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Valencia-Avellan, M, Slack, R, Stockdale, A orcid.org/0000-0002-1603-0103 et al. (1 more author) (2018) Evaluating water quality and ecotoxicology assessment techniques using data from a lead and zinc effected upland limestone catchment. Water Research, 128. pp. 49-60. ISSN 0043-1354

https://doi.org/10.1016/j.watres.2017.10.031

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Accepted Manuscript

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PII: S0043-1354(17)30868-0

DOI: 10.1016/j.watres.2017.10.031

Reference: WR 13288

To appear in: Water Research

Received Date: 11 April 2017

Revised Date: 10 October 2017

Accepted Date: 14 October 2017

Please cite this article as: Valencia-Avellan, M., Slack, R., Stockdale, A., George Mortimer, R.J., Evaluating water quality and ecotoxicology assessment techniques using data from a lead and zinc effected upland limestone catchment, *Water Research* (2017), doi: 10.1016/j.watres.2017.10.031.

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Legend

- Sampling sites
- Mine structures (e.g. dressing floors, adits, shafts, smelt mills)
- ---- Waterways
- Elevation (m)













Evaluating water quality and ecotoxicology assessment techniques using data from a lead and zinc effected upland limestone catchment

4	
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14

15 Abstract

Point and diffuse sources associated with historical metal ore mining are major causes of 16 17 metal pollution. The understanding of metal behaviour and fate has been improved by the integration of water chemistry, metal availability and toxicity. Efforts have been devoted to 18 19 the development of efficient methods of assessing and managing the risk posed by metals to aquatic life and meeting national water quality standards. This study focuses on the 20 21 evaluation of current water quality and ecotoxicology techniques for the metal assessment of 22 an upland limestone catchment located within a historical metal (lead ore) mining area in 23 northern England. Within this catchment, metal toxicity occurs at circumneutral pH (6.2-7.5). 24 Environmental Quality Standards (EQSs) based on a simple single concentration approach 25 like hardness based EQS (EQS-H) are more overprotective, and from sixteen sites 26 monitored in this study more than twelve sites (> 75%) failed the EQSs for Zn and Pb. By 27 increasing the complexity of assessment tools (e.g. bioavailability-based (EQS-B) and 28 WHAM- F_{TOX}), less conservative limits were provided, decreasing the number of sites with

predicted ecological risk to seven (44%). Thus, this research supports the use of
bioavailability-based approaches and their applicability for future metal risk assessments.

31 Keywords: water quality standards, bioavailability, ecotoxicology, metal assessment, river water,

32 *limestone catchment*

33

34 1. Introduction

Humans have impacted upland catchments for centuries through the process of mining. Exposure of metal-bearing minerals to oxygen and water, both subsurface and through dumping of mine wastes above the surface, can result in increased dissolved concentrations of metals in water bodies. Once metals have entered aquatic ecosystems, they interact with a broad spectrum of biotic and abiotic components via dynamic interrelated processes, resulting in a large variety of compounds (Luoma, 2008).

41 In natural waters, the behaviour of metals depends on their speciation. Such species include 42 free ions, inorganic complexes, organic complexes, and metal sorbed to or incorporated 43 within colloids or particulate matter. The formation of metal complexes depends on the concentration of all dissolved components, pH, and ionic strength (Namieśnik and 44 Rabajczyk, 2010). Free metal ions can bind to dissolved organic matter (DOM; particularly 45 46 humic acids), forming complexes that regulate the concentrations of metals available for interaction with organisms (Tipping, 2002). The complexity of these interactions in aquatic 47 environments makes it difficult to measure the distribution of chemical forms. Knowledge of 48 49 chemical speciation is important because the relative distribution of different forms controls 50 metal interactions with organisms, including toxic effects. For instance, dissolved metals, 51 particularly free metal ions, have been related to freshwater ecotoxicity (Campbell, 1995; De 52 Schamphelaere and Janssen, 2002). More recently, the concentration predicted to be bound 53 to humic acid has been found to be a good proxy for both organism body burdens (Stockdale 54 et al., 2010; He and Van Gestel, 2015) and accumulation by aquatic bryophytes (Tipping et 55 al., 2008).

2

56 In the last few decades, a variety of in situ analytical approaches have been developed to understand the dynamics of metals in natural waters (Buffle, 2000). One such technique is 57 Diffusive Gradients in Thin-films (DGT) for assessing a range of active species, depending 58 59 upon their mobility (diffusion coefficients) and kinetics (dissociation rates) across a thin film 60 of polyacrylamide gel (Zhang and Davison, 2015). This technique has being used as a monitoring tool for providing kinetic information on labile metal species in rivers, soil or 61 sediment impacted by mines and coastal waters (Unsworth et al., 2006; Warnken et al., 62 2009). In addition, DGT has been applied to investigate potential metal availability in 63 64 freshwater ecosystems, for instance the accumulation of cadmium in communities of algae 65 (periphyton) has been evaluated by Bradac et al. (2009).

66 In view of the importance of DOM in complexing metals, comprehensive speciation models have been developed for predicting the distribution of chemical species accounting for 67 complexation with inorganic and organic ligands as well as competition for organic binding 68 sites between different metals, and between metals and protons. The Non-Ideal Competitive 69 70 Adsorption (NICA) coupled with a Donnan electrostatic sub-model uses a continuous 71 distribution approach to describe metal and proton bindings to organic matter (Benedetti et 72 al., 1995; Kalis et al., 2006). A discrete site approach is used in the Humic Ion Binding Model 73 VII (Tipping et al., 2011). This is coupled with a inorganic thermodynamic code, the 74 Windermere Humic Aqueous Model (WHAM) (Tipping, 1994). The WHAM code has been 75 successfully applied in a variety of research and regulatory areas related to water quality criteria for zinc and copper in the United States and Europe (Hamilton-Taylor et al., 2011; 76 77 Balistrieri and Mebane, 2014; Balistrieri and Blank, 2008; Cheng et al., 2005). In addition, it 78 has been used across Europe as part of the Critical Loads approach related to atmospheric deposition and soil pollution by cadmium, lead, mercury and copper (de Vries et al., 2007; 79 Hall et al., 2006; Tipping et al., 2003). Over recent decades, WHAM/Model V was 80 81 incorporated into the Biotic Ligand Model approach to improve prediction of metal toxicity to

aquatic organisms and support the implementation of Environmental Quality Standards(EQS).

84 Metal bioavailability and toxicity have long been recognized to be a function of water chemistry (Paguin et al., 2002). Until recently environmental regulations considered only 85 86 hardness-based conditions to derive EQS (Beane et al., 2016). However, latest research 87 has increased understanding of the influence of physicochemical variables in metal 88 speciation, and ecotoxicological studies have extended our knowledge of metal effects on 89 biota. Consequently, bioavailability-based approaches such as the Biotic Ligand Models 90 (BLMs) have been implemented within EQS for chemical and ecological assessment, to 91 evaluate the overall quality of a given waterbody (Niyogi and Wood, 2004). In the UK, the 92 BLMs have been simplified to create a user-friendly method known as metal bioavailability 93 assessment tool (M-BAT), which use pH, DOC and calcium as input data to account for 94 toxicity of single metals in freshwaters within a regulatory context (e.g. Zn-BLM, Cu-BLM, 95 Mn-BLM and Ni-BLM) (WFD-UKTAG, 2014; De Schamphelaere and Janssen, 2002; 96 Steenbergen et al., 2005; Lock et al., 2007). The M-BAT tools parameterised to date include 97 those for Zn, Cu, Mn, Ni and Pb. Some of these tools have been incorporated into a tieredapproach monitoring scheme to implement the bioavailability-based water quality guidelines. 98 99 However, they have limitations for assessing metal mixture effects since they are based on 100 BLMs for specific metal-organism, and they do not include all dissolved chemical species. To 101 overcome this issue, an alternative bioavailability-based model (WHAM- F_{TOX}) developed by 102 Stockdale et al. (2010) offers a plausible option for quantifying mixture toxicity and its 103 potential effects on aquatic organisms. WHAM- F_{TOX} , in common with the BLM, uses 104 organisms as reactants. However, WHAM- F_{TOX} assumes that toxicity is related to non-105 specific binding to organism surfaces rather than specific biotic ligands (Stockdale et al., 106 2010). Several studies have applied WHAM- F_{TOX} for predicting metal toxicity to aquatic biota 107 in laboratory experiments and linking the effects of chemical speciation of metals and protons to species richness of freshwater macroinvertebrates (e.g. Ephemeroptera, 108

Plecoptera and Trichoptera (SR_{EPT})) and zooplankton species diversity (Tipping and Lofts,
2013; 2015; Stockdale et al., 2010; 2014; Qiu et al., 2015).

111 Despite the scientific and regulatory improvements for the establishment of EQSs, achieving 112 good ecological and chemical status of water bodies according to the Water Framework 113 Directive (WDF) (European Commission, 2000) is still a significant challenge (Environment 114 Agency, 2008a). In the UK, the major impediment to meet this aim is related to diffuse 115 pollution, including metal contamination generated by abandoned mines and mined wastes 116 (Jarvis and Younger, 2000). At a national scale, comprehensive data exist for certain areas 117 but there are substantial gaps for other regions, therefore the assessment of the extent and 118 severity of metal pollution from mining activities is partial (Mayes et al., 2009). As a 119 consequence, the prioritisation of threatened sites to address remediation of metal pollution 120 in a logical and cost-effective manner is a difficult task. River Basin Management Plans 121 (RBMP) provide a good framework for the implementation of mitigation measures. But the 122 application of bioavailability-based standards (EQS-B) offers significant alternative for the 123 selection of sites at real ecological risk, reducing the burden of remediation targets for 124 efficiently achieving a better quality of water bodies (Environment Agency, 2008b; Harmsen 125 and Naidu, 2013).

126 This study aims to evaluate different approaches used to assess water quality. Focusing on 127 an upland limestone catchment affected by historical mining, we compare hardness based 128 EQS with approaches that consider more detailed water chemistry such as BLM based EQS and WHAM- F_{TOX} . Data from a single biological survey were used to give context to the 129 results from the chemical approaches. Additionally, we evaluate the ability of the DGT 130 131 technique to yield dynamic dissolved concentrations that could be applied to EQS as the 132 WFD allows for water quality criteria to be set based upon dynamic methods incorporating 133 chemical speciation. This contribution complements a companion paper (Valencia-Avellan et 134 al., 2017), which provided an assessment of the geochemical and hydrological processes 135 controlling the main sources of metal pollution.

