

Original Research

Adsorption Mechanism of Heavy Metals in Heavy Metal/Pesticide Coexisting Sediment Systems through Fractional Factorial Design Assisted by 2D-QSAR Models

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Abstract

Resolution V of the 2^{10-3} fractional factorial design method was used to identify the main effects and second-order interaction effects of pollutants (copper, cadmium, lead, zinc, nickel, dimethoate, metalaxyl, atrazine, malathion, and prometryn) on metals adsorption onto sediments, and 2D-QSAR models were established to reveal the relationships between metal ion characteristics and the effects of pollutants on metals adsorption. The effects on Cd adsorption were attributed to the main effects of pollutant factor concentrations, while the effects on Cu, Pb, Zn, and Ni adsorption were from the second-order interaction effect. No interference with Cd adsorption was observed, and the synergistic contribution of the main effects and second-order interaction effects on Ni adsorption was 67.26%. Additionally, the antagonistic contribution rates to Cu, Zn, and Pb adsorption reached 55.31%, 73.16%, and 86.23%, respectively. Significant correlations existed between the main effects and ionization potential (IP), the change in ionization potential (Δ IP) and ion hydrolysis ability of metals, and the second-order interaction effects with atomic number, atomic weight, and polarizing power (Z^2/r) of metals. The electrochemical potential, Δ IP and IP of metals were found to promote adsorption, Z^2/r , electronegativity and atomic radius of metals to inhibit adsorption significantly. Overall, the results provide theoretical support that helps elucidate compound pollution regulation among heavy metal pollutants in complex environmental systems.

Keywords: adsorption mechanism, fractional factorial design, heavy metals, pesticides, 2D-QSAR model

Introduction

The rapid development of industry and agriculture has led to the production of large amounts of heavy metals and pesticides, which have often entered and gradually accumulated in the environment [1-2]. This has caused serious pollution of water [3], the atmosphere [4], and soil [5], and has threatened the health of animals, plants, and humans [6-7]. In addition to their own environmental pollution, the interactions among heavy metals and pesticides may also produce synergistic or antagonistic effects to form compound pollution [8-9]. The combined pollution effects lead to changes in the migration and transformation pathways as well as in the biological effects of pollutants within different environmental media [10].

Sediment consists of tiny particles that move with fluid flow and eventually become a layer comprising a solid particle system at the bottom of an aquatic system. Because of its unique structure, sediment can become enriched with heavy metals and pesticides because it has significant adsorption effects on pollutants in the environment [11-12]. Sediment is the source and sink of many pollutants in aquatic environments [13-14], and an important carrier facilitating the migration of pollutants in natural aquatic systems [15]. Therefore, several researchers have conducted studies evaluating the adsorption behavior of heavy metals and pesticides in sediments, including investigations of heavy metals and pesticide content in different geographical sediments [16-18] and analyses of pollutant adsorption characteristics of sediments exposed to compound pollution [19-21].

To investigate compound pollution, researchers usually use orthogonal design, factorial design, central composite design, and other experimental design methods; however, when the interactions of multiple experimental factors are taken into consideration, the fractional factor design can effectively reduce the scale of the experiment [22]. Wang et al. used resolution III of the 2^{10-6} fractional factorial design method assisted by a complete foldover design and confounding design to explore the adsorption characteristics of dimethoate and a variety of pollutants on sediments [23]. In their study, they solved the alias structure between the main effect and the second-order interaction effect, but could not clearly identify the second-order interaction effect. Gu et al. [24] and Cheng [25] further differentiated the influence of each second-order interaction effect on target pollutants and used resolution V of the 2^{10-3} fractional factorial design method to investigate the cadmium and nickel competitive adsorption characteristics under a heavy metal (copper, cadmium, lead, zinc, and nickel) and pesticide (dimethoate, metalaxyl, atrazine, malathion, and prometryn) coexisting sediment system while focusing on the adsorption characteristics of a single heavy metal in the compound pollution system. However, they were not able to explain the heavy metals adsorption mechanism based on their nature when they coexisted. Therefore, the present study employed a fixed

effect model of resolution V of the 2^{10-3} fractional factorial design method to investigate the adsorption characteristics of copper, lead, and zinc in the system mentioned above. Construction of the 2D-QSAR model helped elucidate the adsorption mechanism of heavy metals in the composite pollution system. Overall, the results of this study provide a theoretical basis for the regulation of heavy metal pollutants in aquatic environments.

Material and Methods

Collection and Preparation of Samples

Sediment samples were collected from the Songhua River located in Jilin City, Jilin Province, China. Samples were taken from the sediment surface layer (0-5 cm in depth) with a bucket-type sampler, and 4 samples from the same sampling point were mixed uniformly into a single sample. All samples were then filtered to remove water, separated from stones and debris, packed in plastic bags, and transported to the laboratory. Finally, samples were naturally air dried, ground, sieved, and transferred into a sealed jar [26].

