

# Heavy Metals in Water and Surface Sediments of the Fenghe River Basin, China: Assessment and Source Analysis

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## Research Article

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# **Heavy metals in water and surface sediments of the Fenghe River Basin, China: Assessment and Source Analysis**

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13 **Abstract**

14 This paper combines environmental science, inorganic chemistry, water quality  
15 monitoring and other disciplines, and uses several representative evaluation methods  
16 (WQI, Pn, I-geo, RI) for heavy metals in water and sediments. A preliminary  
17 assessment and source analysis of heavy metals (Cd, Cr, Fe, Mn, Zn, Cr, Ti, Ni, Cu,  
18 As, Pb, Sr) in water and surface sediments of the Fenghe River Basin, Shannxi  
19 Province, China was carried out in this study. Results indicate that most of the heavy  
20 metals in water are below national water quality standards. Exceptions include Mn,  
21 which exceeds national tertiary standards and Cr, which exceeds national drinking  
22 water standards. Most heavy metals in the sediments exceed the environmental  
23 standard values except Ni. Water quality index (WQI) and Nemero index (Pn) showed  
24 the same trend in contamination levels of sampling sites. According to the Geological  
25 Accumulation Index method (I-geo) and the Potential Ecological Risk Index method  
26 (RI), high concentrations of Cd poses a high ecological risk in some sampling  
27 locations. Pearson Correlation Analysis (CA), Hierarchical Clustering Analysis  
28 (HCA), Principal Component Analysis (PCA) and Positive Matrix Factorization  
29 (PMF) models are used to explore the relationships and sources of heavy metals. In  
30 general, upstream sources are similar, and middle and lower reaches are easily  
31 clustered into a large category except for some specific sampling points. For example,  
32 metals in sampling site FHK mainly come from surrounding residents and farms and  
33 heavy metals attributes in sampling site SLQ relate to the fact that municipal sewage  
34 is collected and treated. The factors or sources of heavy metals in water and sediment

35 are revealed in detail through PMF models. In the water, the average contribution rate  
36 of these four source factors for heavy metals is 36.8%, 11.7%, 9.4% and 42.0%, while  
37 the average proportion of these four factors for heavy metals in sediment is 8.0%,  
38 29.2%, 23.9% and 38.9% respectively. Results show that the main sources of  
39 pollution in the region are urban construction and transportation, electronics industry,  
40 machinery manufacturing, tourism and agriculture. These sectors should therefore be  
41 given sufficient attention in the prevention and management of heavy metal pollution.

42 **Keywords:** Fenghe River Basin, heavy metal, surface water and sediment,  
43 assessment, source pollution

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## 48 **1. Introduction**

49 Heavy metals, as common pollutants in the water environment, are toxic, persistent  
50 and bio-accumulative (Pekey et al. 2004). Excessive doses of heavy metals threaten  
51 the environment and humans through direct exposure and food chain enrichment  
52 (Zhang et al. 2016; Lin et al. 2016). Sediments are sources and sinks of heavy metals.  
53 Factors such as temperature, pH, dissolved oxygen, redox potential and electrical  
54 conductivity control the release of heavy metals from sediments, causing secondary  
55 pollution to water quality (Lin and Chen 1998; Bertin and Bourgm 1995; Li et al.  
56 2014; Ndimele 2012). Heavy metal pollution is caused by natural factors, such as  
57 weathering and riverbed erosion, and human activities, especially mining,  
58 mechatronic industries, urban construction, urban flood, urban transportation, and  
59 agriculture (Mohamed 2007; Mu et al. 2020; Ke et al. 2015; Zhuang and Gao 2015).  
60 Comprehensive analysis of status, potential risks and sources of heavy metals in water  
61 and sediments is therefore essential for environmental control and management.

62 Heavy metal pollution in rivers has aroused widespread concern due to the rapid  
63 development of society (Li and Zhang 2010). By the mid-19th century, about 40% of  
64 Britain's rivers and lakes had been polluted by human behavior (Li et al. 2014).  
65 Approximately 85% of heavy metals are ultimately enriched in surface sediments on a  
66 global scale (Zahra et al. 2014). In Europe, North America, Africa and Asia, heavy  
67 metal pollution has been studied extensively, especially the rivers Rhine and Meuse in  
68 Western Europe (Wijnhoven et al. 2006), the Danube River (Sunjog et al. 2012), the  
69 Mississippi River in America (Grabowski et al. 2001), the Nile River in Egypt

70 (Elbouraie et al. 2010) , the Soan in Pakistan (NAZEER et al. 2014), the  
71 Subarnarekha River in India (Giri and Singh 2014), the Yellow River (Sun et al. 2016)  
72 and the Yangtze Rivers (Guo and Yang 2016) in China. In China, about 6 of the 21  
73 major cities along the Yangtze River (Panzhuhua, Yichang, Nanjing, Wuhan, Shanghai,  
74 Chongqing) have a pollution rate of heavy metals of 65% (Zhang and Shu 2010).

75 Faced with such serious heavy metal pollution, a great deal of research has assessed  
76 aspects of contamination, such as migration and transformation processes,  
77 bioaccumulation, and toxicity (Pandey and Bhattacharya 2016; Ogendi et al. 2010;  
78 Simpson and Spadaro 2016). However, such studies require complex laboratory  
79 analysis which affect data accuracy and precision, are time intensive, and are  
80 inapplicable at large scales. The total concentration of heavy metals can be used to  
81 evaluate the status of heavy metal pollution, potential ecological risks and sources  
82 (Villanueva and Ibarra 2016). In addition to comparison with national water quality  
83 standards, evaluation methods for heavy metals in water are well-established and  
84 include the single factor water quality index method (Pi) and multi-factor Water  
85 Quality Index method (WQI) (Cheng and Dan 2011; Parparov et al. 1992), and the  
86 Nemero index method (Pn) (Tianxiang et al. 2018). Similarly, there are many methods  
87 for evaluating heavy metals in sediments, including representative methods such as  
88 the Enrichment Factor method (EF) (Ergin et al. 1991), the Geological Accumulation  
89 Index method (I-geo) (Muller 1969), the Pollution Load Index method (PLI) (Angulo  
90 1996), the Potential Ecological Risk Index method (RI) (Hakanson 1980) and the  
91 Sediment Quality Guidelines (SQG) (Macdonald et al. 2000). The Composite Element

92 Index has a cooperativity effect and can comprehensively evaluate regional pollution  
93 status and potential risks. Thus, WQI, Pn, I-geo, RI were selected for this study.

94 In the source analysis of heavy metals, Correlation Analysis (CA) is used to  
95 determine significant correlations between heavy metals through the correlation  
96 coefficient (Zhang et al. 2015). The greater absolute value of the correlation indicates  
97 a higher correlation between the two variables. Hierarchical Cluster Analysis (HCA)  
98 is based on the degree of similarity between variables, which is classified into  
99 categories (Facchinelli et al. 2001). Principal Component Analysis (PCA) reduces  
100 dimensionality in data to uncover trends (Kaidao et al. 2012). CA, HCA, and PCA can  
101 be used in combination to assess contaminant sources (Comero et al. 2011). However,  
102 the data required by software (SPSS) is sensitive and needs to be standardized  
103 (Pornsawai et al. 2013). The Positive Matrix Factorization (PMF) model (Anttila et al.  
104 1995) is applied to study the contribution rate of each source more accurately and  
105 quantitatively. An advanced statistical method, PMF uses the least square method to  
106 estimate each value and analyze the uncertainty of each value. Although this model  
107 has been widely used in atmosphere studies (Jaeckels et al. 2007; Saraga et al. 2010;  
108 Yuan et al. 2012), application to pollution in water and sediments is rare.

109 Representative sampling points were selected to systematically study the heavy  
110 metal pollution in the Fenghe River Basin (FRB). The main goals of this study are (1)  
111 to describe the physical and chemical parameters and the concentration of heavy  
112 metals in water and surface sediments, (2) to comprehensively evaluate the pollution  
113 status of heavy metals in the water through the comprehensive Water Quality Index

114 (WQI) and the Nemer Index (Pn), as well as the accumulation status and potential  
115 risks of heavy metals in sediments through the Geological accumulation Index (I-geo)  
116 and Potential Risk Index (RI) methodology, and (3) to qualitatively and quantitatively  
117 analyze the potential sources and contribution rates of heavy metal pollution by  
118 multivariate statistical analysis and the Positive Matrix Factorization (PMF) model.

## 119 **2. Materials and methods**

### 120 **2.1 Study area and sample collection**

121 As the first tributary of the south bank of the Wei River, the largest tributary of  
122 the Yellow River, the Fenghe River has a total length of 78km and a watershed area of  
123 1460km (Huaien et al. 2016). Its main tributaries include the Taipingyu, Fengyu,  
124 Gaoguanyu and the Yu River. Urban expansion and population growth have led to the  
125 implementation of the Hei River Water Diversion Project (Zhang et al. 2018; Wang  
126 and Tang 2012). As a result of upstream tourism, the construction of towns in the  
127 middle and lower reaches and industrial development have led to deterioration of  
128 water quality and ecology in the Fenghe River (Zhang et al. 2014).

129 Eight monitoring stations were investigated in the FRB: Fengyu (FY),  
130 Gaoguanyu (GGY), Taipingyu (TPY), Yurufeng (YRF), Fenghekou (FHK),  
131 Qinduzhen (QDZ), Yanjiaqu (YJQ), and Sanliqiao (SLQ). The layout of sampling  
132 points in the study area is given in **Fig. 1**. Monitoring stations cover representative  
133 control sections of hydrological monitoring stations and major tributary intersections  
134 from upstream to downstream. As three stations consist of sand and gravel, sediment



135 samples were collected from five sampling locations of the Fenghe River in August,  
136 2018. Water samples were collected from eight sampling locations in October and  
137 December 2018. We use the average concentration of heavy metals in water from two  
138 months.

## 139 **2.2 Sample collection and analysis**

140 In this study, water (USEPA 2013) and sediment (USEPA 2014) methods were  
141 used to collect samples. An YSI multi-parameter water quality analyzer (Germany)  
142 analyzed the physicochemical properties (temperature (°C), pH, dissolved oxygen  
143 (DO), redox potential (ORP), electrical conductivity (EC)) of water. Samples (500 ml)  
144 were collected with polyethylene plastic bottles and sealed, and then brought back to  
145 the laboratory under 4 °C. Water samples were filtered through a 0.45µm filter  
146 membrane and put in a 10 ml centrifuge tube. The water samples were centrifuged  
147 using a low-speed centrifuge for 10 min under 3500 r/min, and after 2% HNO<sub>3</sub> was  
148 added and refrigerated for preservation under 4 °C. Surface sediment samples were  
149 sealed in polythene bags and returned to the laboratory. Samples were put in an oven  
150 at 105 °C for 12 h after natural drying, lightly crushed, sieved through 200-mesh  
151 nylon sieve and sealed in a polyethylene bag.

152 Pretreatment of sediment samples: The 50mg samples were weighed and placed  
153 in polytetrafluoroethylene tanks. Then, 0.5 ml HNO<sub>3</sub> and 1 ml HF were added into the  
154 tanks and the samples were put on an electric heating plate for 145 °C to remove  
155 silicon until the samples were dry. In addition, 1 ml HNO<sub>3</sub> and 1 ml HF were  
156 continuously added into the tanks for 5 h under 145 °C. The treated samples were

157 cooled to room temperature overnight and then steamed for 40 min. A few drops of  
158 H<sub>2</sub>O<sub>2</sub> were added to remove the organic matter and 1 ml HNO<sub>3</sub> was added to steam.  
159 Finally, 2.5ml 40% HNO<sub>3</sub> were added into the samples for 4h, and then the samples  
160 were fixed to 50ml with 40% HNO<sub>3</sub>. Total heavy metal content in the treated samples  
161 was determined by Inductively Coupled Plasma Mass Spectrometer (ICP-MS).

