Electrical Resistivity Tomography as a Support Tool for Physicochemical Properties Assessment of Near-Surface Waste Materials in a Mining Tailing Pond (El Gorguel, SE Spain)

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Abstract: The legacy of the mining industry has left a large number of tailing ponds in the Cartagena–La Unión mining district exposed to water and wind erosion, which causes serious environmental and health problems and requires remediation. Before applying any remediation technique, an intensive sampling of the materials infilling the pond is required to determine the geochemistry of the pond, which will condition the remediation process. However, sampling the large number of tailing ponds that compose the district could be expensive. Thus, the main objective of this study is to evaluate the usefulness of electrical resistivity tomography (ERT) as a non-invasive tool to provide an image of spatial subsurface resistivity distribution and its relation to the physicochemical composition of near-surface mine wastes. To achieve this objective, three short ERT profiles were conducted, and 12 samples in each profile were collected at different depths for its geochemical characterization. Several non-linear regression models were fitted to predict physicochemical properties and metal concentrations from electrical resistivity measures. As a result, a high resistivity area was depicted in the ERT profiles G2 and G3, while the low resistivity ERT profile G1 was also obtained in accordance with the site’s surficial characteristics. Relationships among low resistivity values and high salinity, clay content, high metal concentrations, and mobility were established. Specifically, calibrated models were obtained for electrical conductivity, particle sizes of 0.02–50 µm and 50–2000 µm, total Zn and Cd concentration, and bioavailable Ni, Cd, and Fe. The ERT technique was shown to be a useful tool for the approximation of the location and distribution of the highest ranges of fine particle sizes, moisture, and, to a lesser extent, metal accumulation in the near-surface waste materials.

Keywords: ERT method; heavy metal; regression model; remediation; tailing pond

1. Introduction

The mining of metalliferous mineral deposits and associated smelting and metal processing activities have led to perturbations in surface environments [1]. Indeed, metal-based ore processing generates a large amount of mine tailings, which are watery sludges composed of medium- to
fine-size-grain material [2]. Due to natural leaching from these reactive materials, contamination of the environment close to the deposits often occurs [3]. In the Cartagena–La Unión mining district (South-East, Spain), the legacy of more than 2500 years of intensive mining activity has left a large number of tailing ponds that generate environmental impacts due to leachates with high concentrations of sulfides and metallic elements [1,4–7]. Their wastes are characterized by significant amounts of metals such as Pb, Zn, Cu, and Cd, high salinity, acidic pH levels, and low soil organic matter content [8]. Therefore, mining pond characteristics make the natural establishment of vegetation difficult. In fact, tailing ponds are waste storage structures with bare surfaces completely exposed to erosion agents that promote the transport of those contaminants to the surrounding areas [9]. It has been demonstrated that plant cover provides physical protection to minimize wind erosion and the surface runoff of soil materials [10].

To appropriately implement phytostabilization techniques, a precise and representative physicochemical characterization of the tailing ponds, required to collect a high, representative number of samples from these ponds, is required, not only from the surface but also underneath by means of core sampling acquisition tools or mechanical drilling rigs [6]. However, to accomplish this adequate physicochemical characterization through conventional sampling techniques such as boreholes, auger drilling, and superficial sampling grids, a high budget and time-consuming work are required [11,12]. Therefore, it would be interesting to minimize those drawbacks with the assistance of geophysical techniques, especially electrical resistivity tomography (ERT), since there have been revolutionary improvements in the resistivity method in recent years, and two-dimensional (2-D) surveys are now routinely conducted in many study fields [13].

The electrical resistivity method is one of the oldest and most commonly used geophysical exploration methods [14]. Indeed, electrical resistivity tomography is a non-invasive technique that enables the gathering of a large number of resistivity data in a short period of time, and it has been widely used in environmental and engineering [15–19], hydrological [20,21], archaeological [22,23], and mineral exploration [24–26] surveys. In addition, it has been used for image structures from the millimeter to kilometer scale [27,28].

Most of the studies carried out on tailing ponds using ERT have been focused on identifying the tailings’ internal structure, defining the morphology of the tailing–bedrock contact, highlighting the occurrence of faults or cracks that could put the pond stability at risk, and pinpointing preferential pathways for heavy metal transport or acid mine drainage [6,7,11,12,29,30]. However, the use of ERT to establish relationships among electrical resistivity data and physicochemical variables such as moisture, heavy metal concentration, pH, particle size distribution, and Al/Si as a proxy for clay content in tailing ponds or soils severely polluted by mining has not been widely studied [6,31].

