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Running head: MMME Comparison of Four Modeling Approaches

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METAL MIXTURE MODELING EVALUATION PROJECT: 2. COMPARISON OF FOUR MODELING APPROACHES

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ABSTRACT

2 As part of the Metal Mixture Modeling Evaluation (MMME) project, models were developed by the National Institute of Advanced Industrial Science and Technology (Japan), the 3 U.S. Geological Survey (USA), HDR HydroQual, Inc. (USA), and the Centre for Ecology and 4 5 Hydrology (UK) to address the effects of metal mixtures on biological responses of aquatic organisms. A comparison of the 4 models, as they were presented at the MMME Workshop in 6 7 Brussels, Belgium (May 2012), is provided herein. Overall, the models were found to be similar 8 in structure (free ion activities computed by WHAM; specific or non-specific binding of metals/cations in or on the organism; specification of metal potency factors and/or toxicity 9 10 response functions to relate metal accumulation to biological response). Major differences in 11 modeling approaches are attributed to various modeling assumptions (e.g., single versus multiple 12 types of binding site on the organism) and specific calibration strategies that affected the selection of model parameters. The models provided a reasonable description of additive (or 13 nearly additive) toxicity for a number of individual toxicity test results. Less-than-additive 14 toxicity was more difficult to describe with the available models. Because of limitations in the 15 available datasets and the strong inter-relationships among the model parameters (log K_M values, 16 potency factors, toxicity response parameters), further evaluation of specific model assumptions 17 and calibration strategies is needed. 18

19

Key words (<5 words): Biotic ligand model, Concentration addition, Metal bioavailability,
Metal toxicity, WHAM-F_{TOX}

23 Note to the editor and reviewers: This is one of 11 manuscripts under consideration for an ET&C Special Section on Metal Mixtures. The Section includes an introduction, a technical 24 background, a comparison of multiple modeling approaches, a lessons-learned manuscript, and 25 26 seven manuscripts on specific modeling and interpretation approaches. While each manuscript should be able to stand alone, the individual manuscripts are interrelated and cross-reference 27 each other. If another cross-referenced, submitted manuscript is essential to complete the review 28 29 of the present manuscript, please request the other manuscript from the Corresponding Guest Editor, copying the handling editor. The Corresponding Guest Editor for the series is Eric Van 30 Genderen (evangenderen@zinc.org). Any unpublished material provided to assist your review 31 must also be treated in confidence. 32

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INTRODUCTION

In regulatory applications, metal-mixture toxicity has generally been modeled by Toxic Unit 35 (TU) or other additive approaches that are based on water-exposure concentrations [1]. Data 36 reviews [2,3] have shown that additive approaches based on dissolved-metal concentrations are 37 not always sufficient in predicting mixture toxicity. Rather, metal-mixture toxicity tests have 38 shown a wide range of organism responses with no clear patterns in additive and non-additive 39 behavior. As part of an effort to address metal-mixture toxicity, several quantitative models have 40 been developed to evaluate responses of aquatic organisms to metal mixtures and ultimately to 41 provide a priori predictions of toxicity. Two modeling frameworks have been considered for this 42 purpose: the Biotic Ligand Model (BLM) (as first presented by Di Toro et al. [4]) and WHAM-43 44 F_{TOX} [5].

45 In the BLM framework, metal bioavailability is evaluated by considering competitive interactions of metals and cations for binding to dissolved organic matter (DOM) and inorganic 46 ligands (e.g., HCO₃⁻, CO₃²⁻, Cl⁻) using the Windermere Humic Aqueous Model (WHAM). 47 Competitive binding of metals and cations is also assumed to occur at binding sites on or in 48 biological organisms, which are referred to as the "biotic ligand(s)." The accumulation of metal 49 on the biotic ligand is then correlated to the toxic response of the organism (e.g., using a logit 50 response function). The BLM has been used by various investigators over the past decade to 51 develop predictive models for acute and chronic toxicity in single-metal exposures [6,7]. More 52 recently, several BLMs have been developed or revised for metal mixtures [8,9,10,11]. In these 53 models, metals have either been assumed to exhibit similar joint action (with toxicity expressed 54 in terms of concentration addition of metal accumulation on the biotic ligand) [8,9,10], or 55 56 independent joint action (with toxicity expressed in terms of a multiplicative function of the responses to the individual metals) [11]. 57

WHAM-F_{TOX} [5] was specifically developed to address the effects of metal mixtures on 58 aquatic organisms. This approach uses WHAM to evaluate competitive interactions of metals 59 and cations on DOM and inorganic ligands. In contrast to the BLM, WHAM-F_{TOX} does not 60 explicitly consider competitive binding of metals and cations to a biotic ligand. Rather, the 61 model assumes that non-specific accumulation of metabolically-active metals by the organism is 62 proportional to metal concentrations predicted to accumulate on humic acid (HA, when exposed 63 to the same exposure water (as calculated by WHAM). Accumulated metal is related to toxicity 64 using the F_{TOX} function, which is obtained by multiplying the calculated humic-bound metal and 65 proton concentrations (mmol g⁻¹) by cation-specific potency factors and then summing the results 66 67 over all cations. The resulting F_{TOX} value is correlated to the toxic response of the organism

using a linear-threshold response function, in which the effect thresholds depend only on theorganism (i.e., independent of water chemistry).

The Metal Mixture Modeling Evaluation (MMME) project was initiated to assess the current 70 capabilities of BLM- and WHAM-F_{TOX}-type models in predicting metal-mixture toxicity and to 71 promote the continued development of various modeling approaches [12]. As part of that 72 initiative, models were developed/refined and tested by researchers from (1) the National 73 Institute of Advanced Industrial Science and Technology (AIST) in Ibaraki, Japan; (2) the U.S. 74 Geological Survey (USGS) in Seattle, Washington and Boise, Idaho, USA; (3) 75 HDR HydroQual, Inc. (HDR) in East Syracuse, New York, USA; and (4) the Centre for Ecology 76 77 and Hydrology (CEH) of the Natural Environment Research Council in Lancaster, UK. The 78 purpose of this paper is to summarize and compare the 4 models, as they were presented at the MMME Workshop in Brussels, Belgium in May 2012. Revised versions of the initial AIST, 79 USGS, HDR and CEH models are presented in this issue [8,9,10,11,13]. Because the published 80 version of the USGS model differed substantially from the 2012 version evaluated in this article 81 and because the 2012 version of the USGS model is not otherwise available, it is included for 82 reference in Supporting Information File SI-2. 83 84

