PSO-SVM applied to SWASV studies for accurate detection of Cd(II) based on disposable electrode

Zhao Guo^{1,2}, Wang Hui^{1,2}, Yin Yuan^{1,2}, Liu Gang^{1,2*}

 Key Laboratory of Modern Precision Agriculture System Integration Research, Ministry of Education of China, China Agricultural University, Beijing 100083, China;
 Key Laboratory of Agricultural Information Acquisition Technology, Ministry of Agricultural of China, China Agricultural University, Beijing 100083, China)

Abstract: Square wave anodic stripping voltammetry (SWASV) is an effective method for the detection of Cd(II), but the presence of Pb(II) usually has some potential and negative interference on the SWASV detection of Cd(II). In this paper, a novel method was proposed to predict the concentration of Cd(II) in the presence of Pb(II) based on the combination of chemically modified electrode (CME), machine learning algorithms (MLA) and SWASV. A Bi film/ionic liquid/screen-printed electrode (Bi/IL/SPE) was prepared and used for the sensitive detection of trace Cd(II). The parameters affecting the stripping currents were investigated and optimized. The morphologies and electrochemical properties of the modified electrode were characterized by scanning electron microscopy (SEM) and SWASV. The measured SWASV spectrograms obtained at different concentrations were used to build the mathematical models for the prediction of Cd(II), which taking the combined effect of Cd(II) and Pb(II) into consideration on the SWASV detection of Cd(II). The proposed mathematical models rely on an improved particle swarm optimization-support vector machine (PSO-SVM) to assess the concentration of Cd(II) in the presence of Pb(II) in a wide range of concentrations. The experimental results suggest that this method is suitable to fulfill the goal of Cd(II) detection in the presence of Pb(II) (correlation coefficient, mean absolute error and root mean square error were 0.998, 1.63 and 1.68, respectively). Finally, the proposed method was applied to predict the trace Cd(II) in soil samples with satisfactory results.

Keywords: square wave anodic stripping voltammetry (SWASV), particle swarm, support vector machine, screen-printed electrode, heavy metals, Cd detection, soil pollution

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1 Introduction

As one of the most toxic heavy metals in the

environment, cadmium poses a great threat to human health due to its non-biodegradability and persistence^[1]. The main route of exposure to cadmium for humans is through plants by way of soil, directly or indirectly entering the human body^[2]. Thus, developing a rapid, simple, accurate, and reliable method for the on-site detection of cadmium in soil is very desirable^[3-5].

Various techniques including inductively coupled plasma optical emission spectrometry (ICP-OES)^[6], UV-Vis spectroscopy^[7], atomic fluorescence spectroscopy^[8], surface-enhanced Raman spectrometry (SERS)^[9], inductively coupled plasma mass spectrometry (ICP-MS)^[10], atomic absorption spectroscopy (AAS)^[11],

Received date: 2016-09-23 Accepted date: 2017-06-23 Biographies: Zhao Guo, PhD candidate, research interests: Agricultural Electrification and Automation, Email: 1252624780@ qq.com; Wang Hui, PhD candidate, research interests: agricultural electrification and automation, Email: wanghuilunwem@gmail. com; Yin Yuan, Master, research interests: agricultural electrification and automation, Email: 1144729812@qq.com.

^{*}Corresponding author: Liu Gang, Professor, research interests: agricultural electrification and automation, Key Laboratory of Modern Precision Agriculture System Integration Research, Ministry of Education, China Agriculture University, Beijing 100083, China. Tel: +86-10-62736741, Email: pac@cau.edu.cn.

ion chromatography^[12] and colorimetric analysis^[13] have been used for heavy metals analysis. However, these methods have several drawbacks such as time-consuming analysis, high cost, demanding for complex instrumentation and unsuitability for on-site analysis. Among these methods, anodic stripping voltammetry (ASV) is one of the superior electrochemical methods due to its ease of operation, low cost, reliability, inexpensive instrumentation, quick analysis and fast response^[14,15]. However, the presence of other heavy metal ions (HMs) may influence the stripping responses of the target HMs^[4,16-18], which would decrease the detection precision of the direct calibration model.

