

1	Predicting cadmium adsorption on soils using WHAM VI
2	Zhenqing Shi ^a , Herbert E. Allen ^{a,*} , Dominic M. Di Toro ^a , Suen-Zone Lee ^b , Diego M.
3	Flores Meza ^a , and Steve Lofts ^c
4 5	^a Center for the Study of Metals in the Environment, Department of Civil and Environmental Engineering, University of Delaware, Newark, DE 19716, USA.
6 7	^b College of the Environment, Chia Nan University of Pharmacy and Science, Tainan, Taiwan.
8 9	^c Centre for Ecology and Hydrology, Lancaster Environment Centre, Lancaster LA1 4AP, United Kingdom
10	[*] Corresponding author. Tel.: 1 302 831 8449; fax: 1 302 831 3640.

11 e-mail address: allen@ce.udel.edu.

12 Abstract

13 Cadmium (Cd) adsorption on 14 non-calcareous New Jersey soils was investigated 14 with a batch method. Both adsorption edge and isotherm experiments were conducted 15 covering a wide range of soil composition, e.g. soil organic carbon (SOC) concentration 16 ranging from 0.18% to 7.15%, and varying Cd concentrations and solution pH. The SOC 17 and solution pH were the most important parameters controlling Cd partition equilibrium 18 between soils and solutions in our experimental conditions. The Windermere humic 19 aqueous model (WHAM) was used to calculate Cd adsorption on soils. The effect of 20 solution chemistry (various pH and Cd concentrations) on Cd adsorption can be well 21 accounted for by WHAM. For different soil compositions, SOC concentration is the most 22 important parameter for Cd binding. Only a fraction of SOC, the so-called active organic carbon (AOC), is responsible for Cd binding. We found a linear relationship between 23 24 SOC and AOC based on the adsorption edge data. The linear relationship was validated 25 by the independent data sets: adsorption isotherm data, which presumably can be used to 26 predict Cd partition equilibrium across a wide range of soil compositions. The modeling 27 approach presented in this study helps to quantitatively predict Cd behavior in the 28 environment.

29 Keywords: cadmium, adsorption, WHAM, active organic matter, partition, modeling

30 Introduction

Adsorption of Cd on soils is an important process controlling Cd behavior in the 31 32 environment. Soils consist of different components including soil organic matter (SOM), 33 metal (hydro)oxides and clay minerals that are responsible for Cd binding. Different 34 components in soils may contribute to Cd adsorption to different extents. Due to the 35 limitations of experimental techniques (Zachara et al., 1992), it is very difficult to experimentally identify the contribution of individual adsorbents to the control of trace 36 37 metal adsorption on soil particles. The partition coefficient K_p has been widely used to 38 characterize the equilibrium relationship of trace metals between solution and solid 39 phases. Empirical relationships have been derived by relating K_p to key soil and solution 40 parameters (e.g. total metal concentration, SOM content, pH, etc.) through multiple 41 regression analysis (Janssen et al., 1997; Impellitteri et al., 2002; McBride et al., 1997; 42 Sauvé et al., 2000, 2003). For example, the distribution of trace metals between solid and 43 solution phases could be calculated based on pH, total metal content and organic matter 44 (Sauvé et al., 2000).

45 Compared with traditional statistical regression models, more recently mechanistic models have been used to describe metal partitioning between soils and solutions (Weng 46 47 et al., 2001; Gustafsson et al., 2003; Tipping et al., 2003; Lumsdon, 2004; Ponizovsky et 48 al., 2006), including the Windermere humic aqueous model (WHAM) (Tipping, 1994). 49 SOM has been reported to be the major component among soil components accounting 50 for metal binding. Furthermore, it has been reported that only a fraction of SOM, which is 51 called active organic matter (AOM), is responsible for metal binding (Tipping et al., 2003; 52 Gustafsson et al., 2003; Lumsdon, 2004). The AOM may be due to the difference

between the SOM and the generic humic substance to which the equilibrium models are calibrated. SOM consists of non-humic substances such as amino acids, carbohydrates, and lipids, as well as humic substance, a series of high-molecular-weight, brown to black substances formed by secondary synthesis reactions (Stevenson, 1994). SOM may also coat on mineral phase and thus some inner part of SOM may be not accessible due to coating (Nachtegaal and Sparks, 2003). However, there are no well-established methods to determine AOM.

