Development of Ion Character Property Relationship (IC-PR) for Removal of 13-Metal Ions by Employing a Novel Green Adsorbent Aerva javanica

Fozia Batool 1,*, Ali Irfan 2, Sami A. Al-Hussain 3, Eida S. Al-Farraj 3, Shahid Iqbal 4, Jamshed Akbar 1, Sobia Noreen 1, Taslim Akhtar 5, Tunzeel Iqbal 6 and Magdi E. A. Zaki 3,*

1 Department of Chemistry, University of Sargodha, Sargodha 40100, Pakistan
2 Department of Chemistry, Government College University Faisalabad, Faisalabad 38000, Pakistan
3 Department of Chemistry, College of Science, Imam Mohammad Ibn Saud Islamic University (IMSIU), Riyadh 13623, Saudi Arabia
4 Department of Chemistry, University of Education Lahore, Jauharabad Campus, Lahore 41200, Pakistan
5 Govt. Associate College for Women, Mandi Bahauddine 50400, Pakistan
6 The Rawalpindi Women University Rawalpindi, Rawalpindi 46000, Pakistan
* Correspondence: foziabatool109@gmail.com (F.B.); mezaki@imamu.edu.sa (M.E.A.Z.); Tel.: +92-344-4747-4109 (F.B.)

Abstract: The novel Aerva javanica absorbent was applied for the removal of thirteen selected metal ions from a distilled water solution of each metal by the batch adsorption method. The optimization remediation parameters of the metal ions for the batch adsorption approach were developed, which were the initial concentrations (60 ppm), contact time (60 min) and pH (7). The basic properties of metal ion affected the adsorption results; therefore, 21 properties of metal ions were selected, which are called “descriptors”. The most significant descriptors were selected that were vital for the adsorption results, such as covalent index, polarizability and ion charge. The developed model equation by the descriptors provided more than 80% accuracy in the predicted results. Furthermore, Freundlich and Langmuir adsorption models were also applied on the results. Constants of the Freundlich and Langmuir models were also used for model generation, and the results revealed the importance of a covalent index for the removal phenomenon of metal ions. The current study provided a suitable Ion Character Property Relationship (IC-PR) for the removal of metal ions, and future predictions can be achieved on the proposed adsorbent with significant accuracy. The ecofriendly and cost effective Aerva javanica absorbent in the batch experimental model of the current study predicted that this novel absorbent can be used for the removal of a wide spectrum of heavy metal ions from different sources of waste waters.

Keywords: adsorption; aqueous media; green adsorbent; Aerva javanica; covalent index; ion charge; IC-PR; polarizability

1. Introduction

The natural occurring environmental resources are produced, recycled and controlled by nature itself. Man exploited natural sources to fulfill his necessities, but due to the ever-increasing population, man has been compelled to prepare naturally occurring materials by artificial ways [1,2]. Unfortunately, natural resources are being consumed and deleted by the growing population, with forests burnt and cut to clear them away to make factories, mills, industries and housing places [3,4]. When the natural environment is being disturbed, then the amount of natural resources is also intruded on accordingly. In order to bring sustainability to life, the environmental resources must be reserved [5–7]. Heavy and trace metals removal is very significant due to their potency of toxicity for the environment and water bodies’ living creatures and inland living organism. The most severe and most...
threatening issue for today’s world is water pollution, which has attracted the attention of the modern world due to shortages of clean water resources [1,2]. The increasing number of industries and traffic exhausts are the major sources of anthropogenic inputs in fresh water supplies [3,4]. Heavy metals such as Cu, as well as Cd, are cumulative poisonous materials that are accounted to be exceptionally toxic and causes of serious environmental degradation [5–7]. All countries in the world seek natural materials for the removal of toxic heavy metals from aqueous media. Natural substances for the removal of heavy metals are largely accessible, commonly by agricultural wastes. These are low-cost materials that are used for the separation of toxic heavy metals and are also environmentally friendly [8]. In industries, chemical techniques are used for the removal of heavy metals [9–11]. One of the most destructive drawbacks of these methods is that they are not economically favorable. The adsorption process is the best method for the removal of toxic heavy metals, and it is also an economically cost-effective way [12–14]. Different agricultural wastes such as sawdust, tree barks, rice husk, pine bark, animal bones, activated charcoal, etc. are economically low-cost materials that are being used as adsorbents for the separation of the toxic heavy metals. Activated charcoal is largely used as an adsorbent for the removal of toxic heavy metals [15–18]. Nowadays, too much commercial use of activated charcoal increases the price, so the activated charcoal as a pollutant removal is not economically executable [19,20]. In the current study, we explore a new sorbent for the efficient removal of metal ions from aqueous media. *Aerva javanica* is commonly called desert cotton, which is widely distributed and available in Pakistan. It is a soft and fibrous material in its structure, which is the best suitable option for its utilization in the adsorption process. The action and mechanism of biosorption is highly complex and hard to make a model from and imitate using conventional mathematical modeling. The resulting relationship of the sorption process is highly nonlinear due to the interaction of more variables [21,22]. The Ion Character Property Relationship (IC-PR) is an advanced way to relate the adsorption phenomenon of agro-waste with the structure of the sorbate. In IC-PR, different properties of the ions are related with the adsorption process to generate a theoretical model for the future prediction of the adsorption of metal atoms on a proposed adsorbent [23]. Adsorption is a physicochemical process, and it is mainly based on different chemical properties such as polarity, charge of the ion and its covalent index. It is reported that the covalent index strongly affects the metal binding potential over the surface of the adsorbent, since, as the covalent index increases, the attachment power of metal ions over the adsorbent also increases [24]. Similarly, ions with different charges have variable adsorption potential. It is a basic prerequisite to relate these properties of metal ions with the adsorption property to develop a true picture of the metal attachment on the adsorbent surface and study the underlying phenomenon. Therefore, IC-PR is a suitable way to generate a relationship between adsorption phenomenon and properties of the elements [25]. In the current study, we performed the adsorption of 13 selected metal ions on novel sorbent *Aerva javanica* and developed IC-PR between metal ions and the proposed adsorbent. The adsorption study for all these metal ions was performed in the laboratory to obtain the optimized conditions for the adsorption parameters. Based on these experimental results, a model was generated to correlate the adsorption of metal ions with different properties [26]. The developed model can be successfully applied on all metal ions with a high accuracy and robustness.

2. Results and Discussion

2.1. Characterization of Sorbents

Before adsorption, the experiment surface of the adsorbent was analyzed for different chemical moieties present on the surface of the adsorbent, which are responsible for the attachment of metal ions on it. The infrared analysis was performed for said purpose.