Experimental Methods and Designs

Five heavy metals – copper (A), cadmium (B), lead (C), zinc (D) and nickel (E) – and 5 pesticides – dimethoate (F), metalaxyl (G), atrazine (H), malathion (J) and prometryn (K) – were selected as the experimental factors. The 5 heavy metals were the target pollutants used to explore the heavy metals characteristics of migration and transformation in the water environment through the competitive adsorption of various pollutants. The concentrations of heavy metals and pesticides were 60-150 $\mu\text{mol/L}$ and 10-25 $\mu\text{mol/L}$, respectively. Resolution V of the 2^{10-3} fractional factorial design method was used to set the concentration of 10 contaminants in each mixed solution, and the parallel samples and blank samples were set with a total of 384 groups. For 256 groups, 20 mL of mixed solution was transferred into a conical flask containing 0.5000 \pm 0.0001 g of sediment samples, while for the other 128 groups, 20 mL of mixed solution was set into a conical flask without sediments, then shaken for 48 h and filtered through a 0.22 μm membrane. The concentration of pesticides and heavy metals in the solution was determined by high-performance liquid chromatography (HPLC) and flame atomic absorption spectrophotometry, and the difference between the concentration of contaminants in the blank solution and the solution after adsorption was used to calculate the heavy metals adsorption in sediments [26]. The experimental data describing the heavy metal adsorption onto the sediments is shown in Table 1.

The principle of resolution V of the 2^{10-3} fractional factorial design method determined the generators of the main and interaction effects of the pollutant factor concentration. The alias structures were as follows.

Table 1. Heavy metal absorption on sediments in heavy metal and pesticide system ($\mu\text{mol/g}$).

No.	Adsorption					No.	Adsorption				
	Cu	Cd	Pb	Zn	Ni		Cu	Cd	Pb	Zn	Ni
1	10.75	16.68	7.66	10.22	5.49	65	6.95	6.39	1.90	9.31	8.60
2	11.29	7.63	7.98	7.58	13.14	66	9.75	1.69	4.07	12.02	20.28
3	11.07	11.43	20.59	7.33	3.25	67	21.59	3.51	17.25	4.33	7.43
4	21.93	11.65	6.45	4.34	4.02	68	23.23	7.47	6.18	8.59	23.31
5	2.79	1.40	0.03	0.21	0.59	69	24.61	10.07	21.31	4.95	1.21
6	12.75	13.95	21.51	8.08	13.22	70	23.38	4.70	6.78	5.78	2.80
7	22.47	5.83	21.36	4.93	5.54	71	22.95	13.84	6.90	8.27	3.65
8	28.80	4.50	9.22	11.86	11.26	72	27.73	20.91	8.03	8.11	19.32
9	12.04	7.71	20.35	8.06	16.18	73	24.07	10.01	19.46	3.91	5.74
10	21.89	11.10	7.76	4.31	8.71	74	9.48	6.72	15.02	10.69	5.38
11	22.88	6.39	7.01	5.16	13.14	75	24.91	3.18	19.05	6.60	1.08
12	20.74	9.71	6.58	5.18	3.06	76	10.52	11.50	17.81	7.07	7.51
13	11.51	6.57	20.14	5.90	5.24	77	9.06	8.24	5.53	13.93	7.97
14	9.60	11.08	5.80	8.13	3.44	78	11.54	6.70	18.68	9.51	5.40
15	10.65	13.72	18.72	6.30	4.79	79	10.73	2.30	5.65	12.08	16.90
16	13.20	6.70	23.81	7.02	14.40	80	9.77	18.50	5.70	6.91	4.93
17	10.58	10.54	18.54	5.87	3.71	81	11.16	14.29	17.88	10.45	16.89
18	22.33	11.58	7.43	6.18	9.76	82	11.09	12.97	20.19	7.77	3.17
19	18.50	13.83	6.18	3.71	6.89	83	24.19	14.98	6.40	6.72	4.00
20	9.67	5.72	5.86	9.61	10.19	84	18.27	2.22	14.26	0.90	2.32
21	15.05	8.89	3.98	3.58	8.81	85	12.23	8.90	19.59	3.75	5.70
22	10.23	6.61	18.65	10.09	5.64	86	22.81	6.65	27.07	7.33	11.12
23	13.54	7.66	24.44	10.14	16.77	87	23.80	9.06	8.14	6.29	2.53
24	11.22	8.85	6.77	10.93	6.25	88	10.21	21.95	5.56	24.99	8.89
25	11.58	17.33	7.99	7.74	14.11	89	10.11	15.95	5.74	7.20	10.94
26	22.63	4.67	21.49	5.02	10.87	90	8.17	6.96	3.52	7.88	5.35
27	10.14	6.64	6.47	12.09	5.55	91	12.47	23.56	6.75	10.74	7.20
28	23.03	5.67	6.98	7.81	4.35	92	22.98	4.91	6.29	5.84	1.00
29	12.61	11.33	7.69	13.85	7.08	93	10.80	16.14	5.82	10.77	5.64
30	23.46	11.12	8.37	5.74	3.25	94	9.68	10.01	6.01	5.21	4.69
31	12.94	17.10	20.06	8.75	4.16	95	19.39	10.59	5.03	6.19	5.12
32	13.36	8.68	22.80	8.44	6.77	96	9.86	9.04	11.00	1.92	4.91
33	24.07	5.63	7.40	7.81	4.13	97	28.48	20.64	22.17	10.33	1.97
34	9.64	7.44	5.43	11.83	4.42	98	23.90	4.09	16.31	4.23	4.59
35	10.31	15.08	7.42	8.09	4.19	99	8.78	7.81	4.99	15.91	2.04
36	17.89	1.05	18.68	2.81	4.43	100	25.24	11.88	20.34	7.18	5.24
37	10.89	7.58	18.26	7.57	6.65	101	11.44	6.03	17.56	9.02	10.39
38	18.73	4.80	19.54	2.81	6.33	102	24.81	4.67	19.32	6.15	2.37
39	11.65	6.78	20.77	7.79	4.59	103	11.43	15.99	19.02	7.51	10.67