162 All reagents used in the experiment were analytical reagents. Calibration curves  
163 were determined with correlation coefficients in the range of 0.9997-0.9999. In order  
164 to ensure the accuracy and precision of data, the standard solution (GSB04-1767-  
165 2004) was provided by the national nonferrous metals and electronic materials  
166 analysis and testing center for quality control. Standard recovery rates ranged from  
167 87.53% to 102.29%.

## 168 **2.3 Evaluation method**

169 Evaluation standards for heavy metals in water refers to the surface water  
170 environmental quality standard GB3838-2002. The Water Quality Index method  
171 (WQI), Nemero Pollution Index method (Pn), Geological Accumulation Index method  
172 (I-geo) and Potential Ecological Risk Index method (RI) were selected to evaluate the  
173 heavy metal pollution situation in the FRB.

### 174 **2.3.1 Water quality index method and Nemero index**

175 The water quality index is divided into "single factor pollution index method"  
176 and "comprehensive pollution index method". The former indicates the pollution level  
177 of a single heavy metal, while the latter takes into account the synergistic effects of  
178 pollutants (Duodu et al. 2016; Yan et al. 2016). The calculation formula was as

179 follows (1) (2):

180 (1) Single factor pollution index:  $P_i = \frac{C_i}{Q_i}$

181 Where  $C_i$  is the heavy metal measured concentration;  $Q_i$  the reference  
182 value of the element.

183 (2) Comprehensive pollution index:  $WQI = \frac{1}{n} \sum_i^n P_i$

184 Where  $P_i$  is the single heavy metal pollution index;  $n$  types of elements.

185 The Nemeru comprehensive pollution index method can reflect current heavy metal  
186 pollution in water and the different contributions of various heavy metals. The  
187 calculation formula was as follows (3):

188 (3)  $P_n = \sqrt{\frac{\max(P_i)^2 + \text{ave}(P_i)^2}{2}}$

### 189 2.3.2 Geological accumulation index method

190 Geological accumulation index method (I-geo) (Muller 1969) is widely used to  
191 evaluate heavy metal pollution in sediments. It reflects both the natural variation  
192 characteristics of heavy metal distributions, but also identifies the impact of human  
193 activities on the environment. The equation is described as below:

194  $I\text{-geo} = \log_2 [C_i / (k * B_i)]$

195 Where:  $C_i$  is the heavy metal measured concentration;  $B_i$  is the geochemical  
196 background value of the heavy metal;  $k$  is the diagenetic coefficient, the value is  
197 taken to be 1.5 in order to explain the possible changes in the environmental  
198 background values;  $I\text{-geo}$  is the geological accumulation index.

### 199 2.3.3 Potential ecological risk index method

200 The index method of potential ecological risk<sup>33</sup> (Hakanson 1980)

201 comprehensively considers the ecological, environmental and toxicological effects  
202 of heavy metals and is calculated as follows:

$$E_r^i = T_r^i \times (C^i / C_n^i)$$
$$RI = \sum_i^n E_r^i$$

205 Where  $C_i$  is the heavy metal measured concentration;  $C_n^i$  is the reference value  
206 of the element;  $T_r^i$  is the toxic reaction coefficient of each element (Cu=Pb=Ni=5,  
207 Cd=30, Cr=2, Zn=1, Ti=1, Mn=2) (Hakanson 1980; Sijin et al. 2015; Islam et al.  
208 2015);  $E_r^i$  is the single element potential ecological risk factor. Classification  
209 standards of WQI, I-geo, and RI are presented in **Table1**.

#### 210 **2.4 Statistical analysis**

211 IBM SPSS Statistic 22.0 was used for statistical analysis. Pearson Correlation  
212 Analysis (CA) is used to analyze the correlation between the various elements.  
213 Hierarchical Clustering Analysis (HCA) is used to group the sampling points  
214 according to the metal concentration of the sampling points. Variables with close  
215 distance were clustered first, followed by the variables with far distances until each  
216 variable is in an appropriate class. Intergroup link method and square Euclidean  
217 distance, recognized as the most stable method of systematic clustering analysis, are  
218 used in this study. Principal Component Analysis (PCA) focuses on explaining the  
219 total variance of each variable through dimensionality reduction (Shin and Lam  
220 2001; Tetsuro et al. 2009). In this study, Principal Component Analysis (PCA)  
221 explores the relationship between elements by extracting a small number of potential  
222 factors and analyzes the similarity of distribution sources of heavy metals.

## 223 2.5 Positive matrix factorization (PMF)

224 Positive Matrix Factorization (PMF), a multivariable factor analysis method  
225 (Paatero and Tapper 1993), is used to analyze the sources of heavy metals in this  
226 study (Peng et al. 2016). As the original sample data matrix  $X(x \times j)$  was decomposed  
227 into the product of factor contribution matrix  $G(i \times k)$  and factor profile matrix  $F(k \times$   
228  $j)$  by the PMF model, as well as the sum of residuals matrix  $E(i \times j)$ . The calculation  
229 equation is as follows:

$$230 \quad X = G \times F + E$$
$$231 \quad x_{ij} = \sum_{k=1}^p g_{ik} \times f_{kj} + e_{ij}$$

232 Where,  $x_{ij}$  is the concentration of sample component  $j$  at the  $i$  sample point  
233 (mg/kg);  $g_{ik}$  is the source  $k$  contribution concentration at  $i$  sampling point (mg/kg);  
234  $f_{kj}$  represents the contribution of source  $k$  to sample component  $j$ ;  $e_{ij}$  represents the  
235 residual, and  $p$  represents the number of source factors.

236 PMF model is iterated by the least square method, and the original matrix  $X$  is  
237 split to obtain the optimal matrix  $G$  and  $F$ , so that the objective function  $Q$   
238 approaches the degree of freedom value, namely  $i \times j$ .  $Q$  is defined as follows:

$$239$$
$$240 \quad Q = \sum_{i=1}^n \sum_{j=1}^m (e_{ij}/u_{ij})^2$$

241 Where,  $u_{ij}$  represents the uncertainty of sample component  $j$ , and  $e_{ij}$  represents  
242 the residual.

## 243 3. Results and discussion

### 244 3.1 Water quality parameters

245 The physical and chemical parameters of water, such as temperature (°C), pH,  
246 dissolved oxygen (DO), redox potential (ORP), and electrical conductivity (EC) are  
247 presented in **Table 2**. Physical and chemical parameters are important to understand  
248 given their role in supporting aquatic life and environmental health. The temperature  
249 range (taken in winter) is from 6.55 to 15.05°C. Sampling sites FY and GGY have  
250 lower values than other sites due to the upstream location and high altitude. The  
251 values of pH ranged from 7.88 to 8.49 (**Table 2**), within limits prescribed by GB3838-  
252 2002. The water quality in the Fenghe River is slightly alkaline. The average range for  
253 DO is between 5.24 and 11.25 mg/l (**Table 2**). Variability in DO may be due to  
254 aerobic and organic decomposition of aquatic organisms. Redox potential is a  
255 measurement index of water redox capacity. Higher values indicate higher  
256 oxidizability. The redox potential ranged from 138.62 to 176.77mV. The electrical  
257 conductivity represents the content of soluble impurities in water, and higher values  
258 represent worse water quality. The lowest value of 0.07 (ms /cm) for electrical  
259 conductivity is located in sampling site FY, which is in the upstream, while the  
260 highest value of 0.95 (ms /cm) appears in sampling site YRF, which may be the result  
261 of discharge from surrounding enterprises and residential wastewater.

### 262 **3.2 Heavy metal concentration in water and sediment**

263 The concentrations of heavy metals in water are shown in **Table 3**. At present,  
264 there are some studies on heavy metals in water, possibly because they precipitate  
265 quickly and tend to accumulate in sediments (Simpson and Batley 2010). However,

266 once secondary contamination occurs, trace concentrations may endanger aquatic  
267 organism, accumulate in the foodweb, and eventually pose a threat to human health  
268 (Salem et al. 2014). The average concentration of heavy metals in water decreases  
269 according to: Sr >Fe> Mn> Zn> Cr >Ti >Ni >Cu> As> Pb >Cd. The highest  
270 concentration of Sr was 322.85 ug/L in sampling site FHK and the lowest  
271 concentration of Sr was 54.12 ug/L in sampling site FY. The average concentration of  
272 Fe was observed with the value of 7.29 ug/L for sampling site FY and 269.15 ug/L for  
273 sampling site TPY. The highest concentration of Fe is slightly below Class I water  
274 quality standards and within the drinking water standard range. The concentration of  
275 Mn ranges from 0.8 ug/L to 239.8 ug/L. The concentrations of Mn in sampling sites  
276 TPY, FHK and QDZ exceeded the three standards of surface water (GB3838-2002).  
277 The average concentration of Zn was between 7.34 and 19.86 ug/L. Zn can be used as  
278 a nutrient at low concentrations, but exceeding the threshold can cause toxicity to  
279 aquatic organisms. Interestingly, the highest value of Zn was observed at sampling site  
280 YRF site, which might be attributed to wastewater discharge from an electronics  
281 factory. The concentrations of Cr were uniformly distributed, ranging from 6.13 to  
282 7.56 ug/L, which were lower than the Class I surface water standards (GB3838 2002)  
283 but higher than the WHO standard (5 ug/L) (**Table 3**). The highest concentration of  
284 Ti, Ni and Pb were 8.54, 3.95, 1.51 ug/L, respectively. They are all observed in  
285 sampling site TPY, probably because it is surrounded by residential areas, machinery  
286 and equipment companies. The concentration range of Cu (1.54-2.92 ug/L), As (0.2-  
287 2.66 ug/L), and Cd (0.03-0.1 ug/L) are lower than Class I water quality standards

288 (GB3838 2002). The highest concentrations of heavy metals are in the middle reaches  
289 of the river, except for Cd. Cd is a highly mobile and toxic element that is easily  
290 released into the environment (Li et al. 2012). In general, concentrations of heavy  
291 metals in water are not high in the downstream, such as sampling site SLQ, due to  
292 pollution source control during recent sponge city development. In general,  
293 concentrations of heavy metals in water are lower than in sediment, possibly due to  
294 the fluidity and dilution of the water (Mohiuddin et al. 2012).

295 The concentration of heavy metals in sediment is shown in **Table 4**. The average  
296 concentration of heavy metals are between 615.25-736.40 mg/kg for Mn, 221.50-  
297 409.00 mg/kg for Sr, 182.20-281.30 mg/kg for As, 74.33-125.10 mg/kg for Zn, 50.66-  
298 79.01mg/kg for Cr, 26.56-33.33 mg/kg for Pb, 18.21-48.10 mg/kg for Cu, 21.06-34.01  
299 mg/kg for Ni, 0.18-0.48 mg/kg for Cd. They decreased successively according to:  
300 Mn> Sr> As >Zn> Cr >Pb> Cu> Ni >Cd. Average concentration of these elements  
301 (Mn, Sr, As, Zn, Cr, Pb, Cu, Cd) exceeds background values of Shaanxi soils. Ni is  
302 lower than the soil environmental background value of 28.8 mg/kg in Shannxi  
303 province, but higher than the average value 25.6 mg/kg for the Wei River (**Table 4**).  
304 The excessive multiples of heavy metal concentration for the maximum point position  
305 concentration of Mn, Sr, As, Zn, Cr, Pb, Cu, Cd are 0.32, 1.46, 24.34, 0.80, 0.26, 0.56,  
306 1.25, 4.14. In addition to the proportion exceeding the standard value of heavy metal  
307 concentration for sampling points were 40% for Cr, Ni and 60% for Cu, as well as  
308 100% for Mn, Sr, As, Zn, Pb, Cd. Pb and As are above average shale values (**Table 4**).  
309 The maximum values of Mn, As, Zn, Cr, Pb, Cu and Ni were found at the sampling



310 site FHK, Ti and Cd in sampling site SLQ, and Sr in sampling site TPY. The FHK  
311 sampling site is close to farmland, residential areas and surrounding materials  
312 technology companies, which reflect complex sources of heavy metals. There are  
313 automobile, printing and titanium companies around the SLQ sampling site, which  
314 may also explain high concentrations of Ti and Cd. At the TPY sampling point, the  
315 mechanical, petroleum, and metallurgical industries may be sources of Sr  
316 contamination.