In this study, we attempt to establish associations between the physicochemical properties of tailings and electrical resistivity data. We propose establishing regression models based on these relationships as a proxy to predict the areas inside the pond, and these could show the highest water retention, the finest particles, and a presence of metals, avoiding huge sampling campaigns. Regression models based on electrical resistivity data have been widely used in complex sedimentary environments to estimate water flows and soil moisture content, usually in crop soils and polluted floodplains [19,32–35]. However, the use of regression to range soil properties and metal concentrations using electrical resistivity as a reference variable is quite new [19,35,36].

Therefore, the primary objective of this study is to evaluate the usefulness of ERT as a non-invasive method to derive variations in the physicochemical properties of the most surficial layers of mine tailing ponds, thus enabling more efficient phytoremediation action, representing environment-related work on former mining sites that have been severely affected.
2. Material and Methods

2.1. Study Area

The Cartagena–La Unión mining district is located in the southeast of the Murcia Province (Spain) (Figure 1). Considered one of the most ancient mining areas of Europe, it developed important industrial activity since the Roman period to the first half of the 20th century, which has caused landscape and environmental damage in the area, especially due to the accumulation of mine waste in ponds and acid mine drainages. More than 80 tailing ponds have been catalogued in the area, and a considerable number of them are characterized by high metal concentrations and similar mineral composition: quartz, magnetite, jarosite, goethite, clinochlore, and gypsum. Primary minerals in natural soils of the mining district are dolomite, kaolinite, hematite, calcite, muscovite, and anorthoclase [37]. Nevertheless, metal content and its disposal are specific to each one.

Figure 1. Location of the study area and detailed view of the electrical resistivity tomography (ERT) profile G1 (a), G2 (b), and G3 (c) and their position on the tailing pond (d).

The climate of the area is semiarid Mediterranean, characterized by an annual average temperature of 18 °C and an annual precipitation of 290 mm [38]. This study is focused on the El Gorguel tailing pond (37°35′ N, 0°52′ W), which presents an area of 7.4 km², a 14 m depth, and a 150,000 m³ volume [4]. It was considered representative of the acidic type of tailing ponds located in the Cartagena–La Unión mining district because it had a similar range of salinity, a lack of vegetation and metal content, low organic matter, and similar mineralogy, and was more affected by water and wind erosion than other tailing ponds in the area [4,39]. In September 2011, a remediation project was carried out, amending the mining wastes with marble sludge and pig slurry in a rate of 4 kg/m² and 3 L/m², respectively (European Commission Framework Project FP7 IRIS) [40].
The studied tailing pond, like other tailing ponds in this mining district, was formed in the first half of the 20th century as a result of the disposal of mine waste generated from mineral processing of Fe-, Pb- and Zn-based sulfide ores. Ponds were formed by the gravitational accumulation of waste materials discharged from wooden sluices with multiple outlets (Figure 2). Water in the tailings was drained through either a decant outlet situated in the center or drainage pipes installed on the perimeter of the pond. Due to gravity, the configuration of the ponds led to the settling of heavy and coarse particles close to the dam, while light and fine materials were concentrated near the water disposal points. As in many mining areas in the world, when a tailing pond was full, another dam was constructed upstream, leading to a series of dams (Figure 2) [41,42].

![Image](image-url)

**Figure 2.** Tailing pond from staged embankment. Tailing segregation zones and soil characteristic as a function of distance from the spilling location [41].

### 2.2. Electrical Resistivity Tomography (ERT)

Electrical resistivity tomography (ERT) was employed as a primary geophysical technique. ERT is suitable for investigating complex subsurface structures with vertical and lateral resistivity changes [18], whose resistivity values, depending on the particular site conditions, are influenced by the variation of important properties of the ground such as moisture, soil salinization, and particle size distribution, of which moisture is considered by far the most influential [43].