5

136 2. Methods

137 2.1. Site description

The study area is located within the Yorkshire Dales National Park, northern England (Figure 1). Hebden Beck is a headwater catchment (12 km long) containing extensive historical lead and zinc mining operations, which flows through limestone bedrock surrounded by sheep pasture and peat-rich moorland (Jones et al., 2013). Evidence of mining and smelting exists along the length of the river system, from mine tailings and slag to mine water drainage adits and abandoned buildings. A detailed description of the study area is provided in (Valencia-Avellan et al., 2017).

145 2.2. Water sampling

Monthly field surveys were conducted from November 2013 to December 2014 at sixteen 146 147 sampling sites. Sampling locations (Figure 1) include a source pool, tributaries and the main 148 river channel, covering the most mining impacted area (5 km²). The detailed sampling strategy and analytical procedures are described in Valencia-Avellan et al. (2017). 149 Parameters recorded in situ such as temperature, pH, dissolved oxygen, conductivity and 150 flow, together with measured concentrations of dissolved metals (defined as filterable with a 151 0.45 μ m filter) (Zn²⁺, Pb²⁺ Cd²⁺, and Cu²⁺), major anions (SO₄²⁻, Cl⁻, NO₃²⁻, PO₄²⁻) and 152 153 dissolved inorganic and organic carbon (DIC, DOC) were considered as model input 154 parameters for metal speciation assessment. Concentrations of dissolved metals were also used to evaluate their association with the presence of labile-inorganic species in the river 155 water. Detection limits for metals in water samples were Zn: 5.1×10^{-7} mol/l and Pb: 2.4×10^{-7} 156 10⁻¹³ mol/l. For statistical analysis, annual average concentrations were calculated for each 157 158 site.

159 2.3. Determination of C_{DGT} in river water

160 Diffusive Gradients in Thin-films (DGT) devices (DGT Research Ltd, Lancaster, UK) 161 incorporated a Chelex metal binding layer, a 0.8 mm thick diffusive hydrogel (polyacrylamide

162 gel crossed by agarose cross-linker) and a polyethersulphone filter membrane. DGT devices 163 were subject to minimal handling and transported to and from the site in acid cleaned zip-164 lock bags. A total of 121 DGT measurements were performed bimonthly from December 2013 to November 2014 across the sampling sites as part of the water sampling surveys. 165 166 Deployment time averaged approximately one month but varied from 18 days to 43 days. 167 Duplicate devices were deployed at two of the sixteen sites and a field blank was processed in the field once per visit. All DGTs were processed in the laboratory within 24 hours: this 168 169 involved separating the layers of the probe using acid-washed Teflon tweezers. Each resin 170 gel (Chelex) layer was placed in a 2 ml polypropylene tube with 0.7 ml of 1 M HNO₃ 171 (ARISTAR) then stored at 4°C. Immediately before ICPMS analysis, the eluent was diluted 172 10x with Milli-Q water (18.2 M Ω ·cm).

173 Labile metal concentrations were determined following the method of Zhang and Davison (1995). First, the mass of metal (M) in the Chelex resin is calculated (Eq. 1), where C_e is the 174 175 concentration of metals in the 1 M HNO₃ elution solution (in µg/l) provided by the ICPMS 176 analysis, V_{qel} is the volume of the resin gel, V_{HNO3} is the volume of HNO₃ added to the resin gel, and f_e is the elution factor for each metal, typically 0.8. Once mass is determined, the 177 concentration of labile metals (C_{DGT}) can be calculated as per Zhang and Davison (1995) 178 179 (Eq. 2), where, *M* is the mass of metal in the Chelex resin, Δg is the thickness of the diffusive 180 gel (0.8 mm) plus the thickness of the filter membrane (0.13 mm), D is the diffusion 181 coefficient of metal in the gel reported at http://www.dgtresearch.com/, t is deployment time 182 and A is the exposure area (3.14 cm^2) .

183
$$M = (C_e (V_{gel} + V_{HNO3}) / f_e) / 1000$$
(1)

184
$$C_{\text{DGT}} = (M \Delta g / (DtA)) / 1000$$
 (2)

185 2.4. Speciation modelling using Windermere Humic Aqueous Model (WHAM/Model
186 VII)

187 WHAM/Model VII calculates the chemical speciation of metals using humic (HA) and fulvic 188 acids (FA) as active DOM components. In this study, default model parameters were used 189 together with the measurements for the following inputs; temperature, pH, dissolved anions (Cl⁻, SO₄²⁻, NO₃⁻ and PO₄²⁻), dissolved metals (Zn²⁺, Pb²⁺, Cu²⁺, Cd²⁺, Fe³⁺, and Al³⁺), 190 191 dissolved inorganic carbon (DIC; input as carbonate) and DOC. For the modelling DOM was 192 assumed to be composed of 50% carbon and the active fraction considered to be 65% fulvic 193 acid, and hence was calculated from DOC values (FA = $DOC \times 1.3$) (Tipping et al., 2008). The activity of both AI and Fe was calculated from the measured total filtered concentration 194 and from the equations derived for AI by Tipping (2005) and for Fe by Lofts and Tipping 195 196 (2011), with the lower of the two values being adopted. This avoids over-estimation of Al or 197 Fe activity in filtrates containing colloidal forms of the metals. Two different scenarios were considered for modelling. Firstly, Al and Fe oxide precipitates were not considered to be 198 199 active with respect to surface chemistry. In a second scenario it was considered that the 200 oxides contain active binding sites and that one mole of oxide has 90 g/mol for Fe (Dzombak 201 and Morel, 1990), and 61 g/mol for Al (Lofts, 2012) of active phase. Concentrations of metals 202 bound to oxides were calculated from the output data. These scenarios were applied to allow 203 comparison with the DGT results.

204 2.5. Estimating metal mixture toxicity using WHAM-F_{TOX}

205 WHAM- F_{TOX} , a parameterised version of WHAM established by Stockdale et al. (2010) was 206 applied to evaluate the combined impact of protons (H⁺) and metals (Al³⁺, Zn²⁺, Pb²⁺ and 207 Cu²⁺). In WHAM- F_{TOX} , the metal toxicity function (F_{TOX}) is considered a product of toxic 208 cation-bound concentrations (v, mmol/g) and the toxicity coefficient of each analyte (α) and *i* 209 refers to each toxic cation (Eq. 3).

(3)

210 $F_{TOX} = \Sigma \alpha_i v_i$

211 Concentrations of v_i were obtained from the WHAM/Model VII modelling as described above without active oxides and with the addition of 10^{-6} g of colloidal humic acid. Toxicity 212 213 coefficients for aluminium (α_{Al} : 2.24), zinc (α_{Zn} : 2.69), lead (α_{Pb} : 2.51) and hydrogen (α_{H} : 1) 214 were adopted from Stockdale et al. (2010). The Pb toxicity coefficient (with a published pvalue of 0.91) was used in absence of an alternative published value. We suggest the 215 application of $\alpha_{\mbox{\scriptsize Pb}}$ with caution as it may be conservative compared with other Pb toxicity 216 217 coefficients reported for trout species (cutthroat trout, 6.7; rainbow trout, 4.6) (Tipping and 218 Lofts, 2015).

For these calculations, the following conditions were applied: i) if F_{TOX} is lower than 2.33 (F_{TOX-LT}) the maximum SR_{EPT} will be 23 and no toxicity occurs (Eq. 4), ii) a graded linear toxic response occurs as F_{TOX} increases (Eq. 5), and iii) if F_{TOX} is higher than 5.20 (F_{TOX-UT}) no species are predicted to be present (Eq. 6). Obtained values were converted to a percentage of the maximum SR_{EPT} (90% percentile). A detailed description of these calculations is provided by Stockdale et al. (2010).

225 If
$$F_{\text{TOX-LT}}$$
, then $SR_{\text{EPT}} = SR_{\text{EPT-max}}$ (4)

226 If
$$F_{\text{TOX}} > F_{\text{TOX-LT}}$$
, then $SR_{\text{EPT}} = SR_{\text{EPT-max}} - \left(SR_{\text{EPT-max}} \times \frac{F_{\text{TOX}} - F_{\text{TOX-LT}}}{F_{\text{TOX-UT}} - F_{\text{TOX-LT}}} \right)$ (5)
227 If $F_{\text{TOX}} > F_{\text{TOX-UT}}$, then $SR_{\text{EPT}} = 0$ (6)

229 2.6. Water quality assessment tools

Tools with different level of complexity were applied to a consistent set of water chemistry 230 231 data to assess their contrasting levels of compliance with their respective standards. Firstly, 232 water hardness was calculated at each sampling point for applying established metal limits 233 under hardness-based standards (Environment Agency, 2011). For the hardness based 234 values a risk characterization ratio (RCR) was estimated dividing the dissolved metal 235 concentrations or Predicted Effect Concentration (PEC) by the calculated hardness-based 236 limits or No Predicted Effect Concentrations (PNEC). Water guality standards were exceeded (i.e., over accepted limits) if $RCR \ge 1$ (RCR = PEC/PNEC). 237

238 Secondly, assessment tools based on the BLM were applied. One tool, known as the Metal 239 Bioavailability Assessment Tool (M-BAT) was used for Zn (and is also parameterised for Cu, 240 Mn and Ni) and the "Final Pb Screening Tool" was used for Pb, these are both available from 241 the UK Water Framework Directive Technical Advisory Group (WFD-UKTAG, 2014). An 242 additional Pb tool is also available from the International Lead Association that combines bioavailability models with detailed species sensitivity distribution (SSD) analyses 243 244 (International Lead Association, 2017). These two Pb tools are referred to as Pb-FST and Pb-SSD for the UK-TAG and ILA versions, respectively. For these BLM based tools, an 245 alternate risk characterisation ratio (RCR) was calculated, where PEC values are the 246 247 bioavailable metals based on dissolved concentrations, and PNEC are the site-specific values obtained from the tools, if RCR \geq 1 water quality standards were exceeded 248 249 (Merrington and Peters, 2013; WFD-UKTAG, 2014). For these calculations, Zn-BAT considered dissolved concentrations and the water chemical parameters Ca²⁺, pH and DOC 250 as input data, Pb-FST only required concentrations for Pb and DOC. In the case of the Pb-251 SSD, it required additional parameters such as Mg²⁺, Na⁺, K⁺, SO₄²⁻, Cl⁻, CO₃²⁻. Thirdly, the 252 253 model WHAM- F_{TOX} derived from WHAM/Model VII was applied to predict the effect of water 254 chemistry on the maximum species richness (SR_{EPT-max}) as described above.