Inorganic components, such as clay minerals and metal oxides/hydroxides, can also be important in controlling Cd reactions in soils (Kabata-Pendias, 2001). The role of different soil components for metal binding may be dependent on the experimental conditions. For example, clay minerals are more important for metal adsorption at low pH, low soil organic matter content and high metal loading (Weng et al., 2004).

65 In this paper, the Cd adsorption edge and adsorption isotherms data for 14 New Jersey soils were analyzed using WHAM VI (Tipping 1998), the latest version of 66 67 WHAM. In WHAM VI, a number of parameters have been recognized as being 68 important for predicting metal partitioning including organic matter, and iron and 69 aluminum competition (Tipping, et al., 2003). In this study, we are seeking a unique set 70 of WHAM input parameters for Cd adsorption that are applicable for different soils at 71 various solution chemistry conditions. The competition effect of Al at different pH ranges 72 was tested.

73 Materials and Methods

74 The 14 New Jersey soils those were used to conduct adsorption experiments are described in Table 1. The sample of Delaware River fill material that was included in the 75 original study (Lee et al., 1996) has not been included in the present analysis as it is not a 76 77 soil. All soils are non-calcareous soils with pH less than 7. These soils cover a variety of 78 soil properties. The soil organic carbon (SOC) concentration ranging from 0.18% to 79 7.15%, which was measured with a Variomax CN analyzer. The ammonium oxalate 80 extractable soil metal (hydro)oxides were presented in Lee et al. (1996). The 81 exchangeable Ca and Mg were extracted with the 0.1 M BaCl₂ for 24 h (Carter, 1993). 82 The Al was extracted with the 0.43 M HNO₃ for 24 h (Dijkstra et al., 2004).

83 1. Adsorption edge experiment

Batch equilibration studies were conducted with soil concentration = 1.00 ± 0.01 g per 100 mL solution and ionic strength = 0.01M with NaNO₃. Cd nitrate concentrations = 1×10^{-5} and 1×10^{-4} M with pH = 3 to 9 and temperature = 25 ± 2 °C. Samples were shaken at 150 rpm for 24 hours and filtered through 25 mm diameter, 0.45 µm membrane filters.

89 2. Adsorption isotherm experiment

The soil samples, 1.0 gram of each, were immersed in 100 mL of 0.01 M NaNO₃. The pH was adjusted in order to approach the desired pHs, 4.0, 5.5, and 7.0 and temperature = 25 ± 2 °C. The initial Cd concentrations ranged from 1.0×10^{-6} to 5×10^{-3} M, and the equilibration time was 24 hours. After 24 hours, the pH was readjusted to the initial value and the samples were filtered through 0.45 µm membrane filters. 95 Cadmium concentrations in the filtered solutions were determined by atomic
96 absorption spectrometry as reported by Lee et al. (1996). The detailed information on
97 experimental procedures can be found in Lee et al. (1996).

98 Modeling Method

99 WHAM VI was used to calculate Cd adsorption equilibrium at different conditions. 100 WHAM is capable of calculating the equilibrium chemical speciation in surface and ground waters, sediments, and soils, especially when the chemical speciation is 101 102 dominated by organic matter (Tipping, 1994; Tipping et al., 2003). WHAM uses Model 103 VI. a computer model, to describe the reactions of ions with humic substances, mainly 104 through complexation which is modified by electrostatic reactions. Previous modeling 105 results have demonstrated that, for most non-calcareous soils except for soils with very low SOM concentrations, the SOM is the major adsorbent for trace metals (Tipping et al., 106 107 2003; Ponizovsky et al., 2006). Preliminary WHAM calculations showed that metal 108 oxides have little effect on Cd adsorption for most of our soils, so only SOM and clay fraction were used in WHAM calculations. The WHAM input parameters include 109 110 particulate humic acid (HA) and fulvic acid (FA) concentrations ([HA] and [FA]), the clay concentration, pH, the CO_2 pressure (pCO₂), and solution cations and anions 111 concentrations or activities including $[Na^+]$, $[Ca^{2+}]$, $\{Al^{3+}\}$ or $[Al^{3+}]$, $\{Fe^{3+}\}$ and $[NO_3^-]$. 112 The pCO₂ was set at $10^{-3.5}$ atm. Ponizovsky et al. (2006) have demonstrated that the 113 variation of pCO₂ from 10^{-4} to 10^{-2} atm had little effect in their model calculations. The 114 115 Mg concentrations were low which had little effect on Cd adsorption. For the solution 116 species, the concentrations were input as measured in the experiments. For the SOM,