85

MODEL DESCRIPTIONS

The 4 models presented at the MMME Workshop were based on previous developments of the BLM and WHAM-F_{TOX}. Although the models were based on differing frameworks, they shared many similarities in their overall structure. These similarities included:

89	• A chemical speciation calculation to compute the free ion activities of metals and major
90	cations based on competitive binding to DOM and inorganic ligands using various
91	versions of WHAM.
92	• An evaluation of competing binding of metals and major cations to 1 or more binding
93	sites on an organism using conventional competitive equilibrium chemistry (by either
94	considering biotic ligand(s) or using metal binding to HA as a surrogate for non-specific
95	metal accumulation by the organism).
96	• A correlation of accumulated metal to toxicity using potency factors and/or toxicity-
97	response functions.
98	However, the 4 models differed in the details of their formulations and in calibration procedures
99	that were used by the different modeling groups in fitting the MMME project datasets. Those
100	datasets are listed in Table 1; formulations of the 4 models are summarized in Table 2, and
101	further details are provided below.
102	AIST model
103	The AIST model followed the BLM framework and considered a single biotic ligand as the
104	binding site for all metals on the organism [8,14]. Free ion activities of metals and other cations
105	were calculated using WHAM VII [15] . For this calculation, DOM was assumed to be 100%
106	fulvic acid (FA) and was set directly equal to the reported dissolved organic carbon (DOC)
107	concentration. Thus, conversion from DOC (mg L^{-1}) to the WHAM input for FA (g L^{-1}) was FA
108	= $0.001 \times \text{DOC}$. Carbonate species were not included in the calculations. WHAM-calculated
109	free ion activities were then used in computing competitive binding of metal(s) and major cations
110	on the biotic ligand. Initial estimates of binding constants (log K_M values) for metals and major

cations to the biotic ligand were obtained from previous studies (e.g., for trout [16,17]; for *Daphnia magna* [18,19]; for *D. pulex* [20,21]).

Toxicity was expressed as a function of the fractional coverage of accumulated metal on the 113 biotic ligand. Following the premise that toxicity is not caused by a metal-induced response but 114 rather by the role of the metal in blocking Ca uptake sites [14], all metals were assumed to elicit 115 equally potent toxicological responses when bound to the biotic ligand. According to this 116 assumption, metals exhibit similar joint action and metal-mixture toxicity can be described by a 117 concentration-additive approach based on the accumulation of total metal on the biotic ligand. 118 Response functions considered in the AIST model included a 2-parameter logit (mortality) 119 function for rainbow trout (Oncorhynchus mykiss), cutthroat trout (O. clarkii) and daphnids, and 120 a 2-parameter linear (growth reduction) function for a freshwater alga (Pseudokirchneriella 121 122 *subcapitata*). For example, the 2-parameter logit function is given as:

123
$$R = \frac{1}{1 + e^{-(a + b \theta_M)}}$$
(1)

where R is the biological response (e.g., fractional mortality or growth reduction), a and b are the 124 logit parameters, and θ_M is the fractional coverage of accumulated metals on the biotic ligand. 125 The AIST model was calibrated using 4 of the 6 MMME calibration datasets (Table 1). For 126 127 each dataset, the model was fit to observed mortality (or growth-reduction) responses by adjusting the logit response-function parameters. For D. magna and algae, log K_M values were 128 129 also adjusted, with a different set of log K_M values used for each organism (Supporting 130 Information File SI-1, Tables S1 and S2). Metal-mixture toxicity was predicted using the calibrated log K_M values and the response-function parameters derived from single-metal 131 132 exposure studies.

133 USGS model

134 The 2012 version of the USGS model also followed the BLM framework and considered a single biotic ligand as the binding site on the organism (see Supporting Information File SI-2). 135 Free ion activities of metals and cations were calculated using WHAM VII and the following 136 137 assumptions: DOM was specified as 2 times the reported DOC concentration, 65% of the DOM was assumed to be active, and DOM was considered to be composed of 10% HA and 90% FA. 138 Thus, conversions from DOC (mg L^{-1}) to WHAM inputs for HA and FA (in g L^{-1}) were: HA = 139 $2 \times 0.65 \times 0.1 \times 0.001 \times DOC$ and FA = $2 \times 0.65 \times 0.9 \times 0.001 \times DOC$. In addition, carbonate species 140 141 were included in the calculations (by specifying pH and alkalinity). For field-collected water samples, free ion activities of Al^{3+} and Fe^{3+} were assumed to be in equilibrium with amorphous 142 143 iron and aluminum hydroxides using solubility relationships [22,23]. WHAM-calculated free 144 ion activities were then used in computing competitive binding of metals and major cations on a single-site biotic ligand. Binding constants (log K_M values) for metals and major cations to the 145 biotic ligand were determined from a re-evaluation of data from single-metal toxicity studies on 146 147 rainbow and cutthroat trout (Supporting Information File SI-1, Table S3). The log K_M values 148 remained constant and did not differ among biological species.

Toxicity was expressed as a function of the fractional coverage of accumulated metal on the 149 biotic ligand. In contrast to the AIST model, metals were assumed to have different potencies 150 151 when bound to the biotic ligand. This effect was included in the model by incorporating a TOX 152 function to account for apparent differences in toxicities of the various metals. In this approach, the TOX function is conceptually similar to the F_{TOX} function [5]. However, 1 major difference 153 is that the TOX function is related to the fractional coverage of metal on the biotic ligand, 154 155 whereas F_{TOX} is expressed as a function of non-specific accumulation of metal on HA (in mmole g⁻¹). 156

Toxic response was then determined in 2 steps. First, a potency factor (α_i) was defined to account for the relative toxicity of different metals when bound to the biotic ligand. This factor was multiplied by the fraction coverage of metal on the biotic ligand (θ_i) to calculate the toxic potency of a specified metal, as defined by the *TOX* function:

161
$$TOX_i = \alpha_i \theta_i$$
 (2)

The model was extended to metal mixtures using a concentration-addition type of approach.
However, because each metal was considered to exhibit a different potency when bound to the
biotic ligand, calculations were based on the summation of *TOX_i* values.

165
$$TOX = \sum_{i=1}^{n} TOX_i = \sum_{i=1}^{n} \alpha_i \theta_i$$
(3)

where *n* is the number of metals in the mixture. Second, a 3-parameter logit function was used to calculate biological response as a function of TOX.