255 2.7. Biological survey

256 A single macroinvertebrate survey was carried out at all sites in August 2014. Samples were 257 collected using a standard kick sampling technique and modified from the Water Framework Directive compliant methodologies (EU-STAR project, 2004). Given the small size of 258 259 watercourses in the survey, one-minute kick samples were taken at thirteen sites, 30 second 260 kick samples were taken at 2 narrow sites and one sweep sample was taken in a pond at the top of the catchment. Kick samples were followed by a stone search to collect organisms 261 262 attached to the substrate. Specimens were analysed live at each site. Several were 263 preserved and taken back to the lab for identification confirmation. Invertebrates were analysed to family level. All relevant families within the sample were scored on a presence or 264

265 absence basis and frequency scale, using the Biological Monitoring Working Party (BMWP) 266 index which rates invertebrates based on their tolerance to organic pollution. BMWPs 267 greater than 100 are associated with unpolluted streams, while the scores of heavily polluted streams are less than 10. In addition, the average sensitivity of the macroinvertebrate 268 269 families known as the Average Score Per Taxon (ASPT) was determined by dividing the 270 BMWP scores by the number of taxa present. Scores for ASPT are considered from 0 to 10, 271 where 10 indicates a unpolluted site containing large numbers of high scoring taxa (Armitage et al., 1983) (Table 1). Impacts of metal pollution might not be reflected by the BMWP and 272 273 associated average score per taxon (ASPT) scoring systems as they were originally 274 developed as indicators of organic pollution (Tipping et al., 2009). Despite this caveat, BMWP scores, number of taxa and ASPT values were compared with metal concentrations 275 276 and outcomes from water quality assessment tools.

277 3. Results

278 General water chemistry for Hebden Beck has been reported in detail by Valencia-Avellan et al. (2017). It is dominated by the underlying geology of the catchment (i.e. buffered with high 279 280 Ca and circumneutral pH values), whilst metal levels are associated with the weathering and 281 erosion of Pb and Zn minerals exposed by past mining. Mine-impacted water is 282 characterised by circumneutral pH (6.3-7.5) and elevated metal concentrations. Maximum concentrations of Pb, Cd, Zn, and Cu were measured in site H12 corresponding to an 283 ephemeral tributary flowing through spoil mine wastes $(1.4 \times 10^{-6}, 4.18 \times 10^{-7}, 6.5 \times 10^{-5}, 4.18 \times 10^{-7})$ 284 3.12×10^{-7} mol/l, respectively) (Table S1). Full chemistry data are included in Tables S1-S3. 285 286 Considering the significantly higher concentrations of Zn and Pb, we focus primarily on these 287 metals. Associations between DIC/DOC, sulfate and metals were identified. DIC was strongly correlated with Ca and Mg ($R^2 = 0.9$), whilst DOC was more closely correlated with 288 levels of Fe, Al, Pb and Cu (0.1 \ge r \le 0.7). Sulfate correlated strongly with Zn and Cd (R² = 289 0.6, 0.7; p < 0.001), possibly associated with the dissolution of sulfide minerals (Table S4). 290

291 3.1. Dynamic metal (C_{DGT}) concentrations

From DGT measurements at each site, annual average $Zn-C_{DGT}$ concentrations ranged from 2.8 × 10⁻⁵ to 4.8 × 10⁻⁹ mol/l and for Pb-C_{DGT} from 1.1 × 10⁻⁷ to 1.1 × 10⁻¹⁰ mol/l. Highest Zn-C_{DGT} identified in perennial tributaries were H7: 2.8 × 10⁻⁵ Zn mol/l and highest Pb-C_{DGT} in ephemeral tributaries H13 and H12: 1.1 × 10⁻⁷ Pb mol/l, while in main channel maximum concentrations were observed in H6: 7.5 × 10⁻⁶ Zn mol/l and H10: 2.3 × 10⁻⁸ Pb mol/l (Table 2).

3.2. Comparison between dissolved metal concentrations and C_{DGT} measurements.

299 DGT measurements were compared with the mean of metal concentrations in water samples calculated from samples taken at both the deployment and retrieval dates. Figure 2 300 shows the linear correlations between measured dissolved metals in water samples and 301 302 metals measured by DGT (M-C_{DGT}). A strong relationship exists between Zn measured dissolved species and $Zn-C_{DGT}$ (R² = 0.75, p < 0.001). For Pb, a moderate relationship was 303 observed between dissolved concentrations and C_{DGT} (R² = 0.59, p < 0.001). Whilst there is 304 correlation between C_{DGT} and dissolved concentrations there is a lesser degree of 305 306 agreement, as indicated by the slopes of 0.46 (Zn) and 0.07 (Pb) (Figure 2).

307 3.3. Metal speciation calculated by WHAM/Model VII

308 3.3.1. Modelling where oxide precipitates are not surface active

Concentrations of predicted organic and inorganic species of Zn, and Pb are shown in 309 Figure 3. Average concentrations of metal-organic complexes (i.e. metal-fulvic acid (FA) 310 complexes) were higher for Zn than Pb, although as a function of total dissolved metal, Pb 311 had a greater fraction present as organic complexes, consistent with a higher Pb-FA model 312 binding constant. Values of organic species ranged for Zn from 2.9×10^{-8} to 1.2×10^{-5} mol/l 313 and for Pb from 2.0 x 10^{-10} to 1.1 x 10^{-6} mol/l, where site H15 presented the highest 314 315 concentrations for both metals. For inorganic species, higher average concentrations occurred for Zn (3.6 × 10^{-7} to 5.8 × 10^{-5} mol/l), and lower for Pb (6.0 × 10^{-10} to 7.2 × 10^{-7} 316

317 mol/l), with the highest concentrations in sites H12 and H13, respectively. Figures 4a and 4c 318 show linear correlation between dissolved inorganic metal species predicted by WHAM and 319 C_{DGT} measurements of Zn and Pb. Good correspondence was observed for Zn (R² = 0.81, p 320 < 0.001) and moderate relationship for Pb (R² = 0.54, p < 0.001).

321 3.3.2. Modelling where oxide precipitates are surface reactive

Precipitation of Fe and Al oxides (with reactive surface) were simulated by using the concentration of precipitated metal to the mass of active phase (Figures 4b and 4d). Similar correlations were obtained when oxides were not surface reactive ($R^2 = 0.81$, $R^2 = 0.51$). The presence of colloidal Pb was estimated showing more affinity to FeOx than AlOx. Concentrations of Pb-FeOx complexes ranged from 5.7 × 10⁻¹¹ to 2.5 × 10⁻⁷ mol/l, while Pb-AlOx complexes from 6.8 × 10⁻¹² to 2.5 × 10⁻⁸ mol/l. For both complexes highest concentrations were identified in H12 (Figure 5).

329 3.4. Water quality and ecotoxicological assessment

330 3.4.1. Environmental Quality Standards: EQS-H and EQS-B

331 Annual averages concentrations were used for assessing EQS-H and EQS-B. The EQS-H standards were exceeded for Zn from 0.2 to 254-fold and for Pb from 0.02 to 40-fold (Table 332 2). Highest Zn exceedances (> 60-fold) occurred in upstream sites surrounded by spoil mine 333 wastes (H15, H14 and H12). These sites and H13, also presented highest Pb exceedances 334 (> 20-fold). The EQS-B standards were exceeded for Zn from 0.8 to 209-fold and for Pb from 335 0.02 to 17-fold. Highest Zn exceedances (> 50-fold) appeared in sites H7 > H12 > H15 336 337 which represent a mine adit, an ephemeral tributary flowing through spoil mine wastes and a 338 reservoir next to a smelting mill. For Pb, highest levels (> 6-fold) were presented in H12 > 339 H5, H15 > H14 (Table 2).

340 3.4.2. Toxicity function: WHAM-F_{TOX}

341 WHAM- F_{TOX} considers humic acid (HA) bound to be a plausible proxy for organism bound 342 metal. Concentrations of metals bound to humic acid ranged for Zn (0.0013 to 0.9 mmol/g)

and Pb (0.0001 to 0.09 mmol/g) (Figure 6). Reservoir (H15), perennial tributaries (H14, H7) and ephemeral tributary (H12) were characterised by high Zn concentrations (\geq 0.5 mmol/g), the same sites (except for H7) presented high Pb concentrations (\geq 0.05 mmol/g).

346 Toxicity contributions (F_{TOX-M}) for each element (Zn, Pb, H and Al) were calculated from 347 Equation 3 (Methods Section 2.5.) and summed to produce a total toxicity function value 348 (Total_ F_{TOX}). Equations 4 and 6 were applied to convert Total_ F_{TOX} to a prediction of the 349 maximum species richness that could be expected based on the chemistry of each stream. 350 Effects of metal mixtures on predicted species richness are shown in Table 3. Reduction of 351 predicted species richness (SR_{EPT} < 23) was identified at several sites. In the main river 352 channel (H10, H6 and H2) and perennial tributary (H11) a slight decrease was presented 353 (SR_{EPT}: 22 to 19) associated with the combined moderate toxicity of H, Zn and Al. A more significant decrease (SR_{EPT}: 18 to 10) was shown in sites H15, H14, H13, H12, H9 and H7 354 355 due to higher contributions from Zn, contributions of H and Al and additionally small contributions from Pb; while the lowest value (SR_{EPT}: 8) occurred in site H4 related to the 356 357 influence of H and Al and not metals.

Contributions of metal toxic effects (F_{TOX}) at different pH conditions were predicted by WHAM- F_{TOX} (Figure 7). Figure 7a demonstrates the effects of heavy metals at spoil runoff or adit sites. Higher toxicity (> 2.3) is related to acidic waters (H4 and H13), while adits or water draining through spoil sites are rich in Zn and Pb (H15, H14, H12, H7). Fig 7b highlights the lack of metal contribution of metals at the lower pH sites (pH < 5) due low metal-humic binding as a result of enhanced competition for binding sites from the elevated H and Al concentrations (and possible lower metal too).