117	only the active fraction, AOM, was used in WHAM VI calculations and the active
118	fraction was assumed to consist of 84% HA and 16% FA (Tipping et al., 2003).
119	Due to the high concentration of Cd in this study, preliminary model calculations
120	showed that the competition of Fe^{3+} was minimal. The competition of Al on Cd
121	adsorption could be different at different pH. The Al activity in the soil solution can be
122	estimated by assuming that the solution was equilibrated with Al hydroxides (Tipping et
123	al., 2003), as described by equation 1.
124	$\log\{Al^{3+}\} = \log^* K_{s0} - 3pH $ (1)
125	Tipping (2005) has further assessed how Al^{3+} activity was dependent on pH and whether
126	Al ³⁺ activity was controlled by Al hydroxides or not. At low pH, solubility control of Al
127	hydroxides may overestimate the Al^{3+} activity and a linear regression equation was
128	proposed as described by equation 2.
129	$\log\{Al^{3+}\} = -1.59\text{pH} + 2.22\tag{2}$
130	Other researchers used acid extraction to estimate the active Al accounting for Al
131	competition to metal binding (Dijkstra et al., 2004).
132	In our modeling for the Al competition effect, we started with the acid extractable Al.
133	The 0.43 M HNO ₃ extractable Al was input as the total Al concentration in the WHAM
134	calculations for adsorption reactions. The model calculations showed that the Al^{3+}
135	•
100	activity calculated by acid extractable Al was too high, as it exceeds the solubility control
136	activity calculated by acid extractable Al was too high, as it exceeds the solubility control of Al hydroxides. Then we used the regression equation (equation 2) to calculate Al^{3+}
136 137	activity calculated by acid extractable Al was too high, as it exceeds the solubility control of Al hydroxides. Then we used the regression equation (equation 2) to calculate Al^{3+} activity at low pH. At pH greater than 5, Al^{3+} activity controlled by Al(OH) ₃ solubility
136 137 138	activity calculated by acid extractable Al was too high, as it exceeds the solubility control of Al hydroxides. The n we used the regression equation (equation 2) to calculate Al^{3+} activity at low pH. At pH greater than 5, Al^{3+} activity controlled by $Al(OH)_3$ solubility with a $log^* K_{s0}$ value of 8.5 can consistently describe the Cd adsorption. However, at pH
136 137 138 139	activity calculated by acid extractable Al was too high, as it exceeds the solubility control of Al hydroxides. The n we used the regression equation (equation 2) to calculate AI^{3+} activity at low pH. At pH greater than 5, AI^{3+} activity controlled by $AI(OH)_3$ solubility with a log [*] K_{s0} value of 8.5 can consistently describe the Cd adsorption. However, at pH <= 5, the regression equation provided too high AI^{3+} activity which resulted in much less

140 Cd adsorption on soils compared with the experimental values, as discussed later. So we 141 tried to estimate the active Al concentration by model fitting of experimental data at pH 142 ≤ 5 , which controls the Al competition for each soil at pH ≤ 5 .

