

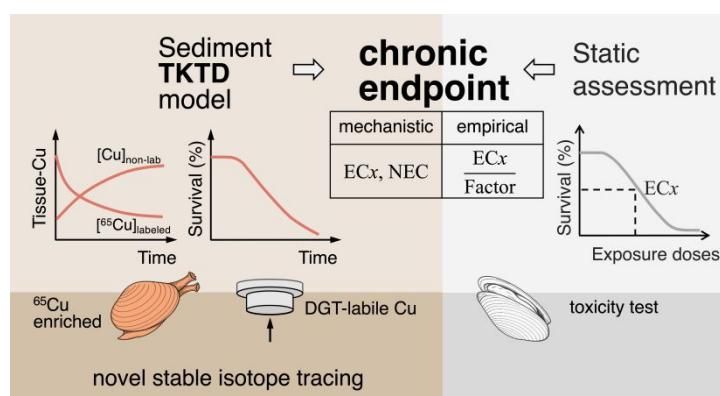
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## Advancing Toxicokinetic-Toxicodynamic Modeling to Assess Kinetic Bioavailability in Sediments

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4 16**TOC Art**

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## Abstract

Traditional sediment risk assessments often rely on fixed-duration dose-response endpoints, which overlook how exposure conditions and organismal response interact over time. Toxicokinetic-toxicodynamic (TKTD) models address this by linking external exposure, internal concentration, and effects, but their application to sediments has been limited by the lack of suitable tracing methods to generate high-quality kinetic data. Here, we developed a novel stable isotope tracing approach tailored for sediment TKTD studies—pre-labeling clams with stable  $^{65}\text{Cu}$  in water prior to Cu-amended sediment exposure. By simultaneously tracking  $^{65}\text{Cu}$  elimination and ambient Cu uptake, we successfully established a sediment TKTD model that reproduced bioaccumulation and toxicity across a wide range of Cu bioavailability driven by speciation differences. The model revealed that *Ruditapes philippinarum*'s Cu elimination and tolerance capability were consistent across exposure conditions, indicating species-specific physiological traits. Cu uptake followed a saturation-type Michaelis-Menten relationship with bioavailable Cu. From these kinetics, we derived mechanistic bioavailability benchmarks—DGT-Cu based  $\text{LC}_{50}$ -chronic =  $46 \mu\text{g L}^{-1}$  and  $\text{NEC}_{\text{DGT}} = 40 \mu\text{g L}^{-1}$ —as chronic thresholds applicable to risk assessment. Overall, this study demonstrates TKTD modeling in sediments using stable isotope tracing, advancing process-based ecological risk assessments and providing a transferable framework for evaluating contaminant bioavailability in estuarine ecosystems.

**Keywords:** Process-based model, sediment risk assessment, Cu speciation, *Ruditapes philippinarum*

**4      39      Synopsis:**

5  
6      40      By designing a novel stable isotope tracing technique, we established a sediment TKTD  
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8      41      model, advancing bioavailability-based risk assessment for contaminants in dynamic  
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10     42      estuarine environments.

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## 1 2 3 4     43    **1. Introduction** 5 6

7     44       In estuarine and coastal ecosystems, sediments can store anthropogenic contaminants  
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9     45       that later remobilize,<sup>1–3</sup> causing prolonged harm to benthic and pelagic communities.<sup>4,5</sup>  
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11     46       Chronic risk assessment is therefore essential for informing sediment remediation, habitat  
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13     47       restoration, and long-term pollution control.<sup>6</sup> Most conventional sediment risk assessments,  
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15     48       however, use static dose-response models to derive the fixed-duration endpoints (e.g., 10-d  
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17     49       EC<sub>x</sub>, where EC<sub>x</sub> represents x% effect concentration).<sup>7–9</sup> When chronic toxicity data are  
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19     50       lacking, these assessments then estimate chronic thresholds indirectly by applying highly  
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21     51       conservative safety factors (often poorly-grounded) to acute values, often without  
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23     52       mechanistic justification.<sup>10,11</sup> Such approaches cannot account for temporal fluctuations in  
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25     53       contaminant bioavailability driven by changing environmental conditions.<sup>12–16</sup>  
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27     54       Process-based toxicokinetic-toxicodynamic (TKTD) models address these limitations  
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29     55       by linking external contaminant concentrations to internal body burdens, and resulting  
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31     56       adverse effects over time.<sup>17</sup> They integrate concentration, exposure duration, and biological  
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33     57       response into a unified framework,<sup>18</sup> allowing chronic thresholds to be derived directly from  
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35     58       time-resolved data. While widely applied in water-only systems,<sup>19–21</sup> TKTD modeling in  
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37     59       sediments is limited,<sup>17</sup> largely because benthic exposures are complex. For benthic  
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39     60       invertebrates, uptake can occur from overlying water, porewater, and ingested particles, thus  
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41     61       bioavailability is strongly influenced by contaminant partitioning and speciation.<sup>22</sup> These  
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43     62       factors make it difficult to obtain the high-quality, process-resolved dataset required for  
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45     63       model parameterization and bioavailability prediction.  
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4       64     Recent studies suggested that isotope tracing is highly effective in laboratory water  
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6       65     exposure experiments for generating kinetic data that are categorially needed for  
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8       66     modeling.<sup>23,24</sup> However, directly amending kilograms of sediments with stable or radioactive  
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10      67     isotopes can be costly and may not reproduce natural binding phases,<sup>25–27</sup> reducing the  
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12      68     realism and practicality of sediment exposure experiments.

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15       69     To overcome these limitations, here we propose a novel isotope-tracing strategy tailored  
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17      70     for sediment studies. Rather than spiking isotopes into sediments, organisms are pre-enriched  
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19      71     with a stable isotope before exposure to contaminated sediments. By tracking the depuration  
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21      72     of the labeled isotope and the simultaneous accumulation of ambient contaminant from  
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23      73     sediments, this method provides the process resolution necessary for TKTD parameterization.

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28       74     To further reduce the model complexity, we characterized the bioavailable contaminant  
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30      75     pool in sediments and incorporated it as a proxy source for the metals available to biota. This  
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32      76     was achieved by a passive sampling technique —diffusive gradients in thin-films (DGT),<sup>28,29</sup>  
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34      77     which quantifies the labile metal species, including both dissolved and dissociable particulate  
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36      78     metal forms.<sup>30</sup> While DGT-labile metals have been tested as a proxy for bioavailable metal  
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38      79     pool in sediment in fixed-duration assessments,<sup>31–34</sup> their integration into kinetic modeling  
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41      80     remains unexplored.

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45       81     Accordingly, the objectives of this study were to develop the TKTD modeling for  
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47      82     sediments by integrating this new isotope-tracing approach with DGT-based bioavailability  
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50      83     characterization. Using copper (Cu) as a model metal contaminant of high ecological  
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53      84     concern,<sup>35–37</sup> we exposed the estuarine benthic clam *Ruditapes philippinarum*, an ecologically

representative species that integrates particle- and water-mediated exposure pathways in surficial sediments. By establishing mechanistic chronic toxicity thresholds for management purposes, this study offers a process-based framework for sediment risk assessment, adaptable to dynamic estuarine environments and capable of evaluating the fate and risk of contaminants under varying environmental scenarios.

## 2. Materials and Methods

### 2.1 Cu-spiked sediment

Plastic containers were cleaned following laboratory protocols to minimize metal contamination ([Note S1, Supporting Information, SI](#)). Coastal water, benthic clams (*Ruditapes philippinarum*), and clean sediments were collected and processed as detailed in [SI Note S1](#).