365 3.6. Evaluation of metal assessment tools

366 EQS-H, EQS-B and WHAM- F_{TOX} were compared to evaluate their relative assessment of 367 potential impact with respect to metal toxicity in aquatic ecosystems (Table2). Multiple sites 368 are necessary as WFD applies to lower order streams than previous legislation, meaning all 369 perennial tributaries may be important from a compliance viewpoint. Each of these sites

370 represents a length of stream therefore an ecosystem. Ephemeral tributaries are assessed 371 as they can represent high input sites into other streams, although these sites may not be 372 significant as aquatic ecosystems as they are more likely to be dry during the productive summer months. From the comparison, EQS-H are more conservative than EQS-B. EQS-H 373 374 showed wider ranges of exceedance for Zn (0.2 to 254-fold) and for Pb (0.02 to 40-fold), while compared to EQS-B for Zn (0.8 to 209-fold) and Pb (0.02 to 17-fold). For both tools, 375 376 the highest Zn failures occurred in reservoir (H15), ephemeral tributary (H12) and perennial tributaries (H14 and H7). For Pb, the same sites, except H7 showed significant failures, in 377 378 addition, using EQS-H an ephemeral tributary (H13) showed exceedance, likewise the 379 perennial tributary (H5) applying the EQS-B approach. For the Pb bioavailability tools (FST 380 and SSD), results showed that the FST approach with fewer chemistry inputs is more 381 conservative than SSD. By using FST, values ranged from 0.1 to 35-fold, while SSD presented ranges from 0.1 to 17-fold. Calculations from WHAM- F_{TOX} showed that at twelve 382 383 sites the water chemistry may reduce the maximum species diversity of macroinvertebrates by between 3 and 65%. All ephemeral tributaries (H12, H13, and H4) together with a 384 385 perennial tributary (H14) and reservoir (H15) showed the largest predicted reductions in 386 species richness of 34% to 65%.

387 3.7. In situ macroinvertebrate survey

388 A total of twenty four macroinvertebrate families were identified across all of the sites. The 389 BMWP scores based on the presence and absence of macroinvertebrates families indicated 390 a moderately impacted water quality (BMWP > 50) for sites H14, H10, H9, H6, H5, H3, H2, 391 H1P and H1 representing mainly main channel sites. While lower scores (BMWP < 25) in sites representing source waters and tributaries (H15, H13, H12, H11, H8, H7 and H4) 392 393 indicated pollution deriving poor water quality (Figure S1), while ASTP scores (< 2) showed sites H13 and H12 as the most polluted (Figure S2), although ephemeral tributaries may 394 395 expected to have reduced or poor biological diversity due to periodic dry periods, particularly 396 in the summer months.

Four macroinvertebrate orders were predominant in the catchment (Figure S3). Based on numbers of total individuals counted, the order Diptera showed 40% represented by the families Psychodidae and Simuliidae (Figure S4), followed by the orders Ephemeroptera (mayfly) with 18% where Baetidae was the most prevalent (Figure S5), Trichoptera (caddisfly) presented 28% with Hydropsychidae as the dominant family (Figure S6), and the order Plecoptera (stonefly) with 11% represented by the family Perlodidae (Figure S7).

403 **4. Discussion**

For aquatic environments, metal assessment has evolved as bioavailability is increasingly 404 considered a critical measure for improving water quality, e.g. under the EU Water 405 406 Framework Directive. A preliminary water quality assessment was carried out where dissolved concentrations of Zn and Pb were assumed to be bioavailable depending on 407 408 ranges of water hardness conditions, and exceedances were calculated from previous-409 established EQS-H Concentrations of Zn and Pb exceeded the limits by 175-fold and 40-fold 410 respectively, however, little consideration is given to their bioavailable forms. Merrington 411 (2016) has indicated that hardness limits have been derived from ecotoxicological tests 412 mainly from crustaceans and fish, where water chemical parameters do not necessarily 413 reflect field conditions. Thus, the use of hardness-based standards may be overly 414 conservative as they might under-represent other important taxa. In addition, they fail to consider chemical speciation in distinguishing sites with potential environmental risk 415 associated with metals. 416

In complex environments such as freshwaters, understanding of the underlying conditions controlling the dynamics and speciation of metals is critical for assessing and managing the risk they pose to aquatic life (Han et al., 2013). This study evaluated the utility of DGT for yielding both time averaged and dynamic Zn and Pb concentrations (potentially a proxy for organism availability) (Diviš et al., 2007; Degryse et al., 2009; Amato et al., 2014; Søndergaard et al., 2011; 2014; Omanović et al., 2015). By comparing, DGT measurements (C_{DGT}) with direct water measurements (M-measured), results indicated that C_{DGT})

concentrations were generally lower than the directly measured values. For Zn-C_{DGT} and 424 Zn-measured a strong linear correlation ($R^2 = 0.75$) was present suggesting the presence of 425 Zn mainly as free ions and readily labile complexes, a slope of 0.47 revealed that DGT 426 427 measurements at higher concentrations failed to reflect dissolved measured concentrations. 428 The better agreement presented at lower concentrations might be caused by the kinetic rate 429 of Zn forms, which will determine its accumulation into the Chelex-resin layer (Warnken et al., 2007). Free Zn²⁺ seems to be rapidly absorbed by the Chelex-resin layer, while the 430 431 supply of labile complexed forms will depended on their degree of lability (Puy et al., 2012). 432 Zhang and Davison (2015) indicated that fully labile complexes might not be completely 433 dissociated within the diffusion layer, thus their uptake in the Chelex-resin could be reduced 434 or not occur. Correlation between Pb-C_{DGT} and Pb-measured was moderate ($R^2 = 0.59$), 435 with a slope of 0.07 indicating poor agreement between the data, possible attributable to the presence of less labile Pb forms (e.g. Pb-FeOx, Pb-DOM) that need more time for diffusion 436 or presence of bigger particles that could be excluded from DGT membranes (Han et al., 437 2013). Considering correlations for both metals, C_{DGT} measurements can be affected by the 438 439 amount of total metal dissolved concentrations present as free ions and their complexation with inorganic or organic substances (Zhang and Davison, 2015). Furthermore, long 440 deployments (> 15 days) might expose DGTs to biological factors like biofouling, and flow 441 442 fluctuations that might decrease the stirring rates, reducing DGT's performance (Davison 443 and Zhang, 1994; Webb and Keough, 2002; Turner et al., 2014).

444 Metals speciation and bioavailability are controlled by a range of different factors such as 445 pH, ionic strength, DOC, hardness, pH and inorganic or organic colloids. Given these 446 factors, the quantity of metal available for organism uptake is not reflected by the total 447 dissolved concentration (Hart, 1981; Luoma, 2008). This study correlated M-C_{DGT} 448 measurements with inorganic metal complexes predicted from dissolved measured 449 concentrations under the absence and presence of active inorganic colloids (e.g. FeOx and 450 AlOx) as they can compete for Zn and Pb binding with DOM. If active oxide colloids are

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excluded from the modelling, Zn is predicted to be largely present as inorganic (e.g. Zn²⁺, 451 452 Zn(OH)⁺, etc.) rather than organic complexes, and highest predicted inorganic concentration was at site H12 (5.8 \times 10⁻⁵ mol/l). Conversely Pb had a greater affinity for organic binding, 453 with highest organic concentration at H15 (10⁻⁶ mol/l) (Figure 5). Both metals showed good 454 linear correlations between inorganic complexes and C_{DGT} for Zn (R² = 0.81), and Pb (R² = 455 0.54), reflecting a modest improvement in the correlation of Zn-C_{DGT} with predicted Zn 456 inorganic complexes than previous correlation between Zn-C_{DGT} with Zn-measured, which 457 can be attributable to the fact that dissolved Zn is present primarily as inorganic and free 458 459 ions, reflecting the absence or low level of Zn-organic complexes in the correlation analysis. The agreement observed for Zn confirms that the total dissolved Zn (mainly inorganic) is 460 approximately equivalent to the dynamic Zn-C_{DGT} (except at high Zn loading, as previously 461 462 discussed). For Pb, there is a discrepancy between Pb-C_{DGT} and dissolved Pb, likely due to larger fractions in other less labile phases (Pb-organic complexes) (Warnken et al., 2008; 463 Han et al., 2013). Outcomes from modelling conditions where oxide precipitates are surface 464 reactive showed similar behaviours for Zn ($R^2 = 0.81$) and Pb ($R^2 = 0.51$) to those obtained 465 466 in the absence of surface activity. The lack of improved agreement to the Pb-inorganic in presence of oxides with the Pb-C_{DGT} (dynamic) concentration implies that the model may be 467 under-predicting the binding to these forms. Results from DGT measurements provided 468 469 useful information about metal dynamics and speciation which is valuable for understanding 470 metal uptake mechanisms. Metal bioavailability studies from Bradac et al. (2009); Yin et al. 471 (2014); (Balistrieri et al., 2012) suggest DGTs as useful tool for understanding the 472 accumulation of metals in freshwater ecosystems (e.g. algae, snails and fish). However, 473 results from this study indicated that DGT is not a suitable substitute for direct water 474 measurements due to complex uptake kinetics of different forms and at higher metal 475 concentrations. These limitations make DGT a potentially useful bioavailability assessment 476 tool, but further work is needed to compare DGT measurements in the field with actual in situ organism metal burdens. Zhang and Davison (2015) have indicated that given the 477 478 complexity of biological system and chemistry of environmental media, the ability of DGT to

479 mimic metal uptake and predict bioavailability is variable. Further research based on the 480 comparison of field data and laboratory experiments under controlled temperature, 481 deployment time and flow rates are suggested for evaluating the applicability of DGTs regarding metal bioavailability in freshwaters. Evidence of marked diurnal ranges as well as 482 483 flow-dependent variations in metal concentrations reported by Jones et al. (2015), also suggests that DGT measurements should be complemented by frequent measurements of 484 metal concentrations (e.g. with auto samplers) to identify the actual variation experienced by 485 DGTs. 486

487 Outcomes from EQS-H, EQS-B and WHAM- F_{TOX} provided different levels of prediction 488 regarding metal bioavailability and toxicity. In comparing EQS-H and EQS-B, the simpler 489 approach (EQS-H) is more conservative than the standard that considered more water 490 chemistry parameters and metal-organism interactions (EQS-B) (Merrington, 2016). 491 However, by contrasting EQS-B and WHAM- F_{TOX} , results suggest that the latter approach is 492 the least conservative of the three approaches. This was clearly identified in sites which 493 exceeded EQS-B like H7 (Zn: 209-fold and Pb: 2-fold) and H5 (Zn: 44-fold and Pb: 9-fold), 494 but with F_{Tox} -SR showing 100% for predicted species richness for both sites. WHAM- F_{TOX} 495 also revealed that major toxic effects can occur at circumneutral pH mostly due to Zn and 496 Pb. Moreover, a limited biological survey agreed with this chemical assessment. The most 497 impoverished sites presented low predicted macroinvertebrate richness (SR_{EPT} \leq 34%) 498 correlating with high concentrations of metals (H12, H13 and H14) or low-pH values (H4) 499 (Table 2). Overall, the orders Diptera and Trichoptera were dominant (40% and 28%) with 500 Chironomidae and Hydropsychidae as the more abundant families. Responses of 501 macroinvertebrates and diatoms to different metal levels in Hebden Beck have been 502 previously evaluated by the Environment Agency (2014). The ecological assessment revealed a relationship between increasing Zn concentrations and reduced biodiversity. 503 504 Likewise invertebrate data showed that responses of species diversity and composition 505 could be related to the influence of complex interactions between metals and acid scenarios.