168
$$R = \frac{1}{\left(1 + e^{-(\beta_1 + \beta_2 \log TOX)}\right)^{\beta_3}}$$
(4)

169 where R is the biological response (e.g., fractional mortality or growth reduction), and β_1 , β_2 , and β_3 are the logit parameters. The model was calibrated using 5 of the 6 MMME calibration 170 datasets (Table 1). For each dataset, the model was fit to observed mortality (or growth-171 172 reduction) responses from single-metal and metal-mixture exposures by adjusting potency factors (α_i) and the 3 logit parameters (β_1 , β_2 , β_3 ; Supporting Information File SI-1, Table S4). 173 174 The potency factors were assumed to be dependent only on the metal, and the logit parameters 175 were considered to be organism-specific in the initial calibration of the model. However, it was necessary to consider organism-specific potency factors in fitting datasets for P. subcapitata 176 growth. 177

178 HDR model

The HDR model extended the BLM approach by considering a separate biotic ligand for each 179 metal (i.e., a Cd-specific, Cu-specific, Pb-specific, and Zn-specific biotic ligand). All metals 180 181 could compete for all binding sites, but only 1 metal was considered to be toxicologically active at a given biotic ligand. The remainder of the model followed BLM calculations for single-metal 182 exposures [24]. Within the HDR model, metal and cation binding to DOM and inorganic ligands 183 were calculated based on WHAM V [25] and the following assumptions: DOM was specified as 184 2 times the reported DOC concentrations, 100% of the DOM was assumed to be active, and 185 DOM was considered to be composed of 10% HA and 90% FA. Thus, conversions from DOC 186 (mg L⁻¹) to WHAM inputs for HA and FA (in g L⁻¹) were: HA = $2 \times 0.1 \times 0.001 \times DOC$ and FA = 187 $2 \times 0.9 \times 0.001 \times DOC$. In addition, carbonate species were included in the calculations (by 188 specifying pH and alkalinity). Because HDR did not explicitly consider the effects of Al or Fe 189 binding. Al^{3+} and Fe^{3+} were not included in the model. 190 In addition to metal and cation binding to DOM and inorganic ligands, the HDR model also 191 192 included simultaneous calculations for metal and cation binding to each of the metal-specific biotic ligands. For each biotic ligand, $\log K_{\rm M}$ values for the toxicologically-active metal and 193 competing cations were obtained from previously-calibrated, single-metal BLMs [24]. For the 194 remaining metals that were not toxicologically-active at a given biotic ligand but could compete 195 for binding to the site, log K_M values were initially set equal to their log K_M values on their 196 toxicologically-active biotic ligands. For example, the log K_M value for Cd on the Cu-specific 197 biotic ligand was set equal to the log K_M value for Cd on the Cd-specific biotic ligand. However, 198 adjustments in some of the log K_M values were made during model calibration. The final log K_M 199 200 values for the HDR model are presented in Supporting Information File SI-1, Table S5.

Toxic response at each of the biotic ligands was correlated to the concentration of metal *'i'*on its toxicologically-active biotic ligand using a 2-parameter logit function:

203
$$R_i = \frac{1}{1 + e^{-(a_i + b_i \log v_i)}}$$
(5)

where R_i represents the biological response (e.g., mortality) due to metal '*i*'; a_i and b_i are the metal-specific logit parameters; and v_i is the concentration of metal '*i*' on its toxicologicallyactive biotic ligand (in nmol g⁻¹). For metal mixtures, the overall response (*R*) was determined by assuming independent joint action and expressing toxicity in terms of a multiplicative function of individual metal responses as:

209
$$R = 1 - \prod_{i=1}^{n} (1 - R_i)$$
 (6)

where *n* is the number of metals in the mixture. This approach is also referred to as responseaddition (see [1] for further discussion).

The HDR model was calibrated using 3 of the 6 MMME calibration datasets (Table 1). The 212 model was calibrated separately for multiple data series within a given dataset by adjusting logit 213 intercepts (a_i) and slopes (b_i) to fit observed mortalities for single-metal exposures [11]. These 214 analyses ultimately provided a global fit of the logit slope and a distribution of logit intercepts 215 216 that were used to quantify the unexplained variance or uncertainty associated with the singlemetal exposure data (Supporting Information File SI-1, Table S6). Metal-mixture toxicity was 217 predicted using the log K_M values and calibrated logit parameters from single-metal exposures. 218 219 The HDR model also considered uncertainty in single-metal toxicity predictions to generate response envelopes for metal-mixture exposures (see [11] for details). Log K_M values for Cu and 220 221 Zn on the Cd-specific biotic ligand were subsequently adjusted to provide a better calibration of the model to the *D. magna* mortality data ("Index 4"). 222

223 CEH model

224 The CEH model was based on WHAM-F_{TOX} [5,26]. In this calculation, WHAM VI [27] was used to be consistent with previous applications of WHAM- F_{TOX} [5,28]. Dissolved organic 225 226 matter was specified as 2 times the reported DOC concentration, 65% of the DOM was assumed to be FA, and the remaining 35% of the DOM was considered to be inert with respect to 227 metal/cation binding. Thus, the conversion from DOC (mg L^{-1}) to WHAM inputs for FA (in g L^{-1}) 228 ¹) was: $FA = 2 \times 0.65 \times 0.001 \times DOC$. Carbonate species were included in the calculations (by 229 230 assuming that total carbonate concentrations were equal to the reported alkalinity). For fieldcollected water samples, free ion activities of Al³⁺ and Fe³⁺ were assumed to be in equilibrium 231 232 with amorphous iron and aluminum hydroxides using solubility relationships [22,29]. A small concentration of WHAM HA (e.g., 10^{-10} g/L) was also included in the calculation. The resulting 233 234 concentrations of metals and protons on the HA were then used in toxicity calculations. In these calculations, metal concentrations on HA included metals that were specifically-bound to HA 235 236 functional groups as well as metals that were nonspecifically-bound by electrostatic interactions 237 and held in close proximity to the HA (i.e., in the Donnan Layer). However, only protons that were specifically-bound to HA functional groups were included in toxicity predictions. Because 238 the CEH model considered a distribution of binding sites in WHAM HA (see [27]), 239 representative composite binding constants were calculated for illustrative purposes by taking the 240 weighted-averages of the K_M values (see Supporting Information File SI-1, Table S7 for details). 241 (Note that log K_M values were also modified for electrostatic corrections in WHAM-F_{TOX}.) 242 Toxic response was determined by assuming that concentrations of metabolically-active 243 244 metals and protons on or in the organism were proportional to their predicted concentrations on WHAM HA in the same exposure water. Toxicity was then determined in 2 steps. First, a 245

potency factor was defined to relate the amounts of accumulated metals and protons to toxic
effect using the F_{TOX} function:

248
$$F_{TOX} = \sum_{i=1}^{n+1} \alpha_i v_i$$
 (7)

where α_i is the relative potency factor, v_i is the concentration of metal and protons on humic acid (in mmol g⁻¹), *n* is the number of metals in the mixture, and *n*+1 is considered to account for proton toxicity. Second, a 2 parameter linear response function was defined to relate toxic response to the *F*_{TOX} function

$$R = 0 \qquad \qquad for \ F_{TOX} < F_{TOX-LT}$$

$$253 \qquad R = \frac{F_{TOX} - F_{TOX-LT}}{F_{TOX-UT} - F_{TOX-LT}} \qquad for \ F_{TOX-LT} \le F_{TOX} \le F_{TOX-UT}$$

$$R = 1 \qquad \qquad for \ F_{TOX} > F_{TOX-UT} \qquad (8)$$

where F_{TOX-LT} represents the lower threshold for toxicity and F_{TOX-UT} represents the threshold for the maximum toxic response.

