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4 **Humic Ion-Binding Model VII: a revised parameterisation of cation-**
5 **binding by humic substances**

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28 Environmental context

29 Natural organic matter exerts a powerful control on chemical conditions in waters and soils,
30 affecting pH and influencing the biological availability, transport and retention of metals. To
31 quantify the reactions, we collated a wealth of laboratory data covering 40 metals and acid-
32 base reactions, and used them to parameterise the latest in a series of Humic Ion-Binding
33 Models. Model VII is now available to interpret field data, and contribute to the prediction of
34 environmental chemistry.

35

36 **Abstract**

37 Humic Ion-Binding Model VII aims to predict the competitive reactions of protons and metals
38 with natural organic matter in soils and waters, based on laboratory results with isolated
39 humic and fulvic acids (HA and FA). Model VII is simpler in its postulated multidentate metal
40 binding sites than the previous Model VI. Three model parameters were eliminated by using
41 a formal relationship between monodentate binding to strong- and weak-acid oxygen-containing
42 ligands, and removing factors that provide ranges of ligand binding strengths. Thus Model VII
43 uses a single adjustable parameter, the equilibrium constant for monodentate binding to strong-
44 acid (carboxylate) groups (K_{MA}), for each metallic cation. Proton-binding parameters, and mean
45 values of $\log K_{MA}$ were derived by fitting 248 published datasets (28 for protons, 220 for
46 cationic metals). Default values of $\log K_{MA}$ for FA were obtained by combining the fitted values
47 for FA, results for HA, and the relationship for different metals between $\log K_{MA}$ and equilibrium
48 constants for simple oxygen-containing ligands. The equivalent approach was used for HA.
49 The parameterised model improves on Model VI by incorporating more metals (40), providing
50 better descriptions of metal binding at higher pH, and through more internally-consistent
51 parameter values.

52

53

54 Introduction

55 The Windermere Humic Aqueous Model (WHAM)^[1,2] incorporating Humic Ion-Binding Model
56 V^[3] or VI^[4] permits the calculation of equilibrium chemical speciation for waters and soils in
57 which natural organic matter plays a significant role. The ion-binding models are based on
58 conventional chemical reactions involving O-containing weak acids, with empirical estimation
59 of the influence of soft ligand atoms (N, S) and electrostatic corrections, and are
60 parameterised from laboratory studies with isolated humic and fulvic acids. The NICA
61 model^[5] is similarly parameterised and provides an alternative picture based on continuous
62 binding-site distributions. Tipping^[2] identified both the Humic Ion-Binding Models and NICA
63 as comprehensive models, meaning that they deal with competitive interactions involving all
64 cations (including H⁺), and take account of ionic strength effects and metal-proton exchange
65 ratios. They seek to represent cation-binding by the complex mixtures that comprise natural
66 organic matter as efficiently as possible, with the minimum number of parameters, in order to
67 be useful in addressing chemical processes in the environment. A different approach to
68 these parameterised models, but also potentially comprehensive, is the “forward modelling”
69 developed by Cabaniss^[6] in which binding is calculated *a priori* from the known or assumed
70 distributed chemistry of humic substances.

71 WHAM has been applied in a variety of research and regulatory areas. Examples include
72 the acidification of soils^[7-14] and surface waters^[15], trace metal behaviour in soils^[16-22], surface
73 waters^[23-31] and groundwaters^[32], lake sediment diagenesis^[33,34], rare earth geochemistry<sup>[35-
74 37]</sup>, iron and manganese geochemistry^[38-41], radionuclide geochemistry^[42-45], organic matter
75 solubility in soils^[46,47], catchment modelling^[48,49], interactions of metals with biota^[50,51],
76 ecotoxicology^[52-59] and Critical Loads^[60-62]. Given this evident utility, it is worthwhile to
77 continue to improve the humic ion-binding model and incorporate new data into its
78 parameterisation. Here we report on activities undertaken towards these goals, namely
79 modification of assumptions about multidentate binding, the fitting of new data, and the
80 introduction of a procedure to obtain more internally-consistent parameters.

81 Changes in binding site formulation were prompted by experience in applying Model VI to
82 new data for the binding of lanthanides, Co and UO₂ by humic and fulvic acids^[63,64]. It
83 became apparent that too strong a pH dependence was predicted by the model at higher pH
84 values, which could be attributed to assumed multidentate sites involving more than one
85 weak-acid ligand (e.g. phenolic oxygen). Therefore we modified the formulation of the array
86 of assumed binding sites, to create Humic Ion-Binding Model VII.

87 Humic Ion-Binding Model VI is parameterised with data for the interactions of cationic metals
88 with isolated humic substances that were available in the late 1990s. Since then, the number

89 of data sets suitable for parameterisation has approximately doubled, with new results
90 notably available for protons, Al, Sc, Cr, Co, Ni, Cu, Zn, Y, Ag, Cd, Hg, MeHg, Pb, UO₂ and
91 the lanthanides. All available data sets were fitted with Model VII to obtain binding
92 parameters.

93 In past work^[2,4] linear free energy relationships (LFERs) were derived to relate model
94 parameters for metal binding to conventional equilibrium constants for simple ligands, and
95 the LFERs were used in some cases to estimate parameters in cases where measured data
96 were not available. We extended this approach, making use of the study of Carbonaro &
97 DiToro^[65] who showed how the Irving-Rossotti^[66] approach could be brought to bear to
98 regularise equilibrium constants.

99

100

101

102 **Modelling**

103 *WHAM*

104 The original version of WHAM was simply a combination of a humic ion-binding model (see
105 below) with an inorganic speciation model^[1]. The latter deals with reactions among the
106 inorganic master species (protons, metal cations, hydroxyl ion, carbonate and phosphate
107 species, sulphate, fluoride, chloride). Ionic strength effects on the inorganic reactions are
108 taken into account using the extended Debye-Hückel equation. Temperature effects on
109 reactions between inorganic species are taken into account using published or estimated
110 enthalpy data; in the absence of experimental information, reactions involving humic
111 substances are assumed to be independent of temperature. A given speciation problem is
112 solved by finding the activities of all the master species at equilibrium, using a combination of
113 algorithms. Inputs to the model are the total concentrations of reactants, as would be
114 obtained by chemical analysis. The model can work with a specified pH, or calculate the pH
115 if the necessary input data are available. The software package currently associated with the
116 model is WHAM6.0 (http://windermere.ceh.ac.uk/Aquatic_Processes/wham/). It includes the
117 Humic Ion-Binding Model VI and the inorganic model, together with further sub-models for
118 the binding of cations to the oxides of Al, Si, Mn and Fe, and to a simple cation-exchanger.

119 *Humic Ion-Binding Model VI*

120 Humic Ion-Binding Model VI is the most important component of WHAM, describing the
121 interactions of protons and metals with natural organic matter. The model was described in
122 detail by Tipping^[4]. It uses a structured formulation of discrete, chemically-plausible, binding
123 sites for protons, in order to allow the creation of regular arrays of bidentate and tridentate
124 binding sites for metals.

