Multivariate-Statistical Assessment of Heavy Metals for Agricultural Soils in Northern China

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The study evaluated eight heavy metals content and soil pollution from agricultural soils in northern China. Multivariate and geostatistical analysis approaches were used to determine the anthropogenic and natural contribution of soil heavy metal concentrations. Single pollution index and integrated pollution index could be used to evaluate soil heavy metal risk. The results show that the first factor explains 27.3% of the eight soil heavy metals with strong positive loadings on Cu, Zn, and Cd, which indicates that Cu, Zn, and Cd are associated with and controlled by anthropic activities. The average value of heavy metal is lower than the second grade standard values of soil environmental quality standards in China. Single pollution index is lower than 1, and the Nemerow integrated pollution index is 0.305, which means that study area has not been polluted. The semivariograms of soil heavy metal single pollution index fitted spherical and exponential models. The variable ratio of single pollution index showed moderately spatial dependence. Heavy metal contents showed relative safety in the study area.

1. Introduction

Soils are critical environments where rock, biology, air, and water interface. Soil pollution has become an important environmental issue in China owing to rapid economic development and industrialization and increasing reliance on agrochemicals in the last few decades [1–3]. Soil heavy metal contents are not only the serious environmental issue but also frequency related to agricultural soil utilization problem [4–6]. Soil heavy metals could be necessary or beneficial to plants at certain levels but toxic when exceeding specific threshold [7–11]. If these elements are absorbed by the plants through the root system, they may enter the food chain and become toxic to humans and animals. The ecological importance of soil heavy metals is closely related to human health due to their high ecological transference potential.

For agricultural soils, the main pollution sources of heavy metals are due to activities such as irrigation using wastewater, pesticides, agricultural fertilizers, and organic manure, disposal of urban and industrial wastes, and atmospheric pollution from motor vehicles and the combustion of fossil fuels. Heavy metals in agricultural soil have become higher than background levels. It has also become a hot spot for the study of the international soil environment science research and ensured its sustainability.

Multivariate statistical approaches including principal component analysis (PCA) and cluster analysis (CA) are the statistical tools used in the elaboration pollution [12, 13]. It has been reported that PCA methods have been widely used in geochemical applications to identify soil pollution sources and distinguish natural versus anthropic contribution [14, 15]. CA is often used coupled to PCA to check results and group
individual parameters and variables [16–20]. The methods of geostatistics use the stochastic theory of spatial correlation for both interpolation and for apportioning uncertainty [21]. Geostatistics is based on the theory of a regionalized variable which uses the technique of semivariogram to measure the spatial variability of a regionalized variable and provides the input parameters for the spatial interpolation of kriging [22, 23]. Geostatistics can also be used to assess the risk of exceeding critical values (regulatory thresholds, soil quality criterion) at unsampled locations and to simulate the spatial distribution of attribute values [21].

The present study is focused on the suburban area of Shijiazhuang city. Zhengding has formed an industrial structure consisting of Shijiazhuang airport, electronic, petrochemical, and highly traffic density undergone a rapid transition from a traditionally agricultural-based to an increasingly industrial-based economy in the last 30 years which could enhance the risk of metal contamination through food chain in the region as heavy metals may enter and accumulate in agricultural soils through atmospheric deposition and irrigation. Early studies in the study area indicated the spatial variability of heavy metal distribution [17].

The study was to analyze possible sources of these heavy metals by multivariate statistical techniques and to assess soil heavy metal contamination using single pollution index and integrated pollution index to identify spatial distribution of single pollution index of soil heavy metal through geostatistical analysis.

2. Materials and Methods

2.1. Study Region and Soil Sampling. The study is focused on the suburban area of Shijiazhuang city, northern China. The region is well known for intense industrial and commercial activities. Soil samples were collected from 100 locations in the Zhengding, an administrative district covering 468 km$^2$ with a population of around 467,000 people. This area has a continental monsoon climate with an average annual temperature 13°C, and an average precipitation of 530 mm. The altitude levels range from 65 m to 105 m within the study area. The main soil type is carbonate cinnamon and Chao soil according to China soil classification system. The samples were collected from agricultural areas (mostly wheat, cereals, and vegetables). The coordinates of sampling locations were recorded with GPS. The sampling locations are shown in Figure 1.

2.2. Soil Analyses. Each sample consists of 10 soil cores (0–20 cm) depth, 1 kg ca total weight, which was collected from within a 10 × 10 m area with the central point corresponding to the defined position for the sample. All of the samples were air-dried at room temperature and sieved by a 2 mm mesh and stored in polyethylene plastic bags for subsequent sample analysis. Metal concentrations for Cu, Zn, Ni, Pb, Cr, Hg, As, and Cd were analyzed after complete dissolution using a mixture of HNO$_3$–HF–HClO$_4$ and heated in a microwave digestion system, using appropriate atomic absorption spectrometric techniques [17]. The accuracy of the procedure was determined by analyzing the certified reference material GB 7475-87 (National Institute of Standards and Technology, China). Quality controls involved analysis of random samples, blank samples, and national standard samples each time.

2.3. Assessment of Soil Contamination. The assessment of soil heavy metal contamination in agro-ecosystem is often the choice of single pollution index (Pi) and the Nemerow...
Table 1: Total variance and component matrices (three factors selected) for heavy metals.

(a) Total variance explained

| Component | Initial eigenvalues | Extraction sums of squared loadings | Rotation sums of squared loadings |
|-----------|---------------------|-----------------------------------|----------------------------------|
|           | Total % of variance  | Cumulative %                      | Total % of variance              | Cumulative %                      |
| 1         | 2.70                | 33.75                             | 3.75                             | 33.75                             |
| 2         | 1.46                | 18.30                             | 30.54                            | 18.30                             |
| 3         | 1.27                | 15.79                             | 36.33                            | 15.79                             |
| 4         | 0.89                | 11.16                             | 47.51                            | 11.16                             |
| 5         | 0.63                | 7.87                              | 55.38                            | 7.87                              |
| 6         | 0.48                | 5.98                              | 61.36                            | 5.98                              |
| 7         | 0.36                | 4.52                              | 65.88                            | 4.52                              |
| 8         | 0.20                | 2.60                              | 100.00                           | 2.60                              |

(b) Component matrices

| Element | Component matrix | Rotated component matrix |
|---------|------------------|--------------------------|
|         | PC1   | PC2   | PC3   | PC1   | PC2   | PC3   |
| Cu      | 0.838 | 0.231 | 0.117 | 0.648 | 0.585 | 0.086 |
| Zn      | 0.804 | 0.236 | −0.325 | 0.875 | 0.207 | 0.008 |
| Ni      | 0.549 | −0.593 | 0.335 | 0.125 | 0.367 | 0.785 |
| Pb      | 0.317 | 0.028 | 0.727 | −0.153 | 0.756 | 0.182 |
| Cr      | 0.692 | −0.355 | −0.148 | 0.565 | 0.114 | 0.544 |
| Hg      | −0.041 | 0.850 | 0.101 | 0.085 | 0.319 | −0.791 |
| As      | 0.428 | 0.381 | 0.380 | 0.205 | 0.637 | −0.156 |
| Cd      | 0.536 | 0.096 | −0.572 | 0.771 | −0.166 | 0.013 |

3. Results and Discussion

3.1. Principal Component Analysis

Summary statistics and normal test of Zhengding datasets were performed and the results were present [17]. PCA can be used to reduce data and to extract a smaller number of independent factors (principal components) to find the relationship among observed variables [16–18]. Principal component analysis (PCA) was applied in the study to have high quality experimental results. The PCA based results for soil heavy metals are listed in Table I. According to the initial eigenvalues, three principal components are selected, accounting for over 67.8% of the total variance. The eigenvalues of the three first extracted factors are greater than one. All of the elements are consequently well represented by these three principal components.

3.4. Statistical Analysis

3.1. Principal Component Analysis. Summary statistics and normal test of Zhengding datasets were performed and the results were present [17]. PCA can be used to reduce data and to extract a smaller number of independent factors (principal components) to find the relationship among observed variables [16–18]. Principal component analysis (PCA) was applied in the study to have high quality experimental results. The PCA based results for soil heavy metals are listed in Table I. According to the initial eigenvalues, three principal components are selected, accounting for over 67.8% of the total variance. The eigenvalues of the three first extracted factors are greater than one. All of the elements are consequently well represented by these three principal components.

The initial component matrix for heavy metals indicates that Cu, Zn, and Cd are associated, showing high values in the first principal component (PC1) which explains 27.3% of the total variance and loads heavily on Cu (0.65), Zn (0.88), and Cd (0.77). Cu and Zn values are controlled by a long-term anthropic activity such as pesticides. The second principal component (PC2) includes univocally Pb and As, which accounts for 20.5% of the total variance. Common sources of lead in soils are manure, sewage sludge, lead-arsenate pesticides, vehicle exhausts, and industrial fumes. The study area has a high vehicular traffic density, constituted by Zhengding international airport and several important railways, expressway...
confirm results and provide grouping of variables [5]. The
from PCA, CA could be used as an alternative method to
3.2. Cluster Analysis. Although not substantially different
alloftheeightvariablesisclearlyrevealed.
inorganic fertilizers. Thus in Table 1, the relationship among
source of Zn, Cu, and Cd due to the application of manure or
enrichment of heavy metals. These practices are an important
body. In addition, normal agricultural practices may cause
maycauseheavymetals such as copper, nickel, zinc, and cad-
mium to accumulate in the topsoil [26, 27]. Common sources
of lead in soils include car exhausts, manure, sewage sludge,
and coal burning. Cd is present in fossil fuel such as coal and
oil and having no functions in plants or animals or human
body. In addition, normal agricultural practices may cause
enrichment of heavy metals. These practices are an important
source of Zn, Cu, and Cd due to the application of manure or
inorganic fertilizers. Thus in Table 1, the relationship among
all of the eight variables is clearly revealed.

| Label | Number |
|-------|--------|
| Cu    | 1      |
| Zn    | 2      |
| Cd    | 3      |
| Ni    | 4      |
| Cr    | 5      |
| As    | 6      |
| Pb    | 7      |
| Hg    | 8      |

Figure 2: Dendrogram results from Pearson correlation coefficients of hierarchical cluster analysis for heavy metals.

(http://www.en.wikipedia.org/wiki/zhengding), for leaded
gasoline widely used. The third principal component (PC3)
is correlated very strongly with Hg loading (−0.79) and also
by Ni loading (0.79), accounting for 20% of the total variance.
Long-term and extensive use of pesticides in farmland
may cause heavy metals such as copper, nickel, zinc, and cad-
mium to accumulate in the topsoil [26, 27]. Common sources
of lead in soils include car exhausts, manure, sewage sludge,
and coal burning. Cd is present in fossil fuel such as coal and
oil and having no functions in plants or animals or human
body. In addition, normal agricultural practices may cause
enrichment of heavy metals. These practices are an important
source of Zn, Cu, and Cd due to the application of manure or
inorganic fertilizers. Thus in Table 1, the relationship among
all of the eight variables is clearly revealed.

3.2. Cluster Analysis. Although not substantially different
from PCA, CA could be used as an alternative method to
confirm results and provide grouping of variables [5]. The
heavy metals concentration data were calculated using the
hierarchical clustering with SPSS software. Figure 2 shows the
CA results for the heavy metals as a dendrogram in the study
area. This figure shows three clusters. (1) Cu, Zn, and Cd are
very well correlated with each other. The farming area has had
several decades of intensive tillage, long-term fertilizer and
pesticide application might be a major source of accumulated
heavy metals in the study area, (2) which is associated with Pb
and As. Pb is mostly found in automobile battery in sufficient
amount; excessive intake Pb can damage the nervous, skeletal,
circulatory, enzymatic, endocrine, and immune systems of
human body. (3) Ni and Hg are commonly associated in a
number of rock types or soil parent materials. The analyzed
results are in good accordance with the findings of the PCA
analysis.

3.3. Hazard Assessment of Soil Heavy Metals. Table 2 shows
that the mean contents of Zn, Hg, and Cd exceed soil
background values, but they are still lower than the Grade II
criteria, which mean that the three metal elements are mainly
affected by anthropogenic sources. The mean value of soil Zn
69.96 mg/kg was higher than its background value 62.0 mg/kg
but it did not exceed the limiting content 300 mg/kg of SEPA.
The average value of Cd 0.15 mg/kg was higher than its back-
ground value 0.075 mg/kg. Phosphate fertilizers have been
well known as the major external source of soil Cd [26, 27].

Especially Hg is 3.49 times the background value and mainly
originated from industry and traffic sources. Cu, Ni, Pb, Cr,
and As concentrations are lower than or approximately equal
to their corresponding background values, which indicates
that these elements are dominated by natural sources and
human activities.

The heavy metal concentrations in Zhengding agricultural
soil are compared with the data reported from other
areas in the world in Table 2 [6, 14, 17, 20]. The mean
content of Cu and As is lower than that for other areas. The
average values of Pb, Hg, and Cd are similar to those of
Beijing and Tianjin, but lower than in other areas. The mean
concentration of Cu, Pb, Cr, As, and Cd was lower than those
reported by Hu, and within the range reported by Qiao.

The Chinese Environmental Quality Standard for soils
(GB15618-1995) [24] and the soil background values of Hebei
were adopted to evaluate the pollution degree. The soil is
mainly alkaline in the investigated area. Grade II criteria for
soil quality are established to protect agricultural production
and to maintain human health. Across the investigated area,
the wide range of soil heavy metal concentrations have been
measured.

The analyzed results indicate that all of the metal con-
centrations are below the Environmental Protection Admin-
istration for soil in China. The PI of soil heavy metals and
the Nemerow P values, range, median values, and CV in
the 100 topsoil samples are shown in Table 3. On average,
the PI indices for Cu, Zn, Ni, Pb, Cr, Hg, As, and Cd
were 0.212, 0.233, 0.417, 0.054, 0.231, 0.076, 0.246, and 0.246,
respectively. Different heavy metal concentrations of single
pollution index are in an order of Ni > Cd > As > Zn >
Cr > Cu > Hg > Pb. The single pollution index showed that
the Ni pollution intensity was strong. Cd, Cr, and Pb are
considered as the most important environmental pollutants
in agricultural soils because of the potential harmful effects
they may have on food quality and health of soil.

The Nemerow integrated pollution index (P) in this area
is 0.305, which is lower than 1, meaning that heavy metal
exposure through the food chain does not have considerable
consequence and is generally safe.

3.4. Semivariogram Analysis. The semivariogram model of
soil heavy metal single pollution index at both orientations
is given in Figure 3. Theoretical models were then employed
to interpret the experimental semivariograms and the model
with the best fitting was chosen [22]. $P_N$ and $P_C$ were fitted
with the exponential model, and the other four heavy metal
single pollution indexes were all best fitted with the spherical
model.

The attributes of the semivariograms for each soil heavy
metal single pollution index were also summarized in Table 4.
All of the Nug/Sill ratios were less than 59.74%, indicating
random heterogeneity. The nugget effect may be caused by
random factors such as data deviations, agricultural activ-
ities, or sample density. Nugget contributions highlighted
the stronger spatial correlation in Table 4. The range of
semivariograms for soil heavy metal single pollution index
ranged from 0.0384 km to 0.1192 km. This confirmed the
rational of the sampling density.
Figure 3: Semivariograms model of soil heavy metal single pollution index.

Table 2: A summary of heavy metal concentrations in agricultural soil for various areas (mg/kg).

| Area            | Cu   | Zn   | Ni   | Pb   | Cr   | Hg   | As   | Cd   | References          |
|-----------------|------|------|------|------|------|------|------|------|---------------------|
| Zhengding       | 21.22| 69.96| 25.04| 18.80| 57.77| 0.08 | 6.16 | 0.15 | This study          |
| Background value| 21.7 | 62.0 | 28.8 | 20.0 | 63.9 | 0.23 | 12.1 | 0.075 | Yang et al. (2009) [18] |
| Grade II        | 100  | 300  | 60   | 350  | 250  | 1.0  | 25   | 0.6   | SEPA (1995) [24]    |
| Beijing         | 26.08| 61.18| 24.01| 18.81| 67.77| 0.079| 7.68 | 0.24  | Hu et al. (2004)    |
| Beijing, Tianjin| 28.2 | 71.0 | NA   | 18.7 | 52.3 | 0.092| 7.9  | 0.145 | Qiao et al. (2011) [15] |
| Alicante        | 22.5 | 52.8 | 20.9 | 22.8 | 26.5 | NA   | NA   | 0.34  | Micó et al. (2006) [6] |
| Ebro            | 17.3 | 57.5 | 20.5 | 17.5 | 20.3 | 0.036| NA   | 0.42  | Rodriguez et al. (2006) |
| Huizhou         | 21.82| 66.15| 20.52| 65.38| 43.01| 0.24 | 12.76| 0.12  | Cai et al. (2012) [9] |

Grade II (the value for protection of agricultural production and human health).
Table 3: Descriptive statistics of soil heavy metals pollution indices.

| Variable | Cu   | Zn   | Ni   | Pb   | Cr   | Hg   | As   | Cd   | P   |
|----------|------|------|------|------|------|------|------|------|-----|
| Min.     | 0.111| 0.154| 0.215| 0.035| 0.131| 0.018| 0.087| 0.15  | 0.091|
| Max.     | 0.332| 0.293| 0.608| 0.117| 0.354| 0.374| 0.396| 0.483 | 0.521|
| Mean     | 0.212| 0.233| 0.417| 0.054| 0.231| 0.076| 0.177| 3.142 | 1.968|
| Kurt.    |      |      |      |      |      |      |      |      | 9.669|
| Skew.    |      |      |      |      |      |      |      |      | 2.769|
| St.d     | 0.034| 0.024| 0.077| 0.011| 0.035| 0.060| 0.060| 0.067 | 0.121|
| C.V.     | 16.04| 10.30| 18.47| 20.37 | 15.15| 78.95| 24.39| 27.24 | 39.67|

St.d: standard deviation, C.V.: coefficient variation.

Table 4: Spatial correlation for heavy metals.

| Model | Nugget $C_0$ | Sill $C + C_0$ | $C_0/(C + C_0)$ (%) | Range (km) |
|-------|---------------|----------------|---------------------|------------|
| $P_{Cu}$ | Spherical | 0.0037 | 0.0064 | 58.40 | 0.0746 |
| $P_{Zn}$ | Spherical | 5.8 e-05 | 0.0001 | 44.25 | 0.05 |
| $P_{Ni}$ | Exponential | 0.0007 | 0.0015 | 45.58 | 0.1192 |
| $P_{Pb}$ | Spherical | 0.0023 | 0.0038 | 59.74 | 0.0613 |
| $P_{Cr}$ | Exponential | 0.0016 | 0.0034 | 48.24 | 0.0384 |
| $P_{Hg}$ | Spherical | 0.0019 | 0.0041 | 46.49 | 0.0521 |

4. Conclusions

The single pollution index, integrated pollution index, and sources of the heavy metals Cu, Zn, Ni, Pb, Cr, Hg, As, and Cd in agricultural topsoil samples collected from Zhengding have been investigated in this work. The mean values of single pollution index and integrated pollution index are less than 1 in the area.

The mean values of Zn, Ni, Pb, Cr, Hg, As, and Cd in the analysed soils do not exceed the limited second grade criteria environmental quality standard for soils in China (GB 15618-1995), which means that the soil in this area is not polluted. Only Zn, Hg, and Cd present higher values in some cases. The mean values of soil Zn, Hg, and Cd were higher than the, respectively, background values. Agrochemical inputs may play the most important role for the input of Zn and Cd. The risk of Hg and Cd accumulation requires further attention and monitoring.

Multivariate statistics is found to be a powerful tool to identify the main factors determining the variability of geochemical data and interpret the measurement results. Variation of Cu, Zn, and Cd concentrations is controlled by anthropogenic intense agriculture activities. The concentrations of Pb and As in agricultural soil are abnormalities mainly affected by aerial deposits from gasoline exhausts, while the concentrations of Ni and Hg in agricultural soil are mainly affected by natural parent material and human activities.

Conflict of Interests

The authors declare that there is no conflict of interests regarding the publication of this paper.

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