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2 **Contrasting Roles of Microbial Heterogeneity and Sediment Heterogeneity in**

3 **Controlling Hyporheic Nitrogen Removal**

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17



18 **Abstract**

19 Bacteria in streambed sediments reduce nitrogen pollution by performing chemical  
20 reactions that can transform pollutants. However, how differences in sediment types (like sandy  
21 vs. clay-rich layers) or the distribution of these bacteria affect the nitrogen removal in streambed  
22 sediments remains unclear. To explore this, we built numerical models to mimic real streambed  
23 conditions and explore how small-scale variations in sediment and microbial communities  
24 influence nitrogen removal. Our results show that the spatial changes in physical and chemical  
25 properties of sediments (e.g., permeability and organic matter content) have little impact on  
26 nitrogen removal. Instead, variations in the distribution of nitrogen-processing microbes are far  
27 more critical. This is because bacteria and the sticky substances they produce can slowly clog the  
28 sediment pores. This clogging limits the movement of pollution to the very top layers of  
29 sediments, making deeper sediment differences irrelevant to overall nitrogen removal. These  
30 results mean that future numerical models aiming to predict nitrogen removal should focus on  
31 mapping where these helpful microbes live, rather than overcomplicating simulations with too  
32 much detail about sediment variations. Simplifying sediment assumptions could make models  
33 more practical while still capturing the key role of microbes in cleaning streams.

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## 36 **1 Introduction**

37 High concentrations of nitrate ( $\text{NO}_3^-$ ) in stream ecosystems poses severe risks to water  
38 quality, impacting both human and ecosystem health (Grant et al., 2018; Gruber & Galloway,  
39 2008). Hyporheic zones (HZs) within streambed sediments are increasingly recognized as  
40 essential biogeochemical hotspots for nitrate attenuation (Boano et al., 2014; Harvey & Gooseff,  
41 2015; Krause et al., 2017, 2022). The high surface area of streambed sediments and nutrient  
42 exchange fluxes across HZs create ideal conditions for microbial communities to flourish,  
43 facilitating diverse biogeochemical reactions that markedly exceed those in the overlying water  
44 column (Chowdhury et al., 2020; Gu et al., 2007). The efficiency of nitrate removal within these  
45 streambed sediments is theoretically governed by solute residence time distributions and  
46 spatially heterogeneous mixing of solutes from different water sources (Krause et al., 2022), and  
47 biogeochemical reaction rates (Zarnetske et al., 2012; Xian et al., 2023). Metabolically relevant  
48 residence time is influenced by the hyporheic flow field, which is intrinsically linked to sediment  
49 permeability, while biogeochemical reaction rates are shaped by many factors such as  $\text{NO}_3^-$   
50 concentration, dissolved oxygen (DO) levels, dissolved organic carbon (DOC) availability, and  
51 microbial activity (Gomez-Velez et al., 2015; Stegen et al., 2018). Denitrification, the primary  
52 biogeochemical process that converts  $\text{NO}_3^-$  to nitrogen gases ( $\text{N}_2$ ,  $\text{N}_2\text{O}$ ), is primarily regulated by  
53 thermodynamic conditions, being inhibited by the presence of oxygen but stimulated by  
54 bioavailable DOC (Martínez-Espinosa et al., 2021; Zarnetske et al., 2011). Denitrification based  
55 nitrate removal necessitates anoxic conditions fostered by prolonged residence times and ample  
56 DOC available to facilitate DO consumption (Sawyer, 2015).

57 Heterogeneous fluvial sediments, encompassing fine-grained (e.g., silt and clay) and  
58 coarse-grained (e.g., gravel and sand) facies, are prevalent in riverine systems (Cardenas, 2004;



59 Sawyer & Cardenas, 2009; Gomez-Velez et al., 2014). It is well-established that this sedimentary  
60 heterogeneity generates substantial spatial variability in both physical (e.g., permeability) and  
61 chemical (e.g., organic matter (OM) content) properties (Briggs et al., 2015; Chen et al., 2025;  
62 Pescimoro et al., 2019; Sawyer, 2015). Fine sediments exhibit organic rich sediments are often  
63 characterized by lower permeability than their coarse grain, low organic counterparts. As  
64 oxygen-rich stream water permeates the streambed sediments, the heterogeneity in physical and  
65 chemical properties promote the formation of anoxic microzones essential for denitrification  
66 (Claret et al., 1997; Knights et al., 2017; Krause et al., 2013). Oxic and anoxic waters converge  
67 at the localized interfaces within the heterogeneous sediments, fostering microsites of elevated  
68 denitrification activity (Sawyer, 2015; Xian et al., 2023; Zhang et al., 2026). Such microscale  
69 mixing enhances nitrate removal efficiency in heterogeneous sediments relative to homogeneous  
70 sediment systems. For example, Sawyer (2015) has indicated that nitrate removal rates in  
71 heterogeneous sediments may exceed those of comparable homogeneous sediments by as much  
72 as sixtyfold. Recent studies (e.g., Barnes et al., 2019; Pescimoro et al., 2019; Ping et al., 2022;  
73 Wallace et al., 2020, 2021; Chen et al., 2024) have further illustrated that mixing processes  
74 within heterogeneous sediments may significantly influence nutrient transformation zones and  
75 nitrogen removal efficacy within riparian and hyporheic zones (HZs).

76         Despite these advances, the mechanisms by which heterogeneous sediments affect nitrate  
77 attenuation in HZs remain not fully understood. Prior studies typically consider only  
78 permeability heterogeneity (e.g., Bardini et al., 2013) or the combined effects of permeability  
79 and OM content (e.g., Barnes et al., 2019; Pescimoro et al., 2019; Sawyer, 2015; Chen et al.,  
80 2024), assuming uniform functional biomass distribution across coarse and fine structures.  
81 However, beyond the physical and chemical heterogeneity of sediments, heterogeneous



82 hyporheic systems also host spatial variability of functional groups of microorganisms  
83 responsible for biogeochemical processes (De Falco et al., 2018). Hyporheic microbiota play  
84 multiple roles in nutrient cycling within HZs (Caruso et al., 2017; Chowdhury et al., 2020; Ping  
85 et al., 2020; Xian et al., 2022; Zhang et al., 2024; Zhang et al., 2026): First, microbial  
86 communities drive numerous biogeochemical transformations, determining the spatial  
87 distribution of reaction rates. Second, microbial biofilms that are aggregations of bacterial cells  
88 embedded in self-secreted extracellular polymeric substances occupy pore spaces on sediment  
89 surfaces, modulating permeability and thus, potentially altering local flow fields. Biofilm growth  
90 introduces complex feedback mechanisms with porous media flow and reactive transport.  
91 Hyporheic flow on one hand influences biofilm development, which in turn modifies nutrient  
92 advection rates. Through these interactions, microbial functional biomass distributions become  
93 heterogeneously organized in both space and time.