2.1.1. Fourier-Transform Infrared Spectroscopy

The surface of the adsorbent was examined using a Fourier-Transform Infrared Spectrometer (Model Shimadzu AIM-8800). Different sorbates adhere to the surface of the
sorbent due to the presence of certain functional groups. The study of these functional groups reveals information on organic functional groups that are involved in adsorption [27–29]. In the range of 4000–450 cm$^{-1}$ wavenumbers, FTIR spectra of the sorbent material were collected; the results are present in Figure 1. It appears that a broad band develops in the 3000–3700 cm$^{-1}$ range, which may be attributable to the $\nu$OH stretching vibration of hydroxyl functional groups, including hydrogen bonding. Hydrogen bonding is indicated by the presence of a broad-banded $\nu$OH group in this frequency range. The hydroxyl group is one of the most important functional groups present on the surface of the adsorbent, as it is responsible for the attachment of sorbate on the surface. The sorbent has a stretching band of $\nu$CH in the 2900–3000 cm$^{-1}$ wavenumber range. The C=O peak appears at 1750 cm$^{-1}$, while the C-O functional group peak appears at 1200 cm$^{-1}$. The $\nu$CN stretching peak can also be seen in spectra at a value of 1049 cm$^{-1}$. At 1645 cm$^{-1}$, vibration caused by secondary amides can be seen. A peak at 1246 cm$^{-1}$ appears due to C-O of the carbonyl functional group. The carbonyl functional group also plays a role in metal ion attachment on the surface [30]. The $\nu$OH group and heteroatoms play an important role in attaching the sorbate to the surface for adsorption [31]. Attachment of metal ions on the surface of the adsorbent occurs mainly due to electrostatic forces developed between metal ions and carbonate functional groups present on the surface of the adsorbent. As far as the complexation attachment of ions to the surface of the adsorbent is concerned, atoms such as nitrogen and oxygen are involved, as they share an electron pair for said purpose. Therefore, $\nu$OH groups as they appear in FTIR spectra are responsible for this attachment [32]. The information obtained through a functional groups analysis helps us to understand possible mechanisms of adsorption. The FTIR spectra of all metal ions were also obtained after adsorption, and changes in position of the functional groups confirm the attachment of metal ions on the adsorbent surface [33].

![Figure 1. FTIR spectra of Aerva javanica.](image)

2.1.2. Batch Adsorption for Parameters Optimization

In this experimental work, different parameters were optimized to get the best adsorption conditions for the remediation of 13 selected metal ions on the adsorbent. For this purpose, the contact time was varied from 30 to 120 min, and the best adsorption was achieved at 60 min, as shown in Figure 2. Initially, by increasing the contact time, a significant rise in adsorption was found for almost all metal ions, but further increases in the contact time have less impact on the parentage adsorption of metals. This factor indicates that the active sites available on the Aerva javanica surface are occupied by the metal ions, and no more vacant places are available for metal attachment [34].
Another important parameter is the pH of the solution while performing adsorption. The pH range was set from 2 to 12, with 1 g of the agro-waste material and 60 min of contact time. The results given in Figure 3 indicated that the adsorption capacity is not significant in the acidic pH range. The adsorption capacity has a positive impact by increasing the pH of the solution. This is because of the fact that, at a low pH, H\(^+\) ions are present in the solution that compete with metals ions for active sites occupation, so the adsorption is low; when the pH was increased, an increase in the adsorption was observed from the 6 to 8 pH range. Further increases produced metal hydroxide, and again, low adsorption capacity was achieved [35].

![Figure 2](image2.png)

**Figure 2.** Effect of contact time (30–120 min) on % adsorption of 13 selected metal ions using Aerva javanica as the sorbent.

![Figure 3](image3.png)

**Figure 3.** Effect of pH on % sorption of metal ions using Aerva javanica as the sorbent (at pH 6, 60 min time and 150 rpm shaking speed).
The initial concentration of sorbate has a dominant affect during the adsorption phenomenon. When the concentration was increased, an increase in the adsorption capacity was observed for almost all metal ions. However, further increase in the concentration shows no significant change in the adsorption, as the active sites are already occupied by metal ions [36]. Although, for some metal ions, an increase in the adsorption capacity was observed, but it was not as significant, and the results are given in Figure 4. Previous studies also reported that the accommodation for sorbates decreases as the concentration is very high due to the unavailability of resident sites [37].

Figure 3. Effect of pH on % sorption of metal ions using *Aerva javanica* as the sorbent (60 ppm each metal concentration, 60 min contact time and 150 rpm shaking speed).

2.1.3. IC-PR of Metal Ions with *Aerva javanica* as Adsorbent

*Aerva javanica* (Desert cotton) was used as an adsorbent to investigate the relationship between distinct metal ion characteristics and their adsorption behavior on the sorbent’s surface. Table 1 summarizes the findings of a regression analysis done by Solver in Microsoft Excel to construct a model equation showing the relationship between each parameter and the percent sorption. The relationship between electronegativity, density, atomic radius/atomic weight, standard reduction potential and covalent index, respectively, was found to be remarkable for X5, X14, X15, X18 and X21. Metal adsorption on the adsorbent requires the use of a covalent index and electronegativity, two key properties. Binding forces are responsible for attaching metals to sorbent surfaces, and these descriptors have a direct impact on them. In the same way, the adsorption potential and density of metals help with their binding. These five descriptors were used to create a model equation that is depicted as Equation (1):

\[
\text{\%Sorption} = 72.37606 + 1.318314 \times 5 + 0.818116 \times 14 + 11.5886 \times 15 - 1.34948 \times 18 + 2.484041 \times 21
\]  

\[(1)\]

This model equation, on the other hand, only achieved 67% adsorption. It was found that using Minitab 17 to run a stepwise multiple regression analysis gave the following results as mentioned in Equation (2):

\[
\text{\%Sorption} = 76.56 - 10.95 \times 17 + 3.54 \times 20 + 4.002 \times 21
\]  

\[(2)\]

\[(N = 13, \text{R-square} = 91.79, \text{predictive } R^2 = 0.8366).\]

A % sorption of 84.34 was achieved with only these three descriptors, with a R-sq value 91.79 and level of significance \(\alpha = 0.05\). The significant descriptors in this equation were...
were the covalent index (X21), polarizability (X20) and ratio of ion charge and square of the ionic radius (X17).