143 For all 14 soils, the adsorption edge data at two initial Cd concentrations were 144 tabulated in Microsoft EXCEL 2000. A version of WHAM VI that can be executed in 145 EXCEL with specific input parameters described previously was used to fit the 146 adsorption edge data. The errors of the percentage of adsorption between WHAM VI 147 calculations and experimental results were calculated. Then the squares of the errors were 148 summed and minimized using the EXCEL 2000 SOLVER program by optimization of 149 model parameters, active organic carbon (AOC) and active Al. The AOC and active Al 150 were obtained separately. The AOC was converted to AOM by multiplying a factor of 1.8, 151 which was further calculated as HA and FA as described before. The data at pH > 5 was used first to obtain the amount of AOC for each soil where the Al^{3+} activity was 152 153 controlled by the Al hydroxide. The amount of AOC was adjusted for each soil until the 154 sum of errors was minimized at pH > 5. Then the data at $pH \ll 5$ was used to obtain the 155 active Al which controls the Al competition at $pH \le 5$. The AOC was used as obtained 156 from model fitting at pH > 5. The total active Al concentration was adjusted until the sum of errors was minimized at $pH \le 5$. The total active Al concentrations for all soils are 157 158 low which precludes the formation of Al hydroxides. The parameters used for WHAM VI 159 calculations are summarized in Table 2.

After obtaining the amount of AOC, we developed a linear relationship between the AOC and SOC. The linear relationship was applied to predict the adsorption isotherms, which were validated by the experimental data.

163 **Results and Discussion**

164 The Cd adsorption changes dramatically from low pH to high pH, less than 10% adsorption at pH < 3, and more than 90% adsorption at pH > 8 for most soils. Generally, 165 166 WHAM VI is able to account for the variation of Cd partitioning as a function of solution 167 pH and Cd concentrations for different soils, and the model fit is consistent with the 168 adsorption edge data. Figure 1 presents the results for three soils. The model curves show 169 discontinuities at pH 5 due to the different approaches for Al competition for pH below 170 or above 5. The model results support the assumption that SOM is the major sorbent for 171 Cd for most of soils in our study. Lee et al. (1996) also reported that SOM plays the most 172 important role in controlling Cd adsorption. By analyzing literature data of more than 70 173 studies, Sauvé et al. (2000) developed a regression equation for Cd partitioning between 174 soils and solutions based on pH, total Cd concentration and SOM concentration. Likewise, the partitioning of Cd between soils and solutions can be predicted by the NICA-Donnan 175 176 model when the organic matter was considered as the major sorbent controlling Cd 177 binding (Benedetti et al., 1996; Lumsdon, 2004). Furthermore by using the regression equation for low pH by Tipping (2005), the model calculated less Cd adsorption. The 178 regression equation was obtained from data at 10 °C and the Al^{3+} activity could vary 179 180 more than one order of magnitude for different soils (Tipping, 2005).

Overall, the amount of AOC for all 14 soils increases with the increase of SOC and the quantity of AOC is linearly related to SOC, with $R^2 = 0.827$ (Figure 2). The linear regression equation has a slope of 0.629 which results in about 63% AOC for all soils. Gustafsson et al. (2003) tried to optimize the fraction of AOM for different soils to model trace metals adsorption using their Stockholm humic model, but they did not find any 186 clear relationship between the optimized values and SOM. Tipping et al. (2003) did not 187 find a clear relationship between active humic substance and SOM in their study. In our 188 study, we focused on the mineral soils, which have much lower SOM than most of the 189 organic soils used by Tipping et al. (2003). The Cd concentrations used in this study are 190 also higher.

191 Different methods have been applied to quantify the reactive organic matter in soils. 192 The amount of AOC may vary depending on the methods used. Base extraction provides 193 a direct way to estimate the humic substances in soils as used by some researchers 194 (Dijkstra et al., 2004; Lumsdon, 2004). Dijkstra et al. (2004) reported that the percentage 195 of the base extractable SOM ranged from 25% to 67% for a variety of soils. To obtain the 196 reactive organic matter, Tipping et al. (2003) adjusted the amount of active humic 197 substances of the soil so that the WHAM calculated pH was the same as the observed 198 value. Weng et al. (2001) used another approach to calculate the AOM in their NICA-199 Donnan model: adjusting the site density of SOM based on the cation exchange capacity 200 measurement of soils and the generic humic substance. Their results showed that the site 201 density of SOM varied between 46% and 16% of that of the generic humic substance, 202 with the average value of AOM for all soils being 31%. Similarly Cances et al. (2003) 203 assumed that 50% organic carbon in soils was reactive in their modeling study. The linear 204 regression equation obtained in this study is based on parameter optimization, which 205 provides a good approximation on reactive organic matter in soils for predicting Cd 206 adsorption on soils using WHAM VI. Whether this regression equation can be applied to 207 other trace metals needs to be further investigated.