94         This study aims to elucidate the effects of microbial heterogeneity and heterogeneity in  
95 sedimentary physical properties (e.g., permeability) and chemical composition (e.g., OM  
96 content)) on nitrate removal in HZs. This will be achieved by establishing a series of numerical  
97 models coupling porous flow, nutrient transport, and biofilm growth in heterogeneous sediments,  
98 comparing results to the equivalent homogeneous counterparts. Our findings reveal that  
99 microbial heterogeneity exerts a decisive influence on enhanced nitrogen removal within  
100 hyporheic zones, while the effect of small-scale sedimentary heterogeneity appears negligible.  
101 These insights challenge the conventional view that sedimentary heterogeneity is the primary  
102 driver of enhanced nitrate removal, highlighting the critical role of microbial heterogeneity in  
103 biogeochemical processes.



## 104 **2 Porous Media Flow, Nutrient Dynamics and Biofilm Growth Models**

105 In this section, we develop a suite of two-dimensional (2-D) reactive transport models for  
106 describing nitrogen dynamics of HZs in heterogeneous streambed sediments. The models are  
107 established by coupling porous media flow, reactive transport of nutrients, biofilm growth and,  
108 most importantly, their feedback mechanisms. Specifically, the processes of porous media flow,  
109 heterotrophic metabolisms of carbon and nitrogen, microbial biomass growth, endogenous decay  
110 of biomass, and bioclogging are simulated for typical small-scale ripple-induced HZs in  
111 heterogeneous sediments.

### 112 2.1 Governing Equations

113 The porous media flow in streambed sediments is simulated with the hydraulic continuity  
114 equation and Darcy's law, given by (Caruso et al., 2017; Zhang et al., 2024):

$$115 \quad \frac{\partial \theta}{\partial t} = \nabla \cdot (K \nabla h) \quad (1)$$

116 where  $\theta$  is the porosity,  $t$  denotes time,  $K$  represents hydraulic conductivity, and  $h$  is the  
117 hydraulic head.

118 The reactive transport of nutrients in HZs is described with the advection-dispersion-  
119 reaction equation:

$$120 \quad \frac{\partial \theta C_i}{\partial t} = \nabla \cdot \theta D \nabla C_i - \nabla \cdot q C + R_i \quad (2)$$

121 where  $C_i$  denotes the solute concentration of solute  $i$ , and  $q$  is the Darcy's velocity,  $R_i$  represents  
122 the source/sink term,  $D$  denotes the hydrodynamic dispersion tensor and is expressed as (Sawyer,  
123 2015):



$$D_{ij} = \alpha_T |v| \delta_{ij} + (\alpha_L - \alpha_T) \frac{v_i v_j}{|v|} + \tau D_0 \delta_{ij} \quad (3)$$

where  $\alpha_L$  and  $\alpha_T$  are the longitudinal and transverse dispersivities, respectively;  $v$  denotes the pore velocity and equals  $q/\theta$ ;  $\tau$  is the tortuosity and is calculated as  $\theta^{1/3}$ ,  $D_0$  represents the molecular diffusion coefficient in water.

The key reactive solutes DOC, DO,  $\text{NO}_3^-$  are considered in the stream and the hyporheic system. The nutrients can be consumed by heterotrophic microorganisms via aerobic respiration and denitrification:



These microbially mediated biogeochemical reactions can be modeled by multiple-Monod kinetic equation (e.g., Gu et al., 2007; Molz et al., 1986). In sediments, the particulate organic carbon (POC) can release DOC by kinetic dissolution, which can be described by first-order kinetics (Gu et al., 2007; Zarnetske et al., 2012; Chen et al., 2023). Thus, the source/sink terms  $R_i$  in Eq. (2) for the species DOC, DO,  $\text{NO}_3^-$  are given as:

$$R_{DO} = -V_O X \frac{C_{DOC}}{K_{DOC} + C_{DOC}} \frac{C_{DO}}{K_{DO} + C_{DO}} \quad (4)$$

$$R_{\text{NO}_3} = -V_N X \frac{K_I}{K_I + C_{DO}} \frac{C_{DOC}}{K_{DOC} + C_{DOC}} \frac{C_{\text{NO}_3}}{K_{\text{NO}_3} + C_{\text{NO}_3}} \quad (5)$$

$$R_{DOC} = R_{DO} + 1.25R_{\text{NO}_3} + \rho_s \alpha_C (C_{POC} - K_d C_{DOC}) \quad (6)$$

where  $C_{DOC}$ ,  $C_{DO}$ ,  $C_{\text{NO}_3}$  and  $C_{POC}$  denote the concentration of DOC, DO,  $\text{NO}_3^-$ , and POC, respectively.  $V_O$ ,  $V_N$  are the maximum microbial uptake rate of DO and  $\text{NO}_3^-$ , respectively.  $X$



143 represents the microbial biomass per bulk volume of sediments.  $K_{DOC}$ ,  $K_{DO}$ ,  $K_{NO_3}$  are the half  
 144 saturation constants of DOC, DO,  $NO_3^-$ , respectively.  $K_I$  is the inhibition constant.  $\rho_s$  denotes the  
 145 bulk density of sediments.  $\alpha_C$  represents the first-order mass transfer coefficient between POC  
 146 and DOC.  $K_d$  denotes the distribution coefficient.

147 The microbial growth is simulated with the growth-death model expressed by (Caruso et  
 148 al., 2017; Ping et al., 2020; Xian et al., 2019):

$$149 \quad \frac{\partial X}{\partial t} = Y \left( 1 - \frac{X}{\rho_X \theta_0} \right) \left[ R_{DO} + 1.25 R_{NO_3} \right] - k_d X \quad (7)$$

150 where  $Y$  is the yield coefficient,  $k_d$  denotes the endogenous decay coefficient,  $\theta_0$  is the initial  
 151 porosity without microbial biomass,  $\rho_X$  represents the density of microbial biomass. The term  $1 -$   
 152  $X/\rho_X \theta_0$  is introduced into Eq. (7) to account for the microbial self-limitation effect due to the  
 153 limited pore spaces (Brovelli et al., 2009; Samsó et al., 2016).