125 Proton dissociation is represented by postulating 8 groups with different acid strengths, the
126 reactions being characterised by intrinsic equilibrium constants, the negative logarithms of
127 which are denoted by $pK_1 - pK_8$. The four most strongly-acid groups (groups 1-4) are referred
128 to as type A groups, and consist mainly of carboxylic acid groups, while the remaining 4 groups
129 (type B) represent weaker acids, such as phenolic acids. The 8 pK_i values are expressed in
130 terms of 4 constants; pK_A and pK_B are the average pK values of the two types of group, and
131 ΔpK_A and ΔpK_B are measures of the spread of the individual pK_i values around the means.
132 Each type A group is assigned an abundance of $n_A/4$ mol g^{-1} humic matter, and each type B
133 group an abundance of $n_A/8$ mol g^{-1} . Thus, within a type, each group is present in equal
134 amounts, and there are half as many type B groups as type A groups. The imposed regularity
135 of the groups facilitates the formulation of bidentate and tridentate sites for metals (Table 1).

136 Metal binding at the type A and B sites is described with average intrinsic equilibrium
 137 constants (K_{MA} , K_{MB}) and associated “spread factors” ΔLK_{A1} and ΔLK_{B1} . Thus K_{MA} is the
 138 average equilibrium constant for the binding of a metal to a type A (carboxylate) group. The
 139 occurrence of bidentate and tridentate sites at the surface of the humic acid or fulvic acid
 140 molecule is calculated probabilistically. Additional binding site heterogeneity is generated by
 141 a parameter, ΔLK_2 , that characterises the tendency of the metal to interact with “softer” ligand
 142 atoms such as N and S. Thus, 9% of the bidentate sites have the logarithms of their binding
 143 constants increased by ΔLK_2 , while 0.9% have increases of $2\Delta LK_2$. For the tridentate sites, the
 144 respective increases are $1.5\Delta LK_2$ and $3\Delta LK_2$. In the standard model, all metal cations (e.g.
 145 Al^{3+} , Cu^{2+} , Hg^{2+}) and their first hydrolysis products ($AlOH^{2+}$, $CuOH^+$, $HgOH^+$) compete with
 146 each other, and with protons, for binding. The combination of multi-denticity and the
 147 increased binding strength of some sites, due to ΔLK_2 , generates many binding sites with a
 148 wide range of affinities. The most abundant (monodentate) sites are the weakest binders,
 149 while the least abundant (tridentate sites enhanced by $3\Delta LK_2$) are the strongest.

150 The intrinsic equilibrium constants are modified by empirical electrostatic terms, incorporating
 151 the electrostatic parameter P , that take into account the attractive or repulsive interactions
 152 between ions and the charged macromolecule. A Donnan sub-model is used to compute
 153 counterion accumulation in the diffuse zone around the molecule; each counterion can be
 154 assigned a selectivity coefficient (K_{sel}), so that accumulation can be made to depend on more
 155 than just the counterion charge; for example, Ca^{2+} can be favoured over Mg^{2+} . The selectivity
 156 coefficients are only used in soil applications where exchanges of major cations on solid-phase
 157 organic matter are important.

158 The maximum number of parameters that can be optimised to describe metal binding is six
 159 (K_{MA} , K_{MB} , ΔLK_{A1} , ΔLK_{B1} , ΔLK_2 , K_{sel}). In practice however, this number can be substantially
 160 reduced. Thus, Tipping^[4] described the setting of a single universal value for ΔLK_{A1} and ΔLK_{B1} ,
 161 and the estimation of ΔLK_2 by correlation with the logarithm of the equilibrium constant for
 162 complex formation with NH_3 ($\log K_{NH3}$) according to the equation;

$$163 \quad \Delta LK_2 = 0.58 \log K_{NH3} \quad (1)$$

164 For dilute systems, as in laboratory experiments, K_{sel} can be set to unity. Finally, K_{MA} and K_{MB}
 165 are strongly correlated. Therefore, the fitting of a new data set can be achieved by adjusting
 166 only K_{MA} , which was the approach taken in the present work. High values of K_{MA} mean that the
 167 metal is strongly bound at the high-abundance “weak” sites. High values of ΔLK_2 mean that
 168 the metal is favoured by the low-abundance “strong” sites, associated, according to the model,
 169 with N or S atoms. If ΔLK_2 is small, the strong sites are not favoured, and binding is
 170 predominantly due to binding at oxygen-containing sites.

171 During the course of developing Model VII from Model VI, we discovered a coding error in
 172 Model VI which means that the parameters ΔLK_{A1} and ΔLK_{B1} were not used as originally
 173 intended, and this means that Model VI was actually different from that described by
 174 Tipping^[4]. This error, which is described in full in the Accessory Material, did not invalidate
 175 Model VI as used with code written by ourselves, since fitting and model applications were
 176 performed consistently. As reported below, ΔLK_{A1} and ΔLK_{B1} are entirely absent from Model
 177 VII.

178 *Humic Ion-Binding Model VII*

179 Model VII is identical to VI with respect to its formulation of total monodentate binding sites,
 180 proton binding constants and electrostatic effects. It differs from Model VI with respect to
 181 metal cation binding, in that the arrangement of multidentate sites has been modified, and
 182 some parameters eliminated.

183 In Model VI there can be four parameters that describe monodentate metal binding, namely
 184 $\log K_{MA}$, $\log K_{MB}$, ΔLK_{A1} and ΔLK_{B1} . Tipping^[2] noted that the relationship between $\log K_{MB}$ and
 185 $\log K_{MA}$, i.e. mean equilibrium constants for binding to weaker and stronger acid sites, was
 186 roughly as expected on the basis of data for simple ligands, i.e. binding of a given metal (also
 187 the proton) to weak-acid groups such as phenolic OH groups is consistently stronger than to
 188 carboxylate groups. The work of Carbonaro & Di Toro^[65] showed this much more generally,
 189 and demonstrated that relative binding strengths of a given metal to different oxygen-
 190 containing ligands are interrelated. Therefore we defined $\log K_{MB}$ formally by the equation;

$$191 \quad \log K_{MB} = \log K_{MA} \times (pK_B / pK_A) \quad (2)$$

192 In other words, the relative binding strengths for a given metal are the same as those for the
 193 proton. We attempted to apply this idea also to the spread factors ΔLK_{A1} and ΔLK_{B1} , i.e. to
 194 predict them from $\Delta pK_{H,A}$ and $\Delta pK_{H,B}$. However, when fitting the data for metal binding, we
 195 found that the spread factors could be entirely eliminated, i.e. it was preferable to fix $\log K_{MA}$
 196 at the same value for each of the four type A sites, and $\log K_{MB}$ at the same value for each of
 197 the four type B sites.