506 In early and current assessment, the family Baetidae (O: Ephemeroptera) was dominant, 507 possibly due to its tolerance to different levels of pollution. A study by Mebane et al. (2015), 508 indicated that the species most sensitive to pollution are the last to re-colonise after recovery 509 from Cu contamination. In our study less sensitive families (Perlodidae, Nemouridae, 510 Baetidae) are widespread throughout the catchment. More sensitive families like 511 Glossomidae, Ephemerellidae and Heptageniidae are mainly confined to main channel sites 512 (H2, H6, H10), although Heptageniidae was also present in a metal impacted site H14 513 (predicted species richness of 55% of the theoretical maximum), another metal sensitive 514 family, Psychodidae was present in less metal impacted sites, such as H1P, showing a predicted species richness of 100%. These results are consistent with the F_{TOX} species 515 predictions, however, current interpretations are generalised as we only identified 516 517 macroinvertebrates to family level, and species sensitivity is complex, where differences in pollution tolerance occur within different taxa (Clements et al., 2013). 518

519 From a regulatory perspective, bioavailability based standards (EQS-B) offer a more 520 ecologically-based approach to address water quality failures than traditional hardness 521 corrections (EQS-H) (Peters et al., 2016), due to their parameterisation with organism 522 toxicity data or field species observations. In this study, both standards facilitated the 523 screening of sites with potential metal risk, however, EQS-B identified sites where dissolved 524 metal concentrations and speciation mechanisms (M-C_{DGT}) were related. Although, 525 bioavailability-based models have been developed on field data and applied in the 526 establishment of water quality standards for metals, their applicability for routine regulatory purposes need to be tested, especially for sites where physicochemical conditions (e.g. pH, 527 hardness, DOC) are near or outside of the boundaries of the model (Rüdel et al., 2015; 528 529 WFD-UKTAG, 2009a). In this study, some sites presented pH and Ca values below Zn-BLM limits, thus default limits of this tool (pH = 6 and Ca = 3mg/l) were used in Zn assessment, 530 531 consequently the use of complementary models like WHAM and WHAM F_{TOX} can offer more 532 complete toxicity predictions under original physico-chemical conditions. Because

533 WHAM F_{TOX} considers protons within its formulation, it can be applied to any combination of 534 water chemistry and metal mixtures. Bioavailability based EQS were implemented for 535 England and Wales in 2014 (DEFRA, 2014). For a practical implementation of these tools a 536 compliance scheme was suggested by the UK Technical Advisory Group, which includes: i) 537 comparison between metal concentrations with EQS-H, ii) application of BLM tools (e.g. Zn-BLM, Pb-BLM) to predict site-specific bioavailability (EQS-B), and iii) local refinement by 538 539 using models that require larger number of input parameters (e.g. full M-BLM or WHAM-VII) (Bio-met, 2015; WFD-UKTAG, 2009b). Further research is desirable, especially in the 540 541 generation of toxicity data to improve the significance of toxicity coefficients (e.g. α_{Pb}). 542 Nevertheless, the combination of EQS-B and WHAM- F_{TOX} could provide an improved 543 rational assessment which is essential for the purposes of considering remediation activities. Vulnerable sites can be prioritised under a holistic scenario (water chemistry, metal 544 concentrations and interactions between metal and organism) where remedial actions can 545 546 be focused on reducing metal bioavailability for the purpose of improving ecological 547 endpoints.

548 This study has evaluated different methods for the assessment of the chemical and ecological status of Hebden catchment regarding metal pollution. The first method (EQS-H) 549 550 was a simple and well-established technique, which provided a very conservative regulatory 551 limit that may be overprotective as seen when calculations are performed with tools 552 designed specifically to account for speciation and organisms interactions. The second (EQS-B) and third (WHAM- F_{TOX}) method based on more rigorous bioavailability tools (e.g. 553 554 Zn-BLM and Pb-SSD), provided less conservative limits but supported by the normalisation 555 of the ecotoxicity data to the site specific water chemistry. Through the evaluation of these approaches a better understanding of metal speciation and dynamics was possible, 556 557 supporting the prediction of toxic effects from metal mixtures and the identification of sites with potential biological risks. Furthermore, the biological survey provided a context for 558

relating the response of aquatic organisms to river water chemistry and pollutants, which areuseful for future ecological assessments.

561 **5. Conclusions**

- 562 The use of bioavailability-based standards is needed to fully assess the impacts of metal 563 pollution and implement regulatory actions for delivering good ecological status.
- In carboniferous catchments, metal toxic effects occur at circumneutral pH. Zn is mainly
 present as free ions and readily labile complexes while Pb is present as less labile forms
 (e.g. organic complexes and colloids).
- Metal assessment through the EQS-H approach is more conservative than EQS-B and WHAM- F_{TOX} . The incorporation of bioavailability will better address some of the processes that control toxic metals effects. In addition, catchment management will be improved as the bioavailability approach will aid in the identification of key reaches where remediation may be required.
- Metal concentrations estimated from DGT provided useful information for the assessment
 of Zn and Pb speciation and for understanding metal uptake mechanisms. However,
 results suggest that further research is required for evaluating the applicability of this tool
 in the assessment of metal bioavailability under field conditions.

576 Acknowledgments

577 This research was funded by the National Secretariat for Higher Education, Sciences, 578 Technology and Innovation of Ecuador (SENESCYT in Spanish) through a PhD scholarship 579 granted to M. Valencia-Avellan in June 2013 (Grant No. 82-ARG5-2013). We thank Dr Vicky 580 Hirst for her expertise in the performance of the biological survey and the Environment 581 Agency for sharing scientific reports and supporting this research. We also thank the 582 anonymous reviewers for their helpful comments.

583 **References**

584 Amato, E.D., Simpson, S.L., Jarolimek, C.V. and Jolley, D.F. (2014) Diffusive Gradients in 585 Thin Films Technique Provide Robust Prediction of Metal Bioavailability and Toxicity in 586 Estuarine Sediments. Environmental Science & Technology 48(8), 4485-4494.

587 Armitage, P.D., Moss, D., Wright, J.F. and Furse, M.T. (1983) The performance of a new 588 biological water quality score system based on macroinvertebrates over a wide range of 589 unpolluted running-water sites. Water Research 17(3), 333-347.

Balistrieri, L.S. and Blank, R.G. (2008) Dissolved and labile concentrations of Cd, Cu, Pb,
and Zn in the South Fork Coeur d'Alene River, Idaho: Comparisons among chemical
equilibrium models and implications for biotic ligand models. Applied Geochemistry 23(12),
3355-3371.

594 Balistrieri, L.S. and Mebane, C.A. (2014) Predicting the toxicity of metal mixtures. Science of 595 the Total Environment 466-467, 788-799.

596 Balistrieri, L.S., Nimick, D.A. and Mebane, C.A. (2012) Assessing time-integrated dissolved 597 concentrations and predicting toxicity of metals during diel cycling in streams. Science of the 598 Total Environment 425, 155-168.

599 Beane, S.J., Comber, S.D.W., Rieuwerts, J. and Long, P. (2016) Abandoned metal mines 600 and their impact on receiving waters: A case study from Southwest England. Chemosphere 601 153, 294-306.

Benedetti, M.F., Milne, C.J., Kinniburgh, D.G., Van Riemsdijk, W.H. and Koopal, L.K. (1995)
Metal Ion Binding to Humic Substances: Application of the Non-Ideal Competitive Adsorption
Model. Environmental Science & Technology 29(2), 446-457.

Bio-met. (2015) Technical guidance to implement bioavailability-based environmental quality
 standards for metals. [Online]. Available from: <u>http://bio-met.net/wp-</u>
 <u>content/uploads/2016/10/FINAL-TECHNICAL-GUIDANCE-TO-IMPLEMENT-</u>
 <u>BIOAVAILABILITYApril-2015.pdf</u>, [Accessed 5 March 2016].

- Bradac, P., Navarro, E., Odzak, N., Behra, R. and Sigg, L. (2009) Kinetics of cadmium
 accumulation in periphyton under freshwater conditions. Environmental Toxicology and
 Chemistry 28(10), 2108-2116.
- 612 Buffle, J.a.H., G. (2000) In situ Monitoring of Aquatic Systems. IUPAC Series in Analytical 613 and Physical Chemistry of Environmental Systems, Wiley, Chichester.
- Campbell, P.G.C. (1995) Metal Speciation and Bioavailability in Aquatic Systems. Tessier,
 A.a.T., D.R., (ed), pp. 45-102, John Wiley & Sons, London.

616 Cheng, T., Schamphelaere, K.D., Lofts, S., Janssen, C. and Allen, H.E. (2005) Measurement
617 and computation of zinc binding to natural dissolved organic matter in European surface
618 waters. Analytica Chimica Acta 542(2), 230-239.

Clements, W.H., Cadmus, P. and Brinkman, S.F. (2013) Responses of Aquatic Insects to Cu
 and Zn in Stream Microcosms: Understanding Differences Between Single Species Tests
 and Field Responses. Environmental Science & Technology 47(13), 7506-7513.

Davison, W. and Zhang, H. (1994) In situspeciation measurements of trace components in natural waters using thin-film gels. Nature 367(6463), 546-548.