The model was calibrated using all 6 MMME calibration datasets (Table 1). For each dataset, the model was fit to observed mortality (or growth-reduction) responses from singlemetal and metal-mixture exposures by adjusting relative potency factors (α_i) and the linear response parameters (F_{TOX-LT} , F_{TOX-UT} ; Supporting Information File SI-1, Table S8). The relative potency factor for H⁺ (α_H) was set equal to 1.0 in the calibration, effectively normalizing the potency of metals relative to that of H⁺. Adjustments in F_{TOX-LT} and F_{TOX-UT} values were also examined in evaluations of the *D. magna* mortality data ("Index 4").

264

METHODS

265 In the present study, model performance was examined by first re-computing results for the 4 models using specifications described in the 'Model Descriptions' section. This step served as 266 an independent check of results presented by the 4 modeling groups and also provided detailed 267 268 model outputs that were subsequently used in model-model comparisons. Three of the larger calibration datasets were considered for this purpose, including:(i) "Index 8", which consisted of 269 114 test results for *P. subcapitata* growth in field-collected water samples spiked with Cd, Cu, 270 Ni, and Zn (see Supporting Information File SI-3); (ii) "Index 4", which consisted of 561 test 271 results for *D. magna* mortality in reconstituted laboratory water spiked with Cd, Cu, and Zn [30]; 272 and (iii) "Index 6", which consisted of 369 test results for cutthroat and rainbow trout mortality 273 in field-collected water samples spiked with Cd, Pb, and Zn [31]. Two additional datasets were 274 considered for model validation: (i) "Index V-1", which consisted of 309 test results for D. 275 276 magna mortality in reconstituted laboratory water spiked with Cd and Zn [30]; and (ii) "Index V-3", which consisted of 96 test results for rainbow trout survival in reconstituted laboratory water 277 spiked with Cd, Cu, Zn [32]. 278

Calculations for the 4 models were performed as follows: For the AIST model, free ion
activities of metals and cations were calculated using WHAM VII [15,33]. The remainder of the
calculation was performed in Excel, with fractional coverage of metal on the biotic ligand
computed from the WHAM-calculated free ion activities and AIST-chosen log K_M values.
Biological responses (e.g., mortality, growth reduction) were then determined as a function of
the fractional coverage of metal(s) using the AIST response functions (Eqn. 1; Supporting
Information File SI-1, Tables S1 and S2).

A similar procedure was followed for evaluation of the USGS model. Free ion activities of metals and cations were calculated using WHAM VII. The fractional coverage of metal on the 288 biotic ligand, the TOX function, and toxicity were computed in Excel using the WHAMcalculated free ion activities and the USGS-chosen log K_M values, potency factors (α_i), and 289 response-function parameters (β_1 , β_2 , β_3) (Eqns. 2-4; Supporting Information File SI-1, Tables 290 291 S3 and S4). For evaluation of the HDR model, free ion activities were calculated using the TICKET 292 model [34] with the WHAM V database. Accumulation of metals and cations on each of the 293 biotic ligand sites (in nmol g(wet)⁻¹) were computed in Excel using the WHAM-calculated free 294 ion activities and HDR-chosen log K_M values (Supporting Information File SI-1, Table S5). 295 Toxic responses at each biotic ligand were determined using the HDR response-function 296 297 parameters (a_i, b_i) (Eqns. 5, 6; Supporting Information File SI-1, Table S6). Finally, for evaluation of the CEH model, free ion activities and concentrations of metals and 298 cations on HA were computed using WHAM VI [27]. Because the concentration of specifically-299 bound protons on HA was not included in the model output of the commercially-available 300 version of WHAM VI, concentrations of specifically-bound protons were estimated as the total 301 302 number of proton binding sites minus the surface charge (in equivalents per gram) minus 2 times the summation of specifically-bound metals and cations. In this calculation, metals and divalent 303 cations are assumed to primarily occupy bidentate and tridentate binding sites on the HA. 304 Toxicity was then computed in Excel using the WHAM-calculated concentrations on HA and the 305 CEH-determined potency factors (α_i) and response-function parameters (F_{TOX-LT} , F_{TOX-UT}) (Eqns. 306 7, 8; Supporting Information File SI-1, Tables S7 and S8). 307 Results for the 4 models were then plotted in comparable formats with mortality (or growth 308 309 reduction) on the vertical axis and fractional coverage on the biotic ligand (θ_M) for the AIST

model, *TOX* for the USGS model, and F_{TOX} for the CEH model on the horizontal axis. This

allowed metals for a large number of single-metal and metal-mixture exposure tests to be plotted and visually compared in a concise and convenient format. Unfortunately, this graphical format is not directly applicable to the HDR model because (i) toxicity response functions for individual metals were not considered to have common logit slopes, and (ii) metal-mixture toxicity was described by independent joint action using a response-additive approach. However, a reasonable comparison was provided by converting the HDR model-predicted responses (*R*) to equivalent *TOX* (*TOX_{equiv}*) values by rearranging Equation 5:

318
$$TOX_{equiv} = e^{-\left(\frac{2.303 \, a}{b}\right)} \cdot \left[\frac{R}{1-R}\right]^{\left(\frac{2.303}{b}\right)}$$
(9)

In this equation, R is the fractional mortality (or growth reduction) for single-metal or metal-319 mixture exposures, a and b are the log-logit intercept and slope, and 2.303 is included because 320 321 HDR used a mixed ln-log function in Equation 5 to describe toxicity. For subsequent calculations, TOX_{equiv} values were computed using response parameters (a, b) for Zn. This 322 effectively normalized results for other metals to the toxicity of Zn. The TOX_{eauiv} function is 323 most appropriate for mortality (or growth reduction) responses near 50%. For responses near 0% 324 325 or 100%, the TOX_{equiv} function is only a crude approximation for metals with log-logit response slopes that are different than Zn. Graphical comparisons of model coefficients and of model 326 results for individual metal-mixture exposure tests were also prepared and analyzed. 327