Chemically modified electrodes (CMEs) represent one of the active research areas in the ASV detection of heavy metals, which can effectively improve the electrochemical properties of electrodes^[19-21]. As a result of the increasing demand for industrial, environmental and food analyses, a variety of screen-printed electrodes (SPEs) have been used for on-site monitoring and point-of-care testing due to their mass production, simple fabrication and low cost^[22-25]. However, the use of nonconductive binders or insulating polymers in the preparation of SPEs may reduce the electronic transmission capability and shelter carbon particles that have electrochemical activity, which will result in the poor kinetics of the heterogeneous reaction^[26]. Ionic liquids (ILs) have many excellent properties such as a good physical and chemical stability, a wide potential window and a high ionic conductivity. The capabilities of ILs for being combined with carbon materials to form conductive composites make them very attractive for electrode modification. In the past decades, many papers have reported the use of CMEs in the ASV detection of $Cd(II)^{[22,27-30]}$. Some of these studies considered the interference of Pb(II) in the SWASV detection of Cd(II); however, most of the reported works only considered the effect of Pb(II) on the SWASV detection of Cd(II) at a certain concentration, instead of various Pb(II) concentrations.

Machine learning method is a subfield of computer science and it is classified as an artificial intelligence method. It can be used in several domains and the advantage of this method is that a model can solve problems which are impossible to be represented by explicit algorithms^[31]. The machine learning models find relations between inputs and outputs even if the representation is impossible; this characteristic allows the use of machine learning models in the case of forecasting problems as described in the previous $report^{[32]}$. Support vector machines (SVMs) are a type of novel machine learning method based on statistical learning theory, which is a new powerful tool for identification and prediction in nonlinear systems^[33]. Particle swarm optimization (PSO) is generally used with SVM for and extraction^[34,35], parameter feature selection optimization of SVM^[36-40] and multi-class classification in SVM^[41]. In this paper, a PSO algorithm was used for the parameter optimization of SVM from training data, i.e., PSO-SVM. Moreover, an improved PSO-SVM model was used to discover, process, and interpret the nonlinear relationships between the stripping peak currents of Cd(II) and Pb(II) and the concentrations of Cd(II) and to create a simple and manageable model for Cd(II) detection. To the best of our knowledge, very few studies have investigated the combination of SWASV and improved PSO-SVM to detect Cd(II) in the presence of Pb(II).

To sum up, in this study, a Bi/IL/SPE was prepared, with N-octylpyridinum hexafluorophosphate (OPFP) as a modifier and binder. Furthermore, an easy but effective method based on the combination of SWASV and improved PSO-SVM was proposed for the optimization of a Bi/IL/SPE sensor to realize the accurate detection of Cd(II) in the presence of Pb(II). The effects of different concentrations of Pb(II) on the stripping voltammetric responses of Cd(II) was also studied to analyze the relationship between the stripping peak currents of Pb(II) and Cd(II) and the concentrations of Cd(II). Consequently, the combination of SWASV and PSO-SVM can be viewed as a source of many potential methods for detecting and quantifying many different types of HMs in various natural samples.

2 Materials and methods

2.1 Reagents and instrumentation

The OPFP IL was purchased from Shanghai Chengjie

Co., Ltd. (Shanghai, China). Graphite powder (spectral purity, size $< 20 \ \mu$ m) and cellulose acetate (CA) were obtained from Sinopharm Chemical Reagent Co., Ltd. (Shanghai, China). Stock solutions of Bi(III), Pb(II) and Cd(II) (1000 mg/L) were purchased from the National Standard Reference Materials Center of China and diluted as required. Acetate buffer (0.1 M) served as the supporting electrolyte for the detection of Cd(II). All other chemicals were of analytical grade and used without further purification. Millipore-Q (18.2 M Ω) water was used for all experiments.

SEM analysis was carried out on a JSM-6701F field emission scanning electron microscope produced by JEOL Ltd. (Japan). SWASV was performed using a CHI660D electrochemical workstation (Shanghai CH Instruments, China). Screen-printed 3-electrode systems, with carbon working electrodes (diameter of 3 mm), carbon counter electrodes and Ag/AgCl reference electrodes, were obtained from the Shanghai Mumei Electronic Technology Co., Ltd. (Shanghai, China). All electrochemical measurements were carried out in a 25 mL cell. A magnetic stirrer was used to stir the test solutions during the deposition steps.