Three short ERT profiles were laid out on the tailing pond to accurately characterize the tailing pond surface to a depth of about 1 m, which is the most crucial layer in remediation or phytostabilization work (Figure 1). Figure 3 depicts the El Gorguel tailing pond, where three ERT profiles are located across the tailing pond and a fourth ERT profile following a lengthwise orientation were conducted as part of a European Commission Framework Project [40]. These ERT profiles identified two different surficial tailing areas according to their electrical resistivity values. Thus, the western side of the tailing pond was characterized by low electrical resistivity values (<8 Ω m), and the eastern side of the tailing pond, close the pond embankment, was characterized by higher electrical resistivity values (>8 Ω m). Therefore, the orientation and position of the small ERT profiles of this study were chosen in order to locate them in these two different regions.
Apparent resistivity measurements were obtained by a Syscal R1 resistivity meter (IRIS Instruments, Orleans, France) with 36 stainless steel electrodes laid out in a transect at constant intervals, with 30 cm electrode spacing and a total length of 10.5 m. We were interested in reaching an investigation horizon of about 1 m depth, since most phytoremediation studies carried out on tailing ponds of this area demonstrated that the main physicochemical variations occurred at that thickness [37,39,40]. In order to combine a good penetration depth, a reasonable vertical and horizontal resolution, the dipole–dipole array was chosen for this study [44,45]. This measurement array has been successfully used in similar studies [17]. The resistivity meter used an injection cycle of 1 s with an output voltage of up to 400 V and rejection filters for 50 and 60 Hz. Noise reduction was obtained by applying continuous stacking, selectable from a minimum value of three to a maximum of six stacks. With these parameters, an average value (and its associated level of uncertainty) of the apparent resistivity was obtained for each quadripole. After the acquisition of the apparent resistivity values, the data were processed to remove bad data. This was done by filtering raw data with low signal values (V/I < 10^6 Ω) or repeatability errors greater than 2%. From a theoretical point of view, the depth of investigation of a measurement depends on the length of the transmitting line AB (current injection electrodes) and on the separation between the transmitting AB line and the receiving MN (potential measurement electrodes) line [46]. Thus, the maximum depth reached for dipole–dipole arrays is estimated to be 0.2·L, L being the total length of the cable [46]. In our study, L was 10.8 m, so the maximum depth was 2.16 m, which is enough to reach resistivity data levels of up to a 1.2–1.5 m depth.

The apparent resistivity datasets from each ERT profile were processed in two stages: (1) initial filtering and removing of the noisy data by means of PROSYS II software (V3, IRIS Instruments,
Orleans, France); (2) data processing through RES2DINV software (V3.57, Geotomo Software, Penang, Malaysia) [47,48]. The code RES2DINV uses the L1 and L2 norms for the data inversion-based processing [13,49]. The method uses a finite element scheme for solving the 2-D forward problem and the blocky inversion method for inverting the ERT data [13,48,49]. The 2D electrical section finally obtained was constituted by true, spatially distributed electrical resistivity data, which are related to the variation in the physicochemical properties of the tailing pond [50].

2.3. Sampling Collection and Geochemical Analysis

Twelve soil samples were collected from each ERT profile (G1, G2, and G3)—four samples at 0–30 cm, four samples at 30–60 cm, and four samples at 60–90 cm—in the El Gorguel tailing pond, for a total number of 36 soil samples, collected using a manual core and packed in plastic bags to be returned to the laboratory.

Samples were dried in a forced-air oven at 45 °C for 72 h and sieved through a 2-mm mesh for physicochemical analysis. A subsample of the sieved sample was ground with an agate mortar (RetschRM 100).

pH was measured in a solution of 1:2.5 soil/deionized water ratio [51], and soil electrical conductivity (S-EC) was measured in a 1:5 soil/deionized water suspension [51], both using selective electrodes. Organic carbon was determined by the dichromate method [51], and calcium carbonate was determined by the volumetric method of Bernard’s calcimeter. Particle size analysis was carried using the Bouyoucos method [52].

The pseudo-total metal concentration was determined by acid digestion of 1 g of ground sample, with 10 mL of nitric acid (HNO₃) and 10 mL of perchloric acid (HClO₄), heated at 210 °C for 90 min in a heating block equipped with a temperature sensor and a time control panel to ensure digestion. To minimize loss by evaporation or volatilization, a recovery system for gases was used in the procedure. After cooling, the samples were passed through a 0.2-µm filter for quantitative analysis, and 0.1 N HCl was added to volume in a 100 mL volumetric flask [53]. The total amounts of metals (Zn, Pb, Cu, Cd, Cr, Ni, Fe, and As) were determined by ICP-OES/MS (Agilent 7500CE, Agilent, Santa Clara, CA, USA).