208 Using the linear regression equation for AOC in Figure 2, WHAM VI was used to predict both Cd adsorption edges and isotherms. The predictions for Cd adsorption edges 209 210 are self-calibrations for the linear regression equation since it is obtained from the 211 adsorption edge data (Figure 3). To further validate the linear regression equation for 212 AOC, WHAM VI predictions were compared with the independent data sets: Cd 213 adsorption isotherms. Figure 4 presents the WHAM VI predictions for the Cd adsorption 214 isotherms together with the experimental data at three different pHs for four typical soils. 215 The model predictions are consistent with the experimental results. The applicability of 216 AOC versus SOC linear relationship to independent data sets supports the validity of our 217 modeling approach.

218 Figure 5 presents a detailed comparison between WHAM VI predicted partition coefficient K_p and experimental K_p in adsorption edge experiments. The results are 219 presented from low pH to high pH. At low pH (2 - 4), WHAM VI over-predicted K_p . 220 221 WHAM predictions are close to experimental values at intermediate pH (4 - 7). At high pH (7-9), more scatter can be observed, indicating greater deviation between WHAM 222 223 predictions and experimental results. The deviations at high pH may be due to the 224 dissolution of SOM, which was not measured in the original experiment and thus not 225 considered in the modeling. Generally almost all deviations are within one order of 226 magnitude. The root mean square error (RMSE) of $\log K_p$ equals 0.41. The comparison 227 between WHAM VI predicted K_p and experimental K_p in adsorption isotherm 228 experiments are presented in Figure 6 for three pHs. The RMSEs were 0.33, 0.23 and 229 0.39 at pH 4.0, 5.5 and 7.0 respectively.

230 The Al competition varied at different pH ranges. It has been reported that, at low pH, Al^{3+} activity may not be controlled by $Al(OH)_3$ but by adsorption/complexation reactions 231 232 (Tipping, 2005). Some researchers have used chemical extractions to quantify the active 233 Al (Gustafsson et al., 2003; Lumsdon, 2004). As pointed out by Gustafsson et al. (2003), 234 the total active Al concentrations are not well quantified by available extraction 235 procedures. Besides the extraction, different optimization methods have been used to obtain the active Al. For example, Lumsdon (2004) used 0.1 M HCl extraction to 236 237 estimate the reactive Al and the fit of Cd adsorption curves was further improved by optimizing the active Al. In our study, if the 0.43 M HNO₃ extractable Al was used as the 238 total Al controlling Al competition for Cd adsorption, the model calculations resulted in 239 higher Al^{3+} activity than that controlled by $Al(OH)_3$. Therefore, we obtained the active Al 240 by optimizing the cadmium adsorption data at low pH. The AI^{3+} activity controlled by 241 242 this active Al is lower than that controlled by Al(OH)₃. This approach provides good 243 model predictions as demonstrated in Figures 5 and 6, together with the solubility control of Al^{3+} at higher pH. Nevertheless accurate measurement of $A^{\hat{p}+}$ activity and 244 characterization of Al phases controlling Al^{3+} activity are desired in order to more 245 accurately predict metal partitioning. 246

The importance of mineral phases such as metal (hydro)oxides on Cd binding has been reported by many studies (Tessier et al. 1985; Zachara, et al., 1992; Trivedi and Axe, 2000). In natural soils, with the competition of SOM for Cd binding, other sorbents besides SOM may be less important with the increase of SOM concentration. Furthermore, the dominant importance of SOM observed in this study is based on laboratory adsorption experiments, which may be different from highly contaminated field soils. The field contaminated soils may experience various contamination processes
and thus the metal speciation could differ significantly (Manceau et al., 1996; Roberts et
al., 2002).