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- 329

RESULTS AND DISCUSSION

Observed mortality for *D. magna* and the corresponding AIST, USGS, HDR and CEH model-calibrated response curves are shown for single-metal (Figure 1A,C,E,G) and metalmixture exposures (Figure 1B,D,F,H) following the graphical formats described above. Similar

333 comparisons for rainbow trout mortality are presented in Figure 2. In most cases, modelcalibrated response curves described the central tendency of observed mortalities for both single-334 metal and metal-mixture exposures. However, scatter in the observed data around the model-335 336 calibrated response curves varied from model to model, with the USGS D. magna results (Figure 1C,D) and the AIST rainbow trout results (Figure 2A,B) showing the largest variations of 337 observed mortality about the model-calibrated response curves. Observed mortality data was 338 also relatively widely scattered around the HDR model-calibrated response curves, particularly 339 for observations near 0% and 100% mortality (Figure 1E,F; Figure 2E,F). However, these 340 differences in part can be attributed to the TOX_{eauiv} approximation (Eqn. 9) in which all metals 341 were assumed to have log-logit slopes similar to that of Zn. Finally, the observed mortality data 342 appeared most closely aligned to the CEH model-calibrated response curves for both D. magna 343 (Figure 1G,H) and rainbow trout (Figure 2G,H). 344 The graphical comparisons in Figures 1 and 2 provide an overview of how well the 4 models 345 were calibrated to observed mortality for 2 of the larger calibration datasets. Additional model-346 model comparisons were made by examining model fits to individual-metal results. For 347 example, USGS, HDR and CEH model results for D. magna mortality in a Cd-only toxicity test 348 (Index 4, Cu-Cd #7-1, with 12.6 µg/L dissolved Cd) are presented in Figure 3A. The 3 models, 349 which computed solution chemistry using 3 different versions of WHAM, predicted 350 approximately 35% of total dissolved Cd was free Cd. However, the fractional coverage of Cd 351 352 on the biotic ligand (or the HA surrogate in the CEH model) varied from 0.24% in the CEH model to 11% in the USGS model. Despite these large differences, the associated calibration of 353 potency factors and/or response-function parameters resulted in similar predictions of mortality 354 355 by the 3 models. This finding demonstrates the strong inter-relationship of $\log K_M$ values,

356 potency factors and response-function parameters, and underscores the latitude that exists in 357 calibrating model parameters with toxicity datasets that include only measures of total dissolved metal concentrations, water chemistry (pH, major ions, alkalinity, DOC) and a select 358 359 toxicological endpoint (e.g., mortality, growth reduction). A similar example is given in Figure 3B for AIST, HDR, and CEH model results for D. 360 magna mortality in a Cu-only toxicity test (Index 4, Cd-Cu #5-2, with 83.8 µg/L of dissolved 361 Cu). In this case, model predictions of free Cu varied from 0.042% to 2.6% of the total dissolved 362 Cu concentration. These differences were due to the version of WHAM that was used in the 363 calculations and to assumptions for DOC composition and carbonate chemistry that were 364 employed by the 3 modeling groups. Differences in model predictions for free Cu were reflected 365 in differences for Cu accumulation on the biotic ligand (or the HA surrogate in the CEH model). 366 367 Despite these large differences, calibration of potency factors and/or response-function parameters again resulted in similar predictions of mortality by the 3 models. This finding 368 demonstrates that the WHAM calculation can also have a large effect on the final calibration of 369 370 log K_M, potency factors, and response function parameters, but again the inter-relationships of the multiple calibration parameters allows for compensation of those differences to produce 371 similar overall predictions among the different modeling approaches. 372 Based on results presented in Figure 3, it is difficult to evaluate calibration strategies in a 373 simple step-by-step procedure. Rather, a more holistic view of the calibration process is needed 374 (see comparison of model calibration parameters in Table 3). In all 4 models, log K_M values for 375 the initial calibration were fixed based on previous studies and were not considered as adjustable 376

appeared to be most constrained, with response parameters allowed to be adjusted only as a

377

parameters. Based on the remaining model parameters, the initial calibration of the AIST model

379 function of the organism (Table 3). This was followed by the USGS model which allowed potency factors (α_i) to be adjusted as a function of only the metal, and response parameters (β_i , 380 β_2, β_3) to be adjusted as a function of only the organism in its initial calibration. Additional 381 flexibility was considered in the initial calibration of the HDR model, which allowed toxicity 382 response parameters (a_i, b_i) to be adjusted as a function of both metal and organism. The CEH 383 384 model provided similar flexibility by allowing potency factors (α_i) to be adjusted as a function of metal and organism, and by allowing small adjustments in response parameters (F_{TOX-LT} , F_{TOX-UT}) 385 386 as a function of only the organism.

The use of a more constrained or a more flexible calibration strategy had a significant 387 effect on the ability of the 4 models to describe mortality (or growth reduction) data (see Figures 388 389 1 and 2). To illustrate this point, USGS and CEH model results for growth reductions of P. subcapitata at pH 6 are given in Figure 4. As shown in Figure 4A, the growth in single-metal 390 and metal-mixture exposures were poorly described by the initial calibration of the USGS model, 391 which was based on potency factors (α_i) that were determined from global fits to the MMME 392 393 calibration datasets. As shown by the USGS modeling group (Supporting Information File SI-2), specification of a separate set of potency factors (α_i) for P. subcapitata at pH 6 produced a much 394 closer correspondence of the model-calculated response curve and observed growth reductions 395 (Figure 4B). By comparison, the CEH model still appears to provide a better description of the 396 observed growth reductions (Figure 4C). The reason is in part due to the inclusion of 397 specifically-bound protons in the CEH toxicity evaluation. For example, in the calculated 398 growth reductions for P. subcapitata at pH 6, proton toxicity accounted for 1.7±0.06 of the 399 400 computed F_{TOX} value (vertical gray bar in Figure 4C). This served to compress the effects of metals to the right of the vertical gray bar. Subtracting the proton contribution from F_{TOX} and re-401

plotting the model response curve and the observed growth reduction provided a different picture
of the variability that may be associated with metal accumulation in the organism (Figure 4D).
Therefore, excluding the extra factor of proton toxicity, which was included in the WHAM-F_{TOX}
calibration, would likely result in some added variability of observed responses around the
model-calculated response curve.