Table 1. Linear regression analysis to show the relationship between % sorption onto Aerva javanica and metal ionic characteristics.

| Sr. No | Model Equation | R  | R²  | R² Adjusted | p-Value | F-Value | Standard Error |
|--------|----------------|----|-----|-------------|---------|---------|----------------|
| 1      | 83.48 + 0.2084 X1 | 0.7082 | 0.5016 | 0.4563 | 0.0067 | 11.0739 | 3.9694 |
| 2      | 84.44 + 0.0571 X2 | 0.6958 | 0.4842 | 0.4373 | 0.0082 | 10.3277 | 4.0308 |
| 3      | 105.94 + 10.80 X3 | 0.6445 | 0.4154 | 0.3623 | 0.0173 | 7.8189 | 4.2990 |
| 4      | 76.52 + 6.79 X4 | 0.6226 | 0.3877 | 0.3320 | 0.0230 | 6.9655 | 4.3999 |
| 5      | 71.86 + 11.81 X5 | 0.6592 | 0.3738 | 0.3146 | 0.0292 | 8.6243 | 4.2490 |
| 6      | 106.87 – 12.79 X6 | 0.6549 | 0.4299 | 0.3771 | 0.0315 | 8.6243 | 4.2490 |
| 7      | 84.17 + 0.0058 X7 | 0.6318 | 0.3991 | 0.3445 | 0.0205 | 7.3080 | 4.3585 |
| 8      | 81.73 + 0.0042 X8 | 0.6929 | 0.4801 | 0.4329 | 0.0086 | 10.1604 | 4.0541 |
| 9      | 93.6887 – 4.9411 X9 | 0.2424 | 0.0588 | 0.0257 | 0.4274 | 0.6872 | 5.4552 |
| 10     | 73.46 + 0.0228 X10 | 0.5871 | 0.3448 | 0.2852 | 0.0348 | 5.7887 | 4.5515 |
| 11     | 93.8452 – 4.6011 X11 | 0.2650 | 0.0702 | 0.0142 | 0.3815 | 0.8309 | 5.4219 |
| 12     | 88.3185 + 0.0320 X12 | 0.2500 | 0.0625 | 0.0224 | 0.4099 | 0.7338 | 5.4443 |
| 13     | 94.06 – 0.278 X13 | 0.4717 | 0.2225 | 0.1518 | 0.1036 | 3.1484 | 4.9580 |
| 14     | 81.49 + 1.250 X14 | 0.8711 | 0.7588 | 0.7369 | 0.0001 | 34.6113 | 2.8714 |
| 15     | 95.90 – 191.7 X15 | 0.8041 | 0.6467 | 0.6146 | 0.0009 | 20.1372 | 3.4321 |
| 16     | 80.09 + 7.01 X16 | 0.6458 | 0.4170 | 0.3640 | 0.0171 | 7.8700 | 4.2931 |
| 17     | 82.57 + 7.19 X17 | 0.6182 | 0.3821 | 0.3260 | 0.0243 | 6.8050 | 4.4197 |
| 18     | 93.1696 + 3.519 X18 | 0.7082 | 0.6372 | 0.6042 | 0.0010 | 19.3254 | 3.8685 |
| 19     | 97.26 – 3.21 X19 | 0.6725 | 0.4257 | 0.3735 | 0.0156 | 8.1566 | 4.2609 |
| 20     | 84.2105 + 1.9073 X20 | 0.5721 | 0.3274 | 0.2662 | 0.0410 | 5.3546 | 4.6115 |
| 21     | 76.95 + 3.762 X21 | 0.8956 | 0.8022 | 0.7842 | 3.4705 | 44.6147 | 2.5007 |

These three descriptors have found a prevailing effect on the adsorption phenomenon. The covalent index is an important descriptor that affects the adsorption rate in terms of the metal ion attachment on the surface of the adsorbent. The covalent index of metal ions is a measure of its capacity to exhibit a covalent interaction with the ligands [38]. The higher the covalent index of a metal ion, the more its adsorption on the surface groups of adsorbents. The importance of this descriptor for IC-PR analysis is described by many other researchers as well [24,39,40]. Similarly, cationic polarizability is one of the most frequent forces expressed in binding events. Since nucleophilic and electrophilic sites available on the adsorbent surface and hydrogen binding are directly affected by the cationic polarizability of ions, metals binding on these active sites decides the significance of adsorption for the particular purpose. Metals with elevated levels of cationic polarizability were found prevailing in adsorption phenomenon as compared with the metals of low polarizability [41,42].

The size of the ions is another factor playing a role in the adsorption of metal ions. The smaller the packing of ions in the form of hard spheres, the better their attraction for the adsorbent site [43–45]. Furthermore, a small size facilitates the adsorption of more ions on the surface of the adsorbent. The ratio of ionic charge and square of the ionic radius were also found as influencing parameters for the present study due to the small sizes of the ions. A probable mechanism of binding is explained by the information present in the IC-PR equation. The underlying mechanism of the binding process can be revealed by careful study of the descriptors.

The parameters for the Langmuir and Freundlich isotherms were also determined by varying the metal ion concentrations, and the results are given in Table 2. The Freundlich adsorption isotherm was found more affective for the present study, with a high value of R-square (R² > 0.9) as compared to Langmuir. The Freundlich adsorption capacity KF was found above 2, which indicates the significant adsorption efficiency of Aerva javanica for all metal ions [46–48]. Therefore, Aerva javanica gives multilayer adsorption of metal ions with a maximum adsorption capacity for lead (10.05) and minimum for potassium ions (2.01).
Table 2. Langmuir and Freundlich isotherm parameters for the adsorption of metal ions onto Aerva javanica.