- De Schamphelaere, K.A.C. and Janssen, C.R. (2002) A Biotic Ligand Model Predicting
 Acute Copper Toxicity for Daphnia magna: The Effects of Calcium, Magnesium, Sodium,
 Potassium, and pH. Environmental Science & Technology 36(1), 48-54.
- de Vries, W., Lofts, S., Tipping, E., Meili, M., Groenenberg, J.E. and Schütze, G. (2007)
 Reviews of Environmental Contamination and Toxicology, pp. 47-89, Springer New York,
 New York, NY.
- 630DEFRA. (2014) Water Framework Directive implementation in England and Wales: new and
updated standards to protect the water environment. Department for Environment, Food &
Rural Affairs. [Online]. Available from:
http://www.gov.uk/government/uploads/system/uploads/attachment_data/file/307788/river-
634
basin-planning-standards.pdf, [Accessed 10 April 2016].
- 635 Degryse, F., Smolders, E., Zhang, H. and Davison, W. (2009) Predicting availability of 636 mineral elements to plants with the DGT technique: a review of experimental data and 637 interpretation by modelling. Environmental Chemistry 6(3), 198-218.
- DGT Research. (2015) *DGT Research: For measurements in water, soils and sediments.* [Online]. Available from: <u>http://www.dgtresearch.com/</u>, [Accessed 1 November 2015].
- 640 Diviš, P., Dočekalová, H., Brulík, L., Pavliš, M. and Hekera, P. (2007) Use of the diffusive 641 gradients in thin films technique to evaluate (bio) available trace metal concentrations in river 642 water. Analytical and bioanalytical chemistry 387(6), 2239-2244.
- Dzombak, D.A. and Morel, F.M. (1990) Surface complexation modeling: hydrous ferric oxide,
 John Wiley & Sons.
- Environment Agency (2014) Ecological sampling to assess the impacts of metal pollutionHebden Beck and Ashfoldside Beck. Unpublished report.
- Environment Agency. (2011) *Chemical Standards Data Base.* [Online]. Available from:
 <u>http://evidence.environment-agency.gov.uk/ChemicalStandards/Home.aspx</u>, [Accessed 30
 March 2016].
- 650Environment Agency. (2008a) Abandoned mines and the water environment. Science project651SC030136-41.[Online].Availablefrom:652http://www.gov.uk/government/uploads/system/uploads/attachment_data/file/291482/LIT_8865379_df7d5c.pdf, [Accessed 20 March, 2017].
- Environment Agency. (2008b) Environmental Quality Standards for trace metals in the
 aquatic environment. Science Report SC030194/SR. [Online]. Available from:
 http://nora.nerc.ac.uk/3209/1/SCH00408BNXX-e-e.pdf, [Accessed 25 March 2016].
- 657 EU-STAR project. (2004) *UK Invertebrate sampling and analysis procedure.* [Online]. 658 Available from: <u>http://www.eu-star.at/pdf/RivpacsMacroinvertebrateSamplingProtocol.pdf</u>, 659 [Accessed 15 April 2016].
- European Commission (2000) Water Framework Directive 2000/60/EC. Official Journal L
 327, 22/12/2000 P. 0001 0073.
- Hall, J.R., Ashmore, M., Fawehinmi, J., Jordan, C., Lofts, S., Shotbolt, L., Spurgeon, D.J.,
 Svendsen, C. and Tipping, E. (2006) Developing a critical load approach for national risk
 assessments of atmospheric metal deposition. Environmental Toxicology and Chemistry
 25(3), 883-890.

Hamilton-Taylor, J., Ahmed, I.A.M., Davison, W. and Zhang, H. (2011) How well can we
predict and measure metal speciation in freshwaters? Environmental Chemistry 8(5), 461465.

Han, S., Naito, W., Hanai, Y. and Masunaga, S. (2013) Evaluation of trace metals
bioavailability in Japanese river waters using DGT and a chemical equilibrium model. Water
Research 47(14), 4880-4892.

- Harmsen, J. and Naidu, R. (2013) Bioavailability as a tool in site management. Journal ofHazardous Materials 261, 840-846.
- Hart, B.T. (1981) Trace metal complexing capacity of natural waters: a review.
 Environmental Technology 2(3), 95-110.
- He, E. and Van Gestel, C.A.M. (2015) Delineating the dynamic uptake and toxicity of Ni and
 Co mixtures in Enchytraeus crypticus using a WHAM-FTOX approach. Chemosphere 139,
 216-222.
- 679 International Lead Association. (2017) *Lead Chronic Biotic Ligand Model Tool.* [Online]. 680 Available from: <u>http://www.ila-lead.org/responsibility/lead-blm-tool</u>, [Accessed 2 July 2016].
- Jarvis, A.P. and Younger, P.L. (2000) Broadening the scope of mine water environmental
 impact assessment: a UK perspective. Environmental Impact Assessment Review 20(1), 85 96.
- Jones, A., Rogerson, M., Greenway, G. and Mayes, W.M. (2015) Zinc Uptake from
 Circumneutral Mine Drainage in Freshwater Biofilms: New Insights from In Vitro
 Experiments. Mine Water and the Environment 34(3), 295-307.
- Jones, A., Rogerson, M., Greenway, G., Potter, H.A.B. and Mayes, W.M. (2013) Mine water
 geochemistry and metal flux in a major historic Pb-Zn-F orefield, the Yorkshire Pennines,
 UK. Environmental Science and Pollution Research 20(11), 7570-7581.
- Kalis, E.J.J., Weng, Dousma, F., Temminghoff, E.J.M. and Van Riemsdijk, W.H. (2006)
 Measuring Free Metal Ion Concentrations in Situ in Natural Waters Using the Donnan
 Membrane Technique. Environmental Science & Technology 40(3), 955-961.
- Lock, K., Van Eeckhout, H., De Schamphelaere, K.A.C., Criel, P. and Janssen, C.R. (2007)
 Development of a biotic ligand model (BLM) predicting nickel toxicity to barley (Hordeum
 vulgare). Chemosphere 66(7), 1346-1352.
- Lofts, S. (2012) User's guide to WHAM7. Hydrology, N.C.f.E.a. (ed), Lancaster, UK.
- Lofts, S. and Tipping, E. (2011) Assessing WHAM/Model VII against field measurements of free metal ion concentrations: model performance and the role of uncertainty in parameters and inputs. Environmental Chemistry 8(5), 501-516.
- Luoma, S., Rainbow, P. (2008) Metal contamination in Aquatic Environments: Science and
 Lateral Management, Cambridge University Press, New York, United States.
- Mayes, W.M., Johnston, D., Potter, H.A.B. and Jarvis, A.P. (2009) A national strategy for identification, prioritisation and management of pollution from abandoned non-coal mine sites in England and Wales. I.: Methodology development and initial results. Science of the Total Environment 407(21), 5435-5447.

- Mebane, C.A., Eakins, R.J., Fraser, B.G. and Adams, W.J. (2015) Recovery of a miningdamaged stream ecosystem. Elementa 3.
- Merrington, G. (2016) Accounting for metal bioavailability in assessing water quality: A step change? Environ Toxicol Chem v. 35(no. 2), pp. 257-265.

Merrington, G. and Peters, A. (2013) *Development and use of the zinc bioavailability tool*(*draft*). United Kingdom Technical Advisory Group (WFD-UKTAG). Report:SC080021/1g_a.
[Online]. Available from: <u>http://www.wfduk.org/resources/category/environmental-standard-</u>
methods-203, [Accessed 15 July 2016].

- Namieśnik, J. and Rabajczyk, A. (2010) The speciation and physico-chemical forms of
 metals in surface waters and sediments. Chemical Speciation & Bioavailability 22(1), 1-24.
- Niyogi, S. and Wood, C.M. (2004) Biotic Ligand Model, a Flexible Tool for Developing SiteSpecific Water Quality Guidelines for Metals. Environmental Science & Technology 38(23),
 6177-6192.
- Omanović, D., Pižeta, I., Vukosav, P., Kovács, E., Frančišković-Bilinski, S. and Tamás, J.
 (2015) Assessing element distribution and speciation in a stream at abandoned Pb–Zn
 mining site by combining classical, in-situ DGT and modelling approaches. Science of the
 Total Environment 511, 423-434.
- Paquin, P.R., Gorsuch, J.W., Apte, S., Batley, G.E., Bowles, K.C., Campbell, P.G.C., Delos,
 C.G., Di Toro, D.M., Dwyer, R.L., Galvez, F., Gensemer, R.W., Goss, G.G., Hogstrand, C.,
 Janssen, C.R., McGeer, J.C., Naddy, R.B., Playle, R.C., Santore, R.C., Schneider, U.,
 Stubblefield, W.A., Wood, C.M. and Wu, K.B. (2002) The biotic ligand model: a historical
 overview. Comparative Biochemistry and Physiology Part C: Toxicology & Pharmacology
 133(1–2), 3-35.
- Peters, A., Schlekat, C.E. and Merrington, G. (2016) Does the scientific underpinning of
 regulatory tools to estimate bioavailability of nickel in freshwaters matter? The Europeanwide environmental quality standard for nickel. Environmental Toxicology and Chemistry
 35(10), 2397-2404.
- Puy, J., Uribe, R., Mongin, S., Galceran, J., Cecília, J., Levy, J., Zhang, H. and Davison, W.
 (2012) Lability Criteria in Diffusive Gradients in Thin Films. The Journal of Physical
 Chemistry A 116(25), 6564-6573.
- Qiu, H., Vijver, M.G., He, E., Liu, Y., Wang, P., Xia, B., Smolders, E., Versieren, L. and
 Peijnenburg, W.J.G.M. (2015) Incorporating bioavailability into toxicity assessment of Cu-Ni,
 Cu-Cd, and Ni-Cd mixtures with the extended biotic ligand model and the WHAM-F tox
 approach. Environmental Science and Pollution Research 22(23), 19213-19223.
- Rüdel, H., Díaz Muñiz, C., Garelick, H., Kandile, N.G., Miller, B.W., Pantoja Munoz, L.,
 Peijnenburg, W.J.G.M., Purchase, D., Shevah, Y., van Sprang, P., Vijver, M. and Vink,
 J.P.M. (2015) Consideration of the bioavailability of metal/metalloid species in freshwaters:
 experiences regarding the implementation of biotic ligand model-based approaches in risk
 assessment frameworks. Environmental Science and Pollution Research 22(10), 7405-7421.
- Søndergaard, J., Asmund, G., Johansen, P. and Rigét, F. (2011) Long-term response of an
 arctic fiord system to lead–zinc mining and submarine disposal of mine waste (Maarmorilik,
 West Greenland). Marine Environmental Research 71(5), 331-341.

Søndergaard, J., Bach, L. and Gustavson, K. (2014) Measuring bioavailable metals using
diffusive gradients in thin films (DGT) and transplanted seaweed (Fucus vesiculosus), blue
mussels (Mytilus edulis) and sea snails (Littorina saxatilis) suspended from monitoring buoys
near a former lead–zinc mine in West Greenland. Marine Pollution Bulletin 78(1–2), 102-109.