### 317 **3.3 Evaluation of heavy metal pollution in water and sediment**

318 Water Quality Index (WQI) evaluation for the FRB is presented in **Fig. 2 a**. The  
319 WQI values of each sampling point from upstream to downstream are 0.15, 0.19,  
320 0.56, 1.39, 0.40, 0.39, 0.74, and 0.43 respectively. Among them, WQI value of  
321 sampling site YRF is within the range of 1~2, which belongs to low pollution. None  
322 of the other sampling sites were contaminated with heavy metals. According to the Pn  
323 of each sampling point (**Fig. 2 b**), YRF and YJQ sampling points are highly polluted.  
324 The pollution levels of TPY and YJQ sampling points are equivalent with moderate  
325 pollution indicators. The Pn of FHK and QDZ sampling points are 0.99 and 0.89,  
326 indicating low pollution levels (between 0.7 and 1). Nemero index evaluation is  
327 basically consistent with an overall pollution trend of:  
328 YRF>YJQ>TPY>SLQ>FHK>QDZ>GGY>FY. The overall pollution trend of the  
329 FRB is that the middle reaches are the most seriously polluted, followed by the  
330 downstream and the upstream. However, the pollution degree obtained by different

331 evaluation methods is somewhat different. For example, the WQI of YRF site shows  
332 low pollution while Pn shows high pollution. The WQI of YJQ, TPY and SLQ sites  
333 are no-pollution while Pn shows high-pollution, medium-pollution, and medium-  
334 pollution respectively. The WQI of FHK and QDZ sites are no-pollution while Pn  
335 shows low pollution. WQI reflects the average pollution level of multiple heavy  
336 metals. Pn considers the average pollution level and the influence of high heavy metal  
337 concentration.

338 The calculated Geological Accumulation Index (I-geo) and Potential Ecological  
339 Risk Index (RI) of heavy metals in sediment are summarized in **Fig3**. In **Fig3 a**, the I-  
340 geo ranged from -0.70 to 3.89. Among the studied metals, the I-geo values decreased  
341 as: As >Cd >Sr >Pb> Zn >Cu> Mn> Cr> Ni. Among these sample sites, the range of  
342 I-geo values for As was 3.45-4.08, indicating high pollution to very high pollution in  
343 this sediment. I-geo values for Cd and Sr were 0.34-1.78 and -0.17-0.72 respectively,  
344 which indicating no pollution to moderate pollution. I-geo values of other element  
345 indicated no contamination in the basin. The Potential Ecological Risk Index (RI) was  
346 also calculated and used to assess the ecological risk of heavy metals in the sediment  
347 of the FRB (**Fig3 b**). In these sampling points, the trend showed gradually decreasing  
348 pollution as: SLQ> FHK> YJQ> YRF >TPY. The Single factor ecological risk  
349 index( $E_r^i$ ) is divided into 5 levels in total:  $E_r^i < 40$ , low potential risk;  $40 < E_r^i \leq 80$ ,  
350 moderate potential risk;  $80 < E_r^i \leq 160$ , considerable potential risk;  $160 < E_r^i \leq 320$ ,  
351 high potential risk;  $E_r^i > 320$ , very high potential risk <sup>62</sup>(Maanan et al. 2015). According  
352 to the calculation results, the single factor ecological risk index( $E_r^i$ ) of Cd ranges

353 from 56.81 to 154.15, which indicates that Cd has high ecological risks in FHK and  
354 SLQ sample sites and considerable ecological risks in other sites. The  $E_r^i$  of As ranges  
355 from 164.14 to 253.42, indicating high ecological risk. From each sampling point, the  
356 RI values are in the range from 313.24 to 375.59. This result demonstrates that the  
357 study area presents moderate potential risk. According to the calculation results of  
358 WQI and RI, the pollution degrees of water and sediment are not always consistent  
359 with the ecological risk, which may be related to the complex influencing factors of  
360 heavy metal pollution. Therefore, the combination of pollution degree and ecological  
361 risk is useful for comprehensive analysis of heavy metal pollution risk.

### 362 **3.4 Multivariate statistical analysis**

363 The Pearson Correlation Analysis of heavy metals reflects the relationship  
364 between heavy metals, so as to determine the source and migration of heavy metals  
365 (Yi et al. 2012). If the correlation between elements is low, it indicates that heavy  
366 metals are affected by complex factors (Kükrcer et al. 2014). Pearson Correlation  
367 Analysis (CA) between heavy metals in water and sediment is elucidated in the  
368 Pearson Correlation Matrix (**Table 5**). However, significantly positive correlations  
369 ( $p < 0.01$ ) and positive correlations of  $p < 0.05$  were found among different elements.  
370 Heavy metals with high correlation have similar sources. In water, Fe was found to be  
371 significantly positively correlated with Mn, Cr, Ni, Ti, but positively correlated with  
372 Sr. Ti also shows significant positive correlation with Pb. Mn is positively correlated  
373 with Pb. Cr-Ni (0.753), Cr-Cu (0.814), Cr-Sr (0.817), Ni-Sr (0.758), Ni-Pb (0.772),

374 Cu-Sr (0.801), As-Sr (0.802) were also positively correlated (**Table 5**). In sediment,  
375 there was significant positive correlation between Mn and Pb, Cr and Ni, Cu and Zn.  
376 Moreover, Ni and Sr exhibited a significant negative correlation of -0.979. And the  
377 Cr-Cu (0.913), Cr-Zn (0.924), Ni-Cu (0.941) and Ni-Zn (0.946) showed positive  
378 correlations in sediment. However, Cr and Sr exhibited negative correlations of -0.940  
379 (**Table 5**).

380 The dendrogram of Hierarchical Cluster Analysis (HCA) are shown in **Figs. 4a**  
381 and **4b** for water and sediment respectively. In water, all sample sites can be divided  
382 into three main types. Sample sites FY, GGY are grouped into Cluster 1. Cluster 2  
383 includes YJQ, SLQ, YRF, and QDZ. The other sites FHK, TPY fall into Cluster 3.  
384 Results show that monitoring points assigned to the same group which have similar  
385 pollution sources and backgrounds. Group 1 is mainly from the upstream and is  
386 largely uncontaminated. Group 2 is mainly distributed in the middle and lower reaches,  
387 where the YRF sampling points may reflect heavy metals from sewage outlets. Group  
388 3 is clustered together, possibly because the rivers in the middle and upper reaches  
389 were distributed around residents, farmland and orchards. It indicates that there is a  
390 significant difference between the upstream and the middle and downstream, so that  
391 the source of heavy metals can be preliminarily determined. In sediment, the YRF,  
392 YJQ, and TPY samples in the middle stream are classified into Cluster 1, the SLQ and  
393 FHK sample sites were grouped in Cluster 2 and 3 respectively. Heavy metal  
394 pollution in the downstream area is serious, followed by the FHK sample sites, which  
395 is consistent with the result of the Potential Risk Evaluation Index (RI).

396 Total interpretation variance and rotation component matrix of the principal  
397 components analysis (PCA) in water and sediment are presented in **Table 6**. The  
398 extracted eigenvalues of the three principal components (PCs) or factors are all  
399 greater than 0.25. PCA reduced the dimensionality of the initial dataset to 3  
400 components in water and 2 components in sediment, which explained 87.97% and  
401 89.90% of the data variance, respectively. Therefore, the three factors and two factors  
402 play a very important role in explaining the heavy metal pollution in the study area. In  
403 water PC1, Mn, Fe, Ni, Pb, and Ti have the highest loadings and accounts for 42.85%  
404 of the total variance of water samples. Fe, Mn, Ti may reflect the natural distribution  
405 of the Earth's crust. Ni and Pb may be related to the electronics industry and urban  
406 transportation. PC2, Cr, Cu, As, Sr, and Cd have high loadings and account for  
407 33.175% of the variance. As is the single dominant metal for PC3, Zn may come from  
408 pigment, plastics and some commercial activities. In sediment samples, PC1, Cr, Ni,  
409 Cu, Zn, and Sr have the highest loadings and account for 55.76% of the total variance,  
410 while, Cd, As, Ti, Mn, and Pb represent PC2 with loadings that account for 34.14% of  
411 the variance. More detailed classification and factor analysis were explored by PMF  
412 model.

### 413 **3.5 Positive matrix factorization**

414 PMF model is run in order to better determine the source of the heavy metals in  
415 the FRB in Shannxi province. The source apportionment results and factor  
416 contribution percentages of heavy metals in water and sediment are shown in **Figs. 5a**

417 and **5b**. From the diagram, we can see the difference between the concentration of the  
418 elements and the percentage contribution. For example, in water for Factor 1, Sr has a  
419 high concentration but a low contribution rate of 17.5%. However, in sediment for  
420 Factor 1, Cd has a low concentration but a high contribution rate of 76.2%. It does not  
421 matter whether a particular metal exhibits a high concentration in a particular factor;  
422 what matters is whether the metal accounts for a large proportion in that factor.  
423 Although it cannot be denied that there is a direct relationship between concentration  
424 and percentage, it means that the percentage of heavy metals under certain factors is  
425 more significant than the concentration. As a result, heavy metal concentrations and  
426 percentages are shown in the chart through PMF models, which play a significant role  
427 in explaining and analyzing the distribution of pollution sources.

428 In water, Factor 1, which constitutes a moderate 36.8% of the contribution (**Fig. 6**  
429 **a**), has higher loadings of Mn (60.9%), Fe (46.8%) and Ti (50.9%). The background  
430 values of these elements are high, and since they are the dominant elements in the  
431 Earth's crust (Pehlivan 2010), they may be partly derived from natural causes (recent  
432 loess of parent material) (Yang et al. 2014) and partly from companies such as the  
433 chemical machinery. This conclusion is also consistent with the results of PCA and  
434 HCA, and there is a strong correlation between these elements. Factor 2, which  
435 accounted for a lower portion (11.7%) of the contribution (**Fig. 6 a**), mainly includes  
436 As (72.6%), Sr (47.3%) and Ni (33.0%). Ni and As may be associated with some of  
437 the electronics companies that are regional distributed. Sr is an alkaline earth metal,  
438 usually in the form of  $\text{SrSO}_4$  and  $\text{SrSO}_3$  in the Earth's crust<sup>67</sup> (Jie et al. 2014). A large

439 number of digital cathode ray tubes are derived from SR<sup>68</sup> (Hibbins 2000), so the  
440 source of Sr may be from YJQ sampling point. Factor 3, which explains only the  
441 contribution of 9.4% (**Fig. 6 a**), is correlated quite well with Cd (99.8%), and  
442 followed by Cr (37.9%) and Cu (31.3%). Some light industries, such as alloy,  
443 components and lubricants, can result in significant copper and chromium emissions  
444 (Mirbagheri and Hosseini 2005; Yeung et al. 2003). At the same time, Cd is closely  
445 related to fertilizers, pesticides and disinfectants (Mansour et al. 2009). The Fenghe  
446 River runs through the city center and suburbs of Xi'an city, and some enterprises,  
447 residents, farmland and orchards (vineyards and strawberry fields) are distributed  
448 around the sampling site, which may be the main sources of Cu, Cr and Cd pollution.  
449 Factor 4, which accounted for a very large percentage 42.0% of the contribution (**Fig.**  
450 **6 a**), contained Zn (49.1%) and Pb (33.2%). The source of Zn and Pb are mainly from  
451 electronics industry and urban transportation. Leaded gasoline can cause water  
452 pollution of Pb through atmospheric subsidence (Miguel et al. 1997).