| Metal Ion | Langmuir Constants | Freundlich Constants |
|-----------|---------------------|----------------------|
|           | Q                   | b                    | R<sub>L</sub> | r² | K<sub>F</sub> | n | r² |
| Ca<sup>2+</sup> | 17.9917 | 0.3418 | 0.0005 | 0.1178 | 3.2411 | 1.0719 | 0.9327 |
| Cr<sup>3+</sup> | 82.0546 | 0.1109 | 0.0001 | 0.3018 | 7.6718 | 1.0177 | 0.9992 |
| Co<sup>2+</sup> | 105.429 | 0.0611 | 0.0019 | 0.9929 | 6.4224 | 0.9988 | 0.9801 |
| Cu<sup>2+</sup> | 5.57382 | 0.2302 | 0.0003 | 0.0633 | 7.7037 | 1.0106 | 0.9717 |
| Cd<sup>2+</sup> | 38.1898 | 0.4853 | 0.0007 | 0.8354 | 2.0179 | 1.1296 | 0.9997 |
| K<sup>+</sup> | 12.2645 | 0.4853 | 0.0007 | 0.8354 | 2.0179 | 1.1296 | 0.9997 |
| Mg<sup>2+</sup> | 16.5423 | 0.4711 | 0.0005 | 0.7363 | 2.9194 | 1.1169 | 0.9969 |
| Mn<sup>2+</sup> | 11.2281 | 0.4882 | 0.0007 | 0.3414 | 2.4666 | 1.0874 | 0.9505 |
| Na<sup>+</sup> | 15.4492 | 0.3542 | 0.0005 | 0.1631 | 3.5473 | 1.0426 | 0.9405 |
| Ni<sup>2+</sup> | 20.0670 | 0.3541 | 0.0004 | 0.1664 | 4.7566 | 1.0397 | 0.9639 |
| Pb<sup>2+</sup> | 178.043 | 0.0415 | 4.6805 | 0.0005 | 10.057 | 1.0004 | 0.8718 |
| Zn<sup>2+</sup> | 55.3924 | 0.0943 | 0.0002 | 0.0339 | 5.2239 | 0.9968 | 0.9803 |
| Fe<sup>2+</sup> | 32.7997 | 0.1861 | 0.0003 | 0.0488 | 8.8613 | 0.9493 | 0.9869 |

2.1.4. Adsorption Capacity of Aerva javanica through Langmuir and Freundlich Isotherm

The adsorption capacity calculated through Langmuir constant Q was used for a regression analysis of metal ions, and the results are present in Table 3. Only a descriptor that gives R-square values above 50% was the covalent index (X21), and the model Equation (3) generated was:

\[
Q = -57.8 + 36.4 \times X_{21}
\]

Poor results for the relationship of the Langmuir constant represent that, in the present study, monolayer adsorption was suppressed by the multilayer adsorption phenomenon [49–51]. A regression analysis through the adsorption capacity generated by the Freundlich isotherm is given in Table 4. Significant descriptors were X21, X14 and X5, respectively, which are in good agreement with the descriptors calculated by % adsorption, as shown in Equation (4):

\[
KF = -0.57 + 1.841 \times X_{21}
\]
Table 4. Linear regression analysis to show the relationship between the adsorption capacity ($K_F$) of *Aerva javanica* obtained from the Freundlich isotherm and metal ionic characteristics.

| Sc. No | Model Equation | $R$ | $R^2$ | $R^2$ Adjusted | $p$-Value | F-Value | Standard Error |
|--------|----------------|-----|------|---------------|-----------|---------|----------------|
| 1      | 2.457681 + 0.106166 X1 | 0.6361 | 0.4046 | 0.3505 | 0.0194 | 7.478 | 2.461 |
| 2      | 2.938278 + 0.039985 X2 | 0.6279 | 0.3943 | 0.3393 | 0.0215 | 7.162 | 2.482 |
| 3      | 14.15095 + 5.66598 X3 | 0.5960 | 0.3552 | 0.2966 | 0.0315 | 6.062 | 2.561 |
| 4      | 0.519302 + 2.625931 X4 | 0.4243 | 0.1800 | 0.1055 | 0.1483 | 2.416 | 2.888 |
| 5      | 14.15095 + 6.404432 X5 | 0.8211 | 0.6742 | 0.6446 | 0.0005 | 22.77 | 1.821 |
| 6      | 14.76476 + 6.8038 X6 | 0.6142 | 0.3772 | 0.3206 | 0.0255 | 6.664 | 2.517 |
| 7      | 3.503003 + 0.002213 X7 | 0.4248 | 0.1803 | 0.1060 | 0.1478 | 2.423 | 2.888 |
| 8      | 1.942227 + 0.001964 X8 | 0.5644 | 0.3185 | 0.2566 | 0.0444 | 5.143 | 2.633 |
| 9      | 6.86076 − 1.55469 X9 | 0.1344 | 0.0180 | −0.0711 | 0.6613 | 0.2062 | 3.161 |
| 10     | −3.14565 + 0.012247 X10 | 0.5595 | 0.3130 | 0.2506 | 0.0467 | 5.013 | 2.644 |
| 11     | 7.645199 − 2.24156 X11 | 0.2275 | 0.0538 | −0.0344 | 0.4545 | 0.609 | 3.106 |
| 12     | 4.628833 + 0.023834 X12 | 0.3276 | 0.1073 | 0.0261 | 0.2745 | 1.323 | 3.014 |
| 13     | 6.926693 − 0.09931 X13 | 0.3926 | 0.1541 | 0.0773 | 0.1844 | 2.006 | 2.934 |
| 14     | 1.390668 + 0.66069 X14 | 0.7930 | 0.6289 | 0.5952 | 0.0012 | 18.65 | 1.943 |
| 15     | 8.643302 − 9.39911 X15 | 0.6989 | 0.4758 | 0.4282 | 0.0090 | 9.866 | 2.309 |
| 16     | 1.203276 + 3.224654 X16 | 0.5236 | 0.2742 | 0.2082 | 0.0662 | 4.156 | 2.718 |
| 17     | 2.108756 + 3.548835 X17 | 0.5378 | 0.2892 | 0.2246 | 0.0579 | 4.477 | 2.689 |
| 18     | 7.417478 + 1.812341 X18 | 0.7253 | 0.5261 | 0.4830 | 0.0050 | 12.21 | 2.196 |
| 19     | 9.541114 − 1.66236 X19 | 0.5951 | 0.3542 | 0.2955 | 0.0318 | 6.034 | 2.563 |
| 20     | 3.163031 + 0.85409 X20 | 0.4516 | 0.2040 | 0.1316 | 0.1212 | 2.819 | 2.846 |
| 21     | −1.33784 + 2.057201 X21 | 0.8634 | 0.7455 | 0.7224 | 0.0002 | 32.23 | 1.609 |

The model equation for the Freundlich adsorption capacity was generated at a significance level 0.05, and the R-square was 79.20%, as depicted in Table 4. The covalent index alone can perform significant adsorption of the metal ions on *Aerva javanica*, as represented later shown in Table 5. The small $p$-value and standard error obtained for a regression analysis of the Freundlich adsorption capacity with metal ions reflects the applicability of this adsorption isotherm on the present study.