198 Multi-dentate sites are formed in the model by combining mono-dentate (proton-binding)
 199 sites, but the choice of combinations has been found to be important. When fitting new data
 200 for Co, lanthanides and UO_2 with Model VI, we found overestimation of the pH dependence
 201 of binding at $pH > 7$. Examination of model outputs revealed that this was due to the
 202 assumed presence in humic matter of binding sites containing 2 or 3 weak-acid (type B)
 203 groups. Therefore multidentate sites containing more than one type B group are omitted
 204 from Model VII.

205 *Data sets*

206 The number of data sets available to calibrate Model VII was approximately twice as many
207 as for Model VI. For protons, we used 15 data sets for HA (4313 data points in all) and 13
208 for FA (4334 data); several of the sets previously used for Model VI were abandoned
209 because they were incomplete. A total of 107 data sets were available to quantify the
210 binding of 36 different cationic metals by HA (4420 data) and 108 data sets for the binding of
211 34 different metals by FA (4004 data). The grand total of data points was 17,116. Most of
212 the previous metal data sets used by Tipping^[4] were retained, and augmented with additional
213 data sets obtained from the collation of Milne and colleagues^[67,68], and by searching the
214 literature. The data sources are summarised in the Accessory Material.

215 *Data fitting*

216 The model was coded in BASIC, and the Nelder-Mead polytope method used for function
217 minimisation and parameter estimation. Since the previous fitting of proton binding data, a
218 significant number of new studies have been published. The availability of these extra data
219 made it possible to apply stronger acceptability criteria, with the objective of selecting data
220 best suited to provide robust estimates of the proton binding parameters. The criteria were
221 (a) the pH range of the data set had to extend above pH 10, in order to obtain good
222 estimates of the weak acid site binding parameters $pK_{H,B}$ and $\Delta pK_{H,B}$, and (b) each data set
223 had to refer to several ionic strengths, so that a value of the ionic strength dependency
224 parameter P could be calculated for each data set. Fitting involved the optimisation of six
225 parameters, namely the strong acid site density (n_A), the average strong and weak acid
226 binding site pK values ($pK_{H,A}$ and $pK_{H,B}$), the factors giving the spread of pK values around
227 the averages ($\Delta pK_{H,A}$ and $\Delta pK_{H,B}$), and the electrostatic factor P . We found that when fitting
228 individual data sets, adjustment of all the parameters simultaneously produced values that
229 were sometimes physically unreasonable. Therefore we adopted a two stage fitting process.
230 Firstly, $\Delta pK_{H,A}$ and $\Delta pK_{H,B}$ were fixed at the values derived by Tipping^[4], and all the data sets
231 were fitted individually to obtain values of the remaining four parameters, and an overall
232 goodness-of-fit parameter (sum of all squared deviations in HA or FA charge, Z), calculated
233 from all the data sets. Then the spread factors were adjusted and the process repeated
234 iteratively to minimise the goodness-of-fit parameter.

235 Metal binding data were fitted by optimising $\log K_{MA}$, using the default parameters obtained
236 from proton-binding data, and with ΔLK_2 obtained from equation (1). In the great majority of
237 cases, optimisation was done by minimising squared errors in $\log v$, where v is the moles of
238 metal bound per gram of humic matter. In a few cases, the effects of metal binding on
239 measured pH were modelled, and optimisation performed by minimising squared errors in

240 pH. To establish the new arrangement of multidentate sites, we forced a uniform
241 representation of the monodentate sites, while keeping the system as simple as possible. All
242 data sets were fitted with a number of trial multidentate arrangements and universal values of
243 ΔLK_{A1} and ΔLK_{B1} (i.e. the same values for all data sets). The best arrangement of sites (Table
244 1) requires 50 different binding sites rather than the 80 sites of Model VI. Overall fitting was
245 no worse if ΔLK_{A1} and ΔLK_{B1} were both set to zero, enabling these parameters to be
246 eliminated. Thus, for dilute systems, Model VII has only two formal parameters for each
247 cationic metal, namely $\log K_{MA}$ and ΔLK_2 , as opposed to the potential five in Model VI
248 (although this number could be reduced to three in practice).

249

250 Results

251 Default parameter values for proton binding and ionic strength dependence were calculated
 252 as the means of the values obtained from each individual data set. New and previous
 253 parameter values are compared in Table 2, while Figure 1 compares calculated humic and
 254 fulvic net charge as a function of pH for the two parameter sets.

255 Mean values of $\log K_{MA}$ for the different metals are shown in the fourth and fifth columns of
 256 Table 3. The average root-mean-squared deviation in $\log v$ was 0.21 for HA and 0.23 for FA,
 257 and overall the fits with Model VII were marginally better than those with Model VI. Figure 2
 258 shows how Model VII performs better than Model VI for lanthanum.

259 We applied the approach and findings of Carbonaro & Di Toro^[65] to analyse the Model VII
 260 results, by plotting $\log K_{MA}$ against α_O , the slope of the equation of Irving & Rossotti^[66] for
 261 ligands with oxygen donor atoms. Results for HA and FA (Figure 3) show reasonable
 262 correlations between $\log K_{MA}$ and α_O , indicating that HA and FA behave approximately as
 263 expected from simpler ligands with respect to binding at the major oxygen-containing ligand
 264 sites. A plot of $\log K_{MA,HA}$ against $\log K_{MA,FA}$ (Figure 4) falls close to the expected line, which
 265 has a slope of 1.11 on the basis of the pK_A values for HA and FA, which are 4.1 and 3.7
 266 respectively. Thus $\log K_{MA,HA}$ for each metal is expected to be greater than $K_{MA,FA}$ by a factor
 267 of 1.11. The mean ratio ($\log K_{MA,HA} / \log K_{MA,FA}$) for the 33 metals with constants for both HA
 268 and FA was 1.09, supporting this expectation.

269 We used these results to improve estimates of $\log K_{MA}$, and thereby reduce the possibility of
 270 excessive outliers. To derive the default constant for the binding of a given metal to FA we
 271 applied the equation;

$$272 \quad \log K_{MA,FA,def} = \{n_{FA} \log K_{MA,FA,mean} + (n_{HA} \log K_{MA,HA,mean} / 1.11) + (3.81\alpha_O + 0.37)\} / (n_{FA} + n_{HA} + 1) \quad (3)$$

273 where n_{FA} and n_{HA} are the numbers of datasets for FA and HA yielding estimates of the
 274 parameter $\log K_{MA}$, and $\log K_{MA,FA,mean}$ and $\log K_{MA,HA,mean}$ are the mean values obtained from
 275 fitting. Thus, we first weight the mean $\log K_{MA}$ value for FA. Second we weight the results
 276 for HA, taking into account that the average $\log K_{MA}$ for HA is 1.11 times the value for FA
 277 (Figure 4). Then we add a prediction of $\log K_{MA}$ using the equation from Figure 3. Finally the
 278 overall weighted mean is taken. For HA, the same approach leads to

$$279 \quad \log K_{MA,HA,def} = \{n_{HA} \log K_{MA,HA,mean} + 1.11 n_{FA} \log K_{MA,FA,mean} + (3.51\alpha_O + 0.74)\} / (n_{HA} + n_{FA} + 1) \quad (4)$$

280 Consequently the default parameters are more internally consistent. In a number of cases
 281 there is no available value of α_O , while for three cations (Be^{2+} , Fe^{2+} and Ba^{2+}) there is a value
 282 of α_O but there are no data for humic substances. The equations can still be applied under

283 either circumstance, but omitting the missing values. The seventh and eighth columns of
284 Table 3 show the derived default $\log K_{MA}$ values for 40 metallic cations.