To examine the effects of Cu speciation and concentration on bioaccumulation and toxicity, sediments were spiked with three Cu forms representing typical speciation and concentrations in oxic/sub-oxic sediments: (1) basic Cu carbonate ( $\text{Cu}_2(\text{OH})_2\text{CO}_3$ ; Ref-0: 0, Low: 300, High: 1300 mg kg<sup>-1</sup>, dry mass basis), (2) copper chloride ( $\text{CuCl}_2$ ; Low: 300, Medium: 600, High: 900 mg kg<sup>-1</sup>), and (3) Cu adsorbed onto Fe oxyhydroxide (Cu-HFO; Low: 300, Medium: 600, High: 1200 mg kg<sup>-1</sup>).  $\text{Cu}_2(\text{OH})_2\text{CO}_3$ , a common Cu precipitate in natural sediments,<sup>38</sup> represents Cu carbonate and hydroxide species.  $\text{CuCl}_2$ , used in laboratory spiking studies, represents soluble Cu<sup>2+</sup> species and short-term sediment-bound Cu.<sup>25</sup> Cu-HFO, the adsorbed Cu phase, was prepared at half-, full-, and 2-fold oversaturation relative to Fe oxyhydroxide based on a Cu: Fe saturation curve ([SI Figure S1, Note S2](#)). All sediments

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4 106 were spiked, homogenized, and equilibrated for short term (3 days) to preserve the initial  
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6 107 speciation, then equilibrated under seawater for 24-h before the clam exposure experiments  
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8 108 ([SI Notes S1](#)).  
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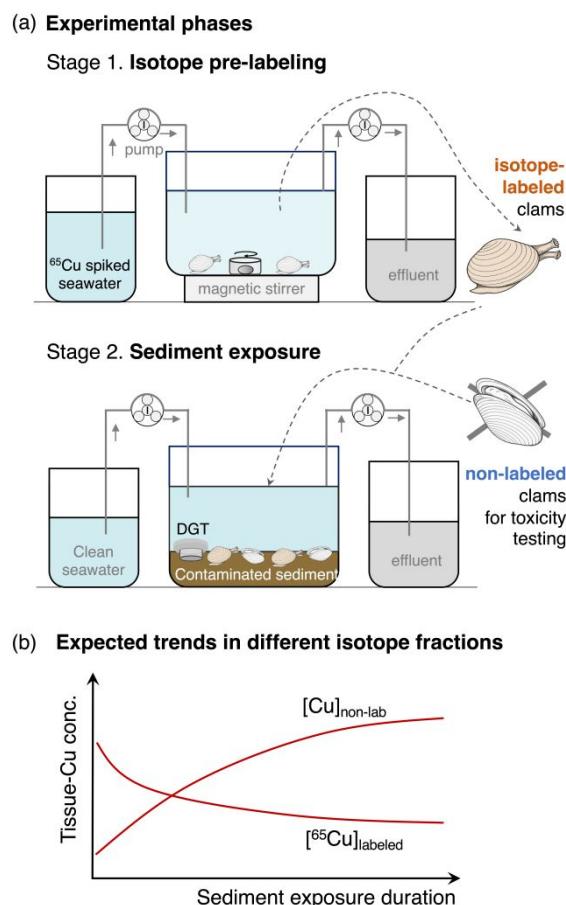
11  
12 109 **2.2 TKTD measurements using stable-isotope tracing for sediment contamination**  
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16 110 We first developed a novel stable-isotope tracing technique to generate a TKTD dataset  
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18 111 in sediments, which involves two main stages ([Figure 1a](#)): (1) isotope pre-labeling of clams  
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20 112 and (2) exposure to Cu contaminated sediments.  
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25 113 *2.2.1 Isotope-labeling stage*  
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27  
28 114 Isotope-labeling was achieved by exposing the clams to  $^{65}\text{Cu}$ -enriched seawater (purity  
29  
30 115 99.7%, purchased from Isoflex USA) to elevate the tissue  $^{65}\text{Cu}$  levels. In a flow-through  
31  
32 116 system, 240 clams were held in a 20-L chamber with  $10 \mu\text{g L}^{-1} {^{65}\text{Cu}}$  for 7 days ([Figure 1a](#)),  
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34 117 with water renewed twice daily. Daily water samples were acidified with “1+1”  $\text{HNO}_3$  to pH  
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36 118 < 2 for  $^{65}\text{Cu}$  analysis. During labeling, clams were transferred daily to a separate container for  
37  
38 119 1 h feeding, rinsed twice with clean seawater to remove algae, and returned to the system.  
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44 120 To evaluate the effectiveness of  $^{65}\text{Cu}$  labeling, 10 clams were sampled on Days 1, 3, and  
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46 121 7. Soft tissues were immersed in  $1 \text{ mmol L}^{-1}$  EDTA for 1 min to remove loosely-associated  
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48 122 metals, rinsed, frozen ( $-20^\circ\text{C}$ ), freeze-dried, digested, and analyzed for Cu isotopes. Ten non-  
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50 123 labeled clams were also processed and analyzed for background Cu composition.  
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**Figure 1.** Schematic illustration of experimental procedures and isotope tracing principle. (a) Overview of the stable isotope tracing approach, consisting of the isotope pre-labeling stage and the contaminated sediment exposure stage. (b) Expected trends in tissue concentrations of labeled  $^{65}\text{Cu}$  ( $[^{65}\text{Cu}]_{\text{labeled}}$ ) and ambient non-labeled Cu ( $[\text{Cu}]_{\text{non-lab}}$ ).

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#### 130 2.2.2 Sediment exposure stage

131 After labeling, clams were rinsed and transferred to Cu-spiked sediment chambers for  
132 sediment exposure (Figure 1a). Sediment exposure was conducted in 27 plastic chambers (30  
133  $\times 20 \times 17$  cm, 3 replicates  $\times$  9 Cu spike treatments). Each chamber was filled with 1-cm  
134 sediment layer and 10-cm overlying seawater. The water was connected to a flow-through  
135 system with 24-h turnover. Continuous aeration was supplied to ensure water mixing. To  
136 further reduce Cu accumulation in the overlying water, a manual water renewal was  
137 performed daily during clam and sediment sampling.

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4 138 At the commencement of sediment exposure (Day 0), thirty clams (20 labeled and 10  
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6 139 non-labeled) were introduced to each chamber. Non-labeled clams were marked for  
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8 140 identification. Within 30 minutes, clams burrowed partially or completely into sediment.  
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12 141 Exposure lasted up to 14 days or until complete mortality, which occurred in two CuCl<sub>2</sub>  
13  
14 142 treatments and one Cu-HFO treatment. For toxicokinetic (TK) assessment, two isotope-  
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16 143 labeled clams per replicate were sampled on Days 0.5, 1, 2, 4, 6, 8, 11, and 14. For  
17  
18 144 toxicodynamic (TD) assessment, non-labeled clam survival was monitored every 12 h, with  
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20 145 dead individuals removed.  
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26 146 During clam exposure, piston-type chelex-DGT samplers (DGT® Research Ltd.) were  
27  
28 147 deployed in Cu-spiked sediments on Days 0, 2, 4, 6, 8, and 11, and retrieved on Days 2, 4, 6,  
29  
30 148 8, 11, and 14, respectively. These samplers were inserted face-down to ~0.5 cm sediment  
31  
32 149 depth to accumulate Cu from porewater and loosely sediment-bound sediment fractions (SI  
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34 150 Note S3). After retrieval, samplers were rinsed and stored for further processing. In the  
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36 151 reference sediment, DGTs remained deployed for the full 14-d duration.  
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41 152 Alongside DGT retrieval, sediment samples (~ 5 g) were collected from multiple  
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43 153 chamber locations, homogenized, and centrifuged (3000 × g, 5 min). The resulting  
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45 154 supernatant porewater was filtered (0.45 µm), acidified (pH < 2), and stored for Cu analysis.  
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47 155 Remaining sediment in the centrifuge tube was preserved for extraction-based speciation  
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49 156 analysis (Section 2.3 and SI Note S4).  
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53 157 Overlying water was sampled daily before and after water renewal. From each  
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55 158 treatment, 3 mL per replicate was filtered, combined, acidified, and stored for analysis.  
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3     159 **2.3 Sampling and analysis**

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6     160 Bulk clean sediment was characterized for particle size and total recoverable metal  
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8     161 (TRM) concentrations. Sediments from exposure chambers were analyzed for Cu speciation:  
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10    162 TRM, dilute-acid extractable metals (AEM), acid volatile sulfide (AVS), and a modified  
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12    163 BCR sequential extraction (exchangeable, reducible, oxidizable, and residual fractions).<sup>39–41</sup>  
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15    164 TRM, AVS, and full BCR were measured at the start and end of exposure; AEM and the  
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17    165 first-step BCR extraction for exchangeable fraction were determined at multiple time points.  
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21    166 More details on sediment analysis are detailed in [SI Note S4](#).