2.1.5. Model Validation

In IC-PR development, the real challenge is to develop such a model that can successfully predict the activity of new compounds. Application of the model in this regard is judged by its validation both internal as well as external. A satisfactory internal validation gives permission to the model for applicability on training and test sets, and external validation shows its approach for the prediction of new compounds activity [22,52,53]. For the present model, internal validation was observed by the $Q^2$ value, and for external validation, the predictive $R^2$ was calculated. A high value of $Q^2$ (0.8769) reflects good internal validation, and $R^2$ was 0.8366, showing predictive ability of the model for new molecules.

3. Materials and Methods

3.1. Materials

*Aerva javanica* (Desert cotton) was gathered from various locations in the Sargodha District of Pakistan. Following collection, the sample was thoroughly cleaned with deionized water to eliminate any dust or surface contaminants that had accumulated. In the beginning, the sorbent was dried in an open container at ambient temperatures and then in an electric oven (Model, LEB-1-20) at 105 degrees Celsius for 24 h to completely remove all of the moisture content. The dried sorbent was separated and processed separately, with the proper particle size (10-mesh) being separated by sieves and stored for subsequent examination.

3.2. Chemicals

All of the reagents, solvents and chemicals utilized in this study were of analytical reagent grade, and they were purchased from Sigma-Aldrich (Darmstadt, Germany) or Merck (Darmstadt, Germany). To make working solutions, a series of dilutions with double-distilled water was performed on the standard solutions. Analytical reagent grade metal salts were used for the adsorption experiments as well.
3.3. Characterization of Sorbent

Characterizing the sorbent is required to determine the various physical and chemical characteristics that influence the adsorption.

3.4. Fourier-Transform Infrared Spectroscopy (FTIR)

A Fourier-Transform Infrared Spectrophotometer was used to identify functional groups in the sorbent’s structure (Model Shimadzu AIM-8800). The detection of these functional groups aids in the identification of the nature of the binding interactions between the sorbate and sorbent surface that cause the sorbate to be adsorbed on its surface [28]. It was decided to employ the diffuse reflectance infrared technique (DRIFT) and use KBr as the background reagent.

3.5. Parameter Optimization by Batch Adsorption Experiment

Batch adsorption experiments with sorbate aqueous solutions were used to identify the optimum removal conditions. For this, a standard solution of 100 mL of each metal was used in a 250-mL Erlenmeyer flask with a known sorbent amount, shaking speed and agitation time. The sorbate concentration (30–90 mg/L), pH (3–12) and contact time (40–120 min) were studied as optimizable parameters. After shaking for a particular time, sorbent samples were taken at appropriate intervals (40–120 min) and filtered to remove contaminants. The parameters were set one by one, and the sorption phenomenon was observed in order to determine the parameters that were most appropriate for adsorption [54]. An optimized set of parameters obtained from the first step of the analysis were used for the Ion Character Property Relationship (IC-PR).

3.6. Adsorption Study

To evaluate the removal efficiency of the sorbent, filtered sorbate solutions (as mentioned above) were subjected for analysis. To analyze the residual amount of metal ions in a solution, the relevant solutions were subjected to FAAS (AA 6300, Schimadzu, Japan). The fuel and oxidant used were acetylene and air, respectively, and background correction was performed by a Deuterium Discharge Lamp. The sorption of metal ions was calculated as shown in Equations (5) and (6), respectively [55].

\[
\%\text{Sorption} = \frac{C_i - C_e}{C_i} \times 100 \tag{5}
\]

Equation (5) depicts the linear Langmuir model, where \(C_e\) and \(C_{ads}\) are the equilibrium and adsorbed concentrations, respectively. The adsorption capacity is represented by \(Q\), a constant, while the adsorption energy is represented by \(b\).

\[
\log C_{ads} = \log K_F + \frac{1}{n} \log C_e \tag{6}
\]

\(K_F\) shows the adsorption capacity, while \(1/n\) represents the adsorption intensity in Equation (6), which is the linearized form of the Freundlich isotherm.

3.7. Selection of Descriptors

For IC-PR, it is critical to use descriptors that accurately capture the unique properties of the elements. There were 21 descriptors chosen to connect the adsorption properties with the metal ionic characteristics. As you can see in Table 1, these descriptors and their respective symbols are listed. We first studied the implications of these descriptors for the attachment of metal ions on the surface of the adsorbent and then selected important descriptors to develop the QSPR model. The interaction of elements with other compounds is mostly determined by their physical and chemical properties, which are chosen as descriptors. From these selected descriptors, we select those that only show the maximum interaction for the purpose [56].
3.8. Ion Character Property Relationship (IC-PR)

The ion character property relationship was developed between the adsorption phenomenon and metal properties based on these descriptors. Several software packages, including Minitab 17 and Microsoft Excel 2010, were used in the development of the relationships, as well as the performance of the regression analysis. Using Minitab 17, the correlation coefficient, F-statistics, t-statistics, \(p\)-value and \(R^2\) adjusted were calculated and predicted by putting the level of significance at 0.15. The stepwise selection of the descriptors was carried out in order to identify the best fitting model, and then, the single-step analysis was carried out using the descriptors that were selected. These descriptors were utilized in the development of the regression equation.

3.9. Regression Equation

Stepwise regression was also performed using Microsoft Excel 2010 tools to calculate the standard error and model equation for each descriptor, as depicted in Equation (7):

\[
\%\text{Sorption} = 15.90 + 24.61 \times 6 + 82.9 \times 16 + 4.087 \times 19 + 0.903 \times 21 + 9.523 \times 22
\]

In Table 5, the characteristics of 13 metals ions are given to develop correlation:

| Symbol | Properties                  | Ca\(^{2+}\) | Cr\(^{3+}\) | Co\(^{2+}\) | Cu\(^{2+}\) | Cd\(^{2+}\) | K\(^{+}\) | Mg\(^{2+}\) | Mn\(^{2+}\) | Na\(^{+}\) | Ni\(^{2+}\) | Pb\(^{2+}\) | Zn\(^{2+}\) | Fe\(^{3+}\) |
|--------|-----------------------------|-------------|-------------|-------------|-------------|-------------|---------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|
| X1     | Atomic Number               | 20          | 24          | 27          | 29          | 48          | 19      | 12          | 25          | 11          | 28          | 82          | 30          | 26          |
| X2     | Atomic Weight               | 40.07       | 51.99       | 58.93       | 63.54       | 112.41      | 39.09   | 24.3        | 54.93       | 22.98       | 58.69       | 207.2       | 65.39       | 55.84       |
| X3     | Atomic Radius (Å)           | 1.92        | 1.25        | 1.25        | 1.35        | 1.48        | 2.27    | 1.6         | 1.37        | 1.86        | 1.25        | 1.54        | 1.31        | 1.24        |
| X4     | Oxidation State             | 2           | 3           | 2           | 2           | 2           | 1       | 2           | 2           | 1           | 2           | 2           | 2           | 2           |
| X5     | Electronegativity           | 1           | 1.6         | 1.8         | 1.9         | 1.7         | 0.8     | 1.2         | 1.5         | 0.9         | 1.8         | 1.9         | 1.6         | 1.8         |
| X6     | Covalent Radius (Å)         | 1.74        | 1.18        | 1.16        | 1.17        | 1.48        | 2.03    | 1.36        | 1.17        | 1.54        | 1.15        | 1.17        | 1.25        | 1.17        |
| X7     | Melting Point (℃)           | 839         | 1857        | 1495        | 1083        | 321         | 649     | 1244        | 98          | 1453        | 1083        | 419         | 1555        |             |
| X8     | Boiling Point (℃)           | 1484        | 2672        | 2870        | 2567        | 765         | 760     | 1962        | 883         | 2732        | 2567        | 907         | 2750        |             |
| X9     | Ionic Radius (Å)            | 0.99        | 0.52        | 0.75        | 0.73        | 0.97        | 1.38    | 0.72        | 0.46        | 1.02        | 0.69        | 1.19        | 0.74        | 0.65        |
| X10    | Ionization Energy (kJ/mol)  | 590         | 653         | 757         | 745         | 866         | 418     | 736         | 716         | 489         | 736         | 804         | 904         | 762         |
| X11    | Crystal Radius (Å)          | 1.14        | 0.76        | 0.54        | 0.71        | 0.92        | 1.52    | 0.86        | 0.81        | 1.16        | 0.7         | 1.49        | 0.74        | 0.69        |
| X12    | Electron Affinity (kJ/mol)  | 2.4         | 64.3        | 63.7        | 118.4       | 0           | 48.4    | 0           | 0           | 53          | 112         | 35          | 0           | 15.7        |
| X13    | Atomic Volume (cm\(^3\)/mol)| 25.97       | 7.23        | 6.7         | 7.1         | 13.1        | 45.36   | 0           | 7.39        | 23.68       | 6.6         | 18.27       | 9.2         | 7.1         |
| X14    | Density (g cm\(^{-3}\))    | 1.55        | 7.19        | 8.9         | 8.96        | 8.65        | 0.86    | 1.74        | 7.43        | 0.97        | 8.9         | 11.4        | 7.14        | 7.86        |
| X15    | Atomic Radius/Atomic Weight | 0.018       | 0.024       | 0.021       | 0.021       | 0.013       | 0.058   | 0.065       | 0.025       | 0.08        | 0.02        | 0.007       | 0.02        | 0.022       |
| X16    | Ion Charge                  | 1.04        | 2.4         | 1.6         | 1.48        | 1.35        | 0.44    | 1.25        | 1.45        | 0.53        | 1.6         | 1.29        | 1.52        | 1.61        |
| X17    | Ion Charge/(Atomic Radius)\(^2\) | 0.54    | 1.92        | 1.28        | 1.09        | 0.91        | 0.19    | 0.78        | 1.06        | 0.28        | 1.28        | 0.84        | 1.16        | 1.3         |
| X18    | Standard Reduction Potential (V) | −2.87  | −0.74       | 0           | 0.34        | −0.4        | −2.93   | −2.37       | 0           | −2.71       | −0.25       | −0.13       | −0.76       | −0.44       |
| X19    | (Atomic Radius)\(^2\) (Å)  | 3.66        | 1.56        | 1.56        | 1.82        | 2.19        | 5.15    | 2.56        | 1.87        | 3.45        | 1.56        | 2.37        | 1.71        | 1.53        |
| X20    | Cationic Polarizing Power   | 2.08        | 7.2         | 3.2         | 2.96        | 2.7         | 0.44    | 2.5         | 2.91        | 0.53        | 3.2         | 2.59        | 3.05        | 3.22        |
| X21    | Covalent Index              | 1.92        | 3.2         | 4.05        | 4.87        | 4.27        | 1.45    | 2.3         | 3.08        | 1.5         | 4.05        | 5.55        | 3.35        | 4.01        |

4. Conclusions

The study performed the adsorption of metal ions on low-cost agro-waste from metal-infested water samples. The best-suited adsorption parameters were achieved after the experimental work (60 ppm initial concentration, 60 mint contact time and 7 pH). The results were correlated with the properties of metal ions that are responsible for metal attachment on the surface of an adsorbent. A computer model based on the experimental results was generated to develop the ion characteristics and property relationship. This model not only demonstrates the adsorption of selected metal ions used during experiments but also predicts the adsorption potential of even those metals for which the experiment was not performed. Therefore, based on the generated model, we can predict the adsorption of a broad range of metal ions on selected novel adsorbents. The current study is a fruitful addition for environmental remediation with cost-effective adsorbents and future predictions through model generation.
Author Contributions: Conceptualization, F.B. and A.I.; methodology, F.B. and T.A.; software, T.A., T.I. and F.B.; validation, S.A.A.-H., F.B. and S.N.; formal analysis, F.B. and E.S.A.-F.; investigation, T.I., J.A. and T.A.; resources, E.S.A.-F., S.I. and A.I.; data curation, J.A. and S.A.A.-H.; writing—original draft preparation, F.B. and S.N.; writing—review and editing, M.E.A.Z. and A.I.; visualization, S.N. and A.I.; supervision, S.I.; funding acquisition, S.A.A.-H. and A.I. and project administration, F.B. and M.E.A.Z. All authors have read and agreed to the published version of the manuscript.

Funding: The authors extend their appreciation to the Deanship of Scientific Research at Imam Mohammad Ibn Saud Islamic University for funding this work through Research Group no. RG 21-09-76.

Institutional Review Board Statement: Not applicable.

Informed Consent Statement: Not applicable.

Data Availability Statement: The data are available in the manuscript.

Acknowledgments: The authors acknowledge the technical support provided by Hi-Tech lab, University of Sargodha, regarding the analysis of the samples for the heavy metals reported in this study.

Conflicts of Interest: The authors declare no conflict of interest.