453 In sediment, the proportion of the last three factors is relatively uniform. Factor 1,  
454 which makes up the minimum proportion 8.0% of the contribution (**Fig. 6 b**), is  
455 composed essentially of Cd (72.6%). As in the case of Factor 3 in water by PMF  
456 analysis, Cd occupies the largest contribution rate. It is obvious that Cd is related to  
457 agronomic activity (fertilizer, pesticide, insecticide and disinfectant)<sup>73,74</sup> (Wang et al.  
458 2015; Jian-Long et al. 2014), which is mainly distributed in the middle and lower  
459 reaches of the study area. Factor 2, which presented 29.2% of the contribution (**Fig. 6**  
460 **b**), is constituted by Cu (49.5%), Zn (34.7%) and Ni (33.0%). Factor 3, which

461 accounted for 23.9% of the contribution (**Fig. 6 b**), is made up of Cu (31.0%) and Zn  
462 (22.2%). These metals are mainly related to tourism in the upstream of the study area,  
463 the construction of towns in the middle reaches and the industry in the downstream.  
464 Each element of this factor accounts for a small proportion, indicating that it should  
465 be given less attention in future pollution control plans. Factor 4 accounted for 38.9%  
466 of the contribution (**Fig. 6 b**), and most elements are included in this factor. For  
467 example, it includes Sr (96.8%), As (63.8%), Pb (56.5%), Mn (55.6%), Ti (47.8%)  
468 and Cr (31.9%). These metals are primarily derived from human activities and are  
469 concentrated in the automobile, printing, metallurgy, materials, and petroleum  
470 machinery industries (Yang et al. 2013; Chowdhury and Maiti 2016; Loska and  
471 Wiechuła 2003).

#### 472 **4. Conclusion**

473 We present a preliminary assessment and source analysis of heavy metal  
474 elements in water and surface sediment in the FRB. The concentration, distribution,  
475 enrichment characteristics, potential risks and sources of heavy metals were taken into  
476 account comprehensively. The results show that most metals do not meet national  
477 water quality standards, except for Mn which exceeds the Class III water quality  
478 standard and Cr which is higher than the drinking water standard. However, all the  
479 elements in the sediments except Ni were higher than the soil environmental  
480 background value in Shaanxi province. The results of WQI and Pn show that the most  
481 serious pollution is in sampling site YRF, possibly due to direct emissions from  
482 sewage outlets. I-geo indicates that As is highly concentrated in the sediments.  $E_r^i$  and



483 RI show that Cd in sediment has a high ecological risk at sampling sites FHK and  
484 SLQ and a moderate ecological risk at other sites. As in sediment has high ecological  
485 risk for all sampling stations. Results for CA, HCA, and PCA confirm the results of  
486 water and sediment analysis, although there are a few differences due to the  
487 government's control for downstream sites and some special pollution sources such as  
488 factories, businesses, residents, town building and agricultural activity (fields and  
489 orchards) in the middle reaches. The factors or sources of heavy metals in water and  
490 sediment are revealed in detail through PMF models. In the water, the average  
491 contribution rate of these four source factors for heavy metals is 36.8%, 11.7%, 9.4%  
492 and 42.0%, while the average proportion of these four factors for heavy metals in  
493 sediment is 8.0%, 29.2%, 23.9% and 38.9% respectively. Most of the river basin is  
494 influenced by human activities, mainly town construction and transportation, some  
495 light industry, machinery and electronics material companies, tourism, and agronomy.  
496 Future control and prevention should focus on these aspects in order to provide better  
497 ecological and environmental benefits for the FRB.

## 498 **5. Declarations**

### 499 **5.1 Ethics approval and consent to participate**

500 All procedures performed in studies involving human participants were in  
501 accordance with the ethical standards of the institutional and/or national research  
502 committee and with the 1964 Helsinki Declaration and its later amendments or  
503 comparable ethical standards. The study was approved by National Key R&D

504 Program of China.

## 505 **5.2 Consent for publication**

506 Not applicable for that section.

## 507 **5.3 Availability of data and materials**

508 The data that support the findings of this study are available from Eight water  
509 quality monitoring stations in the Fenghe River Basin but restrictions apply to the  
510 availability of these data, which were used under license for the current study, and so  
511 are not publicly available. Data are however available from the authors upon  
512 reasonable request and with permission of Eight water quality monitoring stations in  
513 the Fenghe River Basin.

## 514 **5.4 Competing interests**

515 The authors declare that they have no known competing financial interests or  
516 personal relationships that could have appeared to influence the work reported in this  
517 paper.

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## 528 **5.6 Authors' contributions**

529 All authors contributed to the study conception and design. Material preparation,  
530 data collection and analysis were performed by Aidi Huo, Daniel Nover and Shuxin  
531 Kang. Table and Figure editing are performed by Meimei Zhou, Jiqiang Lyu and  
532 Pingping Luo. The first draft of the manuscript was written by Chengyi Xu and all  
533 authors commented on previous versions of the manuscript. All authors read and  
534 approved the final manuscript.

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### Table Caption List

731 **Table 1** Classifications of heavy metal degree of WQI, Pn,I-geo and RI.

732 **Table 2** Water quality parameters of Fenghe river basin, Shannxi Province

733 **Table 3** Heavy metal average concentration (ug/L) in water sample of Fenghe river

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735 **Table 4** Heavy metal concentration (mg/kg) in sediment sample of Fenghe river basin

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737 **Table 5** Pearson Correlation matrix between heavy metal concentrations in water and

738 sediment

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740 components/factors of the Fenghe river in water and sediment samples

741

742 **Table 1** Classifications of heavy metal degree of WQI, Pn,I-geo and RI.

WQI		Pn		I-geo		RI <sup>743</sup>		
classes	Scope	Contamination degree	Scope	Contamina tion degree	Scope	Contamination degree	Scope	Ecological <sup>744</sup> risk
0	<1	Unpolluted	≤0.7	Unpolluted	0	Unpolluted	<110	low risk
1	1 ~2	Slightly polluted	0.7~1	Slightly polluted	0~1	Unpolluted to moderately	110~200	Moderate risk
2	2~3	Moderately polluted	1~2	Moderately polluted	1~2	Moderately polluted	200~400	Moderately risk
3	≥3	Heavily polluted	≥2	Heavily polluted	2~3	Moderately to highly polluted	≥400	Very high risk
4					3~4	Highly polluted		
5					4~5	Highly to very highly polluted		
6					>5	Very highly polluted		

745 **Table 2** Water quality parameters of Fenghe river basin, Shannxi Province

746

Sites	Temperature(°C)	Ph	Dissolved oxygen (DO mg/l)	Redox potential(ORP mV)	Electrical conductivity (ms/cm)
FY	6.55	8.29	10.88	149.85	0.07
GGY	9.98	8.49	11.25	138.62	0.14
TPY	13.20	7.82	5.24	144.25	0.46
YRF	15.05	7.88	6.22	151.43	0.95
FHK	13.62	7.94	8.40	146.37	0.62
QDZ	12.60	8.06	9.80	145.92	0.67
YJQ	11.85	8.40	9.30	146.72	0.52
SLQ	11.30	8.46	9.89	176.77	0.46
Average	11.77	8.17	8.87	149.99	0.49
±SD	2.44	0.26	2.01	10.73	0.27

747

748

749 **Table 3** Heavy metal average concentration (ug/L) in water sample of Fenghe river  
 750 basin and the reference value in water (ug/L).

Sites	Mn	Fe	Cr	Ni	Cu	Zn	Cd	Pb	As	Ti	Sr
FY	0.80	7.29	6.23	0.75	1.54	10.72	0.10	0.75	0.20	0.98	54.12
GGY	4.41	34.29	6.17	0.78	1.40	11.06	0.03	0.70	0.43	1.17	104.28
TPY	239.80	269.15	7.24	3.95	1.93	11.26	0.05	1.51	0.81	8.54	281.75
YRF	99.47	227.15	7.53	2.25	2.29	19.86	0.05	1.25	2.66	4.56	262.50
FHK	134.70	175.20	7.56	3.03	2.92	8.47	0.04	0.93	1.79	3.32	322.85
QDZ	119.15	250.15	7.53	2.29	2.11	7.34	0.03	0.76	2.66	4.02	311.80
YJQ	47.85	118.70	6.69	2.08	2.13	12.81	0.03	0.91	2.48	2.42	320.80
SLQ	20.38	70.78	7.14	1.85	2.26	15.50	0.03	0.99	2.00	1.71	283.50
Average	83.32	144.09	7.01	2.12	2.07	12.13	0.04	0.98	1.63	3.34	242.70
±SD	76.59	94.57	0.54	1.00	0.44	3.74	0.02	0.26	0.95	2.31	97.21
GB3838-2002 limit value(1)*	10	300	10	---	10	50	1	10	50	---	---
WHO(2004)**	300	1000	5	70	2000	5000	3	10	10	---	---
CMC, acute***	---	---	16	470	---	120	1.8	65	340	---	---
CMC, chronic****	---	1000	11	52	---	120	0.72	2.5	150	---	---

751 \* GB3838-2002, Environmental Quality Standard for Surface Water (GB3838-2002) (2002) China.

752 \*\*WHO (2004), Guidelines for Drinking Water Quality, third ed. World Health Organization, Geneva.

753 \*\*\*CMC, acute (Criterion maximum concentration in freshwater for National Recommended

754 Aquatic Life Criteria) <https://www.epa.gov/>(EPA 2009)

755 \*\*\*\*CMC, chronic (Criterion continuous concentration in freshwater for National Recommended

756 Aquatic Life Criteria) <https://www.epa.gov/> (EPA 2009)

757



758 **Table 4** Heavy metal concentration (mg/kg) in sediment sample of Fenghe river basin  
 759 and the reference background value in water (mg/kg).

760

Sites	Mn	Cr	Ni	Cu	Zn	Cd	Pb	As	Sr
TPY	672.50	50.66	21.06	19.25	74.33	0.18	30.69	262.10	409.00
YRF	708.20	62.21	26.43	27.94	81.07	0.23	31.80	281.30	313.20
FHK	736.40	79.01	34.01	48.10	125.10	0.33	33.33	264.80	221.50
YJQ	652.00	67.00	30.03	30.14	96.56	0.23	28.63	261.20	243.70
SLQ	615.25	61.29	23.68	18.21	74.90	0.48	26.56	182.20	347.25
Average	676.87	64.03	27.04	28.73	90.39	0.29	30.20	250.32	306.93
±SD	42.33	9.20	4.58	10.76	19.12	0.11	2.38	34.83	68.38
Soil background value (Shannxi)	557	62.5	28.8	21.4	69.4	0.094	21.4	11.1	166
Wei River(Shannxi)	NA	91.3	25.6	21.2	66.1	NA	21.4	NA	NA
Yellow River(China)	NA	41- 128	NA	NA	NA	NA	26- 78	14-48	NA
ASV	NA	90	68	45	NA	0.3	20	13	NA

761 ASV, average shale value (Turekian and Wedepohl 1961);

\* Significant to probability level 0.05.

\*\* Significant to probability level 0.01.

<b>Water</b>	Mn	Fe	Cr	Ni	Cu	Zn	Cd	Pb	As	Sr	Ti
Mn	1										
Fe	0.894**	1									
Cr	0.679	0.835**	1								
Ni	0.946**	0.850**	0.753*	1							
Cu	0.402	0.484	0.814*	0.601	1						
Zn	-0.177	-0.015	0.095	-0.079	0.051	1					
Cd	-0.168	-0.348	-0.355	-0.281	-0.302	-0.024	1				
Pb	0.748*	0.644	0.493	0.772*	0.255	0.450	-0.042	1			
As	0.161	0.526	0.703	0.317	0.661	0.292	-0.553	0.110	1		
Sr	0.574	0.712*	0.817*	0.758*	0.801*	0.034	-0.629	0.412	0.802*	1	
Ti	0.958**	0.873**	0.591	0.887**	0.230	0.019	-0.143	0.859**	0.131	0.489	1
<b>Sediment</b>											
Mn	1										
Cr	0.508	---	1								
Ni	0.572	---	0.966**	1							
Cu	0.783	---	0.913*	0.941*	1						
Zn	0.654	---	0.924*	0.946*	0.976**	1					
Cd	-0.414	---	0.300	0.063	-0.039	0.041	1				
Pb	0.988**	---	0.391	0.467	0.710	0.583	-0.495	1			
As	0.774	---	0.139	0.342	0.468	0.344	-0.873	0.800	1		
Sr	-0.432	---	-0.940*	-0.979**	-0.854	-0.869	-0.097	-0.313	-0.275	1	
Ti	-0.091	---	0.535	0.369	0.207	0.178	0.756	-0.230	-0.453	-0.443	1

768 **Table 6** Total interpretation variance and rotation component matrix of the principal  
 769 components/factors of the Fenghe river in water and sediment samples

Parameters	Water			Sediment	
	PC1	PC2	PC3	PC1	PC2
Mn	0.966	0.185	-0.170	0.616	0.691
Fe	0.819	0.467	-0.032	----	----
Cr	0.575	0.727	0.058	0.991	-0.129
Ni	0.899	0.378	-0.081	0.978	0.063
Cu	0.269	0.796	0.003	0.954	0.271
Zn	-0.027	0.077	0.994	0.944	0.169
Cd	-0.014	-0.698	0.064	0.189	-0.921
Pb	0.861	0.011	0.475	0.507	0.771
As	0.032	0.918	0.235	0.264	0.924
Sr	0.436	0.866	-0.011	-0.936	0.043
Ti	0.982	0.071	0.037	0.468	-0.719
Eigenvalue	4.713	3.648	1.315	5.576	3.414
% of total variance	42.848	33.165	11.953	55.760	34.144
Cumulative % total variance	42.848	76.013	87.966	55.760	89.904

770 Extration method: principal component analysis.

771 Factor loadings: Varimax normalized (value rendered in italics are loadings>0.700)

772

773

### Figure Caption List

774 **Figure 1** The Map of the study area with sampling locations.