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24    167 DGT samplers were disassembled, and binding gels digested in 1 mL 1-M HNO<sub>3</sub> for at  
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26    168 least 24 h. The eluates were diluted with 2% HNO<sub>3</sub> for Cu analysis and DGT-labile Cu  
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29    169 concentration was computed based on equations recommended by the manufacture ([SI Note](#)  
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32    170 [S3](#)).

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35    171 Sampled clams were processed to clear gut contents, involving a 2-h depuration in clean  
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37    172 seawater after vigorous shaken (~120 strokes min<sup>-1</sup> for 1 minute) in a closed container  
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40    173 (effectiveness was confirmed by tissue digest clarity). Clams were then dissected, and soft  
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43    174 tissues were freeze-dried, cold-digested overnight in 1.5 mL 65% HNO<sub>3</sub> (ANPEL Laboratory  
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46    175 Technologies, Shanghai), and then hot-digested at 80°C for 8 h. Solutions were diluted with  
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49    176 deionized water for Cu isotope analysis. Standard reference material (SRM 2976, mussel  
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52    177 tissue) was processed identically to verify recovery.

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55    178 Cu isotopes (<sup>63</sup>Cu and <sup>65</sup>Cu) in sediment digest, water samples, tissue digests, and DGT  
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58    179 eluents were quantified by inductively coupled plasma mass spectrometry (ICP–MS,

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4 180 NexION 2000, PerkinElmer). Comprehensive quality assurance and quality control  
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7 181 procedures are described in [SI Note S4](#).  
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10 182 **2.4 Data analysis for isotope tracing**  
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13 183 Cu in isotope-labeled clams was divided into two fractions ([SI Figure S2](#)): the newly  
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184 accumulated  $^{65}\text{Cu}$  ( $[\text{Cu}]_{\text{labeled}}$ ), and the non-labeled Cu fraction ( $[\text{Cu}]_{\text{non-lab}}$ ) that includes  
19 both background Cu and Cu accumulated from sediments with natural isotopic composition.  
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22 186 These fractions were quantified based on isotope-specific analysis ([SI Figure S2](#)  
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24 illustrates the principle). The non-labeled Cu concentration was calculated from the ICP-MS  
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27 reported Cu concentration using  $^{63}\text{Cu}$  as the quantifying isotope:  
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30 189  $[\text{Cu}]_{\text{non-lab}} = [\text{Cu}]_{\text{ICP-MS}}$  (1)  
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33 190 where  $[\text{Cu}]_{\text{ICP-MS}}$  ( $\mu\text{g g}^{-1}$  dry weight) is the tissue concentration from ICP-MS reported  
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35 value, adjusted for digestion volume and dilution.  
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38 192 In contrast, the newly accumulated  $^{65}\text{Cu}$  was calculated as the difference between Cu  
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40 concentrations reported using  $^{65}\text{Cu}$  and  $^{63}\text{Cu}$  as quantifying isotopes, multiplied by 30.8%  
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43 194 (the natural abundance of  $^{65}\text{Cu}$ ):  
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46 195  $[\text{Cu}]_{\text{labeled}} = ([\text{Cu}]_{\text{ICP-MS}} - [\text{Cu}]_{\text{ICP-MS}}) \times 30.8\%$  (2)  
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49 196 where  $[\text{Cu}]_{\text{ICP-MS}}$  ( $\mu\text{g g}^{-1}$  dry weight) is also converted to tissue concentration based on  
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51 digestion and dilution steps. For further methodological details, see our previous  
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54 198 publications.<sup>16,42</sup>  
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3      199 **2.5 TKTD modeling**

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7      200        The toxicokinetic-toxicodynamic (TKTD) model was used to simulate Cu  
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10     201      bioaccumulation and toxicity effects on clam survival.<sup>18,43</sup> The model consists of a TK  
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12     202      module for Cu uptake and elimination, and a TD module linking internal Cu concentrations  
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15     203      to mortality.

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18     204        During sediment exposure,  $[^{65}\text{Cu}]_{\text{labeled}}$  was expected to decline due to elimination,  
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21     205        while  $[\text{Cu}]_{\text{non-lab}}$  was expected to increase due to accumulation from sediment (Figure 1b).  
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24     206        These processes were modeled using a one-compartment TK model:<sup>44,45</sup>

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26     207                   $\frac{dC_{\text{int}}(t)}{dt} = J_{\text{in}}(t) - k_e \cdot C_{\text{int}}(t)$                   (3)

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28     208        where  $C_{\text{int}}(t)$  ( $\mu\text{g g}^{-1}$  dry weight) is the internal Cu concentration at time  $t$  (d),  $J_{\text{in}}(t)$  ( $\mu\text{g g}^{-1} \text{ d}^{-1}$ )  
29  
30     209       <sup>1)</sup> is the Cu uptake rate, and  $k_e$  ( $\text{d}^{-1}$ ) is the elimination rate constant. Growth dilution effect  
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32     210        was negligible compared to Cu efflux (SI Figure S3), and was not considered in the model.  
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37     211        Cu uptake was modeled using DGT-labile Cu ( $C_{\text{DGT}}$ ,  $\mu\text{g L}^{-1}$ ) as a proxy for Cu available  
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39     212        to clams:

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41     213                   $J_{\text{in}}(t) = k_{\text{DGT}} \times C_{\text{DGT}}(t)$                   (4)

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43     214        where  $k_{\text{DGT}}$  ( $\text{L g}^{-1} \text{ d}^{-1}$ ) is the uptake rate constant of DGT-labile Cu. TK modeling was later  
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45     215        improved with a unified uptake rate formula for better prediction across different Cu  
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47     216        speciation and concentrations, as discussed in Section 3.4.2.  
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52     217        Cu-induced mortality in non-labeled clams was simulated with a threshold-based TD  
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54     218        model:<sup>19,44</sup>

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56     219                  
$$\frac{dH(t)}{dt} = \begin{cases} k_m \times (C_{\text{int}}(t) - C_{\text{IT}}) + h_0, & \text{if } C_{\text{int}}(t) > C_{\text{IT}} \\ h_0, & \text{otherwise} \end{cases}$$
                  (5)

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4 220  $S(t) = e^{-H(t)}$  (6)  
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where  $H(t)$  (dimensionless) is the cumulative hazard,  $S(t)$  is the survival rate,  $C_{IT}$  ( $\mu\text{g g}^{-1}$ ) is the internal threshold concentration,  $k_m$  is the mortality rate constant ( $\text{g } \mu\text{g}^{-1} \text{ d}^{-1}$ ), and  $h_0$  ( $\text{d}^{-1}$ ) is the background hazard rate (set to 0 due to no mortality in the reference treatment).

## 2.6 Model parameter estimation

TK parameters ( $k_e$  and  $k_{DGT}$ ) were estimated by fitting model simulations to measured tissue Cu concentrations (as illustrated in [Figure 1b](#)). Specifically,  $k_e$  was fitted to  $[^{65}\text{Cu}]_{\text{labeled}}$ . With  $k_e$  fixed,  $k_{DGT}$  was then fitted to  $[\text{Cu}]_{\text{non-lab}}$ . TD parameters  $C_{IT}$  and  $k_m$  were estimated by fitting the whole TKTD model to clam survival in the  $\text{CuCl}_2$  and Cu-HFO treatments where significant toxicity was observed.