Table 1. Continued.

No.	Adsorption					No.	Adsorption				
	Cu	Cd	Pb	Zn	Ni		Cu	Cd	Pb	Zn	Ni
40	20.75	8.10	6.65	4.39	7.51	104	19.83	2.39	16.61	3.05	2.68
41	23.08	5.35	7.73	3.55	22.53	105	25.98	6.26	23.86	2.26	1.81
42	7.85	15.38	32.17	2.15	16.60	106	23.81	12.38	19.47	7.08	2.81
43	18.01	8.97	18.45	3.35	9.11	107	13.32	12.56	3.95	14.22	21.73
44	23.22	4.59	7.89	6.90	7.39	108	12.58	9.62	7.66	11.75	10.11
45	18.87	5.57	17.98	2.11	6.77	109	23.75	12.65	19.60	4.81	0.10
46	12.67	16.07	19.93	7.00	18.01	110	19.88	14.43	4.51	5.94	2.67
47	22.79	2.93	21.19	3.78	0.94	111	9.05	15.29	4.43	10.48	2.42
48	8.85	6.80	4.75	8.17	3.73	112	25.47	6.94	18.64	7.24	3.31
49	10.98	8.34	6.70	11.83	2.72	113	10.97	15.40	17.36	7.47	0.55
50	11.04	5.99	8.42	9.12	7.86	114	25.86	7.64	7.99	1.44	9.34
51	21.35	3.94	19.05	4.97	4.70	115	11.88	6.65	17.19	6.37	2.39
52	23.48	3.61	21.69	5.33	5.82	116	10.23	7.15	17.53	3.29	1.41
53	25.55	7.71	7.58	8.41	9.71	117	19.93	4.37	5.83	1.43	0.88
54	10.08	7.47	6.47	7.81	3.44	118	22.23	5.49	6.32	5.66	5.04
55	21.55	4.79	18.44	2.41	2.45	119	9.73	9.35	17.42	9.58	2.09
56	12.96	27.48	22.58	5.38	12.80	120	20.35	3.85	19.10	1.44	8.19
57	9.81	5.71	6.17	6.73	3.67	121	33.75	10.78	30.60	8.71	8.03
58	12.29	8.38	20.16	6.34	4.42	122	9.30	5.73	4.86	6.11	4.19
59	11.64	0.89	20.53	1.99	1.67	123	21.23	8.08	5.28	8.74	5.81
60	1.56	4.75	0.53	3.13	6.93	124	8.30	12.68	4.47	8.65	4.70
61	23.68	9.71	21.05	4.73	3.54	125	9.71	16.01	6.15	9.18	6.78
62	20.45	3.52	21.20	1.58	7.08	126	21.02	10.48	5.36	4.18	0.80
63	18.98	7.14	21.53	1.96	5.91	127	22.30	7.66	5.85	7.78	3.26
64	0.52	11.56	6.06	6.73	2.64	128	23.58	5.55	6.31	4.66	6.74

Aliases of the main effects of the pollutant factor concentration: A = EFHJ, B = CFGH, C = BFGH, D = GHJK, E = AFHJ, F = BCGH = AEHJ, G = BCFH = DHJK, H = BCFG = AEFJ = DGJK, J = AEFH = DGHK, K = DGHJ.