154 With biomass growth, the porosity  $\theta$  changes with time and is the function of  $\theta_b$ :

$$155 \quad \theta = \theta_0 - \theta_b \quad (8)$$

156  $\theta_b$  represents the microbial biomass volume per volume of sediments, and can be calculated by  
 157  $X/\rho_X$ . The sediment hydraulic conductivity  $K$  also varies in response to biofilm-induced changes  
 158 in porosity. Here we adopted the biofilm-plug formation model (Vandevivere, 1995) to describe  
 159 the relationship between  $K$  and  $\theta$ :

$$160 \quad \frac{K(\theta_r)}{K_0} = F(\theta_r) \theta_r^2 + [1 - F(\theta_r)] \frac{K_{\min} / K_0}{[K_{\min} / K_0 + (1 - \theta_r)(1 - K_{\min} / K_0)]},$$

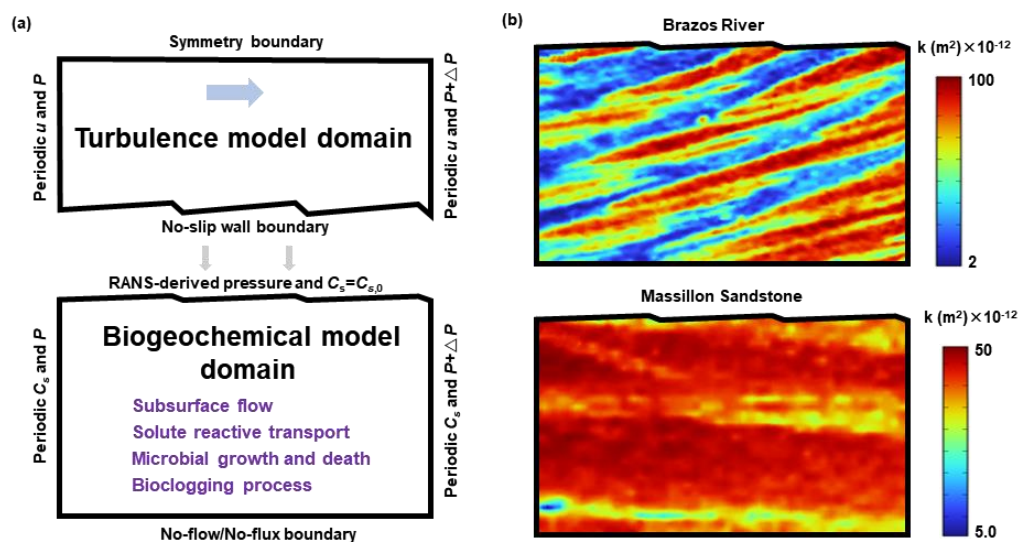
$$F(\theta_r) = \exp \left[ -0.5 \left( \frac{1 - \theta_r}{1 - \theta_{rc}} \right)^2 \right] \quad (9)$$



161 where  $\theta_r$  represents the relative porosity and equals to  $\theta/\theta_0$ ,  $K_{min}$  is the minimum hydraulic  
162 conductivity corresponding to  $\theta_r=0$  at which pore spaces are completely occupied by biomass.  
163  $\theta_{rc}$  denotes a critical value of  $\theta_r$ , and  $\theta_r=\theta_{rc}$  corresponds to the transition from the biofilm  
164 formations (phase I) to the plug formations (phase II) because of biofilm shearing in phase I then  
165 forming plug in phase II (Seifert and Engesgaard, 2012; Vandevivere, 1995).

## 166 2.2 Model Initial and Boundary Conditions, and Parameterisation and Scenario Simulations

167 The model simulations are setup for repeating ripple-induced HZs (Figure 1a). The  
168 water-sediment interface is assigned as a constant pressure boundary for porous media flow and  
169 an open boundary for solute transport in the porous media. The pressure profile along the water-  
170 sediment interface is derived through simulations of turbulent flow over the repeating ripples by  
171 numerically solving the Reynolds-averaged Navier-Stokes equations in steady state using the  $k$ -  
172  $\omega$  closure scheme (See Cardenas & Wilson (2007) for the further details of the model  
173 establishment and validation). The lateral sides for the solute transport are implemented as a  
174 spatially periodic boundary for both pressure and concentration, with a prescribed pressure drop  
175  $\Delta P$  that is consistent to the turbulence model. No flow and no flux are assigned to the bottom  
176 boundary. Two scenarios of stream nutrients, i.e., pristine ( $C_{s0}$ ; DOC=5 mg/L,  $\text{NO}_3^-$ =2 mg/L)  
177 and polluted ( $C_{s0}$ ; DOC=20 mg/L,  $\text{NO}_3^-$ =10 mg/L) and two stream flow velocity conditions  
178 ( $U=0.2$  m/s, 0.8 m/s) are considered. The numerical model parameters governing porous media  
179 flow, solute reactive transport, microbial dynamics (growth and death), and biofilm-induced  
180 bioclogging were calibrated using experimental data from the laboratory-scale flume study  
181 conducted by Quick et al. (2016). The detailed validation procedures and outcomes presented in  
182 Section 3.1, and the model parameters are summarized in Table 1.



183

184 Figure 1. Modeling scheme. (a) The simulated processes, boundary condition configurations for  
 185 the stream water turbulent flow (top part), and the solute reactive transport in streambeds  
 186 (bottom part). (b) Two realistic permeability field adopted in the present study.

187 The datasets of spatial distribution of initial permeability and OM content are  
 188 respectively created based on the images of the permeability field of the Massillon Sandstone  
 189 and the climbing ripple deposits of the Brazos River from the Figures 1a and 1c of Sawyer &  
 190 Cardenas (2009) (Figure 1b). The permeability field of the Massillon Sandstone indicates a  
 191 slightly sloping (dips from  $10^\circ$  to  $20^\circ$ ) and stratified heterogeneous sedimentary structure with a  
 192 variance of 0.15. For the Brazos River, the image of climbing ripple deposits shows sloping  
 193 bimodal deposits consisting of organic-scare sand and organic-rich silt with a variance of 1. We  
 194 converted the color images into greyscale images. For the Massillon Sandstone, for the greyscale  
 195 value of 1 and 0, the log-permeability values of -11.5 to -10.3 were respectively assigned, and  
 196 the OM content of 0.5% and 0.1% were respectively given. These are typical values for sandy  
 197 sediments (Sawyer, 2015). For the Brazos River, the log-permeability values of -10 to -12 and



198 the OM content of 0.2% and 2% are respectively given to the greyscale value of 1 and 0. Linear  
199 interpolation was applied to derive values between the upper and lower limits for the  
200 intermediate greyscale values. The two datasets represent low and moderate heterogeneity in  
201 permeability ( $K$ ) and OM content, respectively (Bardini et al., 2013; Sawyer & Cardenas, 2009).  
202 While the variances in  $\ln(K)$  and OM content are smaller compared to those reported in studies  
203 of large-scale heterogeneous fluvial systems (e.g., Genereaux et al., 2008; Kalbus et al., 2009),  
204 this study focuses on small-scale heterogeneity within fluvial bedforms associated with a single  
205 depositional facies, thereby isolating the effects of localized sediment variability.