285 *Comparisons of outputs from Models VI and VII*

286 Differences between the models with respect to multidentate binding site arrangements and
287 proton binding parameters generally lead to somewhat larger values of $\log K_{MA}$, especially for
288 HA (Table 4). Differences will also have arisen from the use of additional data, and the new
289 procedure for deriving default constants. The best-defined values of $\log K_{MA}$ are those of Cu,
290 for which $\log K_{MA}$ for Model VII exceeds the Model VI value by 0.4 for HA but only 0.1 for FA;
291 these differences can be used as references to compare metals between models (Table 4).
292 Thus, for HA the differences in $\log K_{MA}$ between Models VII and VI are largest for Cr(III), Mn,
293 Fe(III), Ba, Eu and Th, while for FA, those for Cr(III), Fe(III), Ba and Th are relatively large.
294 Only for Ca binding by FA is the difference in $\log K_{MA}$ appreciably smaller than that for Cu.

295 Further comparisons can be made from calculated binding isotherms, examples of which are
296 presented in Figure 5. For both HA and FA, Model VII predicts weaker binding of Al at high
297 pH, because of the removal of binding sites containing two or three type B (weak acid)
298 groups. A similar effect is seen for Eu, except that Model VII predicts stronger binding at low
299 pH, but weaker at pH 8; Am and Cm also behaved like this. There is little change in the
300 prediction of copper binding by either HA or FA. Zinc binding hardly differs between the
301 models for HA, while Model VII predicts slightly stronger binding for FA, due to the new
302 procedure for estimating default values of $\log K_{MA}$.

303 The new Model VII parameterisation leads to changes in predicted competition effects. Due
304 to the complexity of competition reactions, and the large number of potential combinations of
305 metals, generalised analysis of the results is not possible. However, some illustrative
306 examples are given in Figure 6 of the effects on Cu and Zn binding of Mg, Al and Ca, three
307 important competitors in typical soils and waters. The predictions of the two models do not
308 differ very much for Al; only for Cu binding by HA is there a noticeable change, with weaker
309 competition leading to lower concentrations of Cu^{2+} . In the case of HA, competition by both
310 Mg and Ca for both Cu and Zn is calculated to be considerably stronger when Model VII is
311 applied, reflecting higher $\log K_{MA}$ values for the alkaline earths. For FA, competition by Mg
312 and Ca is weaker towards Zn, but stronger towards Cu.

313

314

315 Discussion

316 Because Models VI and VII are identical with respect to the six parameters (n_A , P , pK_A , pK_B ,
317 ΔpK_A and ΔpK_B) that together describe total binding site density, electrostatic effects and
318 proton binding affinity, the new parameter values (Table 2) reflect the use of new more
319 extensive data on proton dissociation from humic substances, especially for HA. The main
320 differences are that, in comparison with Model VI, the Model VII HA and FA have slightly
321 higher site densities, the FA type A groups are weaker acids, the HA type B groups are
322 stronger acids, and the HA electrostatic factor is smaller. These led to the differences in
323 calculated proton dissociation as a function of pH shown in Figure 1.

324 With regard to metal binding, Model VII is appreciably simpler than Model VI, having fewer
325 combinations of monodentate sites to make multidentate sites (Table 1), a formalised
326 relationship between $\log K_{MA}$ and $\log K_{MB}$ (equation 2), and with the spread factors (ΔLK_{A1}
327 and ΔLK_{B1}) set to zero. There is an apparent inconsistency in that the model requires the
328 equilibrium constants for metal-binding not to vary within the type A and B groups, but to
329 differ between the type A and B groups. Full application of the model of Carbonaro & Di
330 Toro^[65] would mean that ΔLK_{A1} and ΔLK_{B1} were non-zero, proportional to ΔpK_A and ΔpK_B
331 respectively. However, we obtained appreciably better fits if such parallelism was not
332 invoked. Inspection of the Carbonaro-Di Toro plots of $\log K_{ML}$ vs $\log \log K_{HL}$ shows that in
333 several cases there is a lower local slope in the range of carboxylate groups ($3 < \log K_{HL} < 5$)
334 indicating less relative variation in the $\log K_{ML}$ values than in $\log K_{HL}$, and so setting ΔLK_{A1} to
335 zero is perhaps defensible. There are insufficient data to judge this for higher $\log K_{ML}$, \log
336 K_{HL} , and ΔLK_{A2} . The values of $\log K_{MA}$ can be considered chemically reasonable in that they
337 are similar to equilibrium constants for the equivalent reaction of metallic cations with simple
338 carboxylate ligands such as lactic acid, as demonstrated for Model VI^[4]. But it is perhaps
339 worth re-emphasising that $\log K_{MA}$ values *per se* do not describe binding to humic
340 substances; rather, they predict binding to weak-acid groups (via equation 2), and are the
341 basis for the appreciably greater constants that apply to multidentate sites, which can be
342 further increased due to soft-ligand effects (equation 1).

343 Default Model VI parameters for different metallic cations were derived simply by taking the
344 averages of the calibrated values of $\log K_{MA}$, a procedure which implies that samples of
345 humic substances used in laboratory experiments have been taken from a range of different
346 materials in the field, so that the average $\log K_{MA}$ is the best overall estimate, and the range
347 of possible values can be characterised by the standard deviation of the $\log K_{MA}$ values. This
348 is satisfactory when a reasonable number of different data sets can be analysed, but may
349 produce an unrepresentative $\log K_{MA}$ if data for the metal in question come from only one or

350 two data sets. The new method of establishing default constants presented here (equations
351 3 and 4) makes greater use of relationships among the different metals, and between HA
352 and FA, and draws directly upon relationships between WHAM parameters and equilibrium
353 constants for well-defined ligands via the Irving-Rossotti slope α_O . This both improves the
354 reliability of the parameters and makes the parameter set more internally consistent. In
355 addition, the approach means that $\log K_{MA}$ can be estimated from constants for a wider range
356 of simpler ligands.

357 In 19 cases for HA and 13 for FA there are at least two data sets per metal, and these can be
358 used to compute standard deviations in $\log K_{MA}$. These range from 0.04 to 1.19, with means
359 of 0.33 and 0.32 for HA and FA respectively. The standard deviations reflect differences in
360 several factors, including the source of the humic substances, experimental methods and
361 experimental conditions. The standard deviations for Cu are relatively low, 0.24 for HA and
362 0.21 for FA, and given that techniques for quantifying Cu binding are better than those for
363 other metals, they probably reflect mostly humic variability. A standard deviation of 0.3 in \log
364 K_{MA} might reasonably be adopted as a standard when applying the model to estimate
365 uncertainty in field predictions.