256 Conclusions

257 WHAM VI has been successfully applied to describe Cd adsorption on different soils 258 with a wide range of soil compositions and different solution chemistry conditions. 259 WHAM VI is able to account for the effect of major solution and soil parameters that 260 control the Cd adsorption on soils. For most of soils in this study, SOC appears to be the 261 major soil component controlling Cd adsorption. Only a fraction of SOC, called AOC, is 262 responsible for Cd binding. The AOC concentration is linearly related to the SOC concentration. Al competition effect should be considered differently at different pH 263 264 ranges. At bw pH, the Al competition was controlled by an active fraction of soil Al rather than the solubility of the Al hydroxides. With the model parameters presented in 265 this study, we can predict Cd adsorption behavior in soils using WHAM VI when SOC is 266 the dominant sorbent in soils. 267

268 Acknowledgements

269 This research has been supported by the U.S. Environmental Protection Agency and270 the Center for the Study of Metals in the Environment (CSME), University of Delaware.

271 **References**

- Benedetti, M. F., Van Riemsdijk, W. H., Koopal, L. K., Kinniburgh, D. G., Gooddy, D.
 C., Milne, C. J., 1996. Metal ion binding by natural organic matter: From the
 model to the field. Geochim. Cosmochim. Ac. 60, 2503-2513.
- Cances, B., Ponthieu, M., Castrec-Rouelle, M., Aubry, E., Benedetti, M. F., 2003. Metal
 ions speciation in a soil and its solution: experimental data and model results.
 Geoderma 113, 341-355.
- Carter, M. R., 1993. Soil sampling and methods of analysis. Lewis Publishers, Boca
 Raton, FL.
- Dijkstra, J. J., Meeussen, J. C. L., Comans, R. N. J., 2004. Leaching of heavy metals from
 contaminated soils: an experimental and modeling study. Environ. Sci. Technol.
 38, 4390-4395.
- Gustafsson J. P., Pechová P., Berggren D., 2003. Modeling metal binding to soils: the
 role of natural organic matter. Environ Sci. Technol. 37, 2767-2774.
- Impellitteri, C. A., Lu, Y. F., Saxe, J. K., Allen, H. E., Peijnenburg, W. J. G. M., 2002.
 Correlation of the partitioning of dissolved organic matter fractions with the
 desorption of Cd, Cu, Ni, Pb and Zn from 18 Dutch soils. Environ. Int. 28, 401410.
- Janssen, R. P. T., Posthuma, L., Baerselman, R., Hollander, H. A. D., Van Veen, R. P. M.,
 Peijnenburg, W. J. G. M., 1997. Equilibrium partitioning of heavy metals in
 Dutch field soils. II. Prediction of metal accumulation in earthworm. Environ.
 Toxical. Chem. 16, 2479-2488.
- 293 Kabata-Pendias A, 2001. Trace elements in soils and plants. CRC Press, Boca Raton, FL
- Lee, S. Z., Allen, H. E., Huang, C. P., Sparks, D. L., Sanders, P. F., Peijnenburg, W. J. G.
 M., 1996. Predicting soil-water partition coefficients for cadmium. Environ. Sci.
 Technol. 30, 3418-3424.
- Lumsdon, D. G., 2004. Partitioning of organic carbon, aluminum and cadmium between
 solid and solution in soils: application of a mineral-humic particle additivity
 model. Eur. J. Soil Sci. 55, 271-285.
- Manceau, A., Boisset, M. C., Sarret, G., Hazemann, J. L., Mench, M., Cambier, P., Prost,
 R., 1996. Direct determination of lead speciation in contaminated soils by EXAFS
 spectroscopy. Environ. Sci. Technol. 30, 1540-1552.
- McBride, M., Sauvé, S., Hendershot, W., 1997. Solubility control of Cu, Zn, Cd and Pb
 in contaminated soils. Eur. J. Soil Sci. 48, 337-346.