775 **Figure 2** The comprehensive quality index (WQI) and Nemero index (Pn) of heavy  
776 metal in water in Fenghe river basin.

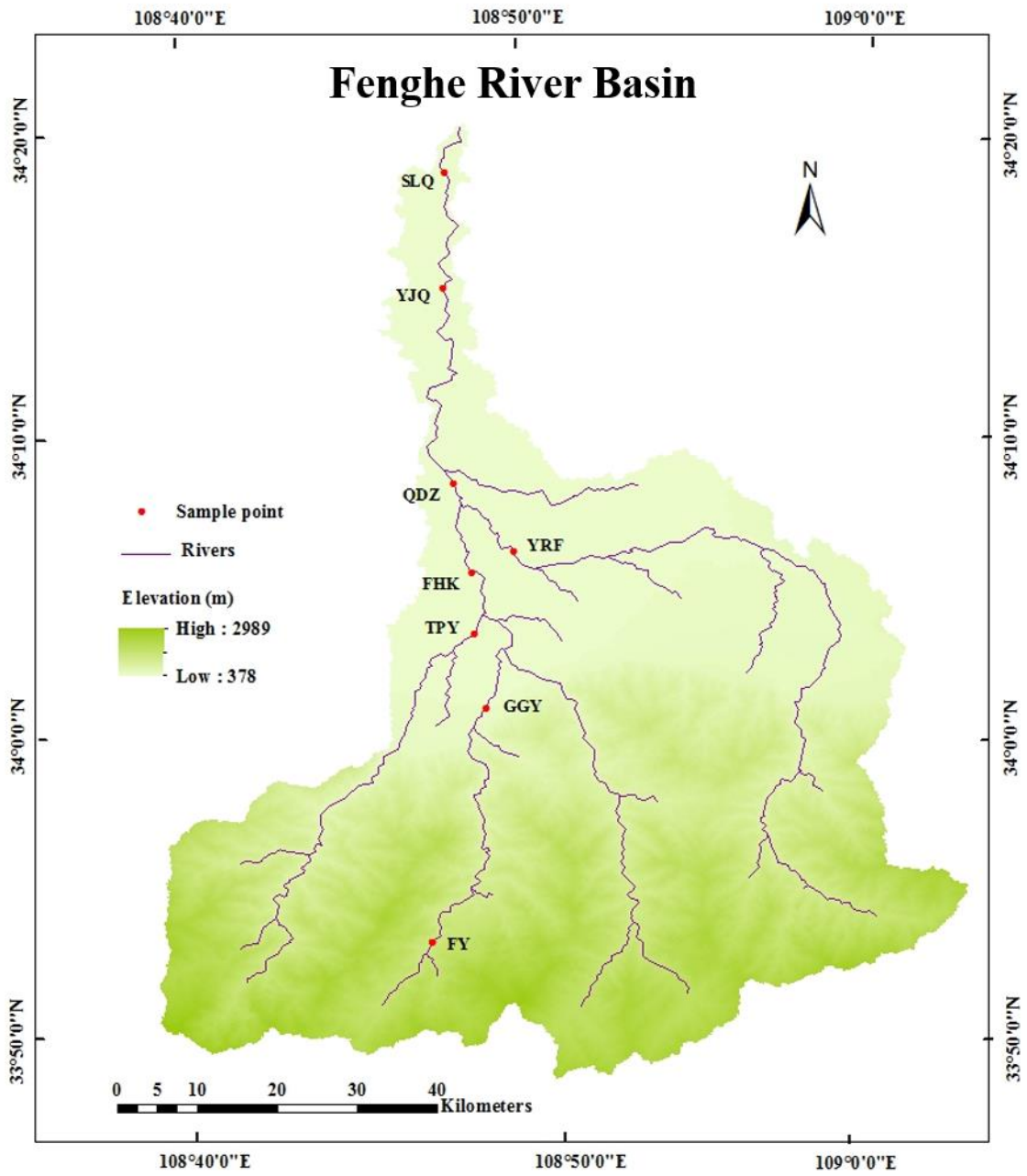
777 **Figure 3** The geological accumulation index (I-geo) and potential ecological risk  
778 index (RI) of heavy metal in sediment in Fenghe River Basin.

779 **Figure 4** Dendrogram of cluster analysis amongst the parameters of Fenghe river in  
780 water(a) and in sediment(b) samples.

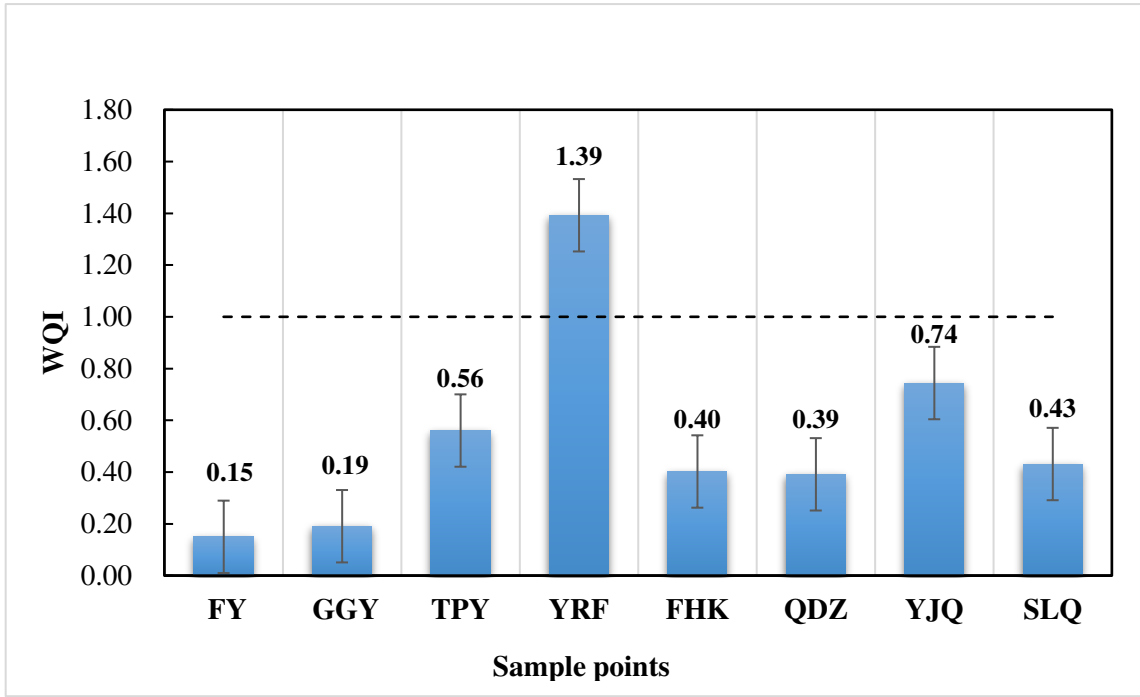
781 **Figure 5** Results of PMF source apportionment modeling for heavy metals in (a)  
782 water and (b) sediment in the Fenghe River, Shannxi.

783 **Figure 6** The average factor contribution ration for heavy metals in (a) water and (b)  
784 sediment in the Fenghe River, Shannxi.

785



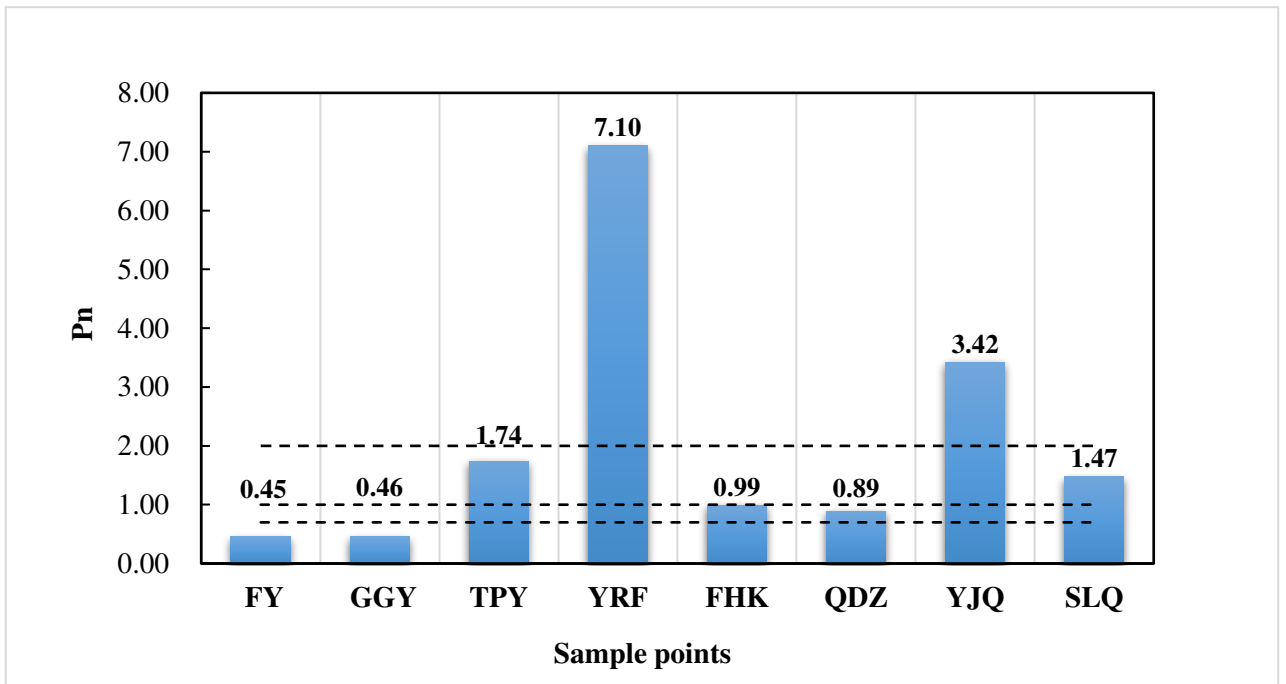
788 **Figure1** The Map of the study area with sampling locations.