Bioavailable metals were obtained using the Lindsay and Norvell’s (1978) [54] and Crock and Severson’s (1980) [55] method of adding diethylenetriaminepentaacetic acid (DTPA) in a 1:2 soil/DTPA ratio. After shaking, samples were filtered and measured by ICP-OES/MS (Agilent 7500CE). Water-soluble metals were obtained according to Buurman et al. (1996) [56] and Frau’s (2000) [57] method of using a 1:5 soil/deionized water ratio. After shaking and filtering, extracts were measured by ICP-OES/MS (Agilent 7500CE). For quality assurance, reagent blanks, as well as certified reference materials (BAM-U110), available from the Federal Institute for Materials Research and Testing Research, were also run with the samples. Recoveries obtained for BAM-U110 were 87% for Pb, 95% for Cu, 95% for Ni, 86% for Cr, 89% for Zn, 97% for Cd, and 81% for As.

2.4. Statistical Analysis

The Kolmogorov–Smirnov normality test at P < 0.05 was used to ensure a normal distribution of the data. This normality assumption was rejected even after log-transformation. As a consequence, non-parametric methods were applied to the data. The Mann–Whitney U test at P < 0.05 was performed to assess significant differences among profiles for each variable. Spearman correlation, denoted by rₛ, was carried out to establish relationships between electrical resistivity data and waste physicochemical properties. The significance of the correlation coefficient was tested at 0.05 level. These statistical analyses were performed with the software SPSS Statistics v.23 (IBM, USA).

In this study, a simple nonlinear regression analysis was carried out to obtain a model that allowed for the prediction of waste properties and metal content in the amended tailing ponds based on electrical resistivity data. The selected-response variable (Y) was the different waste properties or metal concentration, while the waste resistivity was used as an explanatory variable for prediction (X). Boxplot graphs were used to determine outlier data that were excluded from the test. Seventy-five
percent of the total data were randomly selected and used as the calibration set, while the remaining 25% was used as the validation set. Residuals from the calibrated models were satisfactorily checked for the model assumptions of normality (Kolmogorov–Smirnov test), linearity, and homoscedasticity, and fitted models with $R^2 < 0.6$ were rejected. To validate the models, residuals from the estimated variables of the validation set should be within the confidence interval (CI) (at 95%) of the residual distribution of the calibrated model. CI was calculated as ±1.96 times the standard deviation of the residuals in the calibrated model. Modeling was performed with the software SigmaPlot v.12 (Systat Software, Inc., San Jose, CA, USA).

3. Results and Discussion

3.1. Soil Properties, Metal Concentration, and ERT Imaging

The pH in the tailing pond ranged from slightly acidic to moderately alkaline [37], because in a previous rehabilitation project, in order to neutralize the potential acid generation, experimental plots were built, and marble waste (98% carbonates) was applied to the entire pond surface [40,58]. Salinity (S-EC) ranged from very slightly saline to slightly saline [37] due to the coprecipitation of metallic salts with carbonates from the amendments [57]. Inorganic carbon, organic carbon, and nitrogen showed low contents (Figure 4), as were expected in these mining wastes. Coarse-size (50–2000 µm) and medium-size (0.02–50 µm) particles were dominant in the mining waste particle size composition, ranging from 59% to 85% and from 8% to 43%, respectively, while the fine fraction (<2 µm) represents only 4–10% of the total particles.

All metal analyzed exceeded the limit established as a reference of non-polluted soils (NGR) in the Region of Murcia [59], except Cr and Ni, which showed values close to the background level of the area. Therefore, total Zn concentration (mean of the whole pond: 8263 mg kg$^{-1}$) was 92 times higher than the NGR (90 mg kg$^{-1}$), while Pb (3394 mg kg$^{-1}$) and Cd (123 mg kg$^{-1}$) showed values 60 and 50 times higher than their reference values (57 and 0.5 mg kg$^{-1}$, respectively); As (350 mg kg$^{-1}$), considered one of the most toxic elements for human health [60], was over 22 times the NGR (16 mg kg$^{-1}$). Finally, Cu was four-fold higher than its NGR (30 mg kg$^{-1}$). Similar concentrations of Zn, Pb, and Cd were found in tailing ponds from the Cartagena–La Unión mining district [11,61,62].

Regarding the mobility of metals, Zn and Pb had the highest concentration of bioavailable metals, although the ratio between the mean bioavailable concentration and the mean total metal concentration revealed that the most available metal was Cd, followed by Zn, Pb, Cu, Ni, Cr, and As. The water solubility of the metals was closed to zero, except Zn and Cd, which showed slightly higher concentrations in some profiles.

The ERT survey showed great differences among ERT profile G1 and profiles G2 and G3 (Figure 5).

In G1, the whole section showed very low resistivity values (1–4 Ω m), except in a small area located between 30 and 60 cm deep, where resistivity increased to 10 Ω m (Figure 5a). Profiles G2 and G3 showed a more heterogeneous distribution of resistivity that allowed the depiction of different areas within each profile. Resistivity values in the main part of G2 ranged from 25 to 90 Ω m. However, a higher resistivity area (Region 1) can be discerned from the surface to a 50 cm depth (>100 Ω m) as well as a lower resistivity area (10–20 Ω m) on the right side of the profile (region 3) (Figure 5b).

In G3, the highest resistivity area is located between the surface and a 30 cm depth (Region 1), showing values that reach 90 Ω m, which was substantially in contrast with the low resistivity of the entire profile that ranges from 6 to 20 Ω m (Figure 5c). This high resistivity area, or Region 1, is thinner than that region of G2, in agreement with the higher moisture content in G3 compared to that of G2. This surficial higher moisture in G3 could explain, among other factors such as basic pH values in G3, the important, spontaneous growth of vegetation in G3 [63–65].

Similar resistivity values were obtained by Martín-Crespo et al. (2018) [12] from the Brunita tailing pond (La Unión, Spain) and by Acosta et al. (2017) [66] from tailing ponds in the Mazarrón (Spain) mining district, where the same types of ore (pyrite, galena, and sphalerite) were extracted.
Figure 4. Soil properties, metal concentration (mg kg$^{-1}$) (mean ± standard deviation), and results from the Mann–Whitney U-test. Bioav.: Bioavailable; Sol.: Water soluble. Bioav. Cr, Sol. Cu, and Sol. Cr showed values lower than the detection limit of the equipment. Different letters in the same row indicate statistically significant differences between variables ($P < 0.05$).

Particle size distribution is closely related to the electrical conductivity in soils, affecting the amount of surficial electrical charges of the solid soil constituents [67,68]. Thus, clayed size particles, which promote the trapping of a film of electrolyte around them, contribute to a better conduction of the electricity in soils, giving lower values of resistivity than sandy or loamy soils [14,67,68]. It is also related to the capability of clayed size particles for better water retention. In the studied tailing pond, despite the coarse-size (50–2000 µm) and medium-size (0.02–50 µm) particles were predominant, statistically significant differences were found in G1, which showed a higher percentage of fine particles (<0.02 µm) than G2 and G3 profiles (Figure 4). This was in accordance with results obtained for moisture, which showed higher values in G1. Thus, the lower resistivity values found in G1 seemed to be influenced by the higher presence of fine materials, although it was not be possible to distinguish the degree to which moisture and fine particles influence the resistivity. Regardless, the relation established for G1 between observed resistivity values and fine particle sizes was in accordance with that found by Martin-Crespo et al. (2018) [12] and Faz et al. (2013) [69], who explained that, in the Cartagena–La Unión mining district, materials that show electrical resistivities below 8 Ω m correspond to fine-grain tailings, while resistivities ranging from 8 to 150 Ω m correspond to waste rocks.

No differences were found between G1 and G2-G3 in terms of waste physicochemical properties, except for salinity and inorganic carbon content, which showed the highest content in G1. G3, at a 30–90 cm depth, showed a slightly higher content in organic carbon, total nitrogen, and particles sizes
of 0.02–2000 µm than G2. High organic carbon content is usually related to low bulk density in soils, increasing the aggregate stability that suppose a decrease in resistivity [70], as well as the presence of fine particles. Therefore, the differences indicated by these properties in both profiles were shown to be responsible for the lower electrical resistivity indicated by G3 at a 30–90 cm depth. There were no significant variations in the soil properties with depth, except in G1 where pH decreases in depth.

This surficial higher moisture in G3 could explain, among other factors such as basic pH values in G3, the important, spontaneous growth of vegetation in G3 [63–65].

Similar resistivity values were obtained by Martín-Crespo et al. (2018) [12] from the Brunita tailing pond (La Unión, Spain) and by Acosta et al. (2017) [66] from tailing ponds in the Mazarrón (Spain) mining district, where the same type of ore (pyrite, galena, and sphalerite) were extracted.

Figure 5. Electrical resistivity sections of profiles G1 (a), G2 (b), and G3 (c).

No statistical differences were found in the variation in total metal concentration with respect to depth within G1, G2, or G3. However, a comparison between profiles showed statistical differences in total Zn content between G1 and G2, indicating a higher Zn concentration in G1 compared to that in G2. For total Pb and Cu, statistical differences were found between G1 (0–90 cm depth) and G3 (30–90 cm depth), the highest values found in G1.

For total Pb and Cu, statistical differences were found between G1 (0–90 cm depth) and G3 (30–90 cm depth), the highest values found in G1. Cd concentration in G1 (0–90 cm depth) was also statistically different from that in G2 and G3 at a 30–90 cm depth, the highest values of total Cd being found in G1. Moreover, statistical differences were observed for total As between G1 (0–30 cm) and G2 (0–30 cm), although in this case the concentration was higher in G2. Fine particle sizes can retain a large amount of metals because of their high specific surface area [71]. Thus, the highest amount of total metals found in G1 seems to be associated with the highest amount of clayed and fine lime-sized particles in G1.

Indeed, fine particles are generally more reactive than coarse ones [72], while moisture acts as the media where the mobile phase of metals such as metallic sulphates are released being capable of
conducting electricity in soil. Furthermore, the presence/absence of water in soils determine the redox process that affect mobility of metals because the adsorption/desorption processes [73]. Thus, the lowest resistivity values found in G1 primarily seemed associated with the presence of fine particles and moisture, although it could also be influenced, to a lesser extent, by the concentrations of bioavailable and soluble metals, because such metals are more prone to be solved in the pore solution under this conditions. No statistical differences were found among sampling depths in G1, except for Cu, whose bioavailable concentration decreased at a 30–60 cm depth. Similar relations were observed in other tailing ponds from the Cartagena–La Unión mining district (Spain) [69], Uusk (Russia) [74], and Viljakkala (Finland) [75].

3.2. ERT Model

In order to identify associations among electrical resistivity and physicochemical parameters, a Spearman correlation test was performed. The results showed that resistivity was negative and significantly correlated with soil electrical conductivity ($r_s = -0.911; P < 0.01$), particles < 0.02 µm ($r_s = -0.809; P < 0.01$), moisture ($r_s = -0.776; P < 0.01$), bioavailable Cd ($r_s = -0.713; P < 0.01$), TN ($r_s = -0.696; P < 0.001$), soluble Pb ($r_s = -0.636, P < 0.01$), soluble Cd ($r_s = -0.636; P < 0.01$), bioavailable Ni ($r_s = -0.6128, P < 0.01$), total Zn ($r_s = -0.601; P < 0.01$), 0.02–50 µm particles ($r_s = -0.579; P < 0.01$), total Cu ($r_s = -0.578; P < 0.01$), total Cd ($r_s = -0.571; P < 0.01$), total Pb ($r_s = -0.556, P < 0.01$), bioavailable Zn ($r_s = -0.519; P < 0.01$), soluble As ($r_s = -0.505, P < 0.01$), bioavailable As ($r_s = -0.475; P < 0.05$), and total Ni ($r_s = -0.435, P < 0.05$).

A negative correlation suggests that higher water content, fine and medium particle sizes, and the presence of metals, especially in its bioavailable and soluble fractions, are related to lower values of electrical resistivity [65,75,76].

Resistivity was also positive and significantly correlated with 50–2000 µm particles ($r_s = 0.802; P < 0.01$), bioavailable Fe ($r_s = 0.572, P < 0.01$), bioavailable Pb ($r_s = 0.543; P < 0.05$), and soluble Fe ($r_s = 0.512, P < 0.05$), meaning that coarser grain sizes promote higher resistivity values [14,67,68].

Physicochemical parameters significantly correlated with resistivity (with $|r_s| > 0.5$) were selected, and non-linear regression models were fitted to the data. As a result, several model formulations were obtained for the different dependent variables (Figure 6): polynomial-cubic for moisture ($Z = 0.34; P < 0.0001$); polynomial-inverse of 3rd order for soluble Cd ($Z = 0.013; P < 0.0001$); soil electrical conductivity ($Z = 0.018; P < 0.0001$), bioavailable Ni ($Z = 0.033; P < 0.0001$), bioavailable Cd ($Z = 0.196; P < 0.0001$), and soluble Pb ($Z = 0.595; P < 0.0001$); exponential decay-linear combination for total Cd ($Z = 0.10; P < 0.0001$); 0.02–50 µm particles ($Z = 0.23; P < 0.0001$), total Pb ($Z = 0.34; P < 0.0001$), total Cu ($Z = 0.39; P < 0.0001$), and bioavailable Fe ($Z = 0.56; P < 0.0001$), or polynomial-quadratic for particles < 0.02 µm ($Z = 0.013; P < 0.0001$). Only TN, total Fe, bioavailable Zn and Pb, and soluble Fe and As did not achieve all assumptions for model acceptance (residuals normality and homoscedasticity ($Z$), and $R^2 \geq 0.6$) (Figure 6).
Figure 6. Calibrated model for soil properties and metal concentration estimation using resistivity as an explanatory variable ($n = 24$).
The efficiency of each obtained model was validated by applying the corresponding model to each parameter for the validation set. This validation is shown in Figure S1 by plotting the real value of the studied parameter measured in the field and the value estimated with the model validation samples \((n = 9)\).

The best fit of the estimated data was obtained for soil electrical conductivity and bioavailable Cd, with 89% of the estimated data within the confidence interval \((CI)\) at 95%. Seventy-eight percent of the estimated data for 0.02–50 \(\mu m\) and 50–2000 \(\mu m\) particles, total Zn, total Cd, and bioavailable Ni and Fe were within CI, verifying the high quality of the models obtained for these parameters. However, the efficiency of the models obtained for soluble Cd, total Pb, total Cu, <0.02 \(\mu m\) particles, and soluble Pb was constrained because only 67%, 56%, 33%, 22%, and 1%, respectively, of the estimated data were located within the CI. This methodology was similar to that used by other authors [32,34,36], who also faced estimated and real moisture data in plots as well as metals [19].

4. Conclusions

The geochemical characterization showed that metal concentrations, especially for Zn, Pb, and Cd, in the surface area of the El Gorguel tailing pond exceed the legal reference level, with Cd being the most available and water-soluble metal in the pond, especially in the G1 area.

The ERT method has been a useful tool for supporting geochemical surveys for identifying subareas with different materials within tailing ponds. Thus, a high resistivity region was depicted on the surface of G2 and G3, and the G1 materials were catalogued as non-resistive.

By using the statistical model proposed, we estimated significant relations with soil electrical conductivity, bioavailable Cd, medium and coarse particle sizes, total Zn and Cd, and bioavailable Ni and Fe, which allow for a cheaper and efficient way of estimating their concentration ranges in other similar tailing ponds of the Cartagena–La Unión mining district using values of resistivity. Plants species typically used in the phytoremediation of tailing ponds in Mediterranean semiarid climates rarely have very deep roots, so the geochemical state of the material from the surface to a 1 m depth needs to be known.

Thus, ERT can also serve as a proxy for the estimation of ranges of waste properties, because of the relationships observed primarily among the fine particle size distribution and the water retention with low values of electrical resistivity and, to a lesser extent, with the high metal concentration and solubility of the materials infilling the pond. It should be pointed that the relations established between metals and resistivity could be a consequence of the relation between the behavior of metals and the variation in particle size and moisture. Thus, the whole relation should be further studied for applications outside this mining district.

Although the correlations established in this study could be useful for revealing the location and distribution of the highest ranges of metal accumulation at this depth, which helps to avoid hard and expensive field surveys by reducing the number of boreholes and soil samples that need to be collected for physicochemical analysis, such surveys will remain essential for the quantitation and verification of the physicochemical properties of new tailing ponds and will never be replaced by any geophysical technique.

**Supplementary Materials:** The following are available online at http://www.mdpi.com/2075-163X/10/6/559/s1. Figure S1: Relationship between the real parameter and the calculated parameter using the calibrated model obtained for each soil property and metal.

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