Aliases of the second-order interaction effects of the pollutant factor concentration: AB = CEGJ, AC = BEGJ, AD = EFGK, AE = FHJ = DFGK = BCGJ, AF = EHJ = DEGK, AG = DEFK = BCEJ, AH = EFJ, AJ = EFH = BCEG, AK = DEFG, BC = FGH = DFJK = AEGJ, BD = CFJK, BE = ACGJ, BF = CGH = CDJK, BG = CFH = ACEJ, BH = CFG, BJ = CDFK = ACEG, BK = CDFJ, CD = BFJK, CE = ABGJ, CF = BGH = BDJK, CG = BFH = ABEJ, CH = BFG, CJ = BDFK = ABEG, CK = BDFJ,

DE = AFGK, DF = AEGK = BCJK, DG = HJK = AEFK, DH = GJK, DJ = GHK = BCFK, DK = GHJ = AEFK = BCFJ, EF = AHJ = ADGK, EG = ADFK = ABCJ, EH = AFJ, EJ = AFH = ABCG, EK = ADFG, FG = BCH = ADEK, FH = BCG = AEJ, FJ = AEH = BCDK, FK = ADEG = BCDJ, GH = BCF = DJK, GJ = DHK = ABCE, GK = DHJ = ADEF, HJ = AEF = DGK, HK = DGJ, JK = DGH = BCDF.

The above alias structure demonstrates that, when the third-order and above interaction effects are neglected, resolution V of the 2^{10-3} fractional factorial design method can completely distinguish the main effects and the second-order interaction effects of the pollutant factor concentration.

Results and Discussion

The design of experiment (DOE) function in the Minitab software was used to establish the fixed effect model of resolution V of the 2^{10-3} fractional factorial design. The variance in the relationship between the pollutant factor concentration and the adsorption capacity of heavy metals on sediments, as well as the effect value estimation of the main effects and the second-order interaction effects of the pollutant factor concentration on the heavy metal adsorption by sediments were analyzed. According to the P value (significance level $P = 0.05$) of the effect value estimation, the main effects and second-order interaction effects of each pollutant factor concentration were analyzed to determine if the pollutant factor had a significant effect on heavy metals adsorption.

Adsorption Characteristic Analysis of Heavy Metals Based on the Fixed Effect Model

Variance Analysis

In resolution V of the 2^{10-3} fractional factorial design, variance analysis of the main effects and the

second-order interaction effects of the pollutant factor concentration on the adsorption capacity of the heavy metals was conducted using the fixed effects model (Table 2).

The variance analysis was used to test whether the main effects and the second-order interaction effects of the pollutant factor concentration generally had a significant effect on the heavy metals adsorption. The degrees of freedom reflect the number of main effects and second-order interaction effects of pollutant factor concentration that can be evaluated. The P value represents the diminishing index of the result reliability – namely the significance level, which is usually 0.05. As shown in Table 2, the main effects and second-order interaction effects of pollutant factor concentration ($P < 0.05$) have a significant effect on the adsorption of Cu, Pb, and Ni, while the second-order interaction effects ($P < 0.05$) have a significant effect on the adsorption of Zn by sediments, and the main effects ($P < 0.05$) have a significant effect on Cd adsorption. Therefore, it is necessary to screen out the specific main effects and second-order interaction effects of pollutant factor concentrations in order to analyze the mechanism of heavy metals adsorption onto sediments.

Table 2. Variance analysis of the effects of pollutant factor concentrations on heavy metal adsorption.

Heavy metal	Source	DF	SS	MS	F	P
Cu	Main effect	10	1074	107.3666	21.85	0.030
	Second-order interaction	45	364	8.09	1.21	0.016
	Residual error	72	190	2.64		
	Total	127	1628			
Pb	Main effect	10	853	85.33	1.49	0.047
	Second-order interaction	45	515	11.44	1.36	0.000
	Residual error	72	437	6.07		
	Total	127	1805			
Zn	Main effect	10	133	13.32	1.20	0.065
	Second-order interaction	45	619	13.75	36.24	0.000
	Residual error	72	799	11.10		
	Total	127	1551			
Cd ^a	Main effect	10	1488	148.76	45.84	0.000
	Second-order interaction	45	211	4.70	1.45	0.080
	Residual error	72	234	3.25		
	Total	127	1933			
Ni ^b	Main effect	10	710	71.00	18.64	0.000
	Second-order interaction	45	652	14.49	3.80	0.000
	Residual error	72	274	3.81		
	Total	127	1636			

^a Gu et al. (2017), ^b Cheng (2015). DF: degree of freedom; SS: sum of squares; MS: mean square; F: F-test value; P: significance level. The bold terms mean significant at the significance level of 0.05.

Table 3. Estimates and contribution rates of the main and second-order interaction effects of significant pollutant factor concentrations on adsorption of heavy metals.