Steenbergen, N.T., Iaccino, F., de Winkel, M., Reijnders, L. and Peijnenburg, W.J. (2005)
Development of a biotic ligand model and a regression model predicting acute copper
toxicity to the earthworm Aporrectodea caliginosa. Environmental Science & Technology
39(15), 5694-5702.

Stockdale, A., Tipping, E., Lofts, S., Fott, J., Garmo, Ø.A., Hruska, J., Keller, B., Löfgren, S.,
Maberly, S.C., Majer, V., Nierzwicki-Bauer, S.A., Persson, G., Schartau, A.-K., Thackeray,
S.J., Valois, A., Vrba, J., Walseng, B. and Yan, N. (2014) Metal and proton toxicity to lake
zooplankton: A chemical speciation based modelling approach. Environmental Pollution 186,
115-125.

Stockdale, A., Tipping, E., Lofts, S., Ormerod, S.J., Clements, W.H. and Blust, R. (2010)
Toxicity of proton-metal mixtures in the field: Linking stream macroinvertebrate species
diversity to chemical speciation and bioavailability. Aquatic Toxicology 100(1), 112-119.

Tipping, E. (2002) Cation binding by humic substances, Cambridge University Press, Cambridge.

Tipping, E. (2005) Modelling Al competition for heavy metal binding by dissolved organic matter in soil and surface waters of acid and neutral pH. Geoderma 127(3), 293-304.

Tipping, E. (1994) WHAMC—A chemical equilibrium model and computer code for waters,
sediments, and soils incorporating a discrete site/electrostatic model of ion-binding by humic
substances. Computers & Geosciences 20(6), 973-1023.

Tipping, E., Jarvis, A.P., Kelly, M.G., Lofts, S., Merrix, F.L. and Ormerod, S.J. (2009) *Ecological indicators for abandoned mines, Phase 1: Review of the literature. Environment Agency Report:* SC030136/R49. [Online]. Available from:
<u>http://www.gov.uk/government/publications/ecological-indicators-for-abandoned-mines,</u>
[Accessed 30 July 2016].

Tipping, E. and Lofts, S. (2013) Metal mixture toxicity to aquatic biota in laboratory experiments: Application of the WHAM-FTOX model. Aquatic Toxicology 142–143, 114-122.

Tipping, E. and Lofts, S. (2015) Testing WHAM-FTOX with laboratory toxicity data for mixtures of metals (Cu, Zn, Cd, Ag, Pb). Environmental Toxicology and Chemistry 34(4), 780 788-798.

Tipping, E., Lofts, S. and Sonke, J. (2011) Humic Ion-Binding Model VII: a revised
parameterisation of cation-binding by humic substances. Environmental Chemistry 8(3), 225235.

Tipping, E., Rieuwerts, J., Pan, G., Ashmore, M.R., Lofts, S., Hill, M.T.R., Farago, M.E. and
Thornton, I. (2003) The solid–solution partitioning of heavy metals (Cu, Zn, Cd, Pb) in upland
soils of England and Wales. Environmental Pollution 125(2), 213-225.

Tipping, E., Vincent, C.D., Lawlor, A.J. and Lofts, S. (2008) Metal accumulation by stream bryophytes, related to chemical speciation. Environmental Pollution 156(3), 936-943.

Turner, G.S., Mills, G.A., Bowes, M.J., Burnett, J.L., Amos, S. and Fones, G.R. (2014)
Evaluation of DGT as a long-term water quality monitoring tool in natural waters; uranium as
a case study. Environmental Science: Processes & Impacts 16(3), 393-403.

Unsworth, E.R., Warnken, K.W., Zhang, H., Davison, W., Black, F., Buffle, J., Cao, J.,
Cleven, R., Galceran, J., Gunkel, P., Kalis, E., Kistler, D., van Leeuwen, H.P., Martin, M.,
Noël, S., Nur, Y., Odzak, N., Puy, J., van Riemsdijk, W., Sigg, L., Temminghoff, E., TercierWaeber, M.-L., Toepperwien, S., Town, R.M., Weng, L. and Xue, H. (2006) Model
Predictions of Metal Speciation in Freshwaters Compared to Measurements by In Situ
Techniques. Environmental Science & Technology 40(6), 1942-1949.

- Valencia-Avellan, M., Slack, R., Stockdale, A. and Mortimer, R.J.G. (2017) Understanding
 the mobilisation of metal pollution associated with historical mining in a carboniferous upland
 catchment. Environmental Science: Processes & Impacts 19(8), 1061-1074.
- Warnken, K.W., Davison, W. and Zhang, H. (2008) Interpretation of In Situ Speciation
 Measurements of Inorganic and Organically Complexed Trace Metals in Freshwater by
 DGT. Environmental Science & Technology 42(18), 6903-6909.
- Warnken, K.W., Davison, W., Zhang, H., Galceran, J. and Puy, J. (2007) In Situ
 Measurements of Metal Complex Exchange Kinetics in Freshwater. Environmental Science
 & Technology 41(9), 3179-3185.
- Warnken, K.W., Lawlor, A.J., Lofts, S., Tipping, E., Davison, W. and Zhang, H. (2009) In Situ
 Speciation Measurements of Trace Metals in Headwater Streams. Environmental Science &
 Technology 43(19), 7230-7236.
- Webb, J.A. and Keough, M.J. (2002) Quantification of copper doses to settlement plates in
 the field using diffusive gradients in thin films. Science of the Total Environment 298(1–3),
 207-217.
- WFD-UKTAG. Water Framework Directive United Kingdom Technical Advisory Group
 (WFD-UKTAG). [Online]. Available from: <u>http://www.wfduk.org/stakeholders/uktag-work-</u>
 programme, [Accessed 8 March 2016].
- 816 WFD-UKTAG. (2014) *Rivers & Lakes Guide to the Metal Bioavailability Assessment Tool* 817 (*M-BAT*). [Online]. Available from: <u>http://www.wfduk.org/resources/category/environmental-</u> 818 <u>standard-methods-203/tags/bioavailability-assessment-tool-205</u>, [Accessed 28 March 2016].
- WFD-UKTAG. (2009a) Desk top study to evaluate options outside BLM thresholds (low pH and low calcium issues). Project No. SC080021/8d. [Online]. Available from:
 <u>http://www.wfduk.org/resources%20/desk-top-study-evaluate-options-outside-blm-</u>
 <u>thresholds-low-ph-and-low-calcium-issues</u>, [Accessed 24 March 2016].
- WFD-UKTAG. (2009b) Using biotic ligand models to help implement environmental quality
 standards for metals under the Water Framework Directive. Science Report –
 SC080021/SR7b. [Online]. Available from: <u>http://www.wfduk.org/resources%20/using-biotic-</u>
 ligand-models-help-implement-eqs-metals, [Accessed 24 March 2016].
- Yin, H., Cai, Y., Duan, H., Gao, J. and Fan, C. (2014) Use of DGT and conventional methods
 to predict sediment metal bioavailability to a field inhabitant freshwater snail (Bellamya
 aeruginosa) from Chinese eutrophic lakes. Journal of Hazardous Materials 264, 184-194.
- Zhang, H. and Davison, W. (2015) Use of diffusive gradients in thin-films for studies of
 chemical speciation and bioavailability. Environmental Chemistry 12(2), 85-101.

Zhang, H. and Davison, W. (1995) Performance Characteristics of Diffusion Gradients in
Thin Films for the in Situ Measurement of Trace Metals in Aqueous Solution. Analytical
Chemistry 67(19), 3391-3400.

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Evaluating water quality and ecotoxicology assessment techniques using data from a lead and zinc effected upland limestone catchment

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Tables

BMWP score	ASPT score	Water Category	Interpretation
0-10		Very poor	Heavily polluted
11-40	<4	Poor	Polluted or impacted
41-70	>4	Moderate	Moderately impacted
71-100	>5	Good	Clean but slightly impacted
>100	>6	Very good	Unpolluted, un-impacted

Table 1. Scale of water quality categories based on BMWP and ASPT scores.

Table 2. Annual averages and standard deviation (\pm SD) of dissolved metals (M_D) in water samples (mol/l), metal concentrations from DGT devices-C_{DGT} (mol/l) and exceedances of Environmental Quality Standards (EQS) by comparing different tools EQS-H (a), EQS-B (Bioavailability assessment tool-BAT, Final screening tool-FST and Species sensitivity distribution-SSD) and WHAM-*F*_{TOX} (SR_{EPT} = 23 no toxic effects, SR_{EPT} < 23 toxic effects, and reduction of SR_{EPT} indicated as %) for assessing effects of Zn and Pb pollution. RCR: Risk Characterisation Ratio, PEC: Predicted Effect Concentration/ PNEC: Predicted No Effect Concentration of sites are indicated as reservoir (RS), main channel (MC), ephemeral tributaries (ET) and perennial tributaries (PT).