References
1. Muhammad, S.; Usman, Q.A. Heavy metal contamination in water of Indus River and its tributaries, Northern Pakistan: Evaluation for potential risk and source apportionment. Toxic Rev. 2022, 41, 380–388. [CrossRef]
2. Ali, S.; Abbas, Z.; Rizwan, M.; Zaheer, I.E.; Yavaş, İ.; Unay, A.; Abdel-Daim, M.M.; Bin-Jumah, M.; Hasanuzzaman, M.; Kald eris, D. Application of floating aquatic plants in phytoremediation of heavy metals polluted water: A review. Sustainability 2020, 12, 1927. [CrossRef]
3. Bilos, C.; Colombo, J.C.; Skorupka, C.N.; Presa, M.J.R. Sources, distribution and variability of airborne trace metals in La Plata City area, Argentina. Environ. Pollut. 2001, 111, 149–158. [CrossRef]
4. Koch, M.; Rotard, W. On the contribution of background sources to the heavy metal content of municipal sewage sludge. Water Sci. Technol. 2001, 43, 67–74. [CrossRef] [PubMed]
5. Yargholi, B.; Azimi, A.A.; Baghvan d, A.; Liaghat, A.M.; Fardi, G.A. Investigation of cadmium absorption and accumulation in different parts of some vegetables. Am.-Eurasian J. Agric. Environ. Sci. 2008, 3, 357–364.
6. Al Naggar, Y.; Khalil, M.S.; Ghorab, M.A. Environmental pollution by heavy metals in the aquatic ecosystems of Egypt. Open Acc. J. Toxicol. 2018, 3, 555603.
7. Maleva, M.G.; Nekrasova, G.F.; Bezel, V.S. The response of hydrophytes to environmental pollution with heavy metals. Russ. J. Ecol. 2004, 35, 230–235. [CrossRef]
8. Alkhanni, A.; Madrakian, T.; Amini, A.; Karimi, Z. Effect of the impregnation of carbon cloth with ethylenediaminetetraacetic acid on its adsorption capacity for the adsorption of several metal ions. J. Hazard. Mater. 2008, 150, 408–412. [CrossRef]
9. Ku, Y.; Jung, I.-L. Photocatalytic reduction of Cr(VI) in aqueous solutions by UV irradiation with the presence of titanium dioxide. Water Res. 2001, 35, 135–142. [CrossRef]
10. Saravanan, A.; Kumar, P.S.; Jeevanantham, S.; Karishma, S.; Taj sabreen, B.; Yaashikaa, P.R.; Reshma, B. Effective water/wastewater treatment methodologies for toxic pollutants removal: Processes and applications towards sustainable development. Chemosphere 2021, 280, 130595. [CrossRef]
11. Katheresan, V.; Kansedo, J.; Lau, S.Y. Efficiency of various recent wastewater dye removal methods: A review. J. Environ. Chem. Eng. 2018, 6, 4676–4697. [CrossRef]
12. Subramani, A.; Jacangelo, J.G. Emerging desalination technologies for water treatment: A critical review. Water Res. 2015, 75, 164–187. [CrossRef] [PubMed]
13. Hao, X.; Hua, G.; Jiangtao, F.; Dan, W.; Zhengwei, L.; Yu, W.; Wei, Y. Electrochemical oxidation combined with adsorption: A novel route for low concentration organic wastewater treatment. Int. J. Electrochem. Sci. 2019, 14, 8110–8120. [CrossRef]
14. Abd El-Lateef, H.M.; Ali, M.M.K.; Saleh, M.M. Adsorption and removal of cationic and anionic surfactants using zero-valent iron nanoparticles. J. Mol. Liq. 2018, 268, 497–505. [CrossRef]
15. Okpara, O.G.; Ogbeide, O.M.; Ike, O.C.; Menechukw, K.C.; Eijke, E.C. Optimum isotherm by linear and nonlinear regression methods for lead (II) ions adsorption from aqueous solutions using synthesized coconut shell–activated carbon (SCSAC). Toxicin Rev. 2021, 40, 901–914. [CrossRef]
16. Jusoh, A.; Shiung, L.S.; Ali, N.; Noor, M. A simulation study of the removal efficiency of granular activated carbon on cadmium and lead. Desalination 2007, 206, 9–16. [CrossRef]
17. Kang, G.-D.; Cao, Y.-M. Development of antifouling reverse osmosis membranes for water treatment: A review. Water Res. 2012, 46, 584–600. [CrossRef]
18. Er ses, A.S.; Fazal, M.A.; Onay, T.T.; Craig, W.H. Determination of solid waste sorption capacity for selected heavy metals in landfills. J. Hazard. Mater. 2005, 121, 223–232.
19. Abdul, A.; Aberuagba, F. Comparative study of the adsorption of phosphate by activated charcoal from corncobs, groundnut shells and rice-husks. *Aust. J. Technol.* 2005, 9, 59–63.

20. Park, H.G.; Kim, T.W.; Chae, M.Y.; Yoo, I.-K. Activated carbon-containing alginate adsorbent for the simultaneous removal of heavy metals and toxic organics. *Process Biochem.* 2007, 42, 1371–1377. [CrossRef]

21. Prakash, N.; Manikandan, S.A.; Govindarajan, L.; Vijayagopal, V. Prediction of biosorption efficiency for the removal of copper(II) using artificial neural networks. *J. Hazard. Mater.* 2008, 152, 1268–1275. [CrossRef][PubMed]

22. Trospa, A.; Gramatica, P.; Gombar, V.K. The importance of being earnest: Validation is the absolute essential for successful application and interpretation of QSIPR models. *QSAR Comb. Sci.* 2003, 22, 69–77. [CrossRef]

23. Le, T.; Epa, V.C.; Burden, F.R.; Winkler, D.A. Quantitative Structure–Property Relationship Modeling of Diverse Materials Properties. *Chem. Rev.* 2012, 112, 2889–2919. [CrossRef][PubMed]

24. Wang, Z.; Xu, W.; Jie, F.; Zhao, Z.; Zhou, K.; Liu, H. The selective adsorption performance and mechanism of multiwall magnetic carbon nanotubes for heavy metals in wastewater. *Sci. Rep.* 2021, 11, 16878. [CrossRef][PubMed]

25. Rao, B.H.; Asolekar, S.R. QSAR models to predict effect of ionic strength on sorption of chlorinated benzenes and phenols at sediment–water interface. *Water Res.* 2001, 35, 3391–3401. [CrossRef]

26. Can, C.; Jianlong, W.J.C. Correlating metal ionic characteristics with biosorption capacity using QSAR model. *Chemosphere* 2007, 69, 1610–1616. [CrossRef]

27. Belskaya, O.B.; Lavrenov, A.V.; Danilova, I.G.; Kazakov, M.O.; Mironenko, R.M.; Likholobov, V.A. FTIR Spectroscopy of Adsorbed Probe Molecules for Analyzing the Surface Properties of Supported Pt (Pd) Catalysts; IntechOpen: Licko-Senjska, Croatia, 2012.