Parameter estimation was conducted using OpenModel 2.4.2 (University of Nottingham). Initial parameter values were obtained using least-squares fitting (the Marquardt algorithm), and then all parameter values were optimized using Markov Chain Monte Carlo (MCMC) method via the Metropolis-Hastings algorithm.

## 3. Results and Discussion

### 3.1 Sediment characteristics and Cu speciation in spiked sediments

The clean sediments (used for Cu spike and in the reference treatment) were fine-grained (96%  $< 170 \mu\text{m}$ ; [SI Table S1](#)), with metal concentrations well below sediment quality criteria ([SI Table S2](#)). No acid volatile sulfide was detected in the original or Cu-spiked sediment, indicating sub-oxic/oxic conditions. Following spiking, total recoverable Cu (TR-Cu) ranged from 300 to 1300  $\text{mg kg}^{-1}$  across the treatments, with minimal Cu loss during the

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4 241 experiment ([SI Figure S4](#)). BCR sequential extraction revealed that spiked Cu was mainly in  
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6 242 the exchangeable and reducible fractions, accounting for 84% to 98% of TR-Cu ([SI Figure](#)  
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8 243 [S5](#)). These distributions remained stable over time, as did the dilute-acid extractable Cu (AE-  
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10 244 Cu) and BCR-exchangeable Cu ([SI Figure S4](#)). The shallow sediment layer and clam  
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12 245 bioturbation minimized anoxic zones, preventing sulfide formation.  
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18 246 **3.2 Cu concentrations in porewater and overlying water**  
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22 247 Porewater Cu concentrations, measured by instantaneous grab sampling (PW-Cu) and  
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24 248 passive samplers (DGT-labile Cu) varied across treatments, spanning four orders of  
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26 249 magnitude ([SI Figure S6](#)).  $\text{Cu}_2(\text{OH})_2\text{CO}_3$  treatments exhibited a modest increase of porewater  
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28 250 Cu over time, indicating slow dissolution of  $\text{Cu}_2(\text{OH})_2\text{CO}_3$ .  $\text{CuCl}_2$  treatments exhibited  
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30 251 higher and more variable porewater Cu levels. Cu-HFO treatments showed intermediate  
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32 252 porewater Cu concentrations that declined during initial sediment ageing and stabilized  
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34 253 within 4 days. Cu concentrations in the overlying water (OW-Cu) were influenced by Cu  
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36 254 efflux from sediments, water exchange, and manual water renewal ([SI Figure S6 and S7](#)).  
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43 255 The wide gradient of Cu concentrations in porewater across treatments reflects  
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46 256 differences in Cu speciation. More discussion on Cu mobility in the sediment water system is  
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48 257 provided in the [SI Note S5](#). Briefly, PW-Cu and DGT-labile Cu were generally similar in  
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50 258 spiked sediments, suggesting efficient replenishment of porewater Cu from the solid phase—  
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52 259 unlike in the reference sediment.<sup>30</sup> This implies that Cu dissociation in spiked sediments is an  
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54 260 important source of bioavailable metal for benthic organisms, either through direct sediment  
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56 261 ingestion or by sustaining porewater Cu at the organism-porewater interface.

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3    262 **3.3 Threshold-like relationships between Cu bioaccumulation and aqueous Cu metrics**

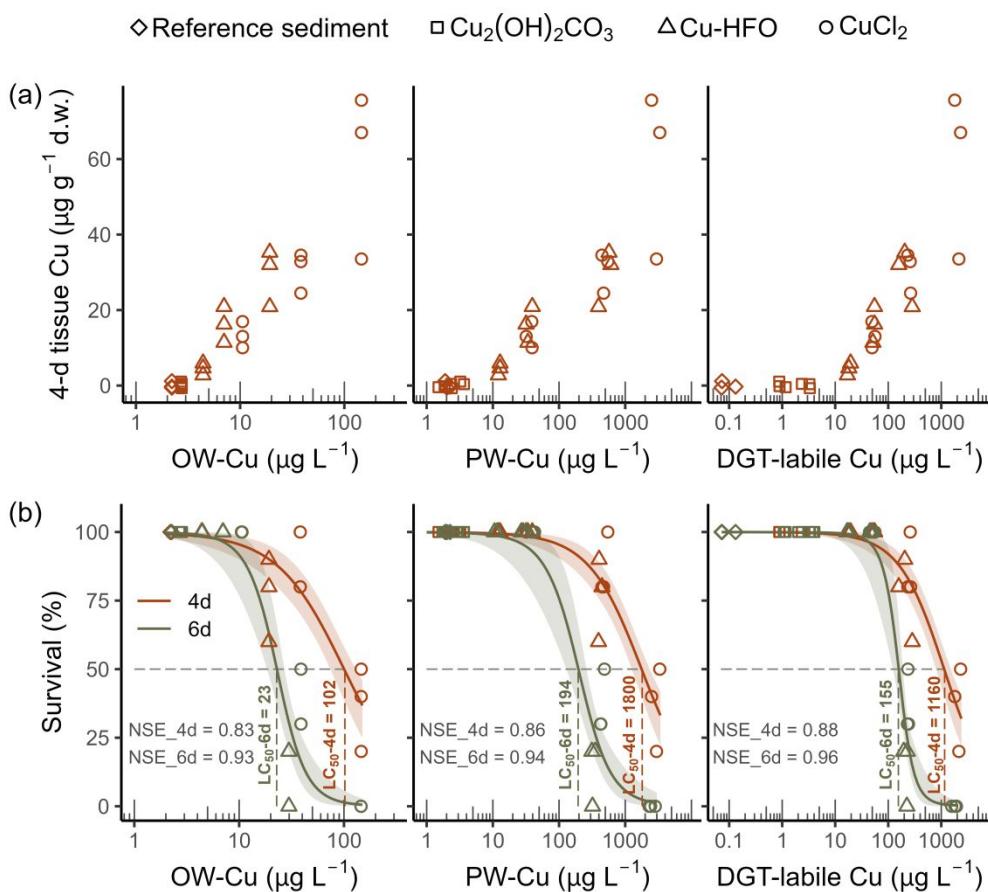
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6    263      Threshold relationships between 4-d tissue Cu and aqueous Cu metrics (OW-Cu, PW-  
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8    264 Cu, and DGT-labile Cu) were evident ([Figure 2a](#)). Tissue Cu remained low until OW-Cu  
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10    265 exceeded  $5 \mu\text{g L}^{-1}$ , or until PW-Cu/DGT-labile Cu surpassed  $10 \mu\text{g L}^{-1}$ . Above these  
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12    266 thresholds, tissue Cu increased sharply ([Figure 2a](#)). Similar threshold responses have been  
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14    267 reported in other benthic invertebrates, particularly for DGT-labile metal fluxes (equivalent to  
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16    268 concentrations through mathematical conversion). For instance, Zinc (Zn) accumulation in  
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18    269 the clam *Tellina deltoidalis* increased markedly when DGT-Zn flux exceeded  $\sim 160 \mu\text{g h}^{-1} \text{ m}^{-2}$ ,  
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20    270 with similar thresholds (OW-Zn =  $32 \mu\text{g L}^{-1}$ , DGT-Zn flux =  $220 \mu\text{g h}^{-1} \text{ m}^{-2}$ ) observed  
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22    271 for *Hyridella australis*.<sup>46</sup> In the midge *Chironomus tentans*, Cu bioaccumulation increased  
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24    272 sharply when DGT-Cu flux exceeded  $\sim 20 \mu\text{g h}^{-1} \text{ m}^{-2}$ .<sup>47</sup>

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26    273      In contrast, no consistent trends were observed with solid-phase Cu metrics—including  
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28    274 TR-Cu, AE-Cu, or BCR-exchangeable Cu ([SI Figure S8a](#))—indicating that they were less  
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30    275 representative of bioavailable Cu.