Heavy metal	Factor	Estimate of effect	P (P<0.05)	Contribution rate (%)	
				synergism	antagonism
Cu	Cu	5.007	0.000	28.62	
	Metalaxyl*Cu	-2.759	0.044		15.77
	Atrazine*Prometryn	2.812	0.041	16.07	
	Malathion*Pb	-2.889	0.036		16.51
	Cu*Pb	-4.030	0.004		23.03
	Total	17.497		44.69	55.31
Pb	Malathion	-3.264	0.017		14.68
	Pb	3.061	0.025	13.77	
	Dimethoate*Cu	-2.852	0.037		12.83
	Metalaxyl*Malathion	-2.793	0.041		12.56
	Malathion*Pb	-4.376	0.002		19.68
	Prometryn*Pb	-3.012	0.028		13.54
	Pb*Ni	-2.877	0.035		12.94
	Total	22.235		13.77	86.23
Zn	Cu	-1.402	0.020		19.36
	Metalaxyl*Zn	-1.246	0.038		17.20
	Atrazine*Ni	1.945	0.001	26.84	
	Prometryn*Zn	-1.470	0.015		20.30
	Cu*Ni	-1.181	0.049		16.30
	Total	7.243		26.84	73.16
Cd ^a	Dimethoate	-0.706	0.030		5.03
	Cu	-2.442	0.000		17.41
	Cd	5.821	0.000	41.49	
	Pb	-1.836	0.000		13.09
	Zn	-1.412	0.000		10.06
	Cu*Zn	1.171	0.000	8.35	
	Zn*Ni	-0.641	0.048		4.57
	Total	14.029		49.84	50.16

Effect Value Estimation and Contribution Rate Analysis

In resolution V of the 2¹⁰⁻³ fractional factorial design, the fixed effect model can be applied to effect value estimation and contribution rate analysis (Table 3) of the main effects and second-order interaction effects of the pollutant factor concentration, which have significant effects on heavy metals adsorption (P<0.05).

As shown in Table 3, the factors having significant main effects of the pollutant factor concentration on Cd adsorption were Cd, Cu, Pb, Zn, and dimethoate,

which contributed 87.08% of the total. Additionally, the significant second-order interaction effects were Cu*Zn and Zn*Ni, which contributed 12.92% [24]. These findings indicate that the effects on Cd adsorption can be attributed to the main effects of pollutant factor concentrations. Moreover, in systems in which heavy metals and pesticides coexist, the effects of heavy metals on Cd adsorption are more obvious. The contribution rates of the main effects and the second-order interaction effects of pollutant factor concentrations on Ni adsorption were 38.09% and 61.91%, respectively [25], showing that the effects on Ni adsorption onto the sediments are due to the second-order interaction effects of pollutant

Table 3. Continued

Ni ^b	Metalaxyl	1.345	0.000	5.82	
	Cu	-0.803	0.023		3.47
	Cd	-0.872	0.014		3.77
	Pb	-0.876	0.013		3.79
	Zn	-0.828	0.019		3.58
	Ni	4.083	0.000	17.66	
	Dimethoate*Atrazine	-0.818	0.020		3.54
	Dimethoate*Prometryn	0.989	0.005	4.28	
	Dimethoate*Cu	0.707	0.044	3.06	
	Dimethoate*Cd	-0.788	0.025		3.41
	Dimethoate*Zn	1.233	0.001	5.33	
	Metalaxyl*Atrazine	-0.759	0.031		3.28
	Metalaxyl*Malathion	-1.087	0.002		4.70
	Metalaxyl*Ni	1.412	0.000	6.11	
	Atrazine*Prometryn	1.147	0.001	4.96	
	Atrazine*Cd	1.895	0.000	8.20	
	Atrazine*Pb	0.755	0.032	3.26	
	Malathion*Cd	-0.739	0.036		3.20
	Malathion*Pb	0.830	0.019	3.59	
	Prometryn*Cd	0.837	0.018		
Pb*Zn	1.154	0.001	4.99		
Total	23.119		67.26	32.74	

^aGu et al. (2017), ^bCheng (2015)

factor concentrations. Moreover, the contribution rates of the main effects of pollutant factor concentrations on the adsorption of Cu, Pb, and Zn were lower than those of the second-order interaction effects, demonstrating that the effects on Cu, Pb, and Zn adsorption are from the second-order interaction effects of pollutant factor concentrations. The above analysis demonstrates that the second-order interaction effects of pollutant factor concentration play a dominant role in the adsorption of heavy metals (Cu, Pb, Zn, and Ni) onto sediments in the same adsorption system; thus, investigating heavy metals/pesticides pollution is of great importance for elucidating the characteristics of heavy metals adsorption onto sediments.