- Nachtegaal, M., Sparks, D. L., 2003. Nickel sequestration in a kaolinite-humic acid
 complex. Environ. Sci. Technol. 37, 529-534.
- Ponizovsky A. A, Thakali S., Allen H. E., Di Toro D. M., Ackerman A. J., 2006. Effect
 of soil properties on Cu release in soil solutions at low moisture content. Environ.
 Toxicol. Chem. 25, 671-682.
- Roberts, D. R., Scheinost, A. C., Sparks, D. L., 2002. Zinc speciation in a smeltercontaminated soil profile using bulk and microspectroscopic techniques. Environ.
 Sci. Technol. 36, 1742-1750.
- Sauvé, S., Hendershot, W., Allen, H. E., 2000. Solid-solution partitioning of metals in
 contaminated soils: Dependence on pH, total metal burden, and organic matter.
 Environ. Sci. Technol. 34, 1125-1131.
- Sauvé, S., Manna, S., Turmel, M. C., Roy, A. G., Courchesne, F., 2003. Solid-solution
 partitioning of Cd, Cu, Ni, Pb, and Zn in the organic horizons of a forest soil.
 Environ. Sci. Technol. 37, 5191-5196.
- Stevenson, F. J., 1994. Humus Chemistry (2nd edition). John Wiley & Sons, New York,
 NY.
- Tessier, A., Rapin, F., Carignan, R., 1985. Trace metals in oxic lake sediments: possible
 adsorption onto iron oxyhydroxides. Geochim. Cosmochim. Ac. 49, 183-194.
- Tipping, E., 1994. WHAM A chemical equilibrium model and computer code for
 waters, sediment, and soils incorporating a discrete site/electrostatic model of ion
 binding by humic substances. Comput. Geosci. 20, 973-1023.
- Tipping E., 1998. Humic Ion Binding Model VI: An improved description of the
 interactions of protons and metal ions with humic substances. Aquat. Geochem. 4,
 3-48
- Tipping, E., Rieuwerts J., Pan G., Ashmore M. R., Lofts S., Hill M. T. R., Farago M. E.,
 Thornton I. E., 2003. The solid–solution partitioning of heavy metals (Cu, Zn, Cd,
 Pb) in upland soils of England and Wales. Environ Pollut. 125, 213-225.
- 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,
 293-304.
- Trivedi, P., Axe, L., 2000. Modeling Cd and Zn sorption to hydrous metal oxides.
 Environ. Sci. Technol. 34, 2215-2223.
- Weng L. P., Temminghoff E. J. M., van Riemsdijk W. H., 2001. Contribution of
 individual sorbents to the control of heavy metal activity in sandy soil. Environ.
 Sci. Technol. 35, 4436-4443.

340	Weng L. P., Wolthoorn A., Lexmond T. M., Temminghoff E. J. M., van Riemsdijk W. H.,
341	2004. Understanding the effects of soil characteristics on phytotoxicity and
342	bioavailability of nickel using speciation models. Environ. Sci. Technol. 38, 156-
343	162.

344	Zachara, J. M., Smith, S. C., Resch, C. T., Cowan, C. E., 1992. Cadmium sorption to soil
345	separates contaminating layer silicates and iron and alumimun-oxides. Soil Sci.
346	Soc. Am. J. 56, 1074-1084.

347 Figure captions

Figure 1. Adsorption of Cd on three soils at the initial Cd concentration of 0.1 mM. The
SOC concentrations are indicated in the figure. Solid lines are WHAM VI fit. The dash
line indicates the model fit with Tipping's regression equation for the highest SOC soil at
pH <= 5.0.
Figure 2. Linear regression of AOC concentration vs. SOC concentration for all 14 New
Jersey soils.
Figure 3. Adsorption edges for Cd adsorption on four soils at two initial Cd

355 concentrations. (a) Downer loamy sand; (b) Birdsboro silt loam; (c) Boonton Bergen

356 County loam; and (d) Boonton Union County loam soils. Solid lines are WHAM VI

357 predictions. The SOC and initial Cd concentrations are indicated in the figure.