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32    276      Consistent with traditional ECx assessments, concentration-effect relationships were  
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34    277 analyzed for the 4-d and 6-d toxicity endpoints, and all three aqueous Cu metrics showed  
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36    278 clear dose-response relationships with survival ([Figure 2b](#)), while sediment-phase Cu  
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38    279 concentrations were poorly correlated ([SI Figure S8b](#)). A two-parameter log-logistic model  
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40    280 ([SI Note S6](#)) described these relationships, with high goodness-of-fit (Nash-Sutcliffe  
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42 efficiency, NSE > 0.83) for all aqueous Cu metrics. The median lethal concentrations (LC<sub>50</sub>)  
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44 declined with exposure duration, consistent with time-dependent toxicity. Specifically, LC<sub>50</sub>  
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values at day 6 ( $LC_{50}$ -6d) were  $23 \mu\text{g L}^{-1}$  (OW-Cu),  $194 \mu\text{g L}^{-1}$  (PW-Cu), and  $155 \mu\text{g L}^{-1}$  (DGT-labile Cu) were lower than the corresponding values at day 4 ( $LC_{50}$ -4d), which were  $102, 1800$ , and  $1160 \mu\text{g L}^{-1}$ , respectively.

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**Figure 2.** Exposure-duration specific static assessment of Cu bioaccumulation and toxicity using aqueous metrics: overlying water Cu (OW-Cu), porewater Cu (PW-Cu), and DGT-labile Cu. (a) Tissue Cu concentrations on Day 4 showing a threshold-like pattern in relation to aqueous Cu concentrations. Points represent mean tissue Cu concentrations from replicate chambers; point shape distinguishes treatments. (b) Clam survival on Days 4 and 6 as a function to aqueous Cu concentrations. Points represent survival in replicate chambers; solid curves represent log-logistic regression, with shaded areas indicating 95% confidence intervals. Vertical dashed lines indicated  $LC_{50}$  values. NSE (Nash–Sutcliffe efficiency) is a goodness-of-fit metric comparing model predictions to observations ( $-\infty$  to 1.00, with 1.00 indicating a perfect fit).

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3 298 **3.4 Toxicokinetic-toxicodynamic (TKTD) modeling**

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6 299 *3.4.1 Cu elimination*

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9 300 After 7 days of aqueous labeling, clams accumulated  $\sim 3.6 \mu\text{g g}^{-1}$  of  $^{65}\text{Cu}$  in tissue

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12 301 ( $[^{65}\text{Cu}]_{\text{labeled}}$ ) ([SI Figure S9](#)). During subsequent sediment exposures,  $[^{65}\text{Cu}]_{\text{labeled}}$  declined

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15 302 exponentially as expected, with the elimination rate slowing over time ([Figure 3](#)). After 14-

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18 303 days, final  $[^{65}\text{Cu}]_{\text{labeled}}$  ranged from  $0.5$  to  $1.3 \mu\text{g g}^{-1}$ , and slightly higher values ( $1.6$ – $1.9 \mu\text{g g}^{-1}$ )

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21 304 <sup>1)</sup> were observed in treatments terminated early due to toxicity ( $\text{CuCl}_2$ -Medium,  $\text{CuCl}_2$ -High,

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23 305 Cu-HFO-High).

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26 306 A one-compartment TK model, assuming first-order kinetics, provided a good fit to the

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29 307 measured  $[^{65}\text{Cu}]_{\text{labeled}}$  in clams ([Figure 3](#)), with the measured and predicted results aligning

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32 308 with the 1:1 line and within a two-fold deviation ([SI Figure S10](#)). The estimated efflux rate

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35 309 constant of  $k_e = 0.152 \pm 0.007 \text{ d}^{-1}$  indicates that  $\sim 15\%$  of tissue Cu was eliminated daily. This

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38 310 consistent elimination rate across treatments—regardless of sediment Cu speciation or

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41 311 observed toxicity—suggests that Cu excretion was largely independent of exposure type or

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44 312 organism health.

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47 313 Efflux rate constants during sediment exposure are rarely reported, yet the  $k_e$  value

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50 314 obtained here aligns with literature estimates for benthic clams during water-based exposure:

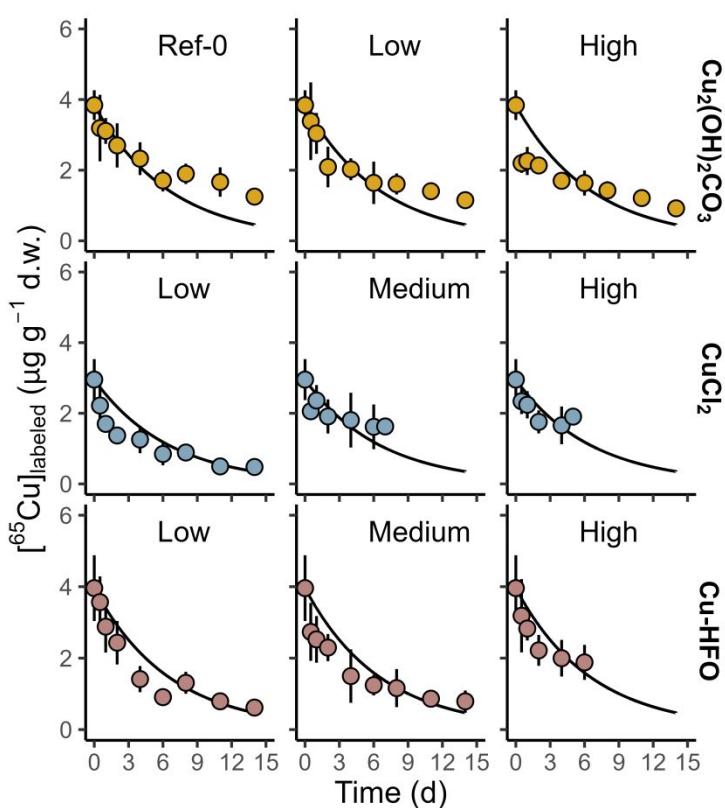
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53 315 *Ruditapes philippinarum* ( $0.147 \pm 0.026 \text{ d}^{-1}$ )<sup>48</sup> and *Ruditapes decussatus* ( $0.143 \text{ d}^{-1}$ ).<sup>49</sup> The

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56 316 similarity of these  $k_e$  values support the interpretation that Cu elimination is a physiological

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59 317 trait largely unaffected by exposure medium.<sup>50,51</sup> Consequently, in the absence of sediment-

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62 318 based  $k_e$  values, estimates from aqueous exposures may serve as reasonable surrogates.

319



**Figure 3.** Consistent  $^{65}\text{Cu}$  depuration kinetics in sediments across treatments, with a unified elimination rate constant ( $k_e = 0.152 \text{ d}^{-1}$ ). Solid points and error bars represent mean remaining pre-labeled  $^{65}\text{Cu}$  concentration in clams ( $[^{65}\text{Cu}]_{\text{labeled}}$ ) and standard deviation of triplicates. Black curves show TK model-predicted  $^{65}\text{Cu}$  concentrations with a unified  $k_e$ .