Next, the 4 models were further evaluated by comparing their final selection of model 407 parameters. Log K_M comparisons for Cd, Pb, and Zn were based on AIST values for rainbow 408 trout, USGS values that were previously determined from cutthroat trout and rainbow trout data, 409 HDR values for the Zn-specific biotic ligand, and average log K_M values for the distribution of 410 binding sites in the CEH model (Supporting Information File SI-1, Tables S1, S3, S5 and S7). 411 From a chemical perspective, the log K_M values for the HDR and CEH models followed an 412 expected increase in metal binding based on affinities of metals to oxygen donor groups on 413 organic acids (Cd < Zn < Pb; Figure 5A). In contrast, the AIST and USGS models had larger 414 binding constants for Cd that were similar to previously-reported log K_M values for Cd (e.g., 415 [35,36,37]). This stronger binding of Cd to biological ligands was attributed to active Ca 416 transport and ionic mimicry in fish gills [35]. Another possible explanation for larger Cd binding 417 418 constants may be that Cd is binding to sulfur (and not oxygen) donor groups in the organism. 419 Toxicity parameters in the 4 models were also compared by combining potency factors (α_i) and response function parameters into a single measure of the lethal accumulation at 50% 420 mortality (LA50) (see Supporting Information File SI-1, Tables S2, S4, S6, and S8). For the 421 comparison, LA50 values were expressed in terms of percent coverage on the biotic ligand or on 422 423 the surrogate HA binding sites for the WHAM-F_{TOX} model. The resulting LA50 values for rainbow trout exposed to Cd, Zn, and Pb ranged from 2 to 3% of the binding sites in the AIST 424

425 and USGS models (Figure 5B). In contrast, the LA50 values for the HDR and CEH models varied more (0.01% for Cd, 7% for Zn, and 2.9% for Pb for the HDR model; 0.05% for Cd, 9.8% 426 for Zn, and 22% for Pb for the CEH model). For comparison, experimentally-derived LA50 427 428 values for Cd, Zn and Pb in rainbow trout studies have ranged from 10% to 64% coverage of strong binding sites on the gill [36,37,38,39]. These values represent the higher end of the 429 model-calculated LA50 values and are not supportive of the very low LA50 values for Cd in the 430 HDR model. However, there are some questions regarding the appropriateness of comparing 431 experimentally-derived LA50 values (which are based on estimates for the density of strong 432 binding sites that have been reported to vary as a function of water chemistry and the specific 433 metal being examined) and model-calculated LA50 values (which are generally based on a 434 binding site density that is considered to be constant across all water chemistries and metals). 435 436 An alternative interpretation of experimentally-derived LA50 values is provided by considering the relative values of measured accumulations on a nmole per gram of fish gill basis. For 437 example, 24-h LA50 measurements for rainbow trout have been reported as 1.1 nmole g⁻¹ (ww) 438 for Cd and 32.8 nmole g⁻¹ (ww) for Pb [36]. This represents a difference of a factor of 30 in the 439 Cd and Pb accumulations on the gill that would elicit a 50% mortality response and is consistent 440 with the lower LA50 values for Cd that were computed by the HDR and CEH models. However, 441 a more appropriate comparison for the CEH model-calculated LA50 values would be body-442 burden measurements for the various metals. 443 A final check on model calibration was performed by examining individual series of mixture 444

toxicity test results for *D. magna* mortality (Index 4). For the Cu-Zn test series #5-4, increases in mortality were observed for *D. magna* exposed to a constant Cu concentration of 100 μ g/L and increasing Zn concentrations (Figure 6A). This behavior is similar to an additive response curve

(see Figure 1 in Meyer et al. [1]). Model-calculated response curves for the AIST, HDR and
CEH models were consistent with the observed trend, with the HDR model corresponding most
closely to the observed data. Differences in the AIST, HDR and CEH model-calculated curves
for the Cu-Zn mixture can be attributed to the calibration of the models and not to the differences
in their formulations.

The Cd-Cu test series #20-3 showed very different behavior (Figure 6B), with observed 453 mortalities for *D. magna* exposed to a constant Cd concentration of 19.5 µg/L and increasing Cu 454 concentrations following a less-than-additive (Case 2) response curve (see Figure 1 in Meyer et 455 al. [1]). The AIST model (which is based on concentration addition) did not predict the observed 456 decrease in mortality as Cu concentrations increased. However, both the HDR model (which is 457 based on independent joint action) and the CEH model (which is based on a FTOX-additive 458 459 approach) predicted a decrease and then an increase in *D. magna* mortality as the Cu concentration increased. Additional adjustments in the both HDR and CEH model calibrations 460 were required to fit the observed response. For the HDR model, this consisted of increasing the 461 log K_M value for Cu binding to the Cd-biotic ligand sites by 4 log units. For the CEH model, 462 global values of F_{TOX-LT} and F_{TOX-UT} that were reported for D. magna were adjusted by 463 optimizing the F_{TOX-LT} and F_{TOX-UT} to the test series #20 data. Although the reported adjustments 464 in F_{TOX-LT} and F_{TOX-UT} were not large (global fit: 1.88 and 2.95; test series #20 fit: 2.61 and 3.18), 465 model responses to the single metal and metal-mixture exposure tests were sensitive to the 466 changes (see Supporting Information File SI-1, Figures S25 and S26). Therefore, questions still 467 remain about how to formulate and calibrate models to reproduce observed responses exhibiting 468 less-than-additive (Case 2) behavior. 469

470	In addition to model calibration evaluations, the calibrated models were used in a blind
471	prediction of mortality for 2 validation studies (Index V-1; Index V-3). Comparisons of
472	observed mortality and model-calculated response curves for the D. magna validation study
473	(Index V-1) are presented in Figure 7. Model-data comparisons for single-metal exposures
474	(Figure 7A,C,E,G) were comparable to the <i>D. magna</i> calibration results presented in Figure 1.
475	Model-calculated response curves for metal mixtures tended to over-predict mortality by factors
476	of 1-2 on the θ_M scale for the AIST model (Figure 7B), factors of 2-5 on the <i>TOX</i> scale for the
477	USGS model (Figure 7D), approximately a factor of 2 on the TOX_{equiv} scale for the HDR model
478	(Figure 7F), and approximately a factor of 2 on the F_{TOX} scale for the CEH model (Figure 7H).
479	Similar results for the rainbow trout validation study (Index V-3) are shown in Supporting
480	Information File SI-1 (Figures S8, S15, S22 and S31). Model-data comparisons tended to show
481	more variability for single-metal exposures. For mixtures, model-calculated response curves
482	tended to over-predict trout mortality by a factor of 3 to 4 on the <i>TOX</i> scale for the USGS model.
483	Model-calculated response curves for the HDR and CEH models were in closer agreement to
484	observed mortality. The AIST model was not considered in the rainbow trout validation test
485	because a rainbow trout log K _M value for Cu was not provided for the AIST model.
486	The overall results of the present study highlighted similarities and differences in 4 models
487	that were developed to describe the effects of single-metal and metal-mixture exposures on
488	biological response (e.g., mortality, growth reduction). The 4 models were calibrated to
489	individual datasets that contained metal-exposure concentrations, water chemistry and biological
490	response data. Because measurements of metal accumulation on a representative biological site
491	were not available, independent calibration of log K_M , potency factors and response-function
492	parameters were not possible. Despite these limitations, calibration of models to single-metal