As shown in Table 3, under the same adsorption system there were several differences in the main effects and the second-order interaction effects of the pollutant factor concentrations, which had synergistic or antagonistic effects on the adsorption of five heavy metals on the sediments. The contribution of the main effects and the second-order interaction effects of the pollutant factor concentration with synergistic effects on Cd adsorption was as high as 49.84%, while the contribution

rate with antagonistic effects was calculated to be 50.16%, indicating that the synergism and antagonism cancel each other out and reach an equilibrium state [24]. These findings indicate that the coexisting system of heavy metals and pesticides has no interference effect on the adsorption of Cd onto sediments. The contribution rates of the main effects and the second-order interaction effects with synergism and antagonism on Ni adsorption reached 67.26% and 32.74%, respectively [25], indicating that the pollutant factor has greater promoting effects on Ni adsorption in the compound pollution system. The synergism contribution rates of the main effects and the second-order interaction effects to the adsorption of Cu, Zn, and Pb onto sediments were calculated to be 44.69%, 26.84%, and 13.77%, respectively, while the antagonism contribution rates were 55.31%, 73.16%, and 86.23%, respectively. These findings indicate that an antagonistic effect plays a dominant role in Cu, Zn, and Pb adsorption on sediments, and that in the heavy metal/pesticide coexistence system, the adsorption of Cu, Zn, and Pb is enhanced by interference with the pollutant factor.

Finally, the adsorption effect values of the five heavy metals revealed that the competitive adsorption capacity

of heavy metals on sediments decreased in the order Ni>Pb>Cu>Cd>Zn.

Adsorption Mechanism Analysis of Heavy Metals based on the 2D-QSAR Model

Studies have shown that the adsorption capacities of heavy metals are related to their physical and chemical properties [27-28]. To further reveal the mechanism of adsorption of heavy metals onto sediments, this study used multiple linear regressions to establish 2D-QSAR models with 11 selected characteristic ion parameters of heavy metals as independent variables and the main effect values, second-order interaction effect values, synergistic effect values, and antagonistic effect values of heavy metals adsorption on sediments as dependent variables. These metal ionic characters represent the structure of atoms in term of mathematical values and could be used to calculate adsorption properties.

Correlation Analysis

The ion characteristic parameters of the heavy metals selected by the model include: atomic number (AN), Pauling ionic radius (r), Pauling electronegativity (Xm), electrochemical potential (ΔE_0), change in ionization potential between the OX and OX⁻¹ states (ΔIP), polarizing power (Z^2/r , Z is the ionic charge), ion hydrolysis ability ($|\log K_{OH}|$, K_{OH} from $M^{n+} + H_2O \rightarrow MOH^{n-1} + H^+$), soft

index (σp), atomic radius (AR), atomic weight (AW), and ionization potential (IP) [29-30]. The ion characteristic parameters of Cu, Cd, Pb, Zn, and Ni and their correlations with heavy metal adsorption effect values are shown in Tables 4 and 5, respectively.

As shown in Table 5, when the significance level was 0.05, the significant correlations between the ion characteristic parameters of heavy metals and the main effect values of heavy metal adsorption were IP, ΔIP , and $|\log K_{OH}|$, and the significant correlations with the second-order interaction effect values of heavy metal adsorption were AN, AW, and Z^2/r . Additionally, ΔE_0 , ΔIP , and IP were significantly correlated with the synergistic effect values of heavy metals adsorption, while Z^2/r , Xm, and AR were significantly correlated with the antagonistic effect values of heavy metals adsorption.

Construction and Evaluation of the 2D-QSAR Model

The 2D-QSAR models between the main effect values, second-order interaction effect values, synergistic effect values, and antagonistic effect values of heavy metals adsorption and their significantly relevant ion characteristic parameters of heavy metals are as follows:

$$Y_{\text{main effect}} = -3.083 + 0.779 \Delta IP + 0.685 IP - 1.619 |\log K_{OH}| \quad (1)$$

Table 4. Partial ion characteristic parameters of heavy metals.

Heavy metal	AN	Xm	ΔE_0	r	$ \log K_{OH} $	σp	Z^2/r	ΔIP	IP	AR	AW
Cu ²⁺	29	1.90	0.16	0.73	8.0	0.104	5.48	12.57	20.29	1.35	63.54
Cd ²⁺	48	1.69	0.40	0.95	10.1	0.081	4.21	7.91	16.91	1.48	112.40
Pb ²⁺	82	2.33	0.13	1.18	7.7	0.131	3.39	7.61	15.03	1.54	207.19
Zn ²⁺	30	1.65	0.76	0.75	9.0	0.115	5.33	8.57	17.96	1.31	65.37
Ni ²⁺	28	1.91	0.23	0.69	9.9	0.126	5.80	10.52	18.17	1.25	58.71
References	McCloskey et al., 1996								Wolterbeek and Verburg, 2001		

AN: atomic number; AR: atomic radius; AW: atomic weight; r: Pauling ionic radius; Xm: Pauling electronegativity; ΔE_0 : electrochemical potential; IP: ionization potential; ΔIP : the change in ionization potential; Z^2/r : polarizing power; $|\log K_{OH}|$: ion hydrolysis ability; σp : soft index

Table 5. Correlation analysis of the ion characteristic parameters of heavy metals and the effect values of heavy metal adsorption.