#### 3.4.2 Cu bioaccumulation

Non-labeled Cu accumulation ( $[\text{Cu}]_{\text{non-lab}}$ ) represents the net result of Cu uptake and

elimination, with sediment Cu concentration and speciation being major influencing factors

(Figure 4). In unspiked reference sediment (Ref-0),  $[\text{Cu}]_{\text{non-lab}}$  remained stable at  $\sim 3.0 \mu\text{g g}^{-1}$

over 14 days, indicating no net Cu accumulation. Unlike the elimination of pre-labeled  $^{65}\text{Cu}$ ,

the background Cu pool was not cleared from tissues, suggesting that it represents a

metabolically inert fraction. This distinction may be due to different Cu storage speciation in

organism tissue: X-ray absorption spectroscopy studies in oysters have found that

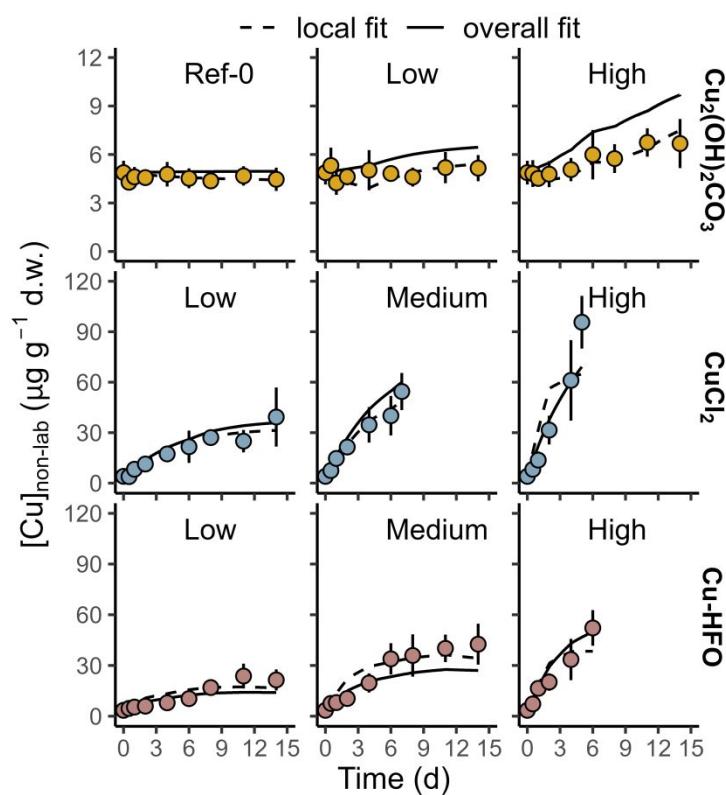
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4 334 background Cu in uncontaminated oysters typically occurs as Cu(I) bound to thio ligands  
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6 335 (more tightly bound), while newly accumulated Cu in contaminated oysters appears as Cu(II)  
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8 336 coordinated to O/N ligands.<sup>52</sup> These differences in speciation likely explain the low  
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10 337 elimination of background Cu.

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14 338 In Cu<sub>2</sub>(OH)<sub>2</sub>CO<sub>3</sub>-spiked sediments, bioaccumulation was limited ([Figure 4](#)): [Cu]<sub>non-lab</sub>  
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17 339 remained stable in Low treatment and comparable to Ref-0, indicating that influx did not  
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19 340 exceed efflux. A slight increase in [Cu]<sub>non-lab</sub> occurred in High treatment (from 3 to 6.2 µg g<sup>-1</sup>  
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22 341 ). In contrast, CuCl<sub>2</sub>-spiked sediments induced substantial and dose-dependent Cu  
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24 342 accumulation ([Figure 4](#)). [Cu]<sub>non-lab</sub> increased to 30, 60, and 90 µg g<sup>-1</sup> within 14, 7, and 6 days  
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26  
27 343 for the Low, Medium, and High treatments, respectively. Cu-HFO treatments showed  
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29 344 intermediate accumulation levels, varying with Cu loading and HFO sorption saturation  
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32 345 ([Figure 4](#)). In Low treatment (half-saturated), [Cu]<sub>non-lab</sub> increased to 25 µg g<sup>-1</sup> over 14 days.  
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34  
35 346 In Medium (saturated) and High (oversaturated) treatments, [Cu]<sub>non-lab</sub> reached 40 and 60 µg  
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37 347 g<sup>-1</sup> over 14 and 6 days, respectively.

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39 42 348 Using the pre-determined efflux rate constant ( $k_e$ ), we first applied a “local fit” approach  
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41 43 349 to model [Cu]<sub>non-lab</sub> individually for each treatment, yielding condition-specific uptake rate  
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43 44 350 constants ( $k_{DGT}$ ), which ranged from 7.2 to 0.01 L g<sup>-1</sup> d<sup>-1</sup> ([SI Table S3](#)). These constants  
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46 351 decreased by three orders of magnitude as  $C_{DGT}$  increased by five orders ([SI Figure S11](#)).  
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49 352 This model described the measured data well within each condition, but its prediction  
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51 353 capability was limited due to the need for concentration-specific calibration of  $k_{DGT}$   
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53 354 parameters. Furthermore, in low-accumulation treatments (Ref-0 and Cu<sub>2</sub>(OH)<sub>2</sub>CO<sub>3</sub>-Low),  
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355 limited accumulation of  $[Cu]_{\text{non-lab}}$  indicates negligible Cu bioavailability, contradicting with  
 356 the high  $k_{\text{DGT}}$  values ([SI Figure S11](#)) implying high Cu bioavailability.

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358  
 359 **Figure 4.** Temporal accumulation of non-labeled Cu in clams, influenced by sediment Cu  
 360 concentration and speciation. Solid points and error bars represent mean non-labeled Cu  
 361 concentrations in clams and standard deviation of triplicates. Curves show TK model  
 362 predictions based on local fits (using condition-specific  $k_{\text{DGT}}$  values) or an overall fit from a  
 363 BLM-based TK model.

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45  
 365 To address these limitations, we developed a biotic ligand model (BLM)-based TK  
 46 framework, modeling the Cu uptake rate ( $J_{\text{in}}$ ,  $\mu\text{g g}^{-1} \text{d}^{-1}$ ) as a function of  $C_{\text{DGT}}$  using a  
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 49 Michaelis-Menten-type relationship ([SI Note S7](#) for model derivation):  
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$$J_{\text{in}}(t) = \frac{J_{\max} \cdot K^{\text{CuBL}} \cdot C_{\text{DGT}}(t)}{1 + K^{\text{CuBL}} \cdot C_{\text{DGT}}(t)} \quad (7)$$
  
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56  
 369 where  $J_{\max}$  ( $\mu\text{g g}^{-1} \text{d}^{-1}$ ) is the maximum Cu uptake rate, and  $K'_{\text{CuBL}}$  ( $\text{L } \mu\text{g}^{-1}$  or  $\text{L mol}^{-1}$ ) is the  
 57 conditional Cu-ligand binding constant.  
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4 371 To account for the metabolically inert background Cu ( $C_{\text{int}}^{\text{bkg}}$ ) and the exchangeable Cu  
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6 372 pool ( $C_{\text{int}}^{\text{exch}}$ ), the total tissue ( $C_{\text{int}}^{\text{tot}}$ ) was modeled as:  
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9 373 
$$\frac{dC_{\text{int}}^{\text{exch}}}{dt} = J_{\text{in}} - k_e \cdot C_{\text{int}}^{\text{exch}} \quad (8)$$
  
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12 374 
$$C_{\text{int}}^{\text{tot}} = C_{\text{int}}^{\text{exch}} + C_{\text{int}}^{\text{bkg}} \quad (9)$$
  