366 Although Model VII represents an improvement on Model VI, its predictions do not differ
367 greatly (Figures 5 and 6). Therefore calculations that have already been run using Model VI
368 are unlikely to be invalidated by the new model, except perhaps for metal binding at alkaline
369 pH. There is probably merit in running both models, and also the NICA model^[5] for new
370 problems, since any differences may provide insights or highlight uncertainty. It should also
371 be borne in mind that “higher” models (such as the CHUM catchment model^[48,49], Critical
372 Limit Functions^[62,70] and WHAM-F_{TOX}^[59]) that use predicted speciation, will have parameter
373 values specific to the chosen Humic Ion Binding Model.

374 In summary, this work has produced a simpler Humic Ion-Binding Model, based on a
375 considerably larger data set, with greater internal consistency, and parameterised for protons
376 and 40 metallic cations. This should improve our ability to predict chemical speciation
377 involving natural organic matter in field situations.

378

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391

392 Table 1. Combinations of monodentate sites making bidentate and tridentate binding sites in
 393 Models VI and VII, expressed in terms of n_A the number of the most strongly-acid groups.
 394 Sites 1 to 4 are type A, present in equal amounts. Sites 5 to 8 are type B, and they total half
 395 of the type A sites. The factor f_{prB} specifies the fraction of the sites that are close enough to
 396 form bidentate sites, and f_{prT} does the same for tridentate sites. The values of f_{prB} and f_{prT} are
 397 0.42 and 0.03 respectively for FA and 0.50 and 0.065 for HA. For each site combination
 398 there are three binding strengths governed by the parameter ΔLK_2 , their fractional
 399 abundances being 0.901, 0.09 and 0.009, so the model has three times as many
 400 multidentate sites as those shown here, 72 in Model VI and 42 in Model VII.

401

Model VI		Model VII	
sites	abundance	sites	abundance
<i>Bidentate sites</i>			
1-2	$f_{prB} \times n_A / 6$	1-2	$f_{prB} \times n_A / 8$
3-4	$f_{prB} \times n_A / 6$	3-4	$f_{prB} \times n_A / 8$
1-5	$f_{prB} \times n_A / 12$	1-5	$f_{prB} \times n_A / 8$
2-6	$f_{prB} \times n_A / 12$	2-6	$f_{prB} \times n_A / 8$
3-7	$f_{prB} \times n_A / 12$	3-7	$f_{prB} \times n_A / 8$
4-8	$f_{prB} \times n_A / 12$	4-8	$f_{prB} \times n_A / 8$
5-6	$f_{prB} \times n_A / 24$		
7-8	$f_{prB} \times n_A / 24$		
<i>Tridentate sites</i>			
1-2-3	$f_{prT} \times n_A / 27$	1-2-5	$f_{prT} \times n_A / 16$
1-2-4	$f_{prT} \times n_A / 27$	1-2-6	$f_{prT} \times n_A / 16$
1-3-4	$f_{prT} \times n_A / 27$	1-2-7	$f_{prT} \times n_A / 16$
2-3-4	$f_{prT} \times n_A / 27$	1-2-8	$f_{prT} \times n_A / 16$
5-6-7	$f_{prT} \times n_A / 216$	3-4-5	$f_{prT} \times n_A / 16$
5-6-8	$f_{prT} \times n_A / 216$	3-4-6	$f_{prT} \times n_A / 16$
5-7-8	$f_{prT} \times n_A / 216$	3-4-7	$f_{prT} \times n_A / 16$
6-7-8	$f_{prT} \times n_A / 216$	3-4-8	$f_{prT} \times n_A / 16$
1-2-5	$f_{prT} \times n_A / 18$		
3-4-6	$f_{prT} \times n_A / 18$		
1-3-7	$f_{prT} \times n_A / 18$		
2-4-8	$f_{prT} \times n_A / 18$		
1-5-6	$f_{prT} \times n_A / 36$		
2-7-8	$f_{prT} \times n_A / 36$		
3-5-7	$f_{prT} \times n_A / 36$		
4-6-8	$f_{prT} \times n_A / 36$		

402

403 Table 2. Default proton binding parameters for humic and fulvic acid.

Parameter	HA		FA	
	Model VI	Model VII	Model VI	Model VII
n_A^*	3.3	3.4	4.8	5.2
pK_A	4.1	4.1	3.2	3.7
pK_B	8.8	8.3	9.4	9.6
ΔpK_A	2.1	2.6	3.3	3.1
ΔpK_B	3.6	3.1	4.9	4.4
P	-330	-196	-115	-119

404 * mmol (gHS)⁻¹

405

406 Table 3. Default cationic metal parameters for Model VII. Values of $\log K_{MA,HA,mean}$ and \log
 407 $K_{MA,FA,mean}$ are averages from data-fitting (n_{HA} and n_{FA} are the numbers of data sets), while \log
 408 $K_{MA,HA,def}$ and $\log K_{MA,FA,def}$ are default values derived with equations (3) and (4), which involve
 409 the Irving-Rosotti parameter α_O . Values of ΔLK_2 were derived with equation (1) using the
 410 compilation of $\log K_{NH_3}$ values by Martell & Hancock^[69].