493 exposure data often provided a reasonable basis for predicting metal-mixture toxicity. This was 494 particularly true for metal-mixtures exhibiting additive (or near additive) behavior. The ability 495 of the models to reproduce less-than-additive behavior posed a greater challenge. This less-than-496 additive toxicity was predicted by the HDR model (which considered independent joint action, 497 but required substantial adjustment of some log K_M values to describe the observed behavior) 498 and by the CEH model (which considered F_{TOX} addition, but required adjustment of F_{TOX} 499 parameters to individual datasets).

These findings indicate that competitive interactions among metals add a level of complexity 500 to toxicity evaluations that will in all likelihood only be appreciated through continued model 501 development. The application of more complex geochemical models (with multiple biotic ligand 502 sites or distributions of log K_M binding sites) may be needed for this purpose. This has led to 503 504 revisions in the 4 modeling approaches, particularly for the AIST and USGS models that were considered in the present study. Revised versions of the AIST, USGS, HDR and CEH models 505 are described in various papers in this issue (see [8,9,10,11,13]). Model calibration remains a 506 507 key issue for the various modeling approaches. Therefore, a further evaluation of specific model assumptions and calibration strategies that were used by the 4 modeling groups is considered in 508 the following paper [40]. 509

510

511

SUPPORTING INFORMATION

512 File SI-1: Modeling parameters and additional analyses (Tables S1 to S8, and Fig. S1 to S31).

513 File SI-2: USGS model description (2012 version).

514 File SI-3: Description of "Index 7" and "Index 8" data sets.

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Indox	Species	Metal mixture / water type	Endpoint	Number of exposures ^b	Source -	Datasets used for model calibration ^c			
mdex						AIST	USGS	HDR	CEH
1	Hyalella azteca Lampsilis siliquoidea	Cd-Cu-Ni-Pb-Zn / Porewater	28-d survival	2/60 2/38	[41]		✓	\checkmark	\checkmark
4	Daphnia magna	Cd-Cu, Cu-Zn / Lab	48-h survival	387/174	[30]	\checkmark	\checkmark	\checkmark	\checkmark
6	Oncorhynchus mykiss	Cd-Pb-Zn / Field	96-h survival	298/71	[31]	\checkmark	\checkmark	\checkmark	\checkmark
7	Pseudokirchneriella subcapitata	Field mixture / Field	72-h growth	7/28	Present Study ^d	\checkmark	\checkmark		\checkmark
8	P. subcapitata	Cd-Cu-Ni-Zn / Field	72-h growth	102/12	Present Study ^d	\checkmark	\checkmark		\checkmark
9	Lactuca sativa	Ag-Cu, Cu-Zn / Hydroponic	4-d root growth	36/202	[42]				\checkmark
V-1	D. magna	Cd-Zn / Lab	48-h survival	132/177	[30]				
V-2	D. magna	Cd-Cu-Zn Lab	48-h survival	3/12	[30]				
V-3	O. mykiss	Cd-Cu-Zn Lab	96-h survival	72/24	[32]				

Table 1. Datasets used in the Metal Mixture Modeling Evaluation project^a

^a See [12] for detailed descriptions.
^b Single-metal or reference exposures / mixture exposures

^c AIST = National Institute of Advanced Industrial Science and Technology, Japan; USGS = U.S. Geological Survey, USA; HDR = HDR|HydroQual, Inc., USA; CEH = Centre for Ecology and Hydrology, UK

^d See Supplemental Information, File SI-3

 F_{TOX} additive^c

5		, , , ,	5	
	AIST	USGS	HDR	СЕН
Solution chemistry	WHAM VII	WHAM VII	WHAM V	WHAM VI
Metal/cation binding to organisms	Competitive binding of metals/cations to a single BL site	Competitive binding of metals/cations to a single BL site	Competitive binding of metals/cations to multiple BL sites	Non-specific accumulation of metals / cations at a distribution of binding sites
Toxicity	Function of fractional coverage of metal on BL	Function of potency and coverage of each metal on BL $(TOX_i = \alpha_i \ \theta_i)^b$	Function of potency and concentration of each metal on its toxicologically-relevant BL	Function of potency and concentration of protons and each metal on WHAM humic acid, assumed proportional to their binding on or in the organism ($F_{TOX} = \alpha_i \ v_i$) ^c
Toxic response	2-parameter logit (or linear) response function	3-parameter logit response function	2-parameter logit response function	2-parameter linear- threshold response function

Table 2. Summary of formulations for the AIST, USGS, HDR, and CEH metal-mixture-toxicity models^a

^a AIST = National Institute of Advanced Industrial Science and Technology, Japan; USGS = U.S. Geological Survey, USA; HDR = HDR|HydroQual, Inc., USA; CEH = Centre for Ecology and Hydrology, UK; BL = biotic ligand; WHAM V, VI, and VII = versions of Windermere Humic Aqueous Model.

Independent action

TOX additive^b

Concentration additive

^b *TOX* = toxicity-response function in USGS model; α_i = potency factor for metal *i*; θ_i = proportion of BL sites occupied by metal *i* (# of sites occupied / # total sites).

^c F_{TOX} = toxicity-response function in CEH model; α_i = potency factor for proton or metal *i*; ν_i = concentration of BL sites occupied by protons or metal *i* (mmol g⁻¹ humic acid).