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15 375 where  $C_{\text{int}}^{\text{bkg}}$  ( $\mu\text{g g}^{-1}$ ) is set as the mean  $[\text{Cu}]_{\text{non-lab}}$  at the beginning of the sediment exposure  
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17 376 phase.  
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20 377 The unified sediment TK model provided an “overall fit” across all exposure conditions,  
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22 378 accurately capturing the relationship between  $J_{\text{in}}$  and  $C_{\text{DGT}}$  ( $\text{NSE} = 0.94$ , [SI Figure S12](#)) and  
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24 379 reproducing  $[\text{Cu}]_{\text{non-label}}$  (overall fit, [Figure 4](#)). The “overall fit” model demonstrated  
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26 comparable performance ( $R^2 = 0.86$  vs. 0.89; [SI Figure S13](#)), offering broader applicability  
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28 380 across sediment conditions.  
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32  
33 382 This model yielded parameter values at  $J_{\text{max}} = 20.2 \mu\text{g g}^{-1} \text{ d}^{-1}$  and  $K'_{\text{CuBL}} = 0.0075 \text{ L}$   
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35 383  $\mu\text{g}^{-1}$  (equivalent to  $10^{5.7} \text{ L mol}^{-1}$ ), which are generally lower than values reported in water-  
36  
37 based BLM studies (see [SI Note S7](#) for comparison). This discrepancy may be due to our use  
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39 384 of DGT-labile Cu rather than the free  $\text{Cu}^{2+}$  activity used in water-based studies, which is  
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41 385 typically two orders of magnitude lower than measured dissolved Cu concentrations.  
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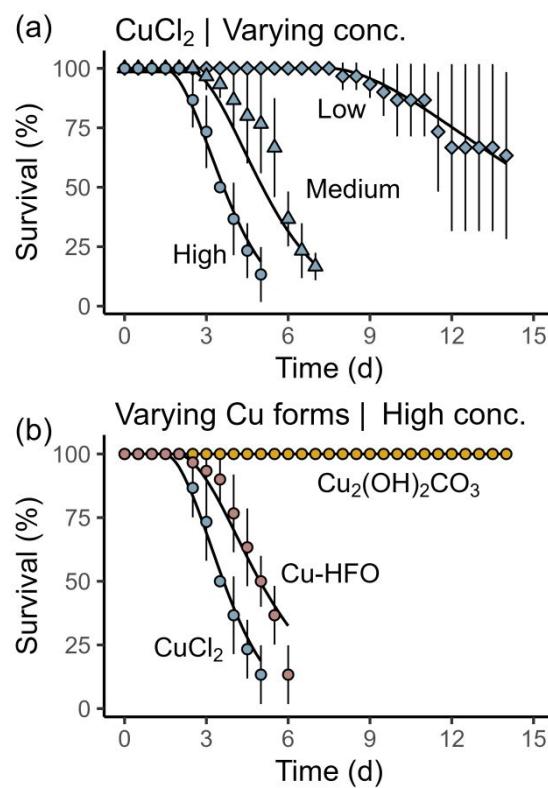
45 387 *3.4.3 Cu toxicity*  
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48 388 Clam survival varied with exposure concentrations and Cu speciation ([Figure 5](#)).  
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50 389 Significant mortality occurred in all  $\text{CuCl}_2$  treatments and in the Cu-HFO-High treatment,  
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52 390 while other treatments showed no observable toxicity ([SI Figure S14](#)). In toxic treatments,  
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54 391 mortality followed a time-dependent pattern, with survival remaining at 100% initially,  
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4 392 followed by a continuous decline over time. Higher Cu exposures led to earlier onset and  
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7 393 faster progression of mortality.  
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9 394 To quantitatively describe the mortality dynamics, we integrated the sediment TK model  
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11 395 with a toxicodynamic (TD) model. Model predictions successfully captured the temporal  
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13 396 toxicity profiles observed in the CuCl<sub>2</sub> and Cu-HFO treatments (NSE = 0.95). The model  
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15 397 used a unified set of parameters: an internal threshold concentration ( $C_{IT} = 30.5 \mu\text{g g}^{-1}$ ) and a  
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17 398 mortality rate constant ( $k_m = 0.0212 \text{ g } \mu\text{g}^{-1} \text{ d}^{-1}$ ). When these parameters were applied to the  
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19 399 non-toxic treatments, the model correctly predicted the absence of mortality, further  
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21 400 validating the TD model ([SI Figure S14](#)).

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23 401 The relatively low  $C_{IT}$  value suggests that *R. philippinarum* was sensitive to Cu  
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25 402 accumulation, consistent with field reports where maximum tissue concentrations in *R.*  
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27 403 *philippinarum* clams were near or below this threshold, ranging from < 4 to < 38  $\mu\text{g g}^{-1}$ .<sup>53-57</sup>  
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29 404 Laboratory bioaccumulation experiments also reported maximum tissue concentrations below  
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31 405 this threshold (< 30  $\mu\text{g g}^{-1}$ )<sup>58</sup> or slightly above it (< 44  $\mu\text{g g}^{-1}$ ).<sup>59</sup> This contrasts with other  
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33 406 benthic clam species in similar habitats showing higher Cu tolerance, such as the razor clam  
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35 407 *Sinonovacula constricta* ( $C_{IT} = 104-140 \mu\text{g g}^{-1}$ )<sup>60</sup> and the estuarine clam *Potamocorbula*  
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37 408 *laevis* ( $C_{IT} = 96$  and 141  $\mu\text{g g}^{-1}$  in separate studies).<sup>20,61</sup>  
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**Figure 5.** Clam survival varied with sediment Cu concentrations and forms. (a) Effect of Cu exposure concentrations. (b) Effect of Cu forms. Solid points and error bars show mean survival and standard deviation of triplicates. Curves represent model fits.

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411 **Figure 5.** Clam survival varied with sediment Cu concentrations and forms. (a) Effect of Cu  
412 exposure concentrations. (b) Effect of Cu forms. Solid points and error bars show mean  
413 survival and standard deviation of triplicates. Curves represent model fits.

### 414 **3.5 Deriving kinetic bioavailability benchmarks with the sediment TKTD model**

#### 415 *3.5.1 Estimating DGT-Cu-based LC<sub>50</sub> in sediments*

416 Using the TKTD model developed in this study (Table 1), we predicted the sediment  
417 DGT-labile Cu concentration expected to cause 50% mortality (i.e., DGT-Cu based LC<sub>50</sub>)  
418 across various exposure durations (R code provided in [SI Note S8](#)).

419 Initially, no 50% mortality occurred within the first 3 days due to Cu uptake being  
420 constrained by the maximum uptake rate (i.e., saturation effect). As exposure duration  
421 increased, the predicted LC<sub>50</sub> values declined sharply, from 3750 µg L<sup>-1</sup> at 4 days, to 231 µg  
422 L<sup>-1</sup> at 6 days, and eventually stabilized around 46 µg L<sup>-1</sup> after 20 days. This steady-state  
423 threshold is thus referred to as LC<sub>50</sub>-chronic value ([Figure 6a](#)).

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4 424 The predicted 6-day LC<sub>50</sub> (231 µg L<sup>-1</sup>) matched the empirically derived value 155 ± 30  
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6 425 µg L<sup>-1</sup> ([Figure 2b](#)). However, the predicted 4-day LC<sub>50</sub> (3750 µg L<sup>-1</sup>) was higher than the  
7  
8 426 empirical value of 1160 ± 320 µg L<sup>-1</sup>, likely due to insufficient dose groups, which  
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10 427 undermined the precision of the static assessments.  
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14 428 Standard toxicity tests typically use fixed exposure durations (e.g., 10-d) to derive acute  
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16 429 endpoints,<sup>39</sup> which differ with duration and hinder cross-study comparisons. Chronic  
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18 430 endpoints indirectly extrapolated from these acute values face the same limitation. In  
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20 431 contrast, the TKTD model provides a time-independent, mechanistically grounded chronic  
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22 432 endpoint—the DGT-Cu-based LC<sub>50</sub>-chronic = 46 µg L<sup>-1</sup> for *R. philippinarum*. This metric  
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24 433 can be integrated into broader ecological risk assessments, such as constructing species  
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26 434 sensitive distribution curves, to estimate hazardous concentrations that protect a defined  
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28 435 proportion of species (e.g., HC<sub>5</sub> for 95% species protection).  
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**Table 1.** Parameters of sediment TKTD Model

Parameter	Unit	Equations or values
$J_{in}$	µg g <sup>-1</sup> d <sup>-1</sup>	$J_{in} = \frac{J_{max} \cdot K^{CuBL} \cdot C_{DGT}}{1 + K^{CuBL} \cdot C_{DGT}}$
$J_{max}$	µg g <sup>-1</sup> d <sup>-1</sup>	20.2 ± 5.9
$K'_{CuBL}$	L µg <sup>-1</sup>	0.0075 ± 0.0013
$k_e$	d <sup>-1</sup>	0.152 ± 0.007
$C_{IT}$	µg g <sup>-1</sup>	30.5 ± 0.2
$k_m$	g µg <sup>-1</sup> d <sup>-1</sup>	0.0212 ± 0.0005

### 438 3.5.2 Estimating DGT-Cu-based NEC in sediments

The DGT-Cu-based no-effect concentration (NEC<sub>DGT</sub>) represents the highest DGT-labile

440 Cu concentration in sediment that does not cause mortality under prolonged exposure. Within

441 the TKTD framework, NEC<sub>DGT</sub> is estimated under two assumptions: (1) Cu bioaccumulation

442 reaches steady state ( $\frac{dC_{int}(t)}{dt} = 0$ ), and (2) the internal Cu concentration remains below the

443 toxicity threshold ( $C_{int} \leq C_{IT}$ ).  $NEC_{DGT}$  is obtained by solving the equation:

$$C_{\text{IT}} = \frac{J_{\max} \times K^{\text{CuBL}} \times \text{NEC}_{\text{DGT}}}{(1 + K^{\text{CuBL}} \times \text{NEC}_{\text{DGT}}) \times k_e} \quad (10)$$

445 Using model parameter values (Table 1), the NEC for DGT-labile Cu was estimated at

446 40 µg L<sup>-1</sup> (Figure 6b). Below this threshold, Cu exposure is predicted to not cause any

447 mortality in *R. philippinarum*.