2.2 Electrode preparation and modification

Briefly, 0.1 g of CA, 1.0 g of OPFP and 4.0 g of the graphite powder were added into a mixed solvent containing 5 mL of cyclohexanone and 5 mL of acetone. Then, a viscous ink was produced by sonicating the obtained composite for 12 h. Next, 7 μ L of the viscous ink was cast onto the SPE surface and left to dry at room temperature. Finally, the electrode surface was smoothed by using a piece of weighing paper to obtain a new uniform surface.

2.3 Cd(II) detection by SWASV in the presence of Pb(II)

The SWASV detection of Cd(II) was carried out in 0.1 M acetate buffer solutions (pH 5.0) in the presence of 600 μ g/L Bi(III), which formed a bismuth film by the in-situ deposition. The measurements on the stripping currents of these two HMs (Cd(II) and Pb(II)) in their mixtures were performed in a wide range of concentrations which ranged from 1 μ g/L to 110 μ g/L. A deposition potential of -1.2 V was applied to the

working electrode for 150 s during the deposition steps under stirring. After a 10 s equilibration period, the SWASV potential scan was conducted from -1.2 V to 0.2 V; the solution was not stirred during these two steps. The potential step, square-wave amplitude and frequency were 5 mV, 25 mV and 25 Hz, respectively. Before the next detection, the modified electrode was activated for 150 s at 0.31 V in a pH 5.0 acetate buffer under stirring to remove the residual metals and the bismuth film on the surface of the working electrode. All of the experiments were performed at room temperature.

2.4 SVM optimized by the particle swarm algorithm

SVM is a robust machine learning technique widely used in establishing both linear and non-linear models^[42]. The performance of SVMs in classification and regression depends on several factors, including the kernel function type and its corresponding parameters, the capacity parameter (C) and the kernel function parameter (γ) . First, the kernel function should be identified to determine the sample distribution in the mapping space. The radial basis function (RBF) was used in this study because of its good general performance, and only a few parameters needing adjusted^[43]. The parameter Ccontrols the trade-off between maximizing the margin and minimizing the training error. The γ of the kernel function greatly affects the number of support vectors (SVs), which is closely related with the prediction performance of the SVM and the training time. Too many SVs can result in overfitting and increase the training time. In addition, γ controls the amplitude of the RBF function and hence controls the generalization ability of the SVM. PSO is an efficient evolutionary computational technique that simulates the behavior of birds searching for food and has been widely used in function optimization^[44-46]. Coupled with the PSO algorithm, SVM was developed as a parameter-free modeling technique that enables the construction of a rational and self-adaptive prediction model according to the performance of the total model. In this paper, The PSO-SVM model was developed to predict the non-linear relationship between the Cd(II) concentration and the stripping peak currents of Pb(II) and Cd(II). The inputs of the model were the stripping peak currents of Pb(II)

and Cd(II), and the concentration of Cd(II) was the output of the model. The accurate detection of Cd(II) in the presence of Pb(II) has been attempted based on the combination of the improved PSO-SVM and SWASV with a Bi/IL/SPE. The prediction performance of the improved PSO-SVM was evaluated by mean absolute error (MAE), root mean square error (RMSE) and correlation coefficient (R^2)^[32,47]. All the PSO-SVMrelated calculations and simulation have been carried out by utilizing MATLAB R2012b (The Mathworks, Inc., USA).

2.5 Soil sample preparation

Soil samples were obtained from farmland in China. The soil samples were pretreated to obtain soil extract solutions according to the methods described in previously published papers^[17,48]. Briefly, the soil samples were placed in an oven to dry for 2 h. The soil samples were then pulverized on a portable soil crusher and subsequently sieved through a 200 mm sieve. The soil samples (1 g) were placed in an extraction bottle and extracted with 40 mL of 0.11 M acetic acid. The mixed samples were put into an end-over-end shaker for shaking 16 h at room temperature. The mixed samples were subjected to centrifugal sedimentation experiments for phase separation. The extracts of HMs in aqueous phase were then filtered by a membrane to remove micro-impurities from the solutions. According to the experimental results of "optimization of experimental conditions", the maximum stripping peak current appeared at pH 5.0 on the Bi/IL/SPE based on SWASV (cf. section 3.1). In order to improve the sensitivity of the SWASV detection of Cd(II) with the Bi/IL/SPE, the pH of extracts solution was adjusted to 5.0 by adding 0.11 M NaOH.