28. Salame, I.; Bandosz, T.J. Surface chemistry of activated carbons: Combining the results of temperature-programmed desorption, Boehm, and potentiometric titrations. *J. Colloid Interface Sci.* 2001, 240, 252–258. [CrossRef]

29. Darweesh, H.H.M.; El-Suoud, M.R.A. Setting, hardening and mechanical properties of some cement/agrowaste composites-Part 4. *Molecules* 2012, 17, 16802–16811. [CrossRef][PubMed]

30. Perry, J.J.; Teich-McGoldrick, S.L.; Meek, S.T.; Greathouse, J.A.; Haranczyk, M.; Allendorf, M.D. Noble gas adsorption in graphite. *J. Environ. Sci. Health Part B* 2018, 53, 276–281. [CrossRef]

31. Fletcher, A.J.; Uygur, Y.; Thomas, K.M. Role of surface functional groups in the adsorption kinetics of water vapor on microporous activated carbons. *J. Phys. Chem. C* 2007, 111, 8349–8359. [CrossRef]

32. Alzahri, M. Adsorption of heavy metals and toxic organics. *Process Biochem.* 2007, 42, 1371–1377. [CrossRef][PubMed]

33. Verhoef, R.W.; Asscher, M. The work function of adsorbed alkalis on metals revisited: A coverage-dependent polarizability approach. *Surf. Sci.* 1997, 391, 11–18. [CrossRef]

34. Shastri, A.; Aminabhavi, T.M. Adsorption of heavy metal ions from aqueous solutions using natural materials: Adsorption and characterisation studies. *J. Environ. Eng.* 2015, 32, 342–349. [CrossRef]

35. Tropsha, A.; Gramatica, P.; Gombar, V.K. The importance of being earnest: Validation is the absolute essential for successful application and interpretation of QSIPR models. *QSAR Comb. Sci.* 2003, 22, 69–77. [CrossRef]

36. Ishola, T.K.; Olanipekun, T.O.; Bolarinwa, T.O.; Oladeji, D.R.; Abubakar, A. Complex Equilibria and Distribution of Metal (II) Ions in Sediment–Water Interface. *Chem. Sci. Int. J.* 2012, 30, 8213–8223. [CrossRef]

37. LeVan, M.D.; Carta, G.; Yon, C.M. Adsorption and ion exchange. *J. Environ. Sci. Health Part B* 2018, 53, 276–281. [CrossRef]

38. Stafiej, A.; Pyrzynska, K. Adsorption of heavy metal ions with carbon nanotubes. *Sep. Purif. Technol.* 2007, 58, 49–52. [CrossRef]
47. Ghorbel-Abid, I.; Trabelsi-Ayadi, M. Competitive adsorption of heavy metals on local landfill clay. Arab. J. Chem. 2015, 8, 25–31. [CrossRef]

48. Motsi, T.; Rowson, N.A.; Simmons, M.J.H. Adsorption of heavy metals from acid mine drainage by natural zeolite. Int. J. Miner. Process. 2009, 92, 42–48. [CrossRef]

49. Thirumavalavan, M.; Lai, Y.-L.; Lin, L.-C.; Lee, J.-F. Cellulose-based native and surface modified fruit peels for the adsorption of heavy metal ions from aqueous solution: Langmuir adsorption isotherms. J. Chem. Eng. Data 2010, 55, 1186–1192. [CrossRef]

50. Altun, O.; Özbekle, H.O.; Doğu, T. Use of general purpose adsorption isotherms for heavy metal–clay mineral interactions. J. Colloid Interface Sci. 1998, 198, 130–140. [CrossRef]

51. Chiban, M.; Soudani, A.; Sinan, F.; Persin, M. Single, binary and multi-component adsorption of some anions and heavy metals on environmentally friendly Carpobrotus edulis plant. Colloids Surf. B Biointerfaces 2011, 82, 267–276. [CrossRef]

52. Roy, K.; Mitra, I.; Kar, S.; Ojha, P.K.; Das, R.N.; Kabir, H. Comparative studies on some metrics for external validation of QSPR models. J. Chem. Inf. Model. 2012, 52, 396–408. [CrossRef]

53. Solov’Ev, V.P.; Kireeva, N.; Tsivadze, A.Y.; Varnek, A. QSPR ensemble modelling of alkaline-earth metal complexation. J. Incl. Phenom. Macrocycl. Chem. 2013, 76, 159–171. [CrossRef]

54. Gayathri, R.; Gopinath, K.P.; Kumar, P.S. Adsorptive separation of toxic metals from aquatic environment using agro waste biochar: Application in electroplating industrial wastewater. Chemosphere 2021, 262, 128031. [CrossRef] [PubMed]

55. Mittal, A.; Kurup, L.; Mittal, J. Freundlich and Langmuir adsorption isotherms and kinetics for the removal of Tartrazine from aqueous solutions using hen feathers. J. Hazard. Mater. 2007, 146, 243–248. [CrossRef]

56. Kovalishyn, V.; Abramenko, N.; Kopernyk, I.; Charochkina, L.; Metelytsia, L.; Tetko, I.V.; Peijnenburg, W.; Kustov, L. Modelling the toxicity of a large set of metal and metal oxide nanoparticles using the OChem platform. Food Chem. Toxicol. 2018, 112, 507–517. [CrossRef] [PubMed]

57. Wolterbeek, H.T.; Verburg, T.G. Predicting metal toxicity revisited: General properties vs. specific effects. Sci. Total Environ. 2001, 279, 87–115. [CrossRef]

58. McCloskey, J.T.; Newman, M.C.; Clark, S.B.J.E.T.; Journal, C.A.I. Predicting the relative toxicity of metal ions using ion characteristics: Microtox® bioluminescence assay. Environ. Toxicol. Chem. Int. J. 1996, 15, 1730–1737. [CrossRef]

59. Miranda, L.S.; Ayoko, G.A.; Egodawatta, P.; Goonetilleke, A. Adsorption-desorption behavior of heavy metals in aquatic environments: Influence of sediment, water and metal ionic properties. J. Hazard. Mater. 2022, 421, 126743. [CrossRef]

60. Zamil, S.S.; Ahmad, S.; Choi, M.H.; Park, J.Y.; Yoon, S.C. Correlating metal ionic characteristics with biosorption capacity of Staphylococcus saprophyticus BMSZ711 using QICAR model. Bioresour. Technol. 2009, 100, 1895–1902. [CrossRef]