Mixture response

	AIST	USGS	HDR	CEH
Binding constants $(log K_M)$	f(metal, organism) ^b	f(metal) ^c	f(metal) ^b	f(metal) ^c
Metal potency factors (α_i)	n/a	f(metal) ^d	n/a	f(metal, organism)
Proton potency factor (α_H)	n/a	n/a	n/a	f(organism)
Response parameters (β_1 , β_2 , β_3 , F_{TOX-LT} , F_{TOX-UT} , etc.)	f(organism)	f(organism)	f(metal, organism)	f(organism)

Table 3. Summary of parameters used in the AIST, USGS, HDR, and CEH metal-mixture-toxicity models^a

^a AIST = National Institute of Advanced Industrial Science and Technology, Japan; USGS = U.S. Geological Survey, USA; HDR = HDR|HydroQual, Inc., USA; CEH = Centre for Ecology and Hydrology, UK.

^b log K_M values for the AIST and HDR models were taken from previously-calibrated, single-metal biotic ligand models. Additional adjustments of log K_M values were made during their studies.

 $^{\rm c}~\log K_M$ values were held constant in the USGS and CEH models and were determined as follows:

USGS: from a re-evaluation of single-metal toxicity data for cutthroat and rainbow trout; CEH: from previous calibration for WHAM VI (a version of the Windermere Humic Aqueous Model), using WHAM humic acid as a surrogate for non-specific accumulation of protons and metabolically-active metals by the organism.

^d In the initial calibration of the USGS model, metal potency factors were considered to be a function of only the metal. A separate set of potency factors was required for the final calibration of the algal dataset.

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643 **Figure Captions**

Figure 1. Model-data comparisons for the effects of single-metal and metal-mixture

- 645 exposures on mortality of *Daphnia magna* in laboratory water with spiked metals. Mortality
- 646 is shown as a function of fractional coverage of metal on the biotic ligand (θ_M) for the AIST
- 647 model (panels **A**, **B**); *TOX* for the USGS model (panels **C**, **D**); *TOX*_{equiv} for the HDR model
- 648 (panels **E**, **F**); and F_{TOX} for the CEH model (panels **G**, **H**). Observed responses (open
- symbols) are compared to the model-calculated response curve for mortality (continuous
- line). Dashed lines represent plus/minus a factor of 2 in the concentration at which a model-
- 651 calculated response occurs. See text for calibration procedures used in each model. Data
- from Meyer et al. [30].
- Figure 2. Model-data comparisons for the effects of single-metal and metal-mixture
- 654 exposures on mortality of rainbow trout (*Oncorhynchus mykiss*) in field-collected water with
- spiked metals. Mortality is shown as a function of fractional coverage of metal on the biotic
- ligand (θ_M) for the AIST model (panels **A**, **B**); *TOX* for the USGS model (panels **C**, **D**);
- 657 TOX_{equiv} for the HDR model (panels **E**, **F**); and F_{TOX} for the CEH model (panels **G**, **H**).
- 658 Observed responses (open symbols) are compared to the model-calculated response curve for
- 659 mortality (continuous line). Dashed lines represent plus/minus a factor of 2 in the
- 660 concentration at which a model-calculated response occurs. See text for calibration
- 661 procedures used in each model. Data from Mebane et al. [31].
- Figure 3. Comparison of model-calculated responses for free metal as a percentage of the
- total dissolved metal, percent accumulated metal on the biotic ligand (or on the WHAM
- humic acid surrogate for generalized binding on or in organisms in the CEH model), and
- 665 percent mortality of *Daphnia magna* for: (A) 12.6 μg/L total dissolved Cd (from Index 4, Cu-

666	Cd #7-1); and (B) 83.8 μ g/L total dissolved Cu (from Index 4, Cd-Cu #5-2). Data from
667	Meyer et al. [30]; indexes are described in Van Genderen et al. [12].

669 exposures on growth reduction of *Pseudokirchneriella subcapitata* at pH 6.0 in field-

Figure 4. Model-data comparisons for the effects of single-metal and metal-mixture

- 670 collected water with spiked metals. Growth reduction is shown for (A) USGS model with
- 671 model parameters from a global calibration to all datasets; (**B**) USGS model with model
- parameters from a calibration to *P. subcapitata* pH 6.0 data; (C) CEH model; and (D) CEH

673 model with the baseline effect of H^+ removed from F_{TOX} . Observed responses for single-

- 674 metal (open symbols) and metal-mixture exposures (closed symbols) are compared to the
- 675 model-calculated response curve for mortality (continuous line). Dashed lines represent
- 676 plus/minus a factor of 2 in the concentration at which a model-calculated response occurs.
- 677 See text for calibration procedures used in each model

- Figure 5. Comparison of (A) metal binding affinity to binding sites on or in the organism
- $(\log K_M)$; and (**B**) lethal accumulations for 50% mortality (LA50) for rainbow trout
- 680 (*Oncorhynchus mykiss*) based on the AIST, USGS, HDR and CEH model calibrations.
- 681 Average log K_M values are given for the CEH model for illustrative purposes. Metals are
- arranged according to their expected affinity to bind to oxygen donor groups [43].
- Figure 6. AIST, HDR and CEH model-data comparisons for the effects of metals on
- 684 mortality of *Daphnia magna* in laboratory water with spiked metal concentrations. Mortality
- is shown: (A) as a function of total dissolved Zn with fixed total dissolved Cu concentrations
- (test series #5-4); and (**B**) as a function of total dissolved Cu with fixed total dissolved Cd
- 687 concentrations (test series #20-3). CEH model-calculated response curves are based on
- 688 F_{TOX-LT} and F_{TOX-UT} values that were optimized to the individual test series ($F_{TOX-LT} = 2.30$,

- 689 $F_{TOX-UT} = 2.71$ for test series #5; $F_{TOX-LT} = 2.61$, $F_{TOX-UT} = 3.18$ for test series #20). Observed 690 responses for metal-mixture exposures (closed symbols) are compared to the model-
- 691 calculated response curve for mortality (dotted, short dashed and continuous lines). Data
- from Meyer et al. [30]; indexes are described in Van Genderen et al. [12].
- Figure 7. Model validation for the effect of single-metal and metal-mixture exposures on
- 694 mortality of *Daphnia magna* in laboratory water with spiked metals. Mortality is shown as a
- function of fractional coverage of metal on the biotic ligand (θ_M) for the AIST model (panels
- 696 **A**, **B**); *TOX* for the USGS model (panels **C**, **D**); TOX_{equiv} for the HDR model (panels **E**, **F**);
- and F_{TOX} for the CEH model (panels G, H). Observed responses (open symbols) are
- 698 compared to the model-calculated response curve for mortality (continuous line). Dashed
- lines represent plus/minus a factor of 2 in the concentration at which a model-calculated
- response occurs. Data from Meyer et al. [30].
- 701





Figure 1.







706 Figure 2.





Figure 3.



Figure 4.





Figure 5.





Figure 7.