206 The heterogeneous models with the microbial growth and the interactions between flow,  
207 nutrients and biofilm growth are called as *Hetero. Sed. & Bio. Growth* model. We also establish  
208 two sets of equivalent homogeneous model. In the first set, called as *Homo. Sed. & Bio. Growth*  
209 model, the equivalent permeability tensors of the heterogeneous sediments and the arithmetically  
210 weighted averages of OM content are used as the initial values. Microbial growth and the  
211 interactions between flow, nutrients and biofilms are included in the model. In the second set,  
212 called as *Hetero. Sed. & Const. Bio.* model, the datasets of heterogeneous permeability and OM  
213 distribution are adopted. Instead of simulating microbial growth processes, the arithmetically  
214 weighted average of the functional microbial biomass distribution of the *Hetero. Sed. & Bio.*  
215 *Growth* model throughout the entire simulated period is applied. The importance of sedimentary  
216 heterogeneity (*Hetero. Sed. & Bio. Growth* vs. *Homo. Sed. & Bio. Growth*) and microbial  
217 heterogeneity (*Hetero. Sed. & Bio. Growth* vs. *Hetero. Sed. & Const. Bio.*) can be isolated by the  
218 comparisons of these experiments. In *Homo. Sed. & Bio. Growth* model, the equivalent  
219 permeability tensors for the Massillon Sandstone  $k_M$  and the Brazos River  $k_B$  were calculated  
220 following the method of Durllofsky (1991):



221 
$$k_M = \begin{bmatrix} 8.36 & -0.0011 \\ -0.0011 & 8.25 \end{bmatrix} \times 10^{-11} \text{ m}^2$$

222 
$$k_B = \begin{bmatrix} 4.87 & 0.44 \\ 0.44 & 4.40 \end{bmatrix} \times 10^{-11} \text{ m}^2$$

223 Eqs. (1-14) are numerically solved by the finite-element software COMSOL Multiphysics. For  
 224 this, 2-D model domains are discretized using triangular elements. The triangular element size is  
 225 0.5 mm at the water-sediment interface and generally increase to 3.5 mm towards the bottom  
 226 boundary with a factor of 1.05. To determine the initial values for model parameterisation, the  
 227 *Hetero. Sed. & Bio. Growth*. models are first run for a period of 180 d to reach a quasi-steady  
 228 state under low stream nutrient concentrations of DOC=2 mg/L, NO<sub>3</sub><sup>-</sup>=1 mg/L, DO=8 mg/L. The  
 229 model simulations are run for a period of 730 d, which is sufficiently long for microbial  
 230 communities to fully acclimate from the initial condition to a new hydro-biogeochemical  
 231 environment. All simulated model scenarios are summarized in Table 2.

232 **Table 1.** Summary of numerical model parameters.

Parameters	2-D flume model	Model simulations	Source
<b><i>Physical parameters</i></b>			
Initial porosity ( $\theta_0$ , -)	0.38	0.38	1
Initial permeability ( $k_0$ , m/s)	$5.5 \times 10^{-4}$	$1 \times 10^{-10}$ - $1 \times 10^{-2}$ $1 \times 10^{-11.5}$ - $1 \times 10^{-10.3}$	2
		$1 \times 10^{-9}$	
<b><i>Reactive transport parameters</i></b>			
Molecular diffusion coefficient of ( $D_m$ , m <sup>2</sup> /s)	$1 \times 10^{-9}$	$1 \times 10^{-9}$	C
Longitudinal dispersivity ( $\alpha_L$ , m)	$1.0 \times 10^{-2}$	$1.0 \times 10^{-3}$	2
Transverse dispersivity ( $\alpha_L$ , m)	$1.0 \times 10^{-3}$	$1.0 \times 10^{-4}$	2



Maximum specific uptake rate for $O_2$ ( $V_{O_2}$ , 1/h)	0.32	0.32	C
Half-saturation constants for DOC ( $K_{DOC}$ , mg/L)	18	18	C
Half-saturation constant for DO ( $K_{DO}$ , mg/L)	5.0	5.0	C
Maximum specific uptake rate for $NO_3^-$ ( $V_N$ , 1/h)	0.28	0.28	C
Half-saturation constant for $NO_3^-$ ( $K_{NO_3}$ , mg/L)	1.0	1.0	C
Inhibition constant ( $K_I$ , mg/L)	0.8	0.8	C
Mass transfer coefficient for DOC ( $\alpha_C$ ; 1/hr)	$5 \times 10^{-4}$	$5 \times 10^{-5}$	3
Distribution coefficient of DOC ( $K_d$ ; L/kg)	10	7500	3
Bulk density ( $\rho_s$ ; kg/m <sup>3</sup> )	1600	1600 <sup>e</sup>	C
Initial organic matter content (POC <sub>0</sub> )	0.15%	0.2%-2% 0.1%-0.5%	3
<b><i>Biofilm growth parameters</i></b>			
Yield coefficient ( $Y$ , -)	0.15	0.15	C
Decay rate coefficient ( $k_d$ , d <sup>-1</sup> )	0.1	0.1	C
Density of biomass ( $\rho_X$ , kg/m <sup>3</sup> )	1.5	1.5	C
<b><i>Biofilm-plug formation model</i></b>			



Minimum relative hydraulic conductivity ( $K_{min}/K_0$ , -)	$1 \times 10^{-4}$	$1 \times 10^{-4}$	C
Critical relative porosity ( $\theta_{rc}$ , -)	0.75	0.75	C

233 Source: 1. Quick et al., 2016; 2. Sawyer & Cardenas, 2009; Bardini et al., 2013; 3. Zarnetske et al., 2012; Pescimoro  
 234 et al., 2019; C. The parameter values are calibrated from the experimental data of Quick et al. et al. (2016).

235 **Table 2.** Summary of the numerical experiments carried out in this study.