358 Figure 4. Adsorption isotherms for Cd adsorption on four soils at three pHs: (a) Downer

loamy sand; (b) Birdsboro silt loam; (c) Boonton Bergen County loam; and (d) Boonton

360 Union County loam soils. Solid lines are WHAM VI predictions. The SOC

361 concentrations and pH values are indicated in the figure. The C_p and C_w represent Cd

362 concentrations in soils and solutions, respectively.

363 Figure 5. Comparison of K_p values predicted by WHAM VI with values from adsorption

364 edge experiments at different pH ranges. Solid line is the 1:1 line and dashed lines

365 indicate \pm one order of magnitude.

Figure 6. Comparison of K_p values predicted by WHAM VI with values from adsorption

367 isotherm experiments at three pHs. Solid line is the 1:1 line and dashed lines indicate \pm

368 one order of magnitude. The pH values are indicated in the figure.



370 Figure 1.



372 Figure 2.











382 Figure 3.













392 Figure 4.



394 Figure 5



397398 Figure 6.

399 Table 1 Selected soil properties [*]	399	Table 1 Selected soil properties [*]
---	-----	---

Soil Name	Particle	Size Dis	tribution	pН	SOC	[Ca]	ECEC
	Sand	Silt	Clay			(BaCl ₂)	
		%		in H ₂ O	%	mg/kg	meq/100g
Birdsboro silt loam	50	32	18	5.7	1.38	769	5.3
Boonton loam (Bergen County)	60	27	13	5.1	3.43	405	4.2
Boonton loam (Union County)	49	35	16	5.1	7.15	394	4.2
Downer loamy sand	87	5	8	4.7	0.74	5	2.3
Dunellen sandy loam	56	30	14	5.6	1.25	573	4.2
Freehold sandy loam (A horizon)	92	2	6	5.2	0.18	66	0.8
Freehold sandy loam (B horizon)	37	42	21	6.4	1.06	671	4.3
Hazen gravelly loam	39	38	23	6.0	2.31	1590	9.3
Lakewood sand	91	3	6	4.2	0.69	1	0.9
Penn silt loam	25	48	27	4.7	1.13	268	3.8
Rockaway stony loam	54	30	16	4.7	2.99	143	2.7
Sassafras sandy loam	45	37	18	5.8	0.62	384	3.1
Washington loam	20	49	31	6.03	2.14	1461	8.9
Whippany silty clay loam	49	16	37	6.17	1.67	1508	9.5

400 * Metal oxides concentration can be found in Lee et al. (1996).

Fraction of active organic carbon (f) and active Al ([Al] _a) at low pH obtained from model fittings									
Soil Freehold Sassafras (A)		Lakewood	Downer	Freehold (B)	Penn	Dunellen			
SOC (%)	0.18	0.62	0.69	0.74	1.06	1.13	1.25		
f	0.600	0.634	0.349	0.598	1.000	0.430	0.806		
$\left[\mathrm{Al} \right]_{a} (\mathrm{M})$	6.41× 10 ⁻⁵	9.16× 10 ⁻⁵	4.00×10^{-5}	1.00×10^{-4}	1.10×10^{-4}	9.75×10^{-5}	1.18×10^{-4}		
Soil	Birdsboro	Whippany	Washington	Hazen	Rockaway	Boonton Bergen	Boonton Union		
SOC (%)	1.38	1.67	2.14	2.31	2.99	3.43	7.15		
f	0.835	1.000	0.894	1.000	0.408	0.688	0.557		
$\left[\mathrm{Al} \right]_{a} (\mathrm{M})$	1.27×10^{-4}	1.00×10^{-4}	2.90×10^{-4}	2.82×10^4	2.11×10 ⁻⁴	2.90×10^{-4}	4.47× 10 ⁴		
WHAM VI input parameters for solution phases									
[Na ⁺] [NO ₃ ⁻] pCO ₂ $\log^* K_{s0}$ (Al) $\log^* K_{s0}$ (Fe(III))		(III)) l	Particle concer	ntration					
М		atm				g/L			
0.01	0.01 0.01 3.16E		8.5	3		10			