Effect value	Pearson correlation											
	AN	AR	AW	r	Xm	ΔE_0	IP	ΔIP	Z^2/r	$ \log K_{OH} $	σp	
Main effect	-0.336	-0.211	-0.323	-0.362	0.088	-0.415	0.852*	0.864*	0.378	-0.944*	-0.077	
Second-order interaction	0.901*	-0.813	0.869*	-0.852	-0.679	0.235	0.604	0.455	0.879*	0.779	-0.249	
Synergistic effect	-0.164	0.052	-0.158	-1.156	0.122	-0.949*	0.877*	-0.863*	0.150	-0.297	-0.357	
Antagonistic effect	-0.828	-0.984*	-0.726	-0.600	-0.960*	0.439	0.737	0.571	-0.947*	0.343	0.105	

* Statistically significant correlation at level of 0.05

Table 6. Statistical parameters of the 2D-QSAR model (n = 5).

Model	R	R ²	R ² _{adj}	SD	F
1	0.966	0.933	0.903	0.852	34.666
2	0.997	0.993	0.973	0.415	49.251
3	0.966	0.934	0.936	0.764	54.716
4	0.993	0.986	0.943	0.729	22.868

R: correlation coefficient; R²: determination coefficient; R²_{adj}: adjusted R square; SD: standard deviations; F: F-test value

$$Y_{\text{second-order interaction}} = -32.714 + 3.294 \text{ AN} + 1.210 \text{ AW} + 2.192 \text{ Z}^2/\text{r} \quad (2)$$

$$Y_{\text{synergistic effect}} = -36.881 - 21.335 \Delta E_0 - 3.847 \Delta \text{IP} + 4.961 \text{ IP} \quad (3)$$

$$Y_{\text{antagonistic effect}} = 29.935 - 0.132 \text{ Z}^2/\text{r} - 1.761 \text{ Xm} - 23.962 \text{ AR} \quad (4)$$

...where $Y_{\text{main effect}}$, $Y_{\text{second-order interaction}}$, $Y_{\text{synergistic effect}}$ and $Y_{\text{antagonistic effect}}$ are the values of the main effect, second-order interactions effect, synergistic effect and antagonistic effect of the adsorption of the five heavy metals. Additionally, ΔIP is the change in ionization potential, IP is the ionization potential, $|\log K_{\text{OH}}|$ is the ion hydrolysis ability, AN is the atomic number, AW is the atomic weight, Z^2/r is the polarizing power, ΔE_0 is electrochemical potential, Xm is the electronegativity, and AR is the atomic radius of heavy metals.

As shown in Table 6, the correlation coefficients (R) of the model were 0.966, 0.997, 0.996, and 0.993 (n = 5, >0.8783), all of which passed the statistical test (P = 0.05). Additionally, the determination coefficients (R²) were 0.933, 0.993, 0.934, and 0.986, and the adjusted R square (R²_{adj}) values were 0.903, 0.973, 0.936, and 0.943, respectively, indicating that the models have good predictive ability [31-32]. Moreover, the standard deviations (SD) of the model were 0.852, 0.415, 0.764, and 0.729, and the F-test values were 34.666, 49.251, 54.716, and 22.868, respectively, demonstrating that the established models had good fits [32].

2D-QSAR Model Analysis

As shown in model (1), a larger ionization potential (IP) and change in ionization potential (ΔIP) were associated with a greater main effect value of heavy metals adsorption. Additionally, a smaller $|\log K_{\text{OH}}|$ value was associated with a stronger hydrolysis ability of heavy metal ions and a greater main effect value of heavy metals adsorption. The IP reflects the affinity and electronegativity of electrons [33] and represents the activity of heavy metal ions and the orbital energy involved in the reaction process. A larger IP value is associated with stronger electron affinity of the heavy metal ions; thus, Cu is more easily adsorbed onto the sediments because it has the highest IP value among

the five heavy metals investigated. This occurs because the main effects of the pollutant factor concentration have greater synergistic effects on Cu adsorption in the contaminant coexistence system. The $|\log K_{\text{OH}}|$ reflects the tendency for connections between heavy metal ions and intermediate ligands (e.g., oxygen-containing groups) [34], with a stronger hydrolysis ability of heavy metal ions being associated with a stronger affinity for intermediate ligands. Kang et al. showed that carboxyl groups play a major role in the adsorption of Pb, while carboxyl and hydroxyl groups play an important role in the adsorption of Cu [35]. As a result, the combination of heavy metal ions and oxygen-containing groups can promote the adsorption of heavy metals onto sediments.