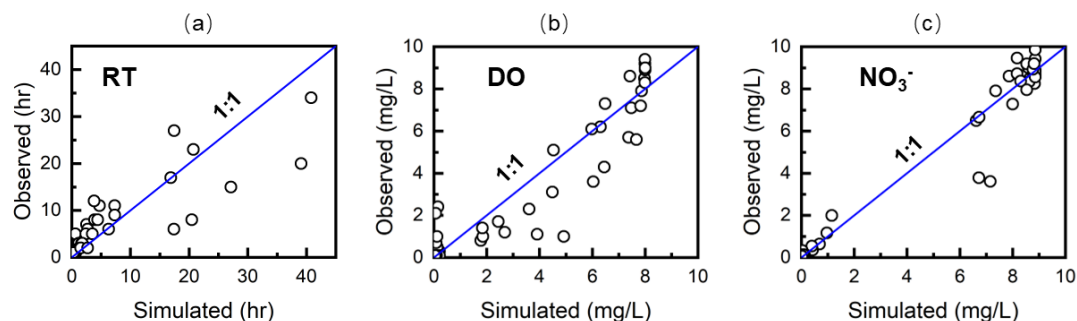
Experiment	Sediment and functional microbial biomass	Parameters
<i>Hetero. Sed. &amp; Bio. Growth model</i>	Heterogeneous sediments, with biomass growth	See Table S1
<i>Homo. Sed. &amp; Bio. Growth model</i>	Equivalent homogeneous sediments, with biomass growth	Based on initial values of <i>Hetero. Sed. &amp; Bio. Growth</i> models, Equivalent permeability tensors and arithmetically weighted averages of OM content are used as initial values.
<i>Hetero. Sed. &amp; Const. Bio. model</i>	Heterogeneous sediments, constant biomass	$K$ is constant and set as initial values of <i>Hetero. Sed. &amp; Bio. Growth</i> model. $X$ is constant and uniform and calculated as arithmetically weighted average of microbial biomass across entire space and time in <i>Hetero. Sed. &amp; Bio. Growth</i> models.



236 **3 Results and Discussions**

237 3.1 Model Validation

238 To ensure the 2D numerical model's reliability, we simulate the nitrate reactive transport  
239 process in the hyporheic zones of the flume experiments by Quick et al. (2016) and compare  
240 with the experimental data. The flume experiment, conducted in 2015, employed artificially  
241 constructed dune morphologies with a wavelength of 0.7 m and amplitude of 0.09 m. The  
242 sediment bed was composed of 90% quarried sand and 10% microbial-inoculated sand, and the  
243 sediment contains 0.15% particulate organic matter (Black cottonwood (i.e., *Populus*  
244 *trichocarpa*) leaves). Over a 13-week experimental duration, nitrate loading was controlled by  
245 injecting  $\text{KNO}_3$  to achieve a surface water concentration of  $10 \text{ mg L}^{-1}$  by day 62. Spatial and  
246 temporal monitoring included continuous DO measurements using in situ sensors, and discrete  
247 sampling of porewater chemistry. See Quick et al. (2016) for the details of the flume experiment.  
248 Model calibration focused on optimizing the parameters of porous flow, biogeochemical reaction  
249 and biofilm growth kinetics by fitting the simulation results to observed data. As portrayed in  
250 Figure 2, the simulated results generally agree well with the experimental data of solute  
251 residence time (RT) distributions, and the distributions of DO and  $\text{NO}_3^-$  concentrations. This  
252 validation indicates the numerical model can effectively capture the coupled hydrodynamic-  
253 biogeochemical dynamics in HZ systems. The calibrated numerical model parameters are listed  
254 in Table 1.



255

256 **Figure 2.** Comparisons between simulated and observed by Quick et al. et al. (2016) of (a)

257 residence time (RT) distribution, (b) DO, and (c) NO<sub>3</sub><sup>-</sup>.

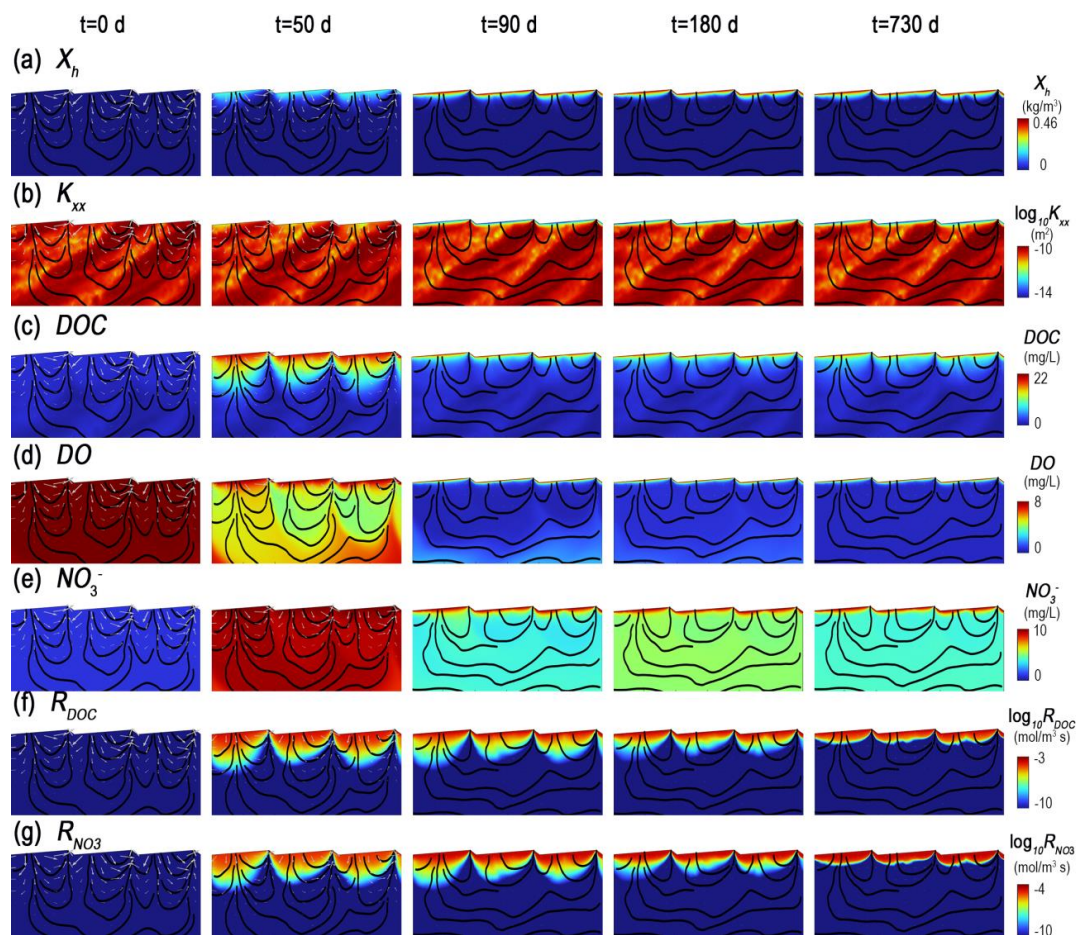
### 258 3.2 Effect of Sedimentary heterogeneity and microbial heterogeneity on Nitrate Removal in HZs

259 The outcomes of the different scenario simulations indicate that, compared to their  
260 respective homogeneous analogues, the spatial variability in functional microbial biomass  
261 presents a decisive role on enhanced removal of nitrate in the hyporheic zones. In contrast,  
262 small-scale heterogeneity of physical streambed properties (i.e., permeability) and chemical (i.e.,  
263 OM content) heterogeneities turned out to have negligible effect.

