distribution to greater persistence of C1 and C2 462 across seasons, extending the window of autotrophic dominance into the late summer and 463 autumn. Such varying seasonal distributions of diel patterns are likely linked to the drought-464 induced changes in catchment and stream environmental factors and their contrasting impacts on 465 in-stream  $NO_3^-$  biogeochemical processes. This also corresponds with the seasonal insights derived based on the daily timeseries (Section 4.2) that showed a shift to greater persistence of 466 467 net autotrophy during the drought years.

#### 468 **5 Conclusion**

469 This study proposed a data-model fusion framework that enables continuous inferences of in-

- 470 stream  $NO_3^-$  biogeochemical processes in large streams and rivers, overcoming major
- 471 methodological constraints of using high-frequency monitoring data under variable flow and
- 472 climatic conditions. Long-term high-frequency estimates in the 6<sup>th</sup>-order lower Bode River

473 further revealed new insights into  $NO_3^-$  dynamics in relation to river ecosystem function, from

- 474 sub-daily to seasonal scales and under a range of flow conditions. These estimates of net in-
- 475 stream retention and release, derived closely from direct *in-situ* measurements, provide useful
- 476 reference values, e.g., for model validation or process conceptualization. Also, identifying and
- 477 deconvoluting shifts in river ecosystem function (between autotrophy- and heterotrophy-
- 478 dominance) are of key importance for understanding the impacts of climate change on water
- 479 quality, ecological status and river function.

480 This high-frequency data-model fusion method can be applied across sites and scales, and

- 481 provides an opportunity for scientists and river managers to capitalize on the emergent wave of
- 482 high-frequency water-quality monitoring. The method can also be extended from small-streams
- 483 under dynamic flow conditions, to river network scale applications coupled with catchment
- 484 models. Moreover, the method can complement high-frequency monitoring for cost-effective
- 485 evaluation of the effectiveness of management practices and river-restoration projects.
- 486

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- 494

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