3 Results and discussion

3.1 Optimization of experimental conditions

To improve the sensitivity for the SWASV detection of Cd(II) in the presence of Pb(II) with the Bi/IL/SPE, the different SWASV parameters were optimized, including the pH of the supporting electrolyte, Bi(III) concentration, deposition potential and the deposition time.

The effect of pH in the range of 3.5-6 on the stripping

responses of the target HMs was investigated, as shown in Figure 1a. The maximum peak current appeared at pH 5.0; therefore, the following experiments were conducted at pH 5.0. Figure 1b shows the effect of the Bi(III) concentration on the peak currents of Cd(II). The peak currents increased with an increasing Bi(III) concentration from 100 μ g/L to 600 μ g/L and then decreased when the Bi(III) concentration exceeded 600 μ g/L. Thus 600 μ g/L was chosen as the optimal Bi(III) concentration. The effect of the deposition potential on the peak currents of Cd(II) over a potential range of -0.8 V to -1.6 V after accumulation for 150 s was investigated, as shown in Figure 1c. When the deposition potential was in the range of -0.8 V to -1.2 V, the peak current increased with the increase of the deposition potential, which may be due to the further reduction of cadmium ions, but when the deposition potential continued to increase on the basis of -1.2 V, the peak current would gradually decrease, which may be due to the occurrence of hydrogen evolution reaction. For further determinations, the accumulation potential of -1.2 V was chosen for HMs preconcentration. As can be seen in Figure 1d, the peak currents increase with the increase of deposition time. However, as the deposition time exceeded 150 s, the peak currents versus time began to increase slowly. In the present study, considering both the sensitivity and efficiency of the Cd(II) detection, a deposition period of 150 s was used in subsequent experiments.

3.2 Effects of Pb(II) on the SWASV detection of Cd(II) based on Bi/IL/SPE

The surface morphologies of the bare SPE and IL/SPE were characterized by SEM. An SEM image of the bare SPE surface was presented in Figure 2a, showing the disordered distribution of graphite flakes, which undoubtedly affected the conductivity of electrode^[49]. Figure 2b illustrates the surface image of IL/SPE, in which a compact and wrinkled paper-like structure can be observed, indicating the good adherence of OPFP with graphite particle. The IL and carbon particles firmly bind together and form a stable solid composition, which not only improves the mechanical stability but also increases the electron transfer rate of the electrode.



Figure 1 Effects of pH, Bi(III) concentration, deposition potential and deposition time on the stripping peak current of $30 \,\mu$ g/L Cd(II) in the presence of $30 \,\mu$ g/L Pb(II)



Figure 2 SEM images of surface morphologies of bare SPE, IL/SPE

Figure 3 compares the peak currents of 40 μ g/L Cd(II) and Pb(II) on the different modified electrodes. As it is shown, the IL/SPE (curve b) exhibites higher stripping peak currents toward both HMs as compared with bare SPE (curve a), possibly attributed to that the high conductive IL enhanced the electron transfer ability and improved the preconcentration efficiency of HMs on the electrode surface^[50]. The highest striping peaks on the detection of Cd(II) and Pb(II) was observed for Bi/IL/SPE (curve c). Such improvement can be attributed to the fact that Bi can form alloys with HMs, which reduces HMs more easily^[51].

Although the higher stripping voltammetry sensitivity of Cd(II) could be obtained using Bi/IL/SPE as compared with bare SPE in the presence of Pb(II). Interactive interference in the determination of HMs, as a common problem, could reduce the detection accuracy of target HMs (in this case, Cd(II)). In order to realize the accurate detection of Cd(II) in the presence of Pb(II), additional studies were performed on the stripping processes of these two metal ions in their mixtures to investigate the effects of different concentrations of

Pb(II) on the SWASV detection of Cd(II) based on Bi/IL/SPE.





Deposition potential: -1.2 V. Concentration of Bi(III): 600 μ g/L

The SWASV voltammograms containing the stripping currents of Cd(II) and Pb(II) in different concentrations were recorded under the optimized conditions, as shown in Figure 4. The SWASV detection was carried out by changing the concentration of one HM while holding the concentration of the other HM constant. As observed from Figure 4, the linearity of the calibration equations of Cd(II) detection in the presence of different concentrations of Pb(II) were practically unaffected by the increasing Pb(II) concentration, but the calibration equation itself was obviously affected by the different concentrations of Pb(II), which indicates that no calibration equation is universally applicable for the detection of Cd(II) due to the existence of different concentrations of Pb(II) in the real samples.