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(a)

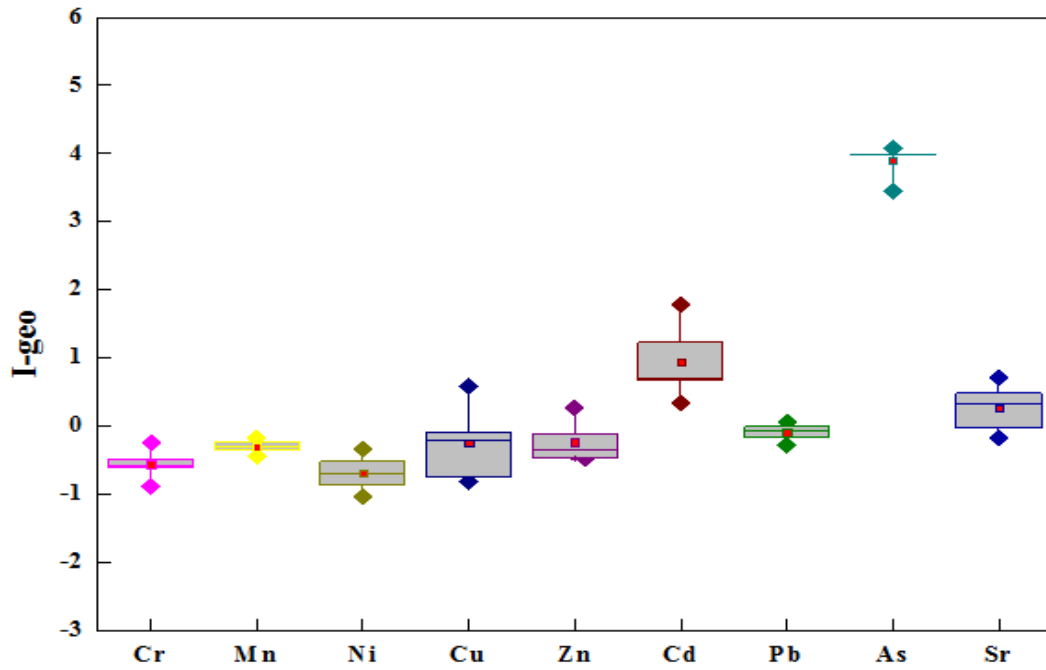


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(b)

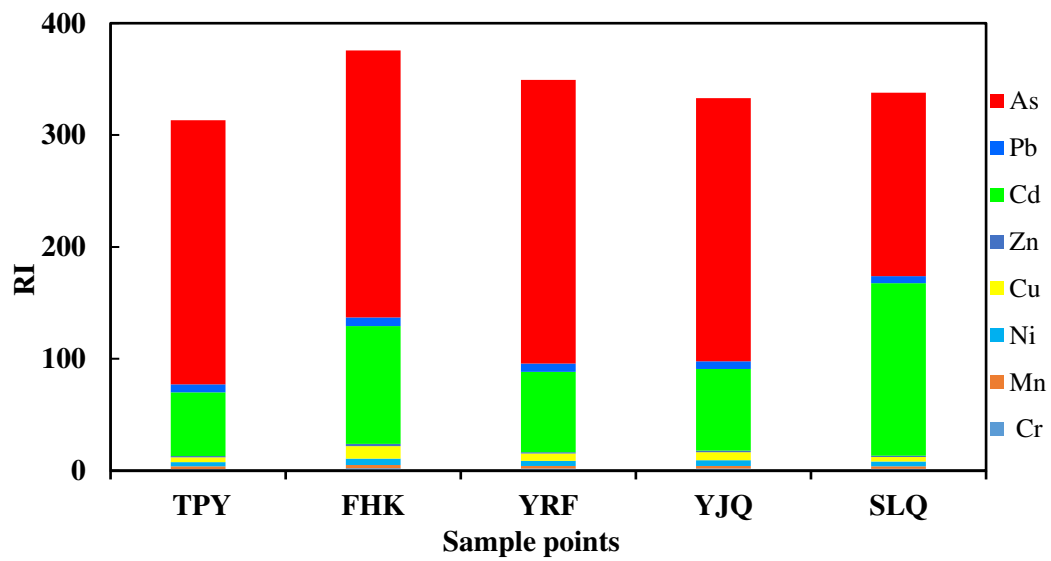
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 794 metal in water in Fenghe river basin.



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(a)



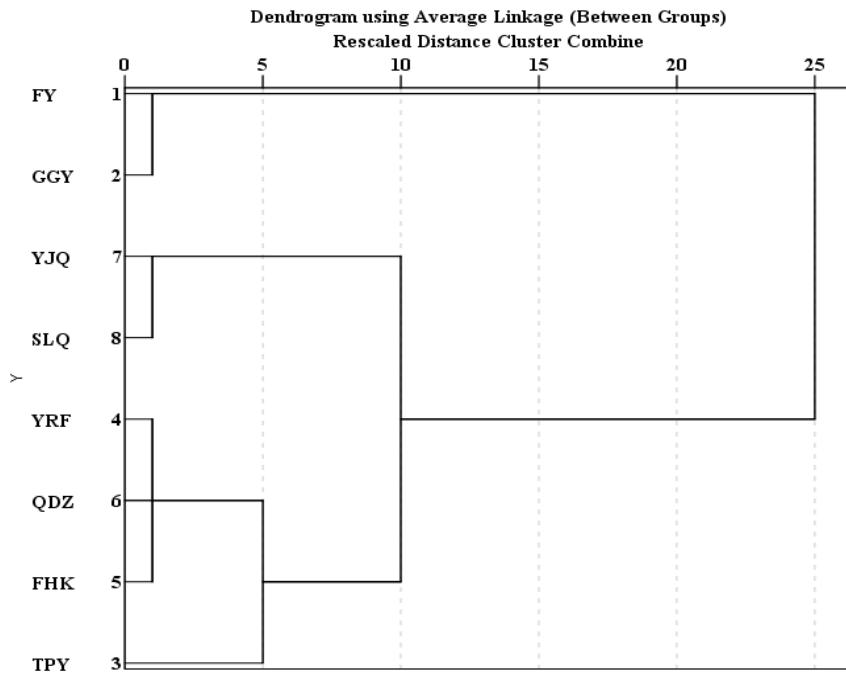
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798

(b)

799 **Figure 3** The geological accumulation index (I-geo) and potential ecological risk

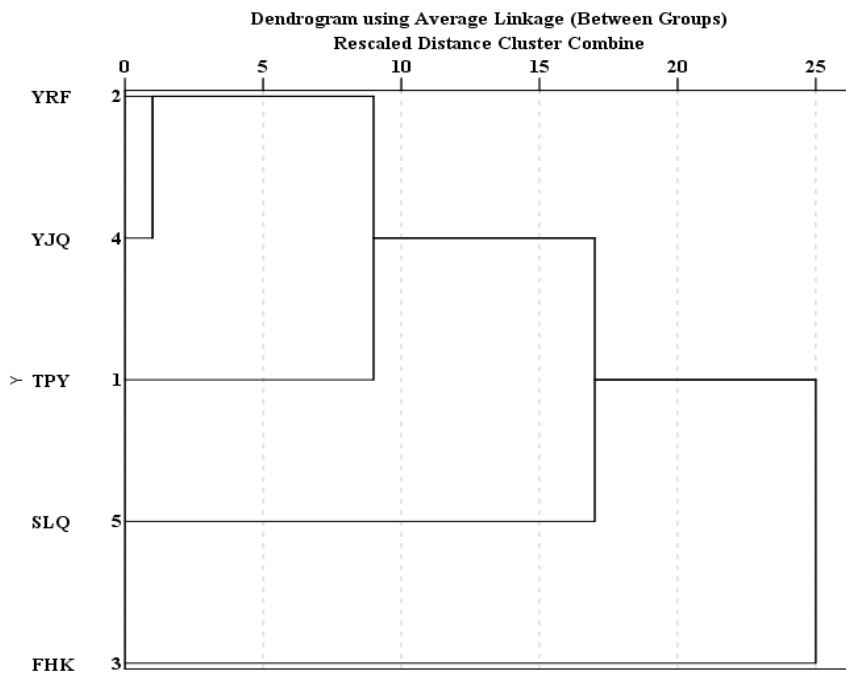
800 index (RI) of heavy metal in sediment in Fenghe River Basin.



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802

(a)



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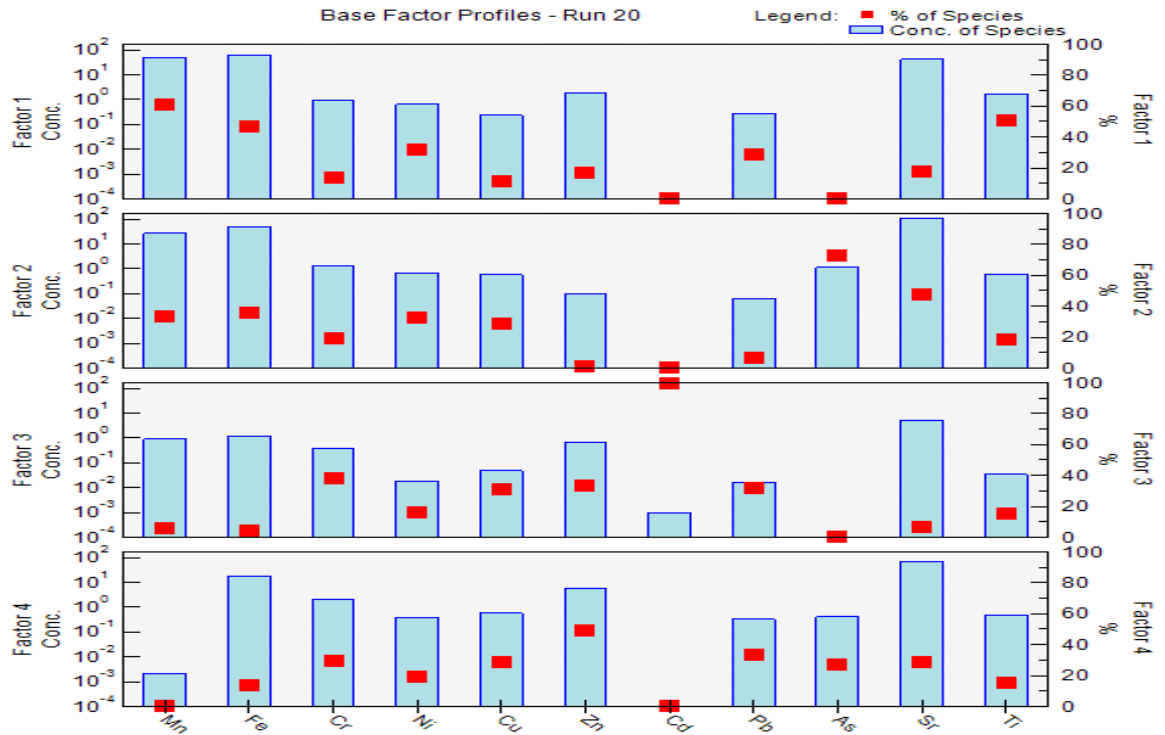
804

(b)

805 **Figure 4** Dendrogram of cluster analysis amongst the parameters of Fenghe river in

806 water(a) and in sediment(b) samples.

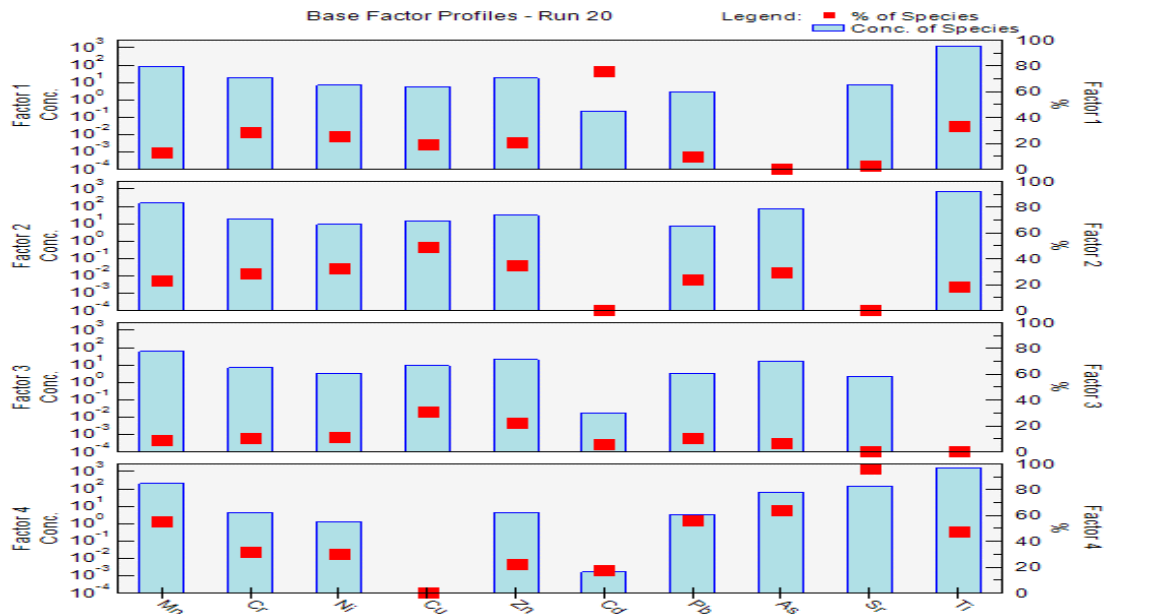




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808

a) Water



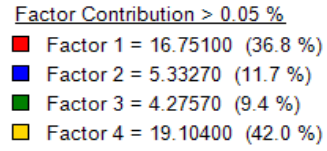
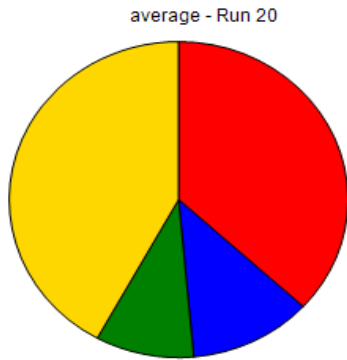
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b) Sediment

811 **Figure 5** Results of PMF source apportionment modeling for heavy metals in (a)

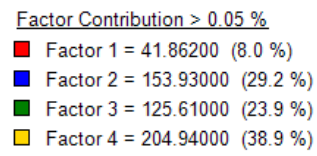
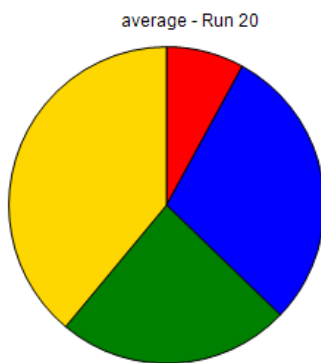
812 water and (b) sediment in the Fenghe River, Shannxi.



813

814

**a) Water**



815

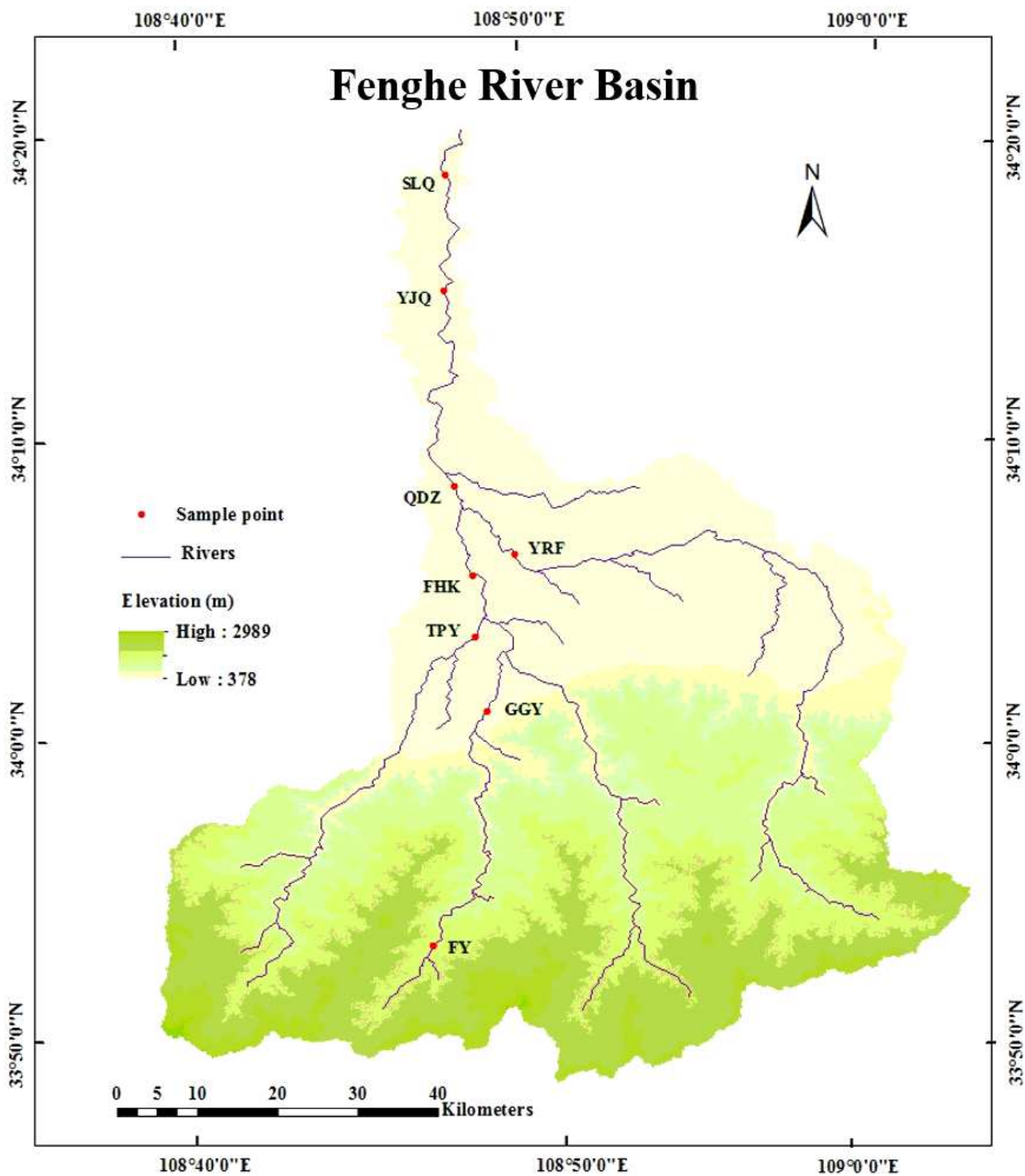
816

**b) Sediment**

817 **Figure 6** The average factor contribution ration for heavy metals in (a) water and (b)

818 sediment in the Fenghe River, Shannxi.

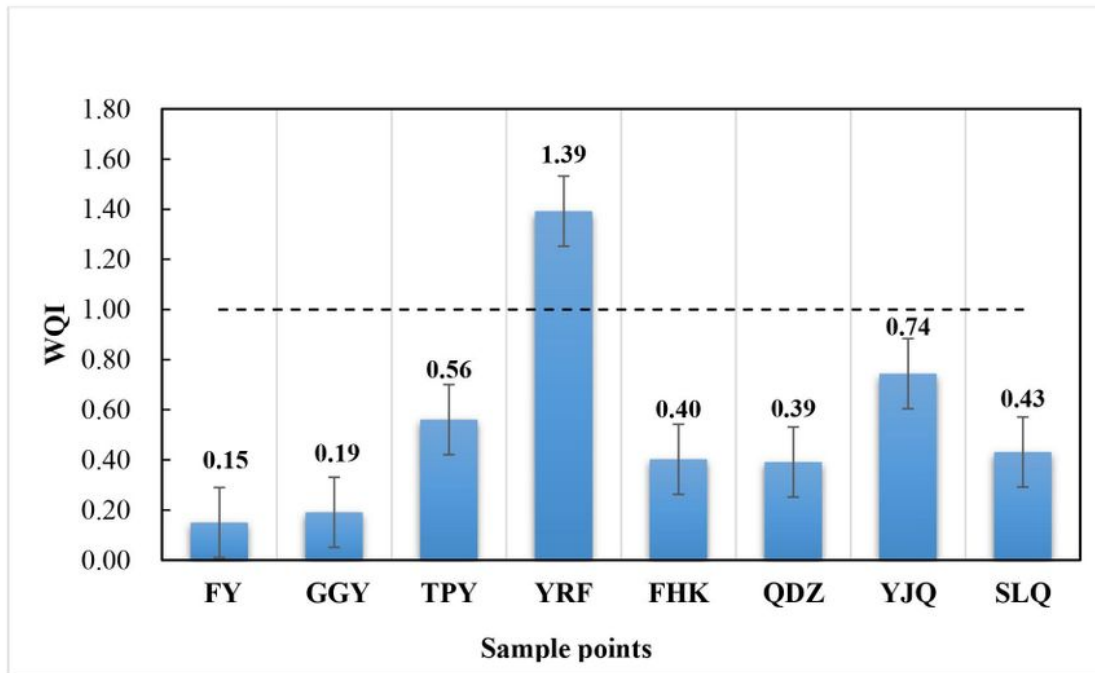
# Figures



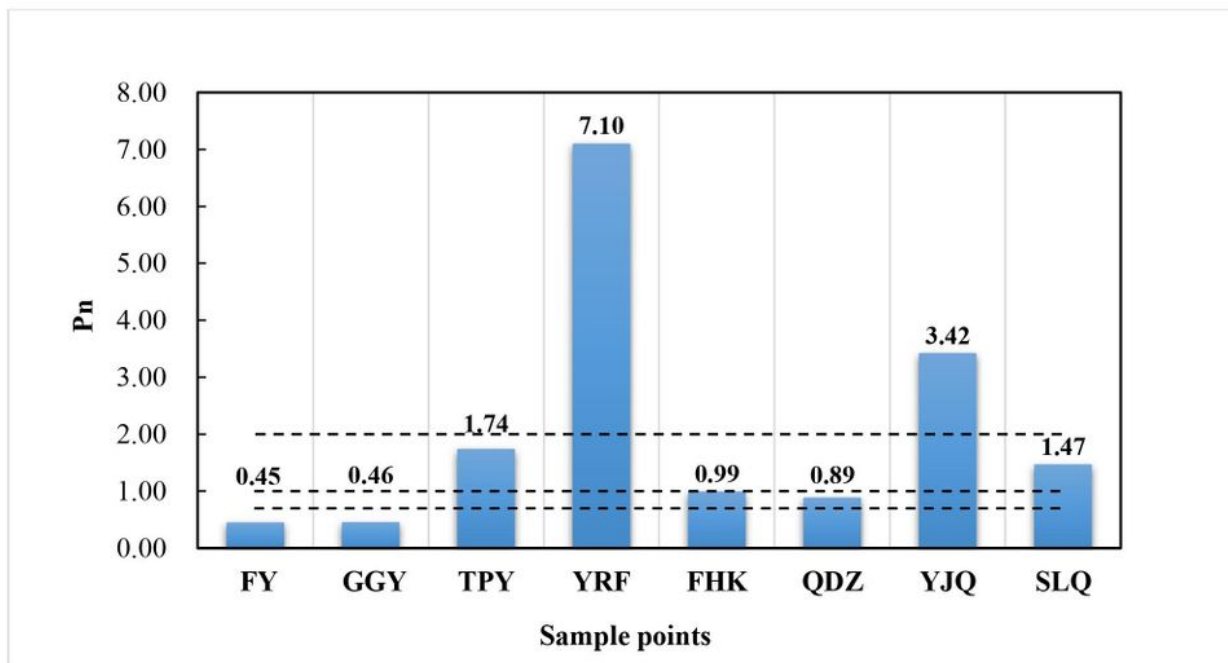
**Figure 1**

The Map of the study area with sampling locations. Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its

authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.



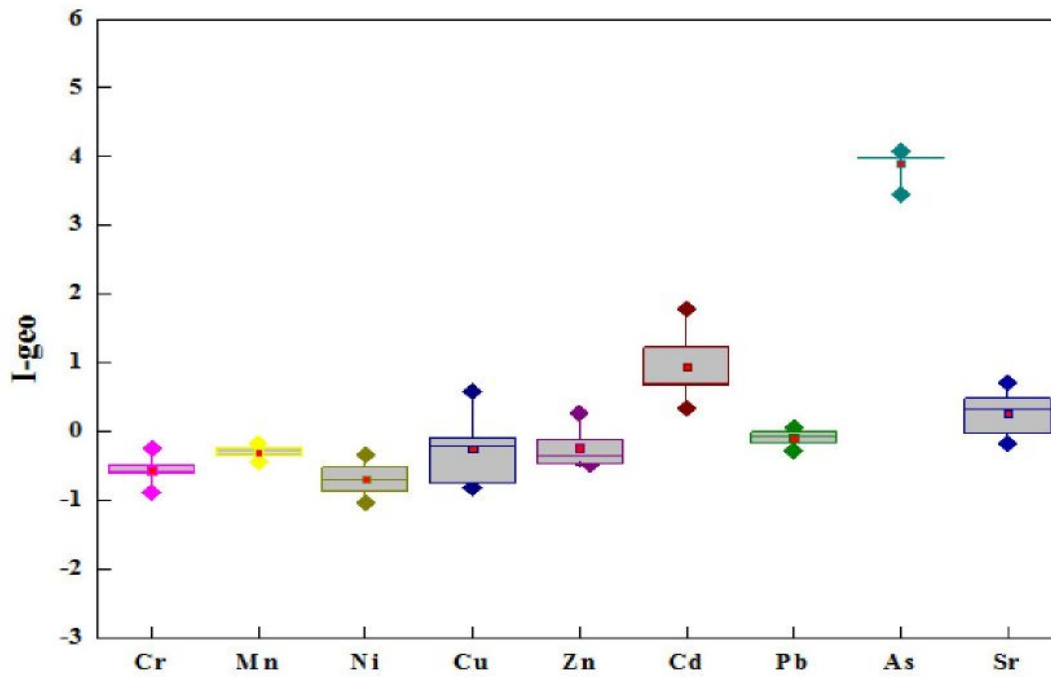
(a)



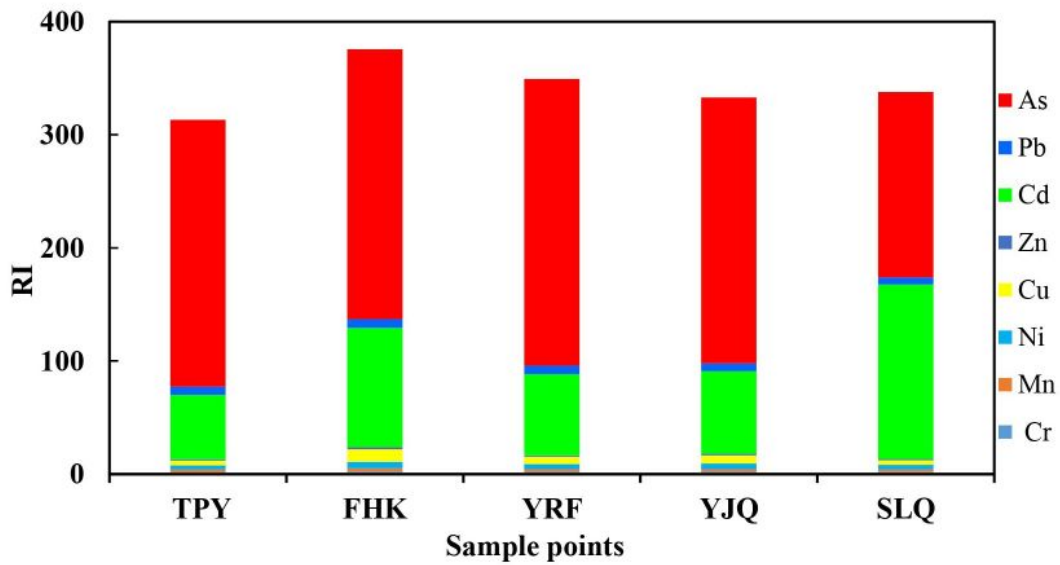
(b)

**Figure 2**

The comprehensive quality index (WQI) and Nemero index (Pn) of heavy metal in water in Fenghe river basin.



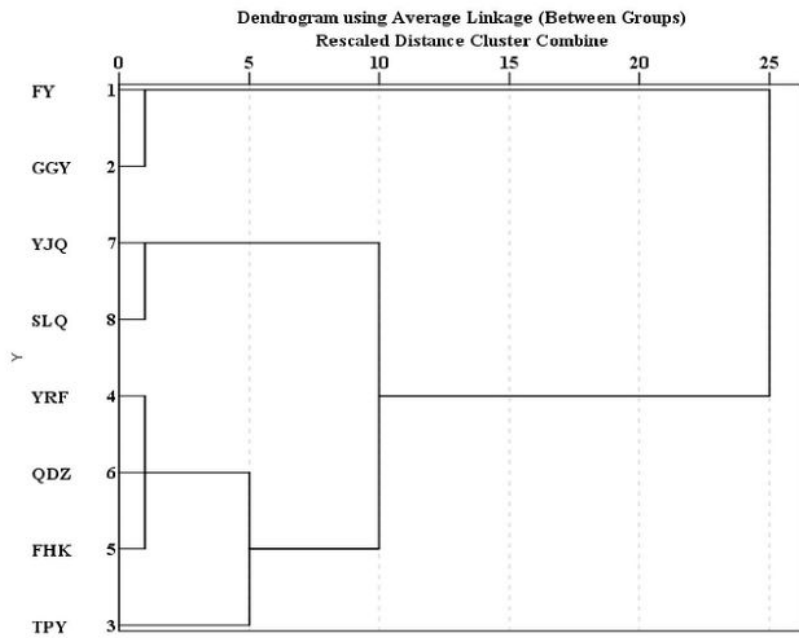
(a)



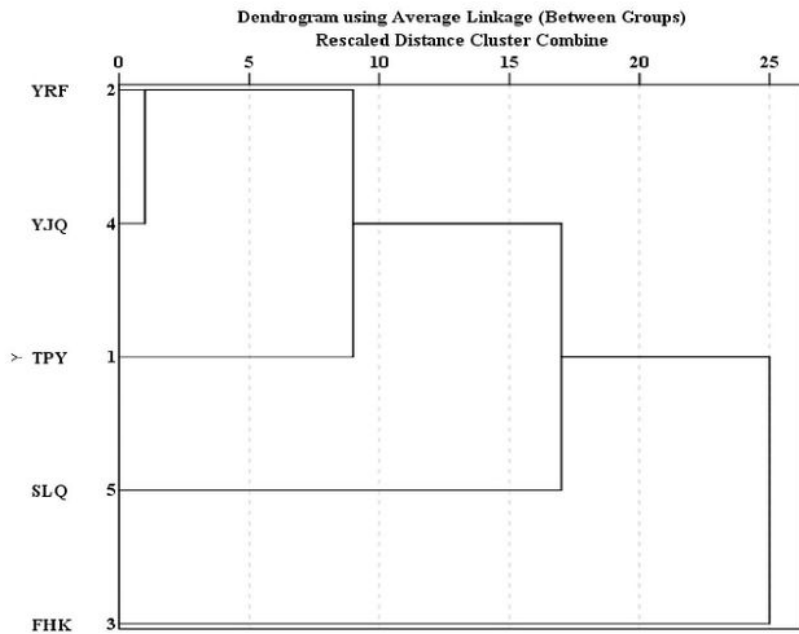
(b)

Figure 3

The geological accumulation index (I-geo) and potential ecological risk index (RI) of heavy metal in sediment in Fenghe River Basin.



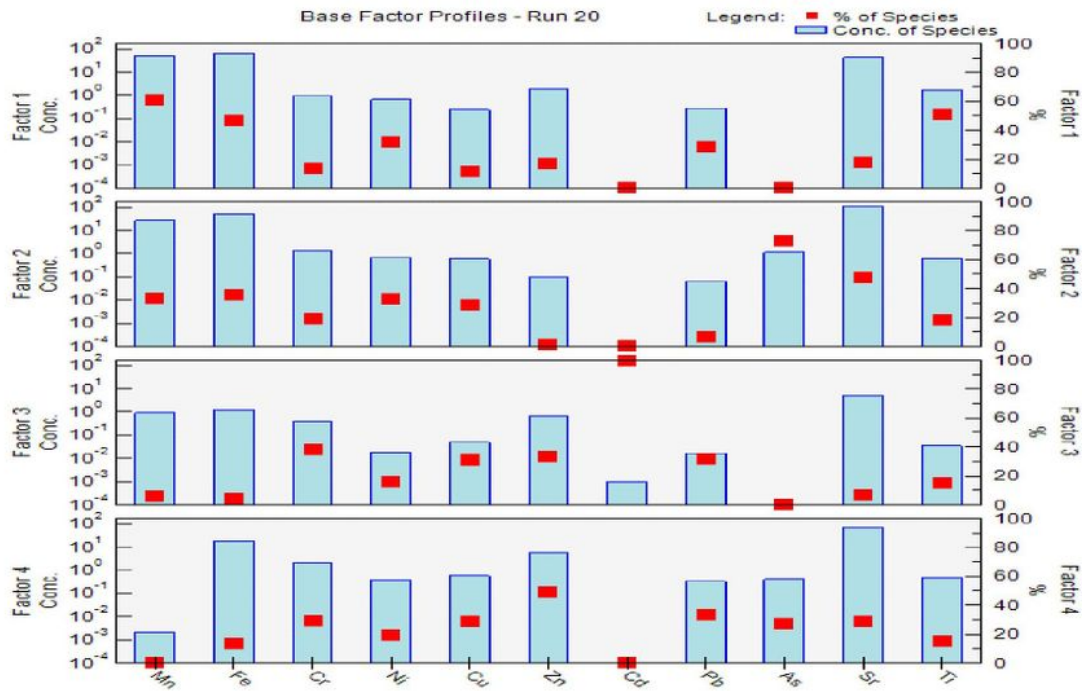
**(a)**



**(b)**

**Figure 4**

Dendrogram of cluster analysis amongst the parameters of Fenghe river in water(a) and in sediment(b) samples.



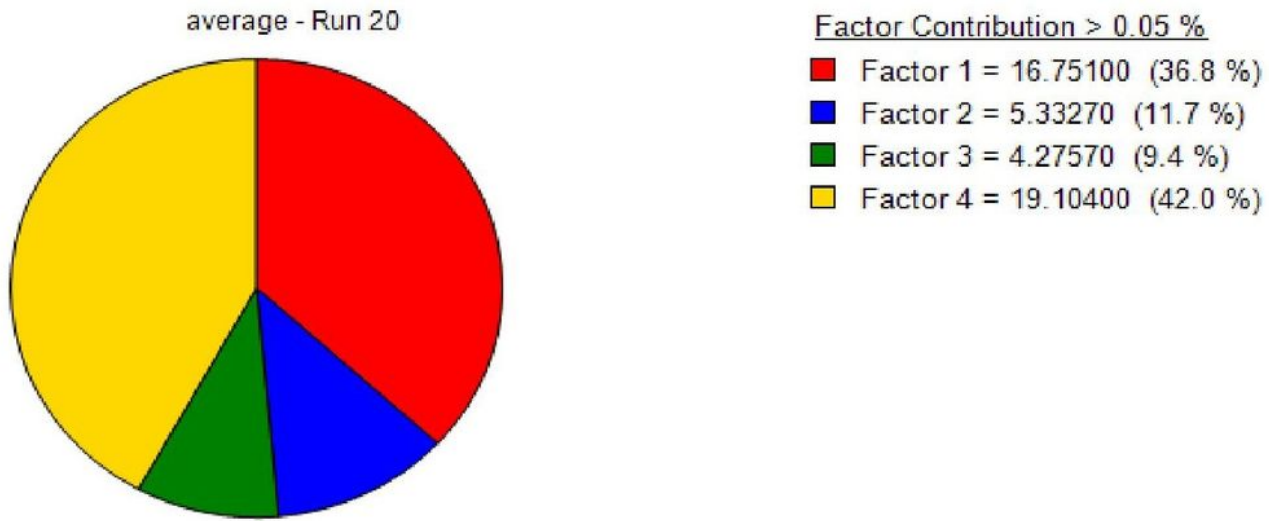
**a) Water**