CHR HIM

		Zn _D -	7n-C	Zn <i>RCR</i>		Pb _D - Ph-Case			Pb RCR	WHAM-F _{TOX}		
Sites	Description	measured (mol/l)	(mol/)	EQS- H	EQS-B	measured (mol/l)	(mol/l)	EQS- H	EQS- B(FST)	EQS- B(SSD)	SR _{EPT}	%SR _{EPT}
H15	RS	3.1×10 ⁻⁵ (±7.2×10 ⁻⁶)	1.2×10 ⁻⁵ (±6.9×10 ⁻⁶)	253.5	54.0	1.3×10 ⁻⁶ (±4.0×10 ⁻⁷)	5.7×10 ⁻⁸ (±5.2×10 ⁻⁸)	38.8	18.9 (d)	8.5	10	43
H14	PT	2.1×10 ⁻⁵ (±5.2×10 ⁻⁶)	1.5×10 ⁻⁵ (±4.4×10 ⁻⁶)	174.7	45.9	7.6×10 ⁻⁷ (±3.6×10 ⁻⁷)	6.1×10 ⁻⁸ (±1.9×10 ⁻⁸)	21.9	15.8 (d)	6.8	13	55
H13	ET	3.2×10 ⁻⁶ (±6.3×10 ⁻⁶)	6.1×10 ⁻⁷ (±3.2×10 ⁻⁷)	25.9	4.1 (b)	9.9×10 ⁻⁷ (±2.6×10 ⁻⁷)	1.1×10 ⁻⁷ (±6.6×10 ⁻⁸)	28.6	10.2 (d)	1.1 (e)	12	52
H12	ET	6.5×10⁻⁵ (±3.0×10⁻⁵)	2.7×10 ⁻⁵ (±4.0×10 ⁻⁶)	85.0	117.6	1.4×10 ⁻⁶ (±4.4×10 ⁻⁷)	1.1×10 ⁻⁷ (±4.0×10 ⁻⁸)	39.5	35.0 (d)	17.1	15	66
H11	PT	6.3×10 ⁻⁶ (±2.1×10 ⁻⁵)	3.3×10 ⁻⁶ (±7.6×10 ⁻⁶)	51.3	8.7	1.5×10 ⁻⁷ (±4.5×10 ⁻⁷)	9.2×10 ⁻⁹ (±1.8×10 ⁻⁸)	4.4	1.6 (d)	0.8	20	85
H10	MC	6.8×10 ⁻⁶ (±2.8×10 ⁻⁶)	5.2×10 ⁻⁶ (±2.5×10 ⁻⁶)	55.6	11.1	3.8×10 ⁻⁷ (±2.0×10 ⁻⁷) ∕	2.3×10 ^{⁻8} (±1.5×10 ^{⁻8})	11.2	6.5 (d)	2.9	19	82
H9	PT	1.5×10 ^{⁻6} (±1.9×10 ^{⁻6})	1.3×10 ⁻⁶ (±5.5×10 ⁻⁷)	12.0	2.1 (b)	1.0×10 ⁻⁷ (±1.1×10 ⁻⁷)	4.6×10 ⁻⁹ (±4.4×10 ⁻⁹)	2.9	1.0 (d)	0.5	18	77
H8	PT	4.1×10 ⁻⁶ (±1.5×10 ⁻⁶)	3.9×10 ⁻⁶ (±2.3×10 ⁻⁶)	3.6	7.4	9.2×10 ⁻⁸ (±3.2×10 ⁻⁸)	1.1×10 ^{⁻8} (±1.1×10 ^{⁻8})	2.6	2.0 (d)	1.1	23	100
H7	PT	4.9×10 ⁻⁵ (±1.3×10 ⁻⁵)	2.8×10 ⁻⁵ (±8.5×10 ⁻⁶)	42.9	209.2	2.1×10 ⁻⁸ (±6.1×10 ⁻⁸)	1.2×10 ⁻⁹ (±1.1×10 ⁻⁹)	0.6	3.0 (d)	1.7	18	76
H6	MC	7.8×10 ⁻⁶ (±1.5×10 ⁻⁶)	7.5×10 ⁻ ⁶ (±1.0×10 ⁻⁶)	10.2	12.8	2.4×10 ⁻⁷ (±9.9×10 ⁻⁸)	8.5×10 ⁻⁹ (±4.4×10 ⁻⁹)	6.9	3.8 (d)	1.7	20	88
H5	PT	1.3×10 ⁻⁵ (±2.9×10 ⁻⁶)	1.3×10 ⁻⁵ (±1.5×10 ⁻⁶)	11.6	43.9	2.6×10 ⁻⁷ (±2.2×10 ⁻⁷)	4.9×10 ⁻⁸ (±1.8×10 ⁻⁸)	7.5	14.6 (d)	8.7	23	100
H4	ET	5.5×10 ⁻⁷ (±1.4×10 ⁻⁷)	2.6×10 ⁻⁷ (±8.9×10 ⁻⁸)	4.5	2.5 (b,c)	2.1×10 ⁻⁸ (±8.2×10 ⁻⁹)	3.7×10 ⁻⁹ (±5.4×10 ⁻¹⁰)	0.6	0.4	0.02 (e)	8	35
H3	PT	<5.1×10 ⁻⁷	1.2×10 ⁻⁶ (±5.1×10 ⁻⁸)	0.2	0.8	8.0×10 ⁻¹⁰ (±9.5×10 ⁻¹⁰)	5.4×10 ⁻¹⁰ (±4.1×10 ⁻¹⁰)	0.02	0.1	0.1	23	100
H2	MC	7.4×10 ⁻⁶ (±1.1×10 ⁻⁶)	7.0×10 ⁻⁶ (±1.0×10 ⁻⁶)	9.7	17.7	1.9×10 ⁻⁷ (±1.4×10 ⁻⁷)	1.3×10 ⁻⁸ (±3.9×10 ⁻⁹)	5.5	4.4 (d)	2.3	22	97
H1P	PT	5.6×10 ⁻⁷ (±1.1×10 ⁻⁶)	4.8×10 ⁻⁹ (±4.5×10 ⁻⁹)	0.7	2.4	3.5×10 ⁻⁹ (±3.1×10 ⁻⁹)	1.1×10 ⁻¹⁰ (±3.0×10 ⁻¹¹)	0.3	0.2	0.1	23	100
H1	MC	2.4×10 ⁻⁶ (±1.1×10 ⁻⁶)	2.5×10 ⁻⁶ (±6.6×10 ⁻⁷)	2.1	4.2	8.0×10 ⁻⁸ (±6.6×10 ⁻⁸)	3.7×10 ⁻⁹ (±3.3×10 ⁻⁹)	2.3	1.8 (d)	1.0	23	100

(a) Limits of EQS based on site-specific hardness, Zn.EQS-H= 1.3×10⁷ (for sites H15, H14, H13, H11, H10, H9, H4); 7.6×10⁷ (for sites H12, H6, H2, H1P) and 1.2×10⁶ (for sites H8, H7, H5, H3, H1). Pb.EQS-H= 3.5×10⁸ for all sites.
(b) pH values were outside the limits of Zn-BAT tool. Default limit (pH=6) was applied for Zn.EQS-B calculations.
(c) Ca concentrations were outside the limits of Zn-BAT tool. Default limit (Ca=3 mg/l) was applied for Zn.EQS-B calculations.
(d) Measured Pb exceeds the site specific PNEC of the Pb-FST tool.

(e)	рН	values	were	outside	the	limits	of	Pb-SDD	tool	(pH=6=8.5).	Pb-PNEC	should	be	interpreted	with	care.
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Table 3. Predicted species richness of macroinvertebrates orders (Ephemeroptera, Plecoptera and Trichoptera) (SR_{EPT}), where SR_{EPT} = 23 indicates no toxic effects and SR_{EPT} < 23 denotes toxic effects. Toxicity values (F_{TOX}) obtained by WHAM- F_{TOX} were calculated from monthly data. Values are expressed as mean and standard deviation (±SD).

Sites	SR _(EPT)	Total_ <i>F_{TOX}</i>	F _{TOX-Zn}	F _{TOX-Pb}	F _{тох-н}	F _{TOX-AI}
H15	10	4.0	1.7(±0.7)	0.2(±0.1)	1.4(±0.4)	0.4(±0.5)
H14	13	3.6	1.6(±0.5)	0.1(±0.0)	1.2(±0.3)	0.3(±0.5)
H13	12	3.7	0.0(±0.1)	0.0(±0.0)	2.6(±0.3)	1.4(±0.3)
H12	15	3.3	2.3(±0.9)	0.2(±0.1)	1.0(±0.3)	0.2(±0.5)
H11	20	2.8	0.1(±0.3)	0.0(±0.0)	1.8(±0.6)	0.3(±0.2)
H10	19	2.8	0.7(±0.3)	0.1(±0.0)	1.2(±0.4)	0.1(±0.1)
Н9	18	3.0	0.1(±0.1)	0.0(±0.0)	2.0(±0.7)	0.4(±0.2)
H8	23	1.7	0.5(±0.2)	0.0(±0.0)	0.9(±0.3)	0.1(±0.1)
H7	18	3.0	2.0(±0.3)	0.0(±0.0)	1.0(±0.2)	0.0(±0.1)
H6	20	2.7	0.8(±0.2)	0.0(±0.0)	1.2(±0.3)	0.2(±0.3)
Н5	23	2.3	0.9(±0.2)	0.1(±0.1)	1.1(±0.2)	0.1(±0.2)
H4	8	4.2	0.0(±0.0)	0.0(±0.0)	3.3(±0.4)	0.9(±0.3)
Н3	23	1.2	0.1(±0.0)	0.0(±0.0)	1.0(±0.3)	0.0(±0.1)
H2	22	2.4	0.6(±0.2)	0.0(±0.0)	1.3(±0.4)	0.2(±0.3)
H1P	23	1.9	0.1(±0.0)	0.0(±0.0)	1.5(±0.6)	0.2(±0.1)
H1	23	1.5	0.3(±0.0)	0.0(±0.1)	0.8(±0.2)	0.0(±0.1)
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Figures Legends

Figure 1. Hebden Beck discharging into the River Wharfe. Black dots represent sampling stations, they are ordered from upper to lower catchment, therefore from H15 to H1. Black arrows indicate the direction of flow.

Figure 2. Measured dissolved metals versus DGT concentrations (C_{DGT}). Panel a, shows the correlation between Zn_D and $Zn-C_{DGT}$. Panel b, shows the correlation of Pb_D with Pb-C_{DGT}. Dotted lines represent a regression line, solid lines represent the ideal 1:1 line.

Figure 3. Calculated concentrations of metal species in the absence of surface active oxide precipitates using WHAM/Model VII.

Figure 4. Relationship between inorganic metal concentrations (mol/l) calculated by WHAM and metal concentrations from C_{DGT} measurements (mol/l) for Zn (panels a and b) and Pb (panels c and d). Inorganic concentrations were predicted in WHAM in the absence (panels a and c) and presence (panels b and d) of active oxides. Dotted lines represent a regression line and solid lines represent the ideal 1:1 line.

Figure 5. Calculated concentrations of Pb species in the presence of surface active oxide precipitates using WHAM/Model VII.

Figure 6. Calculated concentrations of Zn and Pb bound to humic acids (mmol/g).

Figure 7. Variations of toxicity function (F_{TOX}) with pH. Panel a shows total F_{TOX} , which includes the effect of metals derived from mine-affected areas and acid contribution from peat soils. Panel b shows the contribution of only H and Al to F_{TOX} . The horizontal dashed line represents the F_{TOX} threshold (2.3), above which toxic effect are predicted to occur. The upper limit of the plot represents the upper threshold (5.2).

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Highlights

- In carboniferous catchments, Zn and Pb toxic effects occur at circumneutral pH.
- Metal assessment with EQS-H tools is more conservative than EQS-B and WHAM-F_{TOX}.
- Bioavailability-based tools support remediation to reach good ecological status.