metal	n_{HA}	n_{FA}	$\log K_{MA,HA,mean}$	$\log K_{MA,FA,mean}$	α_O	$\log K_{MA,HA,def}$	$\log K_{MA,FA,def}$	ΔLK_2
Be	0	0	-	-	0.433	2.27	2.02	0.99
Mg	1	2	0.98	1.01	0.176	1.14	0.99	0.13
Al	4	4	2.67	2.69	0.607	2.82	2.57	0.46
Ca	8	11	1.19	1.17	0.194	1.26	1.13	0.00
Sc	1	0	3.61	-	-	3.61	3.28	0.39
VO	0	1	-	2.51	-	2.76	2.51	1.74
Cr	1	0	2.52	-	0.818	3.07	2.89	1.97
Mn	2	1	2.21	1.67	0.255	1.98	1.76	0.58
Fe(II)	0	0	-	-	0.287	1.76	1.46	0.81
Fe(III)	2	1	3.19	3.03	0.861	3.37	3.12	2.20
Co	2	8	1.51	1.32	0.306	1.50	1.35	1.22
Ni	2	5	1.6	1.41	0.301	1.60	1.43	1.57
Cu	13	16	2.54	2.07	0.466	2.38	2.16	2.34
Zn	2	4	1.87	1.71	0.304	1.87	1.68	1.28
Sr	1	1	1.49	1.01	0.171	1.32	1.13	0.00
Y	1	1	2.84	2.93	-	3.03	2.76	0.22
Ag	4	1	1.50	1.14	0.177	1.44	1.27	1.91
Cd	10	6	1.61	1.58	0.306	1.67	1.51	1.48
Ba	0	0	-	-	0.158	1.30	0.97	0.00
La	1	1	2.64	2.74	0.414	2.62	2.36	0.11
Ce	1	1	2.68	2.7	0.451	2.66	2.41	0.13
Pr	1	1	2.69	2.74	-	2.85	2.59	0.16
Nd	1	1	2.68	2.71	-	2.83	2.57	0.18
Sm	1	1	2.76	2.81	-	2.93	2.66	0.20
Eu	5	10	2.97	2.61	0.530	2.89	2.62	0.29
Gd	1	1	2.77	2.84	-	2.95	2.68	0.24
Tb	1	1	2.86	2.92	-	3.04	2.76	0.26
Dy	2	1	3.19	2.93	-	3.20	2.91	0.28
Ho	1	1	2.95	2.96	-	3.10	2.82	0.30
Er	1	1	3.03	3.09	-	3.21	2.92	0.32
Tm	1	1	3.09	3.07	-	3.23	2.94	0.35
Yb	1	1	3.12	3.05	-	3.24	2.94	0.37
Lu	1	1	3.17	3.1	-	3.29	2.99	0.39
Hg	3	5	4.1	3.4	0.796	3.84	3.51	5.10
MeHg	4	1	0.53	0.39	-	0.51	0.46	3.60
Pb	9	10	2.39	2.14	0.442	2.37	2.15	0.93
Th	2	0	3.41	-	0.902	3.58	3.34	0.23
UO2	4	4	2.64	2.28	0.621	2.61	2.38	1.16
Am	7	3	2.95	2.74	0.543	2.94	2.68	1.57
Cm	3	1	2.58	1.91	0.537	2.50	2.27	1.57

412 Table 4. Comparison of default log K_{MA} values for Models VI and VII. Values in bold indicate
 413 that the difference between the models is more than 0.2 log units greater or less than the
 414 difference for Cu.

	Model VI		Model VII	
	HA	FA	HA	FA
Mg	0.7	1.1	1.1	1.0
Al	2.6	2.5	2.8	2.6
Ca	0.7	1.3	1.3	1.1
VO	2.5	2.4	2.8	2.5
CrIII	2.2	2.2	3.1	2.9
Mn	0.6	1.7	2.0	1.8
FeII	1.3	1.6	1.8	1.5
FeIII	2.4	2.6	3.4	3.1
Co	1.1	1.4	1.5	1.4
Ni	1.1	1.4	1.6	1.4
Cu	2.0	2.1	2.4	2.2
Zn	1.5	1.6	1.9	1.7
Sr	1.1	1.2	1.3	1.1
Cd	1.3	1.6	1.7	1.5
Ba	-0.2	0.6	1.3	1.0
Eu	2.1	2.4	2.9	2.6
Dy	2.9	2.5	3.2	2.9
Hg	3.5	3.5	3.8	3.5
Pb	2.0	2.2	2.4	2.2
Th	2.8	2.7	3.6	3.3
UO ₂	2.2	2.1	2.6	2.4
Am	2.5	2.6	2.9	2.7
Cm	2.2	2.0	2.5	2.3

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Figure captions

- Figure 1 Proton dissociation calculated with Model VI and VII default parameters; Z is the charge per g of HA or FA. The upper of each pair of plots refers to an ionic strength of 0.1 M, the lower to 0.001M.
- Figure 2 Experimental data of Sonke (2006) for the binding of La(III) by humic acid (open circles) and fulvic acid (closed circles), expressed as $K_{app} = \nu / [La^{3+}]$, where ν is the amount of bound metal in mol g⁻¹. The lines are fits with Models VI (dashed lines) and VII (full lines).
- Figure 3 Fitted log K_{MA} for different metals (individual data sets) plotted against α_0 , the Irving-Rossotti slope.
- Figure 4 Fitted log K_{MA} for HA (average value for each metal) plotted against the corresponding value for FA. The line has the expected slope of 1.11 (see Results). The triangles show data for lanthanides.
- Figure 5 Metal binding isotherms calculated with the default parameters of Models VI and VII. ν (ν) is moles bound per gram FA. Open symbols Model VI, closed symbols Model VII. Circles pH 4, squares pH 6, triangles pH 8.
- Figure 6 Competition by Mg, Al and Ca for Cu and Zn binding by HA and FA; comparison of results with default parameters for Models VI (broken lines) and VII (full lines). The calculations refer to pH 5 for Al and pH 7 for Mg and Ca, and an ionic strength of 0.01 M.

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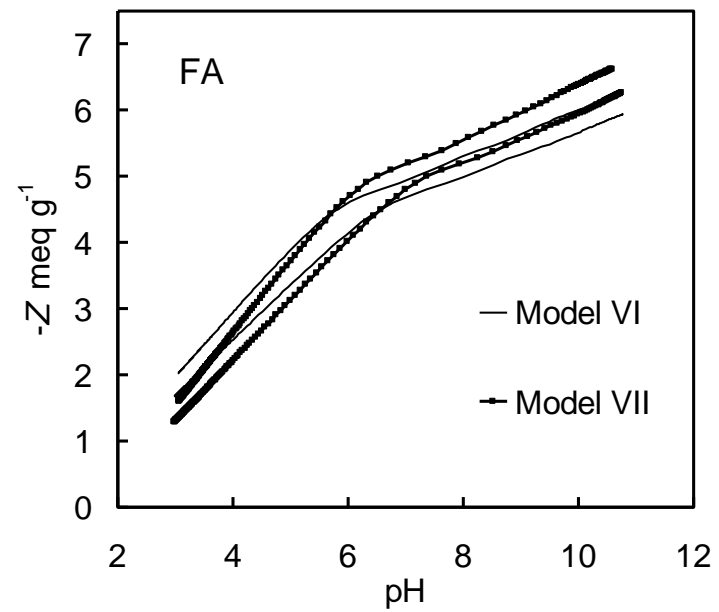
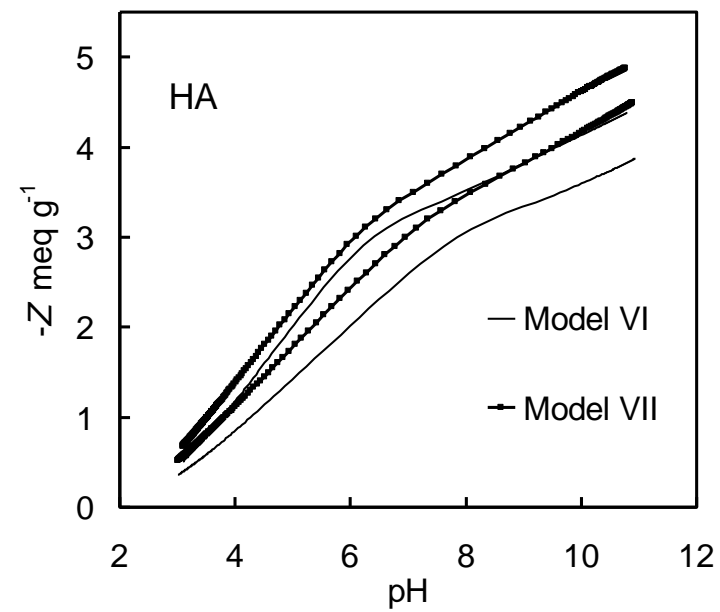
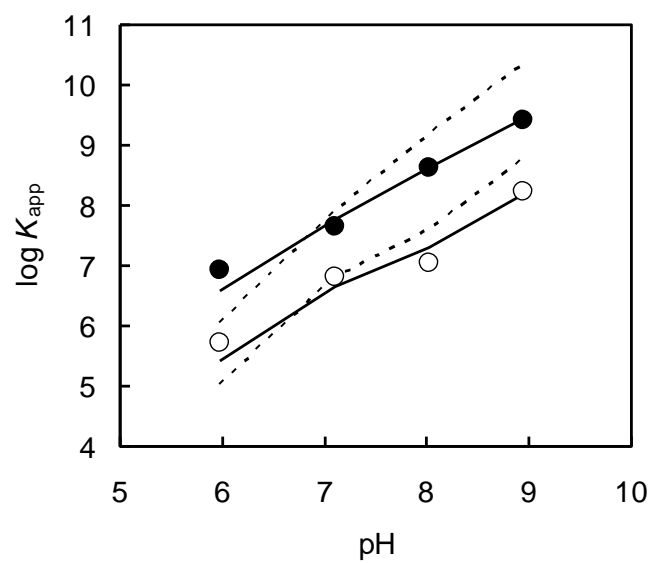