264 Figure 3 illustrates that in the *Hetero. Sed. & Bio. Growth* models parameterized to  
265 represent Brazos River conditions, five hyporheic flow cells driven by ripples develop within  
266 streambed sediments. Under nutrient enrichment from fluvial inputs, microbial biomass  
267 proliferates near the water-sediment interface, eventually forming a bioactive layer at the  
268 streambed surface (Figure 3a). Concurrently, biofilm-induced clogging generates a progressive  
269 reduction of permeability in a layer aligned with this active zone (Figure 3b). In response to such  
270 reduced permeability and intensified microbial nutrient consumption, transport fronts of DOC,  
271 DO, and NO<sub>3</sub><sup>-</sup>, alongside DOC and nitrate reduction hotspots, migrated further toward the



272 sediment-water interface (Figure 3c-3g).

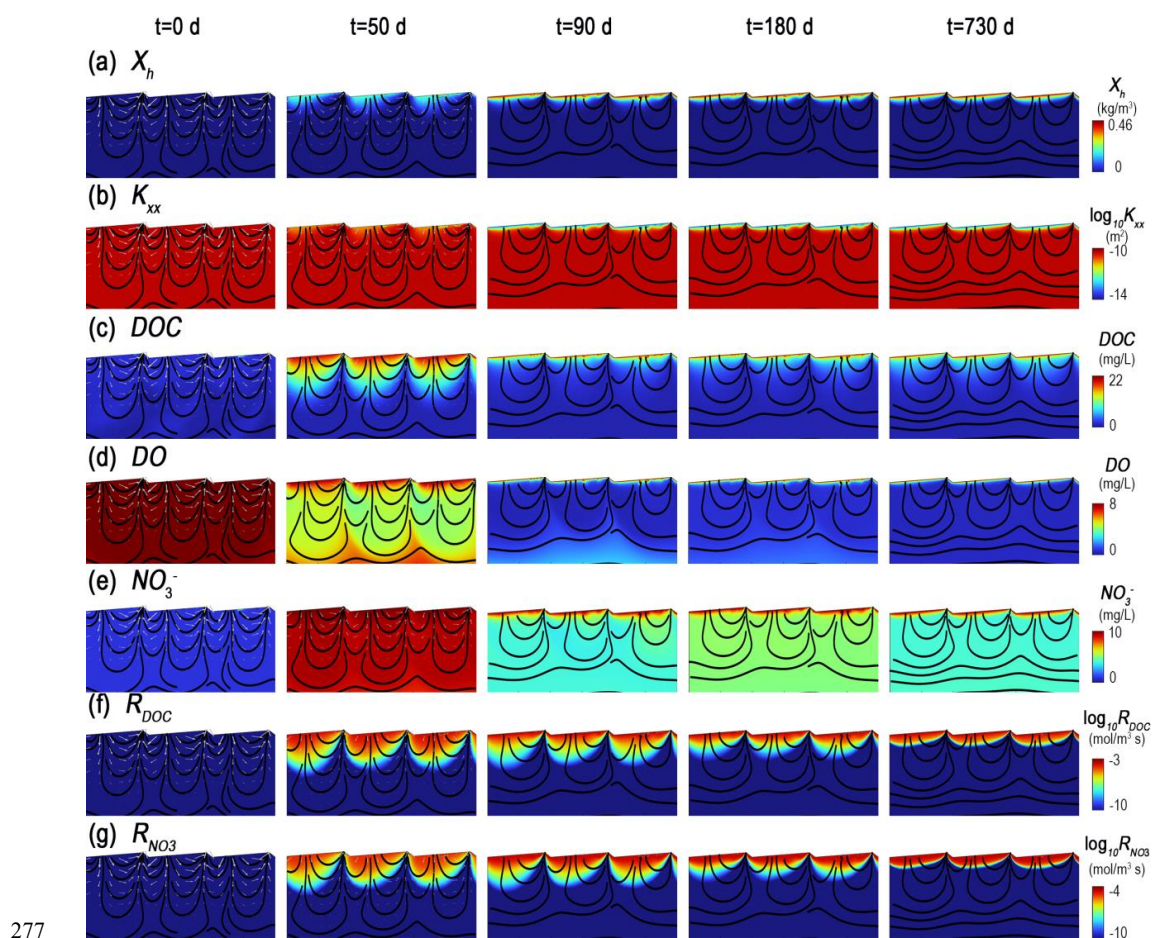


273

274 Figure 3. Evolution of spatial distribution of (a)  $X_h$ , (b)  $K_{xx}$ , (c) DOC, (d) DO, (e)  $\text{NO}_3^-$ , (f)  $R_{DOC}$ ,

275 (g)  $R_{\text{NO}_3}$  over time in the *Hetero. Sed. & Bio. Growth* model of the Brazos River under average

276 stream flow velocity  $U=0.8$  m/s.



277

278 Figure 4. Evolution of spatial distribution of (a)  $X_h$ , (b)  $K_{xx}$ , (c) DOC, (d) DO, (e)  $\text{NO}_3^-$ , (f)  $R_{DOC}$ ,  
 279 (g)  $R_{\text{NO}_3}$  over time in the *Homo. Sed. & Bio. Growth* model of the Brazos River under average  
 280 stream flow velocity  $U=0.8$  m/s.

281 Figure 4 shows the evolutions of the spatial distribution of  $X_h$ ,  $K_{xx}$ , DOC, DO,  $\text{NO}_3^-$ ,  $R_{DOC}$   
 282 and  $R_{\text{NO}_3}$  for the *Homo. Sed. & Bio. Growth* model which considered an equivalent permeability  
 283 and POC content and included microbial growth. By comparing Figure 3 and Figure 4, only  
 284 slight differences occur between the results of *Hetero. Sed. & Bio. Growth* and *Homo. Sed. &*  
 285 *Bio. Growth* models. Specifically, the plume fronts are slightly jagged in the *Hetero. Sed. & Bio.*



286 *Growth* model due to the heterogeneous permeability and POC content, while they are relatively  
 287 smooth in the results of the *Homo. Sed. & Bio. Growth* model because of more homogeneous  
 288 distribution of permeability and POC content. To quantify the effect of sedimentary  
 289 heterogeneity on nitrate removal, we compare the total nitrate removal  $R_N$  by integrating removal  
 290 rate of nitrate across the 2-D model domain over the simulated period. As indicated in Table 3,  
 291 the differences of  $R_N$  between *Hetero. Sed. & Bio. Growth* and *Homo. Sed. & Bio. Growth* model  
 292 over various scenarios range from 2.3%-10.1% for the Brazos River, and 0.6%-7.6% for the  
 293 Massillon Sandstone. This only minor effect of sedimentary heterogeneity on nitrate removal can  
 294 be attributed to the followings: With biofilm growth, the permeability of the sediment near the  
 295 water-sediment interface is gradually reduced (Figure 3b and Figure 4b). In response, the HZ  
 296 size diminishes, and therefore the effect of sediment heterogeneity in the deeper sediments on  
 297 nitrate removal is weakened. In consequence, the small-scale sedimentary heterogeneity  
 298 including spatial variabilities in permeability and OM content only have slight effect on the  
 299 removal of nitrate in HZs.