Figure 4 Voltammograms for Cd(II) ranging from $1.0 \,\mu\text{g/L}$ to $110 \,\mu\text{g/L}$ in different concentrations of Pb(II) ranging from 0 to $110 \,\mu\text{g/L}$ (Deposition time: 150 s. Deposition potential: -1.2 V. Concentration of Bi(III): $600 \,\mu\text{g/L}$)

The obtained results in Figure 5a show that the calibration curve of Cd(II) significantly affected by different concentrations of Pb(II) as compared with the absence of Pb(II). Figure 5b demonstrates the effect of different concentrations of Pb(II) on the stripping response of Cd(II). As observed from Figure 5b, the presence of 20 μ g/L Cd(II) has a significant influence on the stripping response of Cd(II) in the range of 1.0 to 70 μ g/L. When the concentration of Pb(II) was more than 70 μ g/L, the interference of Pb(II) on the stripping

response of Cd(II) still existed and had a more significant effect on the higher concentration of Cd(II). The influence of Pb(II) on the stripping responses of Cd(II) may be due to the close similarity in electrochemical behavior of these two HMs. Furthermore, the presence of Cd(II) may have a negative effect on both the reduction of Cd(II) on the surface of the electrode during the deposition step and the stripping of Cd during the stripping step, because Cd can form multicomponent alloys with Pb and Bi.



Figure 5 Effects of different concentrations of Pb (II) on the fitting curve (a) and stripping peak currents of Cd(II)

The interference of Pb(II) on the stripping responses of Cd(II) would lead to the changes of the calibration equations as compared with the absence of Pb(II), and thereby reduce the prediction accuracy of Cd(II) based on SWASV. These changes of calibration equations are listed in Table 1 which presents the calibration equations of Cd(II) and the Adj. R-Square (R^2) in the presence of different concentrations of Pb(II) as compared with the absence of Pb(II). From Table 1, it can be seen that the calibration equations of Cd(II), including the slope and intercept, were affected by the different concentrations of Pb(II) as compared with the calibration equation of Cd(II) in the absence of Pb(II). Therefore, the presence of Pb(II) will lead to low detection precision of Cd(II) when using the direct calibration model based on SWASV. The presence of different concentrations of Pb(II) leads to a change in the calibration model, which reduced the detection accuracy of Cd(II) to different degrees.

 Table 1
 Linear calibration equations of Cd(II) in the presence of different concentrations of Pb(II)

Pb(II) concentration/ μ g·L ⁻¹	Linear calibration equation (Cd(II))	R^2
0	<i>Y</i> =0.512 <i>X</i> +0.396	0.991
1	Y = 0.456X + 1.17	0.987
5	Y = 0.487X + 1.06	0.999
10	Y = 0.611X + 1.72	0.995
20	Y = 0.713X + 2.34	0.971
40	Y = 0.651X + 2.96	0.971
70	Y = 0.605X + 2.99	0.971
110	Y = 0.746X + 2.14	0.983

3.3 Proposed PSO-SVM model for the detection of Cd(II)

3.3.1 Model training and optimization

SVM was applied to build a non-linear model for the prediction of Cd(II) in the presence of Pb(II), with the stripping peak currents of Pb(II) and Cd(II) as input variables and the concentration of Cd(II) as an output variable. SVM training is an essential step to ensure the prediction accuracy of the model. This step entails the

utilization of a training dataset to train the model. In addition, the testing dataset was used to verify the prediction precision of the proposed SVM model. In this investigation, a searching program named "particle swarm optimization algorithm" was used in the training step to determine the optimized parameter combination of C and γ while preventing the over-fitting of the training dataset. Finally, C and γ were fixed at 64.6741 and 0.38582, respectively.

3.3.2 Establishment and validation of the improved PSO-SVM model

Forty sets of experimental data (training dataset) were used to train the PSO-SVM model using the optimized parameter combination above, and 9 sets of experimental data were used as the testing dataset. The fitness and prediction accuracy of the proposed PSO-SVM model was evaluated by the MAE, RMSE and correlation coefficient^[32,47]. The statistical analysis results show that the prediction accuracy of the proposed model was high because of the reasonable MAE, RMSE and R^2 values for both the training dataset and testing dataset, as shown in Table 2.