Figure 1

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453 Figure 2

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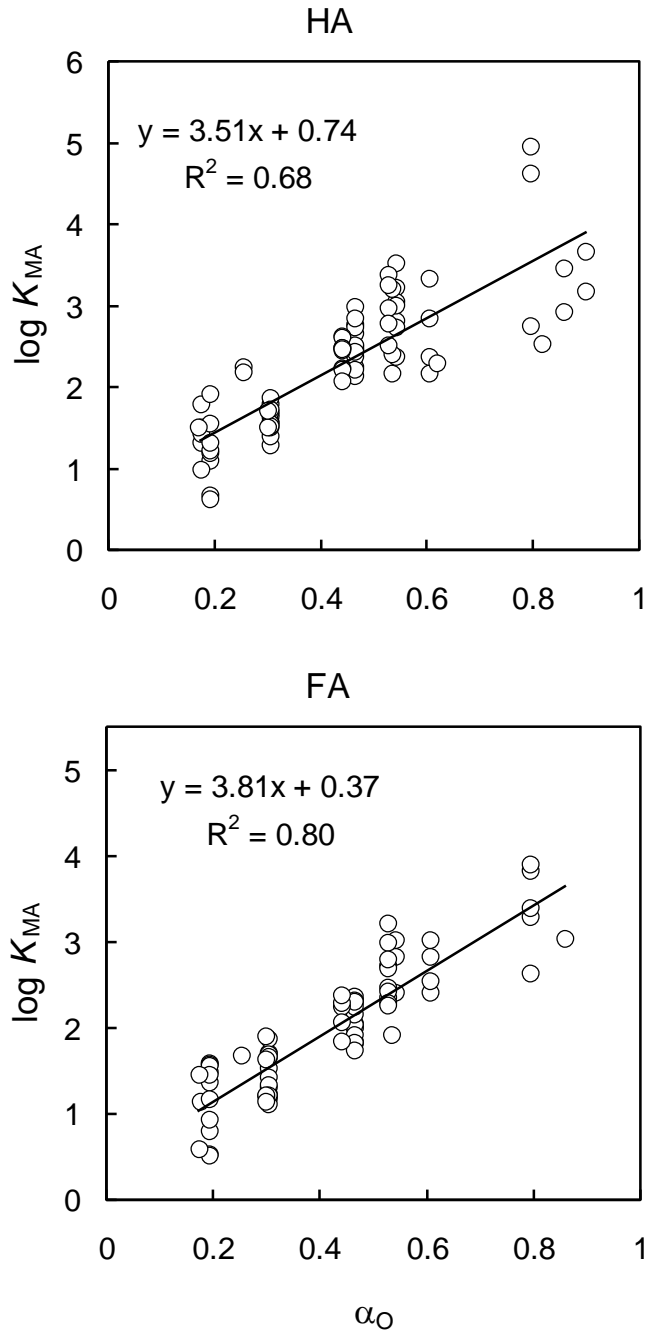
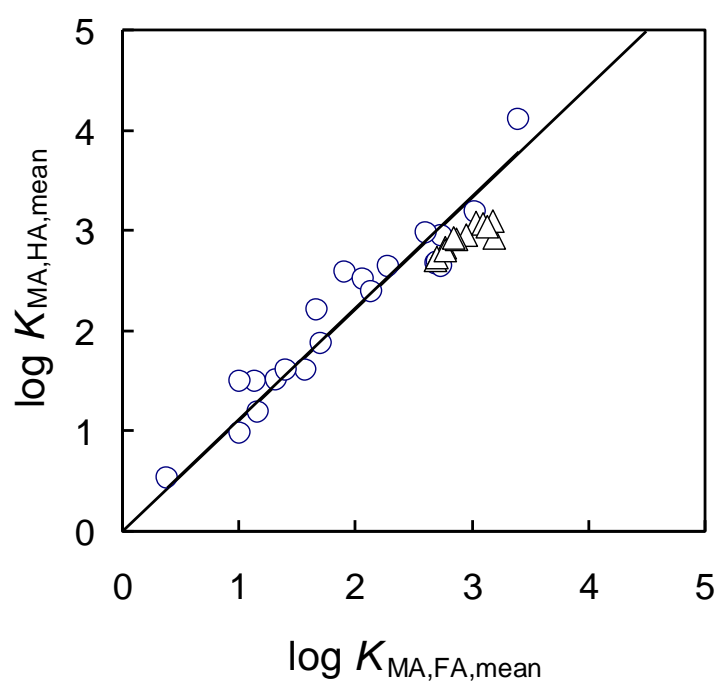