448 This NEC pertains specifically to survival and may not protect against sublethal effects

449 like impaired growth or reproduction, which often occur at lower concentrations.

450 Nonetheless, this survival-based NEC, combined with the operational simplicity and time-

451 integrating capability of the DGT technique, serves as a valuable screening tool.

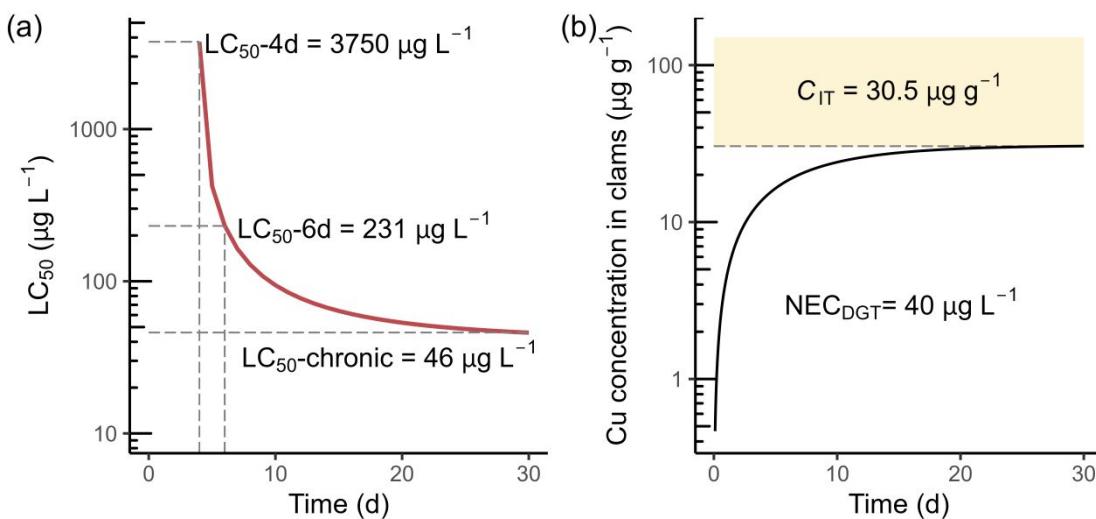
If NEC<sub>DGT</sub> are established for more sensitive species or conservation priorities (e.g.,

453 endangered species) this metric can guide early-stage site assessments, flagging high-risk

454 sediments for further management. Thus, this TKTD-derived NEC bridges mechanistic

<sup>455</sup> toxicology with practical sediment quality diagnostics, enabling more biologically relevant

457



**Figure 6.** Kinetic bioavailability benchmarks from sediment TKTD models for *R. philippinarum*, including DGT-Cu based chronic median lethal concentration (LC<sub>50</sub>-chronic) and no-effect concentration (NEC<sub>DGT</sub>). (a) Time-varying LC<sub>50</sub> curve; (b) Derivation of NEC<sub>DGT</sub> (40 µg L<sup>-1</sup>). Black curve shows Cu accumulation in clams ( $C_{\text{int}}$ ), remaining below the internal threshold level ( $C_{\text{IT}}$ ) during prolonged exposure to DGT-labile Cu concentrations of 40 µg L<sup>-1</sup>.

### 3.6 Bioavailability as a kinetic process: implications for sediment quality benchmarking

Environmental risk assessments typically assume equilibrium conditions,<sup>63</sup> yet

contaminant exposure in real-world is dynamic. Traditional methods for characterizing

sediment contaminant bioavailability, such as porewater analysis or solid-phase extraction,

are widely used for their practicality.<sup>9</sup> However, these methods often oversimplify the

temporal complexity of exposure.

Static assessments, such as short-term bioassays, are constrained by their fixed-time

design and often fail to capture the dynamic nature of contaminant accumulation and toxicity.

The threshold values (ECx) derived from these assessments are dependent on the exposure

duration and may not remain relevant under prolonged exposure.<sup>64–66</sup> Furthermore,

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4 476 bioaccumulation tests alone cannot identify the internal contaminant dose that trigger  
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6 477 toxicity. In our study, a tissue Cu concentration above  $\sim 35 \mu\text{g g}^{-1}$  was associated with over  
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8 478 50% mortality by day 4 ([Figure 2a and 2b](#)). However, this threshold could vary across  
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10 479 different exposure scenarios due to changes in subcellular partitioning patterns, which  
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12 480 influence the proportion of metabolically available versus detoxified Cu.<sup>67</sup> Without  
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14 481 concurrent toxicity testing, these critical thresholds remain unclear. This disconnect may  
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16 482 explain why tissue residue data are underused in sediment risk assessments.<sup>39,50</sup>

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23 483 To address these gaps, we developed a sediment TKTD model that mechanistically  
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25 484 reframe bioavailability as a dynamic, kinetic property. Using DGT-labile Cu to represent the  
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27 485 bioavailable Cu pool, along with a stable-isotope tracing technique, we derived kinetics-  
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29 486 based chronic bioavailability benchmarks, such as DGT-labile Cu-based LC<sub>50</sub>-chronic and  
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31 487 NEC<sub>DGT</sub>. These new benchmarks are grounded in mechanistic processes and account for  
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33 488 prolonged exposure effects, providing more informative metrics than those indirectly  
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35 489 estimated in conventional studies. Notably, the DGT technique integrates time-dependent  
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37 490 measurements of labile metals, combining kinetic relevance with operational simplicity. This  
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39 491 ensures that the derived LC<sub>50</sub> and NEC values are both scientifically robust and compatible  
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41 492 with existing monitoring protocols,<sup>33</sup> making them suitable for inclusion in regulatory  
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43 493 sediment quality assessments.

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53 494 Although our model implicitly assumes constant exposure, it is not a necessity, and the  
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55 495 model can be adaptable to more variable conditions, such as variations in salinity,<sup>45</sup>  
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57 496 temperature,<sup>68</sup> or hydrodynamics<sup>69</sup>, all of which influence metal chemistry and organism

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4 497 physiology. By establishing relationships between these variables with TKTD parameters, the  
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6 498 model can offer more accurate insights into bioavailability dynamics, supporting the  
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8 499 development of context-sensitive, bioavailability-based sediment quality benchmarks that  
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10 500 reflect both exposure duration and environmental variability.<sup>17</sup>  
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15 501 This study introduces a mechanistic framework for assessing the risks posed by metal-  
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17 502 contaminated sediments. The stable-isotope tracing technique generates high-quality data for  
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19 503 resolving bioaccumulation kinetics in sediment environments, while DGT measurements  
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21 504 simplify exposure scenarios, enabling direct modeling without the complexity of multiple  
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23 505 exposure routes. The model's time-integrated, process-based benchmarks enhance the kinetic  
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25 506 understanding of bioavailability, offering a scientifically grounded concept of bioavailability  
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27 507 highly relevant for environmental management in dynamic estuarine ecosystems.  
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37 509 **Notes**  
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39 510 The authors declare that they have no known competing financial interests or personal  
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41 511 relationships that could have appeared to influence the work reported in this paper.  
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