In model (2), the second-order interaction effect value of heavy metals adsorption increased as the ion characteristic parameters increased (AN, AW, and Z^2/r). The polarizing power (Z^2/r) characterizes the intensity of the interaction between heavy metal ions and sediment components, as well as the tendency to form ionic bonds, which also reflects the energy of the electrostatic interaction between heavy metal ions and ligands [36]. When Z^2/r is larger, the interaction of heavy metal ions with sediment components is stronger, and the parameter Z^2/r has a negative linear relationship with $|\log K_{\text{OH}}|$ [33]. Specifically, a larger Z^2/r is associated with a smaller $|\log K_{\text{OH}}|$. In other words, a greater ability of ion hydrolysis is associated with a stronger binding capacity of heavy metal ions and intermediate ligands, so the heavy metals are more favorably absorbed onto sediments. Therefore, a high value of Z^2/r for Ni is beneficial to absorption onto the sediments, and the second-order interaction effects of pollutant factor concentration have the greatest synergistic effect on the adsorption of Ni in the composite contamination system. In addition, the atomic weight (AW) increases as the atomic number (AN) increases, with a greater atomic number or atomic weight of the heavy metal making its adsorption onto sediments more favorable.

As shown in model (3), the coefficients of the parameters ΔE_0 and ΔIP were both negative, indicating that electrochemical potential (ΔE_0) and change in ionization potential (ΔIP) have a negative effect on the synergistic effect value of heavy metal adsorption, to a certain extent, which can inhibit the adsorption of heavy metals onto the sediment. While the coefficient of the parameter IP is positive (meaning that IP has a positive effect on the synergistic effect value of heavy metals adsorption), it can promote the adsorption of heavy metals onto the sediment. Therefore, Pb, which had the lowest value of ΔE_0 , and Cu, which had with the highest IP value, is subject to greater synergism associated with the pollutant factor concentration, resulting in easier adsorption onto the sediments.

As shown in model (4), the coefficients of the parameters Z^2/r , Xm, and AR were both negative, indicating that the polarizing power (Z^2/r), electronegativity (Xm), and atomic radius (AR) have a negative effect on heavy metals adsorption. Specifically, the inhibition of heavy metals

adsorption is weakened; therefore, it is conducive to the adsorption of heavy metals onto the sediments. There are different active sites on the surface of the sediment, and high electronegativity can induce strong interactions between active sites and heavy metal ions, resulting in multiple adsorptions of heavy metal ions [37]. In this study, Pb, Ni, and Cu, which have high electronegativity, were found to be more easily adsorbed onto sediments than Cd and Zn, which have low electronegativity. These findings are consistent with the conclusion that the competitive adsorption capacity of heavy metals occurs in the order Ni>Pb>Cu>Cd>Zn. In addition, a greater value of AR was associated with a greater contacting area of heavy metal ions and sediments; thus, it could promote the adsorption of heavy metals onto the sediments. Pb, which has the largest AR value of the 5 investigated heavy metals, is subject to the lower antagonistic effects of the pollutant factor concentration in the contaminant coexistence system; therefore, it can be more easily adsorbed onto sediments. Chen et al. [33] found that the larger ion radius was favorable for the adsorption of heavy metals in a biological adsorption study of Pb²⁺, Ag⁺, Cr³⁺, Cu²⁺, Zn²⁺, Cd²⁺, Co²⁺, Sr²⁺, Ni²⁺, and Cs⁺.

The above analysis explored the intrinsic relationship between the values of the main effects, second-order interaction effects, synergistic effects, and antagonistic effects on the adsorption of 5 heavy metals and the physical and chemical properties of heavy metals to provide a basis for further investigation of the adsorption mechanism of heavy metals on sediments in a compound pollution system.

Conclusions

In this study, resolution V of the 2¹⁰⁻³ fractional factorial design method with the fixed effect model was used to identify the significant (P = 0.05) main effects and the second-order interaction effects of the pollutant factor concentration on the adsorption of copper, lead, and zinc in a heavy-metals (copper, cadmium, lead, zinc, and nickel) and pesticides (dimethoate, metalaxyl, atrazine, malathion, and prometryn) coexistence system. Further, the effect values and the contribution rates of synergistic and antagonistic effects were calculated, and the competitive capacity of sediments for adsorption of heavy metals was determined. Moreover, the quantitative structure-activity relationship between effect values of heavy metals adsorption and physical and chemical properties of heavy metals was established by 2D-QSAR models to reveal the physical and chemical properties of heavy metals influencing the effect values of heavy metals adsorption in a composite contamination system. The results provided herein will serve as the foundation for further investigation of the transformation of heavy metals and optimal regulation of pollution.

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