Figure 3

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483 Figure 4

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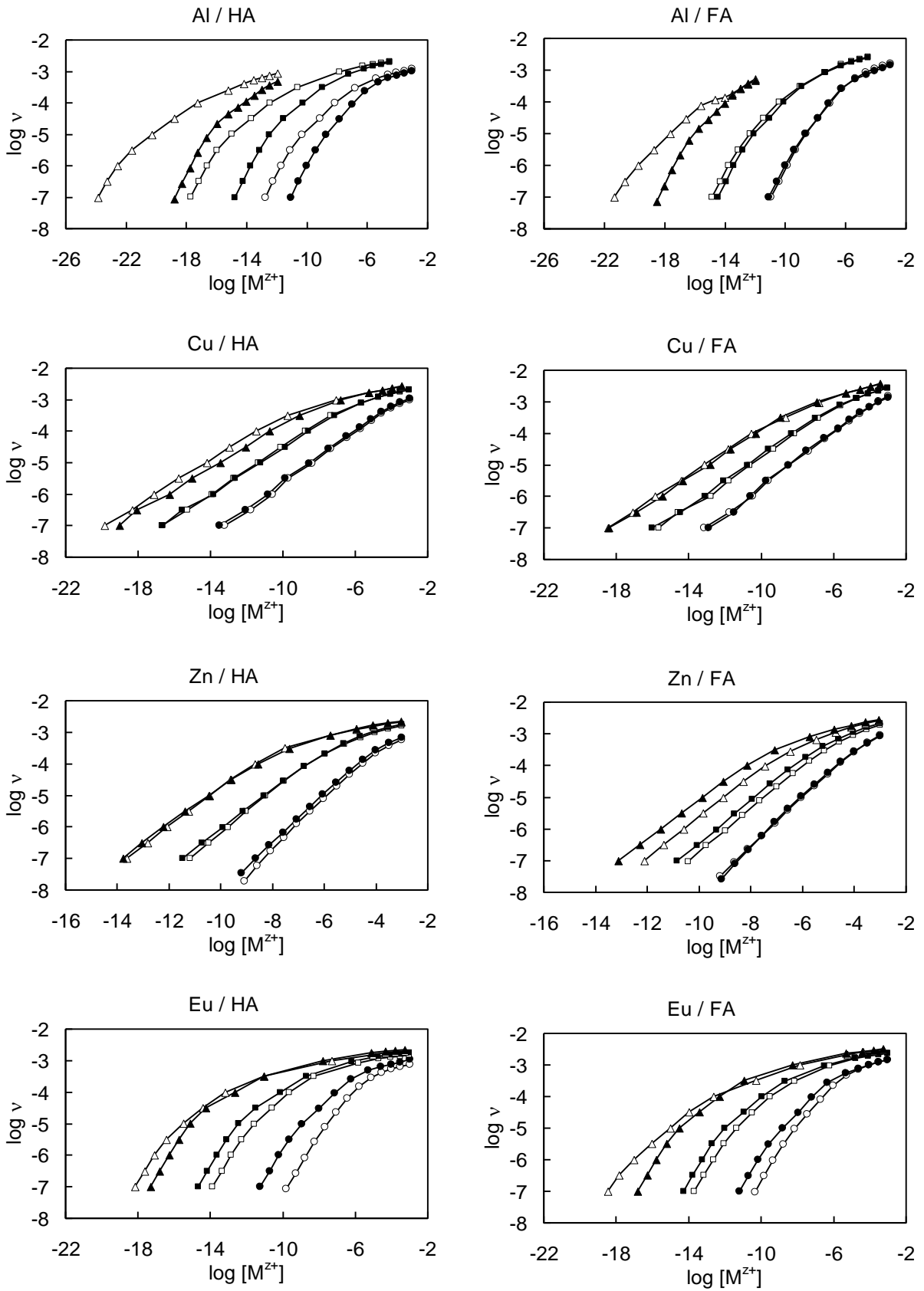


Figure 5

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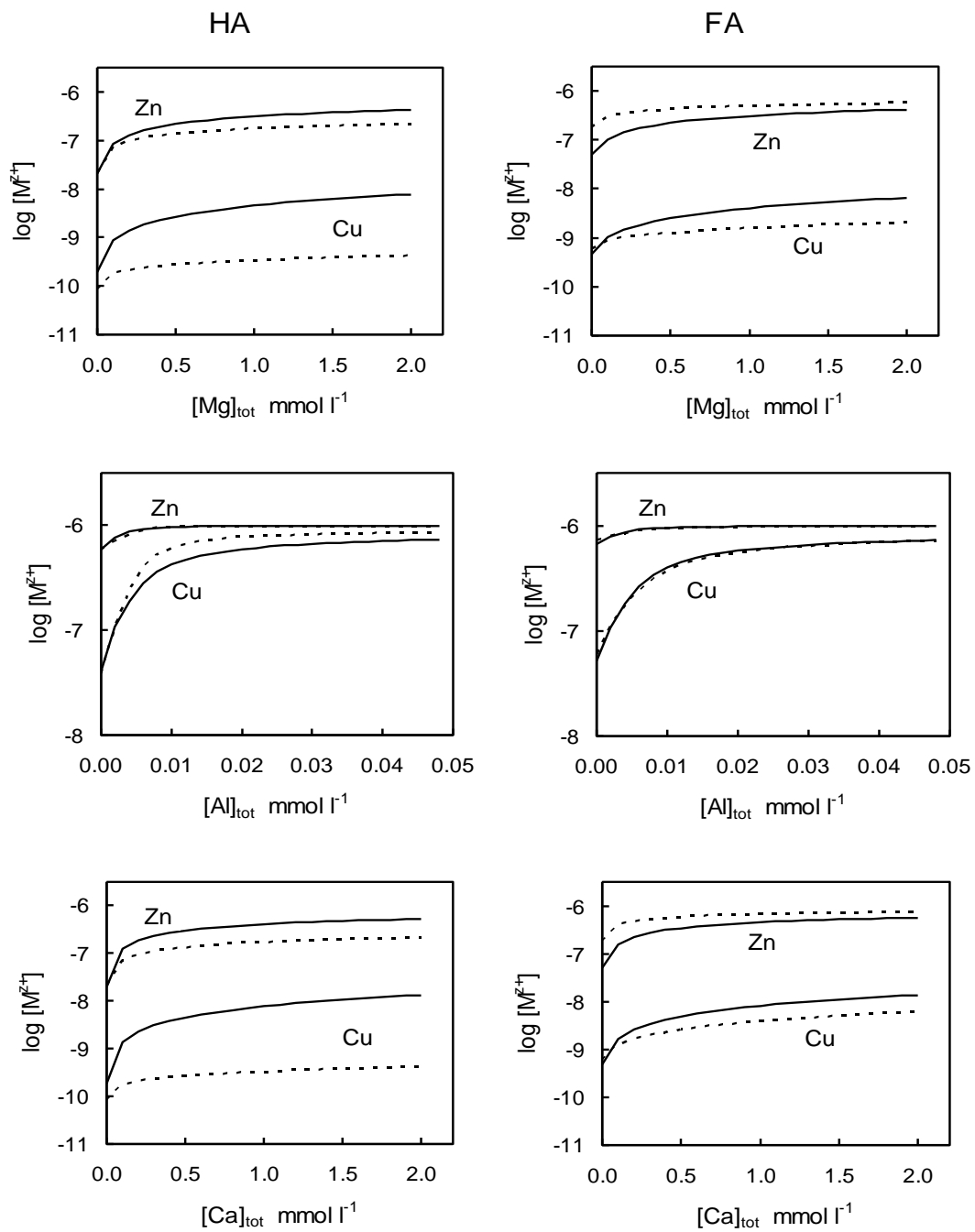


Figure 6