 Table 2
 Prediction results of ANN model on the training dataset and testing dataset

Data set	$MAE/\mu g \cdot L^{-1}$	$RMSE/\mu g \cdot L^{-1}$	R^2
Training set	0.516	0.718	0.999
Testing set	1.63	1.68	0.998

In order to further validate the prediction accuracy of the proposed PSO-SVM model, the prediction results of both the well-trained PSO-SVM model and the direct calibration model with respect to the testing dataset were compared with the actual values, as shown in Figure 6a and Figure 6b. As can be seen in the Figure 6c and Figure 6d, the linear regression analysis results indicate that the correlation (R^2 =0.998) between the predicted values from the PSO-SVM model and the actual values is strong as compared with the direct calibration model (R^2 =0.843).



Figure 6 Comparison of prediction results between the (a) direct calibration model and (b) improved PSO-SVM model. Linear regression analysis of the prediction results obtained from the (c) direct calibration model and (d) improved PSO-SVM model The generalization ability of the developed PSO-SVM parameters (MAE, RMSE and R^2) from the PSO-SVM model was evaluated by comparing the statistical and from the direct calibration model, both of which were

used for the prediction of the testing dataset, as shown in Table 3. Under the optimized conditions, the further statistical analysis results indicate that the improved PSO-SVM model has better prediction precision for the concentration of Cd(II) than the direct calibration model (MAE and RMSE are 1.63 μ g/L and 1.68 μ g/L for the improved PSO-SVM model, respectively, and 14.95 μ g/L and 20.01 μ g/L for the direct calibration model, respectively). This means that the presence of Pb(II) would make a negative influence on the accuracy of the direct calibration model for the prediction of Cd(II) concentration.

 Table 3
 Comparison of prediction results of the direct calibration and PSO-SVM models

Prediction method	$MAE/\mu g \cdot L^{-1}$	$RMSE/\mu g \cdot L^{-1}$	R^2
Direct calibration	14.9	20.0	0.843
PSO-SVM	1.63	1.68	0.998

3.4 Application to real sample analysis

To evaluate the applicability of the presented method, the proposed method was applied to the concrete analysis of Cd(II) in real soil samples. The soil samples were prepared according to Section 2.5, and the experiments were performed by the standard addition method. To improve the stripping peak signals of Cd(II) and Pb(II) eliminate the Cu(II) interference in ASV, and ferrocvanide ions were commonly added to the sample solution^[29]. Our experiment results show that 0.1 mM ferrocyanide is the effective masking agent, which could mask copper effectively without having a detrimental effect on the Pb(II) and Cd(II) responses. Table 4 shows the results obtained from the analysis of the real samples by the proposed method. Satisfactory recovery results were obtained with an average recovery of 98.56%, confirming the availability of the developed method.

	Table 4	Results	of Pb(II)	detection in	soil sam	ole extracts
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Sample No.	Added/ $\mu g \cdot L^{-1}$	Found/ $\mu g \cdot L^{-1}$	RSD	Recovery/%
	-	5.56 ^a	0.32	-
1	4	9.41	0.26	96.25
	8	13.45	0.19	98.625
	-	6.82	0.13	-
2	15	21.83	0.29	100.07
	30	36.37	0.14	98.50
	-	4.94	0.28	-
3	45	49.19	0.21	98.33
	90	94.57	0.17	99.59

Note: ^a SWASV measurements were repeated five times (n=5).

4 Conclusions

In this paper, a novel method combining the SWASV, PSO-SVM model and Bi/IL/SPE was proposed. The proposed PSO-SVM model has been used to analyze the SWASV spectrum for the design of Cd(II) concentration estimating mathematical models within the range of concentrations 1-110 μ g/L. Moreover, a disposable Bi/IL/SPE was fabricated and further used for sensitive detection of Cd(II), which exhibited some advantages over traditional electrodes, such as large surface area, high ionic and electronic conductivity and good sensitivity. To the best of our knowledge, this is the first report of the combination of SWASV and an improved PSO-SVM model using Bi/IL/SPE to predict the concentration of Cd(II) in the presence of Pb(II). The statistical analysis results obtained are reasonable for both the training dataset and the testing dataset, which indicates that the improved PSO-SVM model has better prediction precision than the direct calibration model. The feasibility of the proposed method was further confirmed by the real samples analysis with satisfactory recovery results, which holds great promise for its wide application in environmental and food analysis.

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