300 **Table 3.** Total nitrate removal ( $R_N$ ) of different model simulation scenarios

Stream conditions ( $U$ [m/s], $C$ [mg/L], $\text{NO}_3^{-1}$ [mg/L])	$R_N$ (g/m)			Errors (%)	
	<i>Hetero. Sed. &amp; Bio. Growth</i>	<i>Homo. Sed. &amp; Bio. Growth</i>	<i>Hetero. Sed. &amp; Const. Bio.</i>	<i>Homo. Sed. &amp; Bio. Growth</i>	<i>Hetero. Sed. &amp; Const. Bio.</i>
<i>Brazos River</i> $U=0.2, C_0= 5,$ $\text{NO}_3^{-1}=2$	3.7	3.3	0.8	10.1	79.5
$U=0.2, C_0= 20,$ $\text{NO}_3^{-1}=10$	67.8	64.7	6.1	4.6	91.0



	$U=0.8, C_0=5,$ $NO_3^-=2$	13.9	13.4	3.7	3.4	73.0
	$U=0.8, C_0=20,$ $NO_3^-=10$	176.4	180.4	21.0	2.3	88.1
<i>Massillon</i>	$U=0.2, C_0=5,$ $NO_3^-=2$	2.7	2.7	0.2	0.6	91.1
<i>Sandstone</i>	$U=0.2, C_0=20,$ $NO_3^-=10$	91.3	84.3	2.2	7.6	97.6
	$U=0.8, C_0=5,$ $NO_3^-=2$	20.5	20.6	3.5	0.7	82.8
	$U=0.8, C_0=20,$ $NO_3^-=10$	320.2	344.3	22.6	7.5	92.9

301 Note. Errors= $|R_{N-Baseline} - R_{N-Set 1}$  (or  $R_{N-Set 2}$ )  $| / R_{N-Baseline} \times 100\%$ , where the *Baseline* denotes the *Hetero. Sed. &*  
 302 *Bio. Growth* model, the *set 1* represents the *Homo. Sed. & Bio. Growth* model, and the *set 2* denotes the  
 303 *Hetero. Sed. & Const. Bio.* model.

304 In contrast, significant differences exist in the spatiotemporal distributions of DOC, DO  
 305 and  $NO_3^-$ , and  $R_{DOC}$ ,  $R_{NO_3^-}$  between the *Hetero. Sed. & Bio. Growth* and *Hetero. Sed. & Const.*  
 306 *Bio.* models, which account for the sedimentary heterogeneity while assuming constant and  
 307 uniform functional microbial biomass distributions. As described in Figures 3 and 5, the  
 308 penetration depths of DOC, DO and  $NO_3^-$  are greater in the *Hetero. Sed. & Const. Bio.* models  
 309 than the *Hetero. Sed. & Bio. Growth* models after  $t=50$  d. This is because the microbial biomass  
 310 concentration is assumed uniform and constant, and is much lower in the hyporheic zone for the  
 311 *Hetero. Sed. & Const. Bio.* models than that in the superficial layer of *Hetero. Sed. & Bio.*  
 312 *Growth* models, and thus the biofilm-induced bioclogging effect is much weaker. Accordingly,  
 313 the hotspots of aerobic respiration are more pronounced and the hotspots of denitrification occur

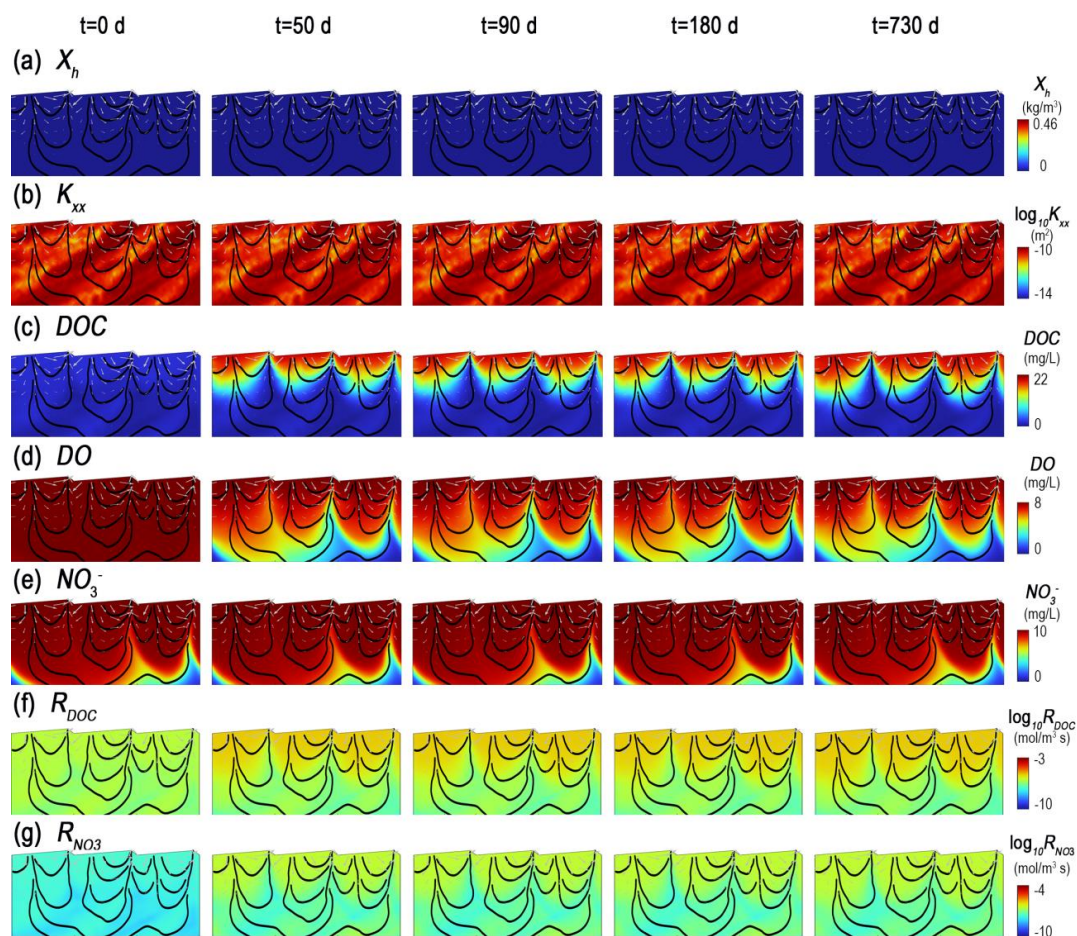


314 deeper in the *Hetero. Sed. & Const. Bio.* models. In comparisons, in the *Hetero. Sed. & Bio.*  
315 *Growth* models, the aerobic respiration mainly in the superficial bioactive layer, while the  
316 denitrification occurs in the shallower sediments. These discrepancies are also quantitatively  
317 evaluated using  $R_N$  in Table 3. The differences of  $R_N$  between the *Hetero. Sed. & Bio. Growth*  
318 and *Hetero. Sed. & Const. Bio.* models in the various scenarios range from 73.0% to 91.0% for  
319 the Brazos River, and 82.6 to 97.6% for the Massillon Sandstone. The  $R_N$  ratio of the *Hetero.*  
320 *Sed. & Bio. Growth* and *Hetero. Sed. & Const. Bio.* models can reach as much as 41.3. Thus,  
321 functional microbial biomass heterogeneity has a substantial impact on the enhanced removal of  
322 nitrate in contrast to the hyporheic zone models assuming constant and uniform microbial  
323 biomass. See Figures S2-S7 for the comparisons of the Brazos River under different stream flow  
324 velocity and nutrient conditions.

325         Some pioneering works on this issue include, for example, Bardini et al. (2013) who  
326 indicated that the sole permeability heterogeneity has negligible effects on nitrogen cycling in  
327 HZs. Further, Sawyer (2015) presented that the combined effects of permeability and OM  
328 content heterogeneity can largely enhance the upwelling groundwater-borne nitrate removal. The  
329 enhanced nitrate removal is attributed to the mixing effects of waters with different chemical  
330 components at the oxic-anoxic interfaces and create the localized anoxic microzones in fine  
331 structures of heterogeneous sediments that can effectively promote denitrification. As such, it is  
332 expected that sedimentary heterogeneity would enhance the nitrate removal of HZs in contrast to  
333 the equivalent homogeneous sediments. However, unexpectedly, the present study shows that the  
334 sedimentary heterogeneity has only small impact on nitrate removal, despite under the existence  
335 of spatial variability in permeability and OM content. Thus, it is unnecessary to take tremendous



336 labor for acquiring the small-scale sedimentary heterogeneity to improve prediction accuracy of  
 337 HZs' nitrate removal models.



338

339 Figure 5. Evolution of spatial distribution of (a)  $X_h$ , (b)  $K_{xx}$ , (c) DOC, (d) DO, (e)  $\text{NO}_3^-$ , (f)  $R_{DOC}$ ,  
 340 (g)  $R_{\text{NO}_3}$  over time in the *Hetero. Sed. & Const. Bio.* model of the Brazos River under average  
 341 stream flow velocity  $U=0.8$  m/s.

342 Compared to the negligible effect to sedimentary heterogeneity, the present study shows  
 343 that spatial heterogeneity in functional microbial biomass momentarily enhances the nitrate  
 344 removal of HZs in relative to the streambed assuming uniform microbial distribution. Under the



345 stream-borne nutrient stimulations, microbial biomass grows mostly near the water-sediment  
346 interface, while it is relatively sparse in the deep sediments. The hotspots of denitrification  
347 develop adjacent to the superficial bioactive layer under the regulation of nutrients, DO and  
348 microbial distributions. The aggregation of functional microbial biomass near the water-sediment  
349 interface importantly increases the denitrification rate across the hotspot as well as the nitrate  
350 removal capacity of streambed sediments, in contrast to the HZs in the streambed using the  
351 common assumption of constant and uniform microbial biomass. By conducting flume  
352 experiments, De Falco et al. (2018) compared the microbial consumption rate of DO in disturbed  
353 (homogeneous biomass distribution) and stationary (concentrated biomass near water-sediment  
354 interface) bed forms. The enhanced effect of the concentrated functional microbial biomass is  
355 documented by the elevated DO consumption rate in the stationary bed forms under various  
356 stream flow velocity. Consequentially, microbial heterogeneity plays a more decisive role on  
357 regulation of nitrate removal of HZs than sedimentary heterogeneity. In field settings, it is  
358 critical to get the spatiotemporal distribution of functional microbial biomass in HZs, especially  
359 near the water-sediment interface, otherwise significant errors of nitrate removal would be  
360 induced by the typical assumption of constant and uniform microbial biomass.

## 361 **5 Conclusions**

362 Under the feedbacks of porous flow, reactive transport of nutrients and microbial growth,  
363 sedimentary heterogeneity including physical (e.g., permeability), chemical (e.g., OM content)  
364 and microbial heterogeneity coexist and interconnect in HZs of streambeds. This study is the first  
365 to reveal the regulations of sedimentary and microbial heterogeneities on nitrate removal of HZs.  
366 For small-scale heterogeneous riverbeds, the influence of sedimentary heterogeneity on nitrate  
367 removal of HZs can be negligible, while the impact of the heterogeneous distribution of



368 functional microorganisms is crucial. These novel discoveries have important implications for  
369 the conceptualization and parameterization of heterogeneous riverbed models. These include the  
370 possibility of using equivalent homogeneous riverbed models as substitutes for actual  
371 heterogeneous riverbeds, thereby overcoming the significant challenges associated with  
372 obtaining sediment heterogeneity information. In contrast, accurate acquisition of information  
373 regarding the heterogeneous distribution of functional microorganisms in HZs is crucial for  
374 improving model estimation accuracy.

#### 375 **Supporting information**

376 Additional details of the model simulation results were displayed in the supporting  
377 information.

#### 378 **Data availability**

379 All raw data can be provided by the first author upon request.

#### 380 **Competing interest**

381 The authors declare that they have no conflict of interest.

#### 382 **Author contribution**

383 Conceptualization: YX, ZW

384 Formal analysis: YX

385 Funding acquisition: YX, ZW

386 Investigation: YX, ZW

387 Methodology: YX

388 Writing-original draft: YX

389 Writing-review and editing: ZW, SK

390 Project administration: ZW



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