A multi-criteria evaluation system for arable land resource assessment

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Abstract This study proposed a multi-criteria evaluation system for arable land resources by combining the soil integrated fertility index (IFI) with a soil cleanliness index (based on heavy metals and metalloid content). A total of 16 typical arable land units in Chongming District, China, were evaluated using the proposed evaluation system based on 104 collected soil samples in 16 towns. The comprehensive soil evaluation scores of arable lands in 16 towns were in the range of 90.7 to 99.2 with a mean of 96.2, indicating that the arable land in all 16 towns was at the level of excellent (≥ 90.0). Lower cleanliness indices had a significant impact on the final evaluation score. In comparison with single-index evaluation systems (i.e., the IFI or soil cleanliness index), the proposed multi-criteria system better reflects the quality of the soil. In the practice of arable land requisition and subsidy policy, the proposed multi-criteria evaluation system not only encourages farmers to preserve arable lands during farming but also helps agricultural authorities make effective and reliable management decisions.

Keywords Multi-criteria evaluation system · Arable land resource · Integrated fertility index (IFI) · Soil cleanliness index · Heavy metals · Comprehensive soil evaluation score

Introduction

Arable land is the basis of agricultural production, and its quality is essential for crop security and ecological sustainability (Stenberg 1999). It represents a key component in the synchronization of urban and rural development. Rapid economic development and industrialization degrade the arable land in China (Hu et al. 2016; Zhao et al. 2014). The joint report on the current status of soil contamination in China, issued by the Ministry of Environmental Protection and the Ministry of Land and

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Resources of the People’s Republic of China in 2014, revealed that more than 19.4% of agricultural soils have been contaminated according to the soil environmental quality limits (MEP 2014; National Environmental Protection Bureau 1995). During the last two decades, a number of studies have shown that heavy metal pollution in soils has been widespread in China (Chen et al. 1999; Hu et al. 2016; Khan et al. 2008; Teng et al. 2010; Zheng et al. 2016). According to the State Environment Protect Agency (SEPA) of China (2006), it is estimated that 12 million tons of grain is polluted by heavy metals every year and a total of 10 million ha of arable land in China has been polluted by heavy metals such as chromium (Cr), zinc (Zn), copper (Cu), and Zn and a metalloid arsenic (As) (Teng et al. 2010).

Besides contamination, urban land expansion also largely encroached upon arable land resources in the last decade. For example, one study has shown that the urban land area in the Beijing–Tianjing–Hebei region (China) expanded by 71% during a 10-year period (1990–2000) (Tan et al. 2005). This problem—the decrease in arable land resource with urbanization and an increase in population—has attracted worldwide attention (Cai et al. 2002; Fazal 2001; She and Xie 2000; Tania et al. 2001). In 1998, China issued an arable land requisition–subsidy balance policy, which promised to subsidize an equal amount of land to farmers when their arable land is requisitioned for non-agricultural use, such as infrastructures for residential and industrial purposes. The aim of this policy was to preserve land resources for agricultural use. However, due to the lack of an effective and reliable arable land evaluation method, the requisition–subsidy balance policy is proven to be ineffective and the loss and degradation of arable land continues (Chen et al. 2015; Hu et al. 2012). Therefore, it is very important to develop a more effective and reliable method for evaluating arable land resources.

The land capability classification released by the US Department of Agriculture in 1961 laid the foundation for the quantitative analysis of arable land resources (Klingebiel and Montgomery 1961). The current international and national quality evaluation systems of arable land tend to focus more on production capability, land potential, and ecological quality and sustainability (Fu and Bai 2015). However, the implementation practices often only consider one single criterion, such as the soil integrated fertility index (IFI) (Breibda et al. 2000; Mu et al. 2018; Wang et al. 2001). The lack of a comprehensive evaluation standard results in noncomparable evaluation results. A more scientific and applicable tool for assessing arable land resources is needed. A practical assessment of arable land resource requires integrated consideration of key soil properties and their spatial and temporal variations (Alaoui et al. 2018). Currently, most of the existing evaluation systems are based on the provincial level yet there is little on the county or town level, although the relationships of available micronutrients in the soil and influencing factors were scale- and location-dependent (Tan et al. 2005; Zhu et al. 2016). This implies that, in order to improve the quality of arable land, different management practices are needed on a smaller scale level.

This study aims to develop a multi-criteria evaluation system for evaluating arable land area by taking into account both the fertility of land and the soil cleanliness index (i.e., metal contamination) as restriction factors. The evaluation system determines arable land resources via a new arable land area correction method, which could provide an effective and reliable method for the evaluation and management of arable land resources.

Materials and methods

Study area and backgrounds

The study area, Chongming District (31.45° to 31.85° N and 121.16° to 121.90° E), is located in the Yangtze River estuary of China. Chongming District includes three islands (Chongming, Changxing, and Hengshan), which possess the largest and most concentrated agricultural land resources as well as the best agricultural environment in Shanghai, China. Since almost half of the area of the present islands is from the reclamation of wetland (Zheng et al. 2016), the quality of the reclaimed soil has been a concern, especially due to contamination by heavy metals and metalloids (Yang et al. 2013; Zheng et al. 2016). With the increasing emphasis on arable land quality and management, a series of studies have been carried out on heavy metals in the soils of Chongming District (Hu et al. 2013; Ma et al. 2015; Sun et al. 2010). Extensive agricultural activity has increased the accumulation of heavy metals (e.g., Cr, Zn, Cu, and Zn and As) in paddy fields and farmland (Zheng et al. 2016). In addition, stubble burning is also regarded as a significant source of heavy metals through atmospheric deposition (Sun et al. 2010). Thus, a multitude of factors
might possibly affects the arable land assessment of Chongming District, China.

Sample collection

Topsoil samples (2 cm to 20 cm) were collected with a bamboo spade in 16 towns of Chongming District in April and July 2016. Each town featured 4 to 7 sampling sites including paddy and upland fields. A total of 104 samples were collected. Figure 1 shows the location of the 16 towns studied in Chongming District. At each sampling site, a 1 × 1 km² sampling grid was randomly selected. Five topsoil cores were collected from each sampling grid, including one central point and four additional points towards the east, west, south, and west. After collection, these five topsoil samples were mixed together to make a single composite sample.

Sample analyses

After transport to the laboratory, soil samples were oven-dried at 60 °C, ground, and passed through a 75-μm (equivalent to no. 200 according to ASTM E11 standards) stainless steel sieve. Soil samples were stored in a desiccator prior to further analyses. Total organic matter (TOM) was estimated by the potassium dichromate (K₂Cr₂O₇) volumetric method (NY/T 1121.6; Ministry of Agriculture 2006) using the K₂Cr₂O₇-sulfuric acid solution as the digestion medium. Available phosphorus (Av-P) was extracted by sodium bicarbonate and determined by the molybdenum–antimony colorimetric method (NY/T 1121.7; Ministry of Agriculture 2014). Available potassium (Av-K) was extracted by ammonium acetate and measured by flame atomic absorption spectrophotometry (NY/T 889; Ministry of Agriculture 2004). The land fertility levels for TOM, Av-P, and Av-K were assessed based on the classification of soil nutrition adopted by the Second National Soil Survey (National Soil Survey Office 1979).

The concentrations of As, Cu, Cr, Pb, and Zn were measured using inductively coupled plasma mass spectrometry (ICP-MS) (PE NexION 300X, PerkinElmer). Prior to ICP-MS analysis, 0.1 g soil samples were digested by 3 mL HNO₃ (65%), 1 mL HF (40%), and 1 mL H₂O₂ (30%) in sealed Teflon vessels in a microwave (PreeKem, TOPEX). After transfer to a volumetric flask, HClO₄ (1 mL) was added to the clear digest to

Fig. 1 The locations of 16 towns (separated with different colors) sampled in Chongming District, China. Their corresponding soil integrated fertility index (IFI), soil cleanliness index (Kᵢ), and comprehensive soil evaluation index scores (CSEI) are shown as columns.
remove the remaining HF. All the acid used in the digestion step was ultrapure and could be used for trace metal analysis. Analytical quality was controlled by using sample replicates, reagent blanks, and an internal standard. The relative standard deviation (RSD) between duplicates was 0.2% to 15.8%. Internal standard solutions including Sc, Ge, In, and Bi were used for ICP-MS analysis to correct the signal bias and drifts caused by the matrix interference. The study did not consider mercury (Hg) and cadmium (Cd), which had concentrations below the detection limit of ICP-MS.

The assessment of the multi-criteria evaluation system

IFI

The assessment of soil fertility is a useful system that helps to improve sustainable land use management. IFI is an effective and important indicator for assessing the quality and degradation of arable land (Mu et al. 2018; Shang et al. 2014). In this study, we calculated the integrated IFI based on TOM, Av-P, and Av-K parameters using a weighted function

\[ IFI = 100 \sum_{i=1}^{n} F_i \times C_i \]

where \( F_i \) is the score of the \( i \)th parameter, which is used to assess soil fertility index, and \( C_i \) is the weight coefficient of the \( i \)th parameter of soil fertility. The weight coefficients for TOM, Av-P, and Av-K were 0.600, 0.200, and 0.200, respectively. In order to minimize the effect of temporary fertilization in the evaluation operation, the weight coefficient of TOM in the matrix was one level higher than that of Av-P and Av-K (Jiao et al. 2014; Lee et al. 2004). The score \((F_i)\) value of each measured soil fertility parameter was calculated by its measured absolute value and a standard scoring function (SSF) (Hussain 1997; Shang et al. 2014). An S-pattern function (Eq. (2)) was used to calculate the SSF values of each parameter (Tian and Xin 2006)

\[ F_i = \begin{cases} 0 & u_i \leq u_t \\ 1/(1+a_i(u_i-c_i)^2) & u_t < u_i < c_i (i=1,2,\ldots,m) \\ 1 & u_i \geq c_i \end{cases} \]  

where \( u_i \) is the measured concentration of soil samples, \( c_i \) is the standard index, \( a_i \) is a constant, and \( u_t \) is the bottom limit of the index. The values of \( a_i \), \( c_i \), and \( u_t \) were derived from expert assessments and analysis by Statistical Package for the Social Sciences (IBM SPSS Statistics 24; Li 2012). For TOM, the values of \( a_i \), \( c_i \), and \( u_t \) are 0.040, 14.4, and 2.00, respectively. For Av-P, the values of \( a_i \), \( c_i \), and \( u_t \) are 0.019, 20.3, and 3.00, respectively. For Av-K, the values of \( a_i \), \( c_i \), and \( u_t \) are 0.0007, 138, and 20.0, respectively.

Soil cleanliness index (K)

The soil cleanliness index \((K)\) was optimized by the coefficient construction method of soil environment quality proposed by Lu et al. (2011). The \( K \) index was evaluated based on a pollution-level determination method

\[
K_i = \begin{cases} 100 & P_C < 0.700 \\ \frac{3-P_C}{3-0.7} & 0.700 \leq P_C < 3.00 \\ 0 & P_C \geq 3.00 \end{cases}
\]

where \( K_i \) is the cleanliness index in the \( i \)th unit and \( P_{Ci} \) is the comprehensive soil pollution index in the \( i \)th unit. The cleanliness index in a given area was calculated by the average value of cleanliness index of all sample sites in the area. The comprehensive pollution index \( (P_C) \) is defined as

\[ P_C = \sqrt{\frac{(P_A^2+P_{\max}^2)}{2}} \]

where \( P_A \) is the mean value of individual pollution indices and \( P_{\max} \) is the maximum value of the individual pollution index. The standards for the levels of pollution were defined as follows: \( P_C \leq 0.700 \), very clean; \( 0.700 < P_C \leq 1.00 \), clean; \( 1.00 \leq P_C < 2.00 \), light pollution; \( 2.00 \leq P_C < 3.00 \), medium pollution; and \( P_C \geq 3.00 \), heavy pollution. The individual soil pollution index \( (P_i) \) is calculated as

\[ P_i = \frac{M_i}{S_i} \]

where \( P_i \) is the individual pollution index, \( M_i \) is the measured value of pollution, and \( S_i \) is the lower limit of the pollution index, which is based on the China agricultural soil standard (GB15618-1995). The expression \( P_i \leq 1.00 \) refers to a qualified individual pollution index that does not exceed the standard limit. The expression \( P_i > 1.00 \) refers to an unqualified individual pollution index that exceeds the standard limit. Both the individual and comprehensive soil pollution indices were calculated according to the technical specifications for the survey and quality evaluation of arable land (NY/
Comprehensive soil evaluation index

The arable land resource value was calculated using a comprehensive soil evaluation index (CSEI), which is defined as

\[
CSEI = \frac{IFI + K}{2} \tag{6}
\]

In this study, a CSEI of > 60 is considered acceptable, 70–80 is good, and > 90 is excellent.

Land area correction method

An “ideal hectare” is defined as a hectare of arable land with a CSEI score of 100. Based on the concept of an ideal hectare, the corrected land area is determined as

\[
S_T = CSEI \times \frac{S}{100} \tag{7}
\]

where \( S_T \) is the value of land area after correction and \( S \) is the measured geometric area of a given area of arable land.

Results

Soil fertility and metal content

The concentrations of TOM of all 16 towns were in the range of 13.3 ± 2.61 g/kg to 22.9 ± 3.45 g/kg (Table 1). According to the Second National Soil Survey, 14 of 16 sampled towns were at TOM level 4 (10 g/kg to 20 g/kg; National Soil Survey Office 1979). The TOM concentration in the towns of Miao (21.7 ± 4.44 g/kg) and Chengqiao (22.9 ± 3.45 g/kg) was higher than 20 g/kg, which belongs to level 3 according to the Second National Soil Survey (Table 1). The range of Av-P content in 16 towns was 159 ± 72 mg/kg to 896 ± 196 mg/kg (Table 1). According to the soil agrochemical standards, Av-P content in all 16 towns was at level 1 (> 40.0 mg/kg, Table 1; Nanjing Agriculture University 1996). For Av-K, 8 of 16 towns were at level 1 (> 200 mg/kg), 6 of 16 towns were at level 2 (150–200 mg/kg), and 2 of 16 towns were at level 3 (100–150 mg/kg, Table 1). In general, due to low TOM content, land fertility was at level 4 in 14 of 16 selected towns (Table 1).

Concerning metals and As concentrations, the mean concentrations of Cr, Cu, Zn, and As were 64.5 mg/kg, 31.9 mg/kg, 86.0 mg/kg, and 11.7 mg/kg, respectively, which are higher than their corresponding mean background values in China of 61.0 mg/kg, 22.6 mg/kg, 74.2 mg/kg, and 11.2 mg/kg, respectively (Table 2; Chen et al. 2015). According to Chinese soil guidelines, the concentrations of Cr, Cu, Zn, and As belong to level 1 (Table 2). The mean concentration of Pb (mean = 22.5 mg/kg) was generally low across the whole island. Individual samples were found to have relatively high metal contents in a few towns when compared with the background concentration of China (CNEMC 1990). For example, Cu was found to be 39.5 ± 4.97 mg/kg in the town of Chengqiao (Table S1); the concentration of As in Jianshe reached 21.6 ± 3.00 mg/kg (Table S1); the concentrations of Cu and Zn in Shuxin were 37.2 ± 13.3 mg/kg and 111 ± 38.4 mg/kg, respectively (Table S1).

Comprehensive soil evaluation indices

Only 2 (Sanxing and Changxing) of 16 towns had IFI values lower than 95.0. All of the other fourteen towns had high IFI values in the range of 95.3 to 100 (Fig. 1). The IFI values for Sanxing and Changxing were 93.7 and 94.7, respectively. By comparing the scores of TOM \( (F_{TOM}) \), Av-P \( (F_{Av-P}) \), and Av-K \( (F_{Av-K}) \), we found that only Av-K showed lower score values with a range of 0.685 to 1.00 (mean of 0.858). For example, the lowest IFI (93.7) was found in Sanxing, with scores of 1 \( (F_{TOM}) \), 0.685 \( (F_{Av-P}) \), and 1 \( (F_{Av-K}) \), respectively.

In general, the soil cleanliness indices \( (K_i) \) of all 16 towns were relatively good, with a range of 84.6 to 100 (mean of 95.3, Fig. 1). The lowest \( K \) value (84.6; Fig. 1) was found in the town of Xianghua with a highest comprehensive pollution index \( (P_C) \) of 2.77 (Fig. 2). There were 10 of 16 towns with \( P_C \) values lower than 0.7 (Fig. 2), suggesting the soils in these towns were at the very clean level based on metal and metalloid contamination. The \( P_C \) values in 4 of 16 towns (Sanxing, Jianshe, Zhongxing, and Xincun) were at the clean level, with a range of 0.72 to 0.99 (Fig. 2). The towns of Miao and Xianghua
had high $P_C$ values of 1.17 and 2.77, respectively (Fig. 2). The soil in Miao was at the light pollution level with a $P_C$ value of 1.17. Xianghua was at the medium pollution level, with a $P_C$ value of 2.77 (Fig. 2).

The means of CSEI (Eq. (6)) of 16 towns were in the range of 90.7 to 99.2 with a mean of 96.2, suggesting that the soils in all of the towns were at the excellent level (Fig. 1). The lowest CSEI value of 90.7 was found in the town of Changxing due to its low IFI (94.7) and $K$ (86.6) values (Fig. 1). The order of towns based on CSEI value was completely different than the orders based solely on the IFI or soil cleanliness index.

### Land area correction

The arable land in the town of Gangyan, with a CSEI value of 99.2, was close to the so-called ideal hectare. A

### Table 1

| Towns  | TOM (g/kg) | TOM level | Av-P (mg/kg) | Av-P level | Av-K (mg/kg) | Av-K level | Soil grade |
|--------|------------|------------|--------------|------------|--------------|------------|------------|
| Xincun | 18.4 ± 1.21| 10–20      | 292 ± 89.2   | > 40       | 270 ± 140    | > 200      | 4          |
| Sanxing| 18.7 ± 3.91| 10–20      | 175 ± 35.1   | > 40       | 177 ± 78.6   | 150–200    | 4          |
| Miao   | 21.7 ± 4.44| 20–30      | 178 ± 66.4   | > 40       | 114 ± 31.5   | 100–150    | 3          |
| Gangxi | 18.9 ± 4.42| 10–20      | 159 ± 72.0   | > 40       | 176 ± 68.4   | 150–200    | 4          |
| Chenguiao| 22.9 ± 3.45| 20–30     | 188 ± 83.4   | > 40       | 151 ± 48.7   | 150–200    | 3          |
| Jianshe| 18.6 ± 3.24| 10–20      | 177 ± 69.0   | > 40       | 271 ± 153    | > 200      | 4          |
| Xinhe  | 14.4 ± 2.26| 10–20      | 207 ± 123    | > 40       | 331 ± 160    | > 200      | 4          |
| Shuxin | 14.9 ± 3.48| 10–20      | 243 ± 200    | > 40       | 269 ± 129    | > 200      | 4          |
| Bu     | 14.3 ± 1.87| 10–20      | 288 ± 225    | > 40       | 288 ± 152    | > 200      | 4          |
| Gangyan| 13.9 ± 3.17| 10–20      | 374 ± 308    | > 40       | 324 ± 128    | > 200      | 4          |
| Xianghua| 15.1 ± 3.39| 10–20     | 367 ± 84.6   | > 40       | 245 ± 181    | > 200      | 4          |
| Zhongxing| 13.3 ± 2.61| 10–20     | 896 ± 196    | > 40       | 259 ± 171    | > 200      | 4          |
| Chenjia | 13.8 ± 1.57| 10–20      | 407 ± 240    | > 40       | 158 ± 69.7   | 150–200    | 4          |
| Hengsha| 17.1 ± 2.44| 10–20      | 290 ± 242    | > 40       | 189 ± 62.8   | 150–200    | 4          |
| Changxing| 15.6 ± 1.16| 10–20     | 221 ± 120    | > 40       | 121 ± 39.2   | 100–150    | 4          |
| Shanghai| 15.2 ± 2.37| 10–20    | 295 ± 114    | > 40       | 159 ± 51     | 150–200    | 4          |

### Table 2

|        | Cr    | Cu    | Zn    | As    | Pb    |
|--------|-------|-------|-------|-------|-------|
| Mean   | 64.5  | 31.9  | 86.0  | 11.7  | 22.5  |
| Maximum| 189   | 56.2  | 179   | 20.0  | 69.3  |
| Minimum| 39.5  | 14.4  | 49.0  | 0     | 14.0  |
| 25th percentile| 52.4 | 22.1  | 71.6  | 3.14  | 21.2  |
| 50th percentile| 56.6 | 25.7  | 81.7  | 7.68  | 23.8  |
| 75th percentile| 61.7 | 31.3  | 93.7  | 10.5  | 26.7  |
| Mean backgrounds in China*| 61.0 | 22.6  | 74.2  | 11.2  | 26.0  |
| Chinese soil guidelines (level 1)| 90.0 | 35.0  | 100   | 15.0  | 35.0  |
| Chinese soil guidelines (level 2)| 200 | 200   | 250   | 30.0  | 300   |

*The background values in China were obtained from China National Environmental Monitoring Center (CNEMC, 1990). The Chinese soil quality categories are defined according to the report by Chinese Environmental Protection Administration (CEPA, 1995)
relatively lower corrected area of arable land, in comparison to their measured geometric area, was found in the towns of Hengsha, Shuxin, Jianshe, Xianghua, and Changxing (Fig. 3). For example, in Changxing, the corrected arable area accounted for only 90.6% of its original measured arable area because of its relatively low fertility and cleanliness indices (Figs. 1 and 3).

Discussions

A number of studies have reported the soil quality of arable land in Chongming District, mainly focusing on the distribution and quality assessment of heavy metals and dissolved OM (Lou et al. 2017; Sun et al. 2010; Wang et al. 2015; Zheng et al. 2016). The previous studies together with the results obtained in the present study show that the soil quality of the agricultural land in Chongming District is generally good (Zhang et al. 2014; Zheng et al. 2016; this study). However, sustained attention and management is still necessary due to the potential risk of heavy metal accumulation and soil degradation problems (Zhang et al. 2014; Table 2 in this study).

Due to rapid economic development, soil pollution by heavy metals has been widespread in China since the late 1970s (Chen et al. 1999). In general, in this study, we found that the heavy metal levels in Chongming District were good, with most of heavy metals (based on mean values of each town)
belonging to the level 1 category of Chinese soil guidelines (CEPA 1995). However, with large variations, the mean concentrations ($n = 104$) of Cr, Cu, Zn, and As were higher than those of the mean backgrounds in China (see Tables 2 and 3). Moreover, the mean concentrations of Cr, Cu, Zn, and As found in the present study were relatively higher than those taken in the previous studies within the same area (Wang et al. 2007; Zheng et al. 2016; Table 3). These results indicate that there is a possibility that the concentration of heavy metals and As has accumulated in recent years. The over-application of pesticides and stubble burning may partly explain the accumulation of heavy metals in agricultural soils in Chongming District (Sun et al. 2010).

As the quality of arable soil changes with agricultural practices and anthropogenic activities, the evaluation system for arable land resources also needs development and renewal with time. The multi-criteria evaluation system integrating the IFI and the soil cleanliness index ($K$) proposed in this study provides a new evaluation method for the arable land resources. When the proposed system was applied to evaluate arable land based on town unit in Chongming District, the CSEI values of the 16 towns were found to range from 90.6 to 99.2 (Fig. 1). The town of Changxing, which had the lowest CSEI value (90.6), was also found to have a low K value of 86.6 (Fig. 1). In this study, in order to calculate the IFI, three parameters (TOM, Av-P, and Av-K) were selected, of which Av-P and Av-K were indicators of nutrient status and TOM influenced the biological activities in the soil habitat. Besides the IFI, the introduction of the soil cleanliness index made the comprehensive evaluation of arable land more reliable. The order of the 16 towns based on the multi-criteria evaluation system (CSEI value) was different from the orders based solely on the IFI or soil cleanliness index, indicating the

| Sampling site (agriculture soils) | Cr        | Cu     | Zn     | As        | Pb        |
|----------------------------------|-----------|--------|--------|-----------|-----------|
| This study                       | 43.8–189  | 20.6–39.5 (31.9 ± 7.56) | 67.8–111 (86.1 ± 19.5) | 0–21.6 (11.7 ± 5.0) | 10.8–34.4 (22.5 ± 7.66) |
| Kermanshah, Iran (Doabi et al. 2019) | 32.0–235  | 10.0–83.0 (46.5 ± 36.5) | 40.0–113 (76.5 ± 36.5) | ND        | ND        |
| Pakhtunkhwa, Pakistan (Khan et al. 2013) | 0.29–0.64 (0.47 ± 0.18) | 0.28–0.61 (0.45 ± 0.17) | 0.20–0.52 (0.36 ± 0.16) | ND        | ND        |
| Telangana, India (Adimalla et al. 2019) | 55.9–135.8 (95.9 ± 40.0) | 12.7–69.6 (41.2 ± 28.5) | 71.3–173 (122 ± 50.9) | 2.40–5.3 (3.85 ± 1.45) | 5.90–26.8 (16.4 ± 10.5) |
| Morocco (Oumenskou et al. 2018) | 16.1–294 (155 ± 139) | 1.46–191 (96.3 ± 95) | 24.5–1272 (648 ± 624) | ND        | 3.40–135 (69 ± 66) |
| Colombia, America (Marrugo-Negrete et al. 2017) | 0.01–0.08 (0.045 ± 0.035) | 12.6–2522 (1267 ± 1257) | 285–2632 (1459 ± 1174) | ND        | 0.02–0.13 (0.075 ± 0.055) |
| Odo-Oba, Nigeria (Adagunodo et al. 2018) | 23.0–341 (182 ± 159) | 3.91–20.7 (12.3 ± 8.39) | 22.8–61.3 (42.1 ± 19.3) | 1.60–3.70 (2.65 ± 1.05) | 19.0–43.9 (31.4 ± 12.5) |
| Serbia (Saljnikov et al. 2019) | 25.6–100 (62.6 ± 37.0) | 20.4–109 (64.8 ± 44.4) | 50.7–125 (87.9 ± 37.2) | 4.89–54.1 (29.5 ± 24.6) | 4.77–171 (88.1 ± 83.3) |
| Guangdong, China (Cai et al. 2019) | 5.70–57.1 (31.4 ± 25.7) | 1.20–48.6 (24.9 ± 23.7) | 25.1–106 (65.6 ± 40.5) | 1.80–25 (13.4 ± 11.6) | 25.6–84.9 (55.3 ± 29.7) |
| Sihui, Guangdong, China (Zhang et al. 2018) | ND | 4.60–62.3 (33.5 ± 28.9) | ND | 3.31–83.1 (43.2 ± 39.9) | 13.3–71.3 (42.3 ± 29) |
| Taiyuan, China (Liu et al. 2015) | 14.6–193 (104 ± 89) | 5.83–274 (140 ± 134) | 169–278.6 (148 ± 131) | 0.62–23.5 (12.1 ± 11.4) | 6.32–73.7 (40.0 ± 33.7) |

Values in the brackets are the mean ± standard deviations
ND not detected
A practical and reliable system for evaluating arable land resources requires the integrated consideration of key soil properties (e.g., fertility and contamination indicators) and their variations across space and time. However, the current arable land evaluation system does not cover all of these aspects (Cui et al. 2010; Wu et al. 2011). Productivity and quality indicators (e.g., fertility parameters) are currently the main parameters that can be used for land evaluation systems (Hao et al. 2015; Liu et al. 2014; Wang et al. 2014). The complex data acquisition process is another hurdle that exists in the poor implementation of current evaluation systems for arable land resource value (Duru et al. 2010; Torbert et al. 2009). For example, in many evaluation systems, data acquisition and processing often require large-scale instrumentation monitoring systems and large sample sizes, which local governments view as significant obstacles. Therefore, the present study’s emphasis on arable land ecology on a small regional scale, combining both the IFI and the soil cleanliness index, has made it feasible to evaluate land resources related to both land fertility and ecological quality.

Returning to the requisition–subsidy balance policy, farmers are the most basic interest group for the nonmarket value of arable land resources, so their needs should be given special attention. In the practice of arable land requisition and subsidy, the subsidy amount can be calculated based on the corrected land area and a predefined price for the ideal hectare. In other words, the subsidy can correspond to the ideal hectare of arable land. This type of subsidy can better reflect the quality of arable land and soils, encouraging the authorities and farmers to pay more attention to the protection of arable land and guide a more reasonable and reliable subsidy policy for arable land. In conclusion, this study provides an easy and effective method to measure arable soil quality and potentially guarantee the quality of agricultural products.

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References

Adagunodo, T. A., Summonu, L. A., & Emetere, M. E. (2018). Heavy metals’ data in soils for agricultural activities. Data in Brief, 18, 1847–1855. https://doi.org/10.1016/J.DIB.2018.04.115.

Adimalla, N., Qian, H., & Wang, H. (2019). Assessment of heavy metal (HM) contamination in agricultural soil lands in northern Telangana, India: an approach of spatial distribution and multivariate statistical analysis. Environmental Monitoring and Assessment, 191(126), 1–15. https://doi.org/10.1007/s10661-019-7408-1.

Alaoui, A., Schwilch, G., Geissen, V., & Fleskens, L. (2018). Heavy metals in agricultural soils from a typical township in Guangdong Province, China: occurrences and spatial distribution. Ecotoxicology and Environmental Safety, 168, 184–191. https://doi.org/10.1016/J.ECOENV.2018.10.092.

Adagunodo, T. A., Sunmonu, L. A., & Emetere, M. E. (2018). Heavy metals contamination features and health risk of soil heavy metals in agricultural soils of Kermanshah Province, Iran. Environment International, 101(2018), 1–15. https://doi.org/10.1016/J.ENVINT.2016.04.042.

Fazal, S. (2001). The need for preserving farmland: a case study from a predominantly agrarian economy (India). Landscape and Urban Planning, 55(1). https://doi.org/10.1016/S0169-2046(00)00134-1.

Fu, G. Z., & Bai, W. Q. (2015). Advances and prospects of evaluating cultivated land quality in China. Resources Sciences, 37(2), 226–236.

Hao, G., Yun, F., Fangfang, L., Yang, L., WenLi, Y., & Yin, Y. (2015). Soil diagnosis and land suitability assessment for vegetation restoration on coal waste piles in Liupanshui, Guizhou, China. International Journal of Mining, Reclamation and Environment, 30(3), 209–216. https://doi.org/10.1080/17480930.2015.1036519.

Hu, Y., Cheng, H., Tao, S., Zheng, R., Zhao, J., Zhou, X., et al. (2012). GIS and RS based assessment of cultivated land quality of Shandong Province. Environmental Monitoring and Assessment, 20(1), 359–369. https://doi.org/10.1007/s11368-014-0976-3.

Hu, X. F., Du, Y., Feng, J. W., Fang, S. Q., Gao, X. J., & Xu, S. (2013). Spatial and seasonal variations of heavy metals in wetland soils of the tidal flats in the Yangtze estuary, China: environmental implications. Pedosphere, 23(4), 511–522. https://doi.org/10.1016/S1002-0160(13)60044-2.

Hu, Y., Cheng, H., & Tao, S. (2016). The challenges and solutions for cadmium-contaminated rice in China: a critical review. Environment International, 92–93, 515–532. https://doi.org/10.1016/J.ENVINT.2016.05.014.

Hussain, I. (1997). Tillage effects on soil properties and crop production in southern Illinois. University of Illinois at Urbana-Champaign.

Jiao, X., Gao, C., Sui, Y., Lü, G., & Wei, D. (2014). Effects of long-term fertilization on soil carbon and nitrogen in Chinese mollisols. Agronomy Journal, 106, 1018–1024. https://doi.org/10.2134/agronj13.0233.

Khan, S., Cao, Q., Zheng, Y. M., Huang, Y. Z., & Zhu, Y. G. (2008). Health risks of heavy metals in contaminated soils and food crops irrigated with wastewater in Beijing, China. Environmental Pollution, 152(3), 686–692. https://doi.org/10.1016/j.envpol.2007.06.056.

Klingebiel, A. A., & Montgomery, P. H. (1961). Land-capability classification. State of New Jersey: soil conservation service. Soil Conservation Service, U.S. Department of Agriculture.

Kowalska, J., Mazurek, R., & Jurek, M. (2018). Heavy metal contamination in agricultural soils of Kermanshah Province, Iran. Environmental Earth Science, 78, 60–72. https://doi.org/10.1007/s12665-019-8099-0.

Liu, Z., Zhou, W., Shen, J., Li, S., & Ai, C. (2014). Soil quality assessment of yellow clayey paddy soils with different
productivity. *Biologia et Fertility of Soils*, 50(3), 537–548. https://doi.org/10.1007/s00374-013-0864-9.

Liu, Y., Huifeng, W., Li, X., & Jinchang, L. I. (2015). Heavy metal contamination of agricultural soils in Taiyuan, China. *Pedosphere: An International Journal*, 25(6), 901–909. https://doi.org/10.1016/S1002-0160(15)30070-9.

Lou, S., Liu, S., Dai, C., Tao, A., Tan, B., Ma, G., et al. (2017). Heavy metal distribution and groundwater quality assessment for a coastal area on a Chinese Island. *Polish Journal of Environmental Studies*, 26(2), 733–745. https://doi.org/10.15244/pjoes/67064.

Lu, J., Li, L., Wu, K., Zhao, H., & Huang, Q. (2011). Cultivated land comprehensive quality evaluation based on agricultural land classification and soil environmental quality evaluation (in Chinese). *Transactions of the CSAE*, 27(2), 323–329.

Ma, C., Zheng, R., Zhao, J., Han, X., Wang, L., Gao, X., & Zhang, C. (2015). Relationships between heavy metal concentrations in soils and reclamation history in the reclaimed coastal area of Chongming Dongtan of the Yangtze River estuary, China. *Journal of Soils and Sediments*, 15(1), 139–152. https://doi.org/10.1007/s11368-014-0976-3.

Marrugo-Negrete, J., Pinedo-Hernández, J., & Díez, S. (2017). Assessment of heavy metal pollution, spatial distribution and origin in agricultural soils along the Sinú River Basin, Colombia. *Environmental Research*, 154, 380–388. https://doi.org/10.1016/j.envres.2017.01.021.

MEP (Ministry of Environment Protection of China). (2014). *Soil pollution status survey bulletin*. Ministry of Agriculture. (2004). Determination of exchangeable potassium and non-exchangeable potassium content in agricultural industry standard, NY/T 889-2004.

Ministry of Agriculture. (2006). Soil testing—part 6: method for determination of organic matter contents in agricultural industry standard, NY/T 1121.6-2006.

Ministry of Agriculture. (2014). Soil testing—part 7: method for determination of available phosphorus in agricultural industry standard, NY/T 1121.7-2014.

Mu, H., Fu, S., Liu, B., Yu, B., & Wang, A. (2018). Influence of soil and water conservation measures on soil fertility in the Beijing mountain area. *Environmental Monitoring and Assessment*, 190(9), 504. https://doi.org/10.1007/s10661-018-6836-7.

Nanjing Agriculture University. (1996). *Soil agrochemical analyses (in Chinese)*. Beijing: China Agriculture University.

National Environmental Protection Bureau. (1995). Environmental quality standard for soils GB 15618-1995.

National Soil Survey Office. (1979). *Provisional technical regulations of the second national soil survey (in Chinese)*. Beijing: China Agriculture University.

Oumenskou, H., El Baghdadi, M., Barakat, A., Aquit, M., Ennaji, W., Karroum, L. A., & Aadaouit, M. (2018). Assessment of the heavy metal contamination using GIS-based approach and pollution indices in agricultural soils from Beni Amir irrigated perimeter, Tadla plain, Morocco. *Arabian Journal of Geosciences*, 11(22), 692–618. https://doi.org/10.1007/s12517-018-4021-5.

Saljnikov, E., Mrvić, V., Čakmak, D., Jaramaz, D., Perović, V., Antić-Mladenović, S., & Pavlović, P. (2019). Pollution indices and sources appointment of heavy metal pollution of agricultural soils near the thermal power plant. *Environmental Geochemistry and Health*, 41, 1–15. https://doi.org/10.1007/s10653-019-00281-y.

Shang, Q., Ling, N., Feng, X., Yang, X., Wu, P., Zou, J., Shen, Q., & Guo, S. (2014). Soil fertility and its significance to crop productivity and sustainability in typical agroecosystem: a summary of long-term fertilizer experiments in China. *Plant and Soil*, 381(1–2), 13–23. https://doi.org/10.1007/s11104-014-2089-6.

She, L. M., & Xie, B. G. (2000). Some considerations about the dynamic balance of the total cultivated land quantity in China. *Research of Agricultural Modernization*, 21(2), 87–90.

Stenberg, B. (1999). Monitoring soil quality of arable land: microbiological indicators. *Acta Agriculturae Scandinavica Section B: Soil and Plant Science*, 49(1), 1–24. https://doi.org/10.1080/09064719950135669.

Sun, C., Bi, C., Chen, Z., Wang, D., Zhang, C., Sun, Y., Yu, Z., & Zhou, D. (2010). Assessment on environmental quality of heavy metals in agricultural soils of Chongming Island, Shanghai City. *Journal of Geographical Sciences*, 20(1), 135–147. https://doi.org/10.1007/s11442-010-0135-8.

Tan, M., Li, X., Xie, H., & Lu, C. (2005). Urban land expansion and arable land loss in China—a case study of Beijing-Tianjin-Hebei region. *Land Use Policy*, 22(3), 187–196. https://doi.org/10.1016/j.landusepol.2004.03.003.

Tania, D., Lopez, T., Aide, M., & John, R. (2001). Urban expansion and the losses of prime agricultural lands in Putero Rico. *AMBIO*, 30, 49–54.

Teng, Y., Ni, S., Wang, J., Zuo, R., & Yang, J. (2010). A geochemical survey of trace elements in agricultural and non-agricultural topsoil in Daxing area, China. *Journal of Geochemical Exploration*, 104(3), 118–127. https://doi.org/10.1016/j.jgeexp.2010.01.006.

Tian, Y. G., & Xin, J. S. (2006). *Guidelines for evaluation of cultivated land fertility (in Chinese)*. Beijing: China Agricultural Science and Technology.

Torbert, H. A., Yakushev, V., Krueger, E., & Kurtener, D. (2009). Evaluation of resources of agricultural lands using fuzzy indicators. *Journal of Agricultural Machinery Science*, 5(4), 469–476.

Wang, J., Zhang, F., Wang, R., Jia, X., & Zhang, C. (2001). Application of integrated fertility index method in evaluating changes in soil fertility. *Rural Eco-Environment*, 17(3), 13–16.

Wang, J., Chen, Z. L., & Wang, C. (2007). Heavy metal content and ecological risk warning assessment of vegetable soils in Chongming Island, Shanghai City (in Chinese). *Environmental Science*, 28(3), 647–653.

Wang, X., Dong, S., Yang, B., Li, Y., & Su, X. (2014). The effects of grassland degradation on plant diversity, primary productivity, and soil fertility in the alpine region of Asia’s headwaters. *Environmental Monitoring and Assessment*, 186(10), 6903–6917. https://doi.org/10.1007/s10661-014-3898-z.

Wang, Y. L., Yang, C. M., Zou, L. M., & Cui, H. Z. (2015). Spatial distribution and fluorescence properties of soil dissolved organic carbon across a riparian buffer wetland in Chongming island, China. *Pedosphere*, 25(2), 220–229. https://doi.org/10.1016/S1002-0160(15)60007-8.

Wu, Z. J., Wei, C. F., & Shang, H. (2011). Research on the Economic Output Value of Cultivated Land Based on
Peasant Household Panel (In Chinese). Economic Geography, 31(9):1516-1522.

Yang, S., Zhou, D., Yu, H., Wei, R., & Pan, B. (2013). Distribution and speciation of metals (Cu, Zn, Cd, and Pb) in agricultural and non-agricultural soils near a stream upriver from the Pearl River, China. Environmental Pollution, 177, 64–70. https://doi.org/10.1016/J.ENVPOL.2013.01.044.

Zhang, X., Yin, S., Li, Y., Zhuang, H., Li, C., & Liu, C. (2014). Comparison of greenhouse gas emissions from rice paddy fields under different nitrogen fertilization loads in Chongming Island, Eastern China. Science of the Total Environment, 472, 381–388. https://doi.org/10.1016/J.SCITOTENV.2013.11.014.

Zhang, J., Li, H., Zhou, Y., Dou, L., Cai, L., Mo, L., & You, J. (2018). Bioavailability and soil-to-crop transfer of heavy metals in farmland soils: a case study in the Pearl River Delta, South China. Environmental Pollution, 235, 710–719. https://doi.org/10.1016/J.ENVPOL.2017.12.106.

Zhao, Q., Wang, Y., Cao, Y., Chen, A., Ren, M., Ge, Y., et al. (2014). Potential health risks of heavy metals in cultivated topsoil and grain, including correlations with human primary liver, lung and gastric cancer, in Anhui Province, Eastern China. Science of the Total Environment, 470–471, 340–347. https://doi.org/10.1016/J.SCITOTENV.2013.09.086.

Zheng, R., Zhao, J., Zhou, X., Ma, C., Wang, L., & Gao, X. (2016). Land use effects on the distribution and speciation of heavy metals and arsenic in coastal soils on Chongming Island in the Yangtze River estuary, China. Pedosphere, 26(1), 74–84. https://doi.org/10.1016/S1002-0160(15)60024-8.

Zhu, H., Hu, W., Bi, R., Peak, D., & Si, B. (2016). Scale- and location-specific relationships between soil available micronutrients and environmental factors in the Fen River basin on the Chinese Loess Plateau. CATENA, 147, 764–772. https://doi.org/10.1016/J.CATENA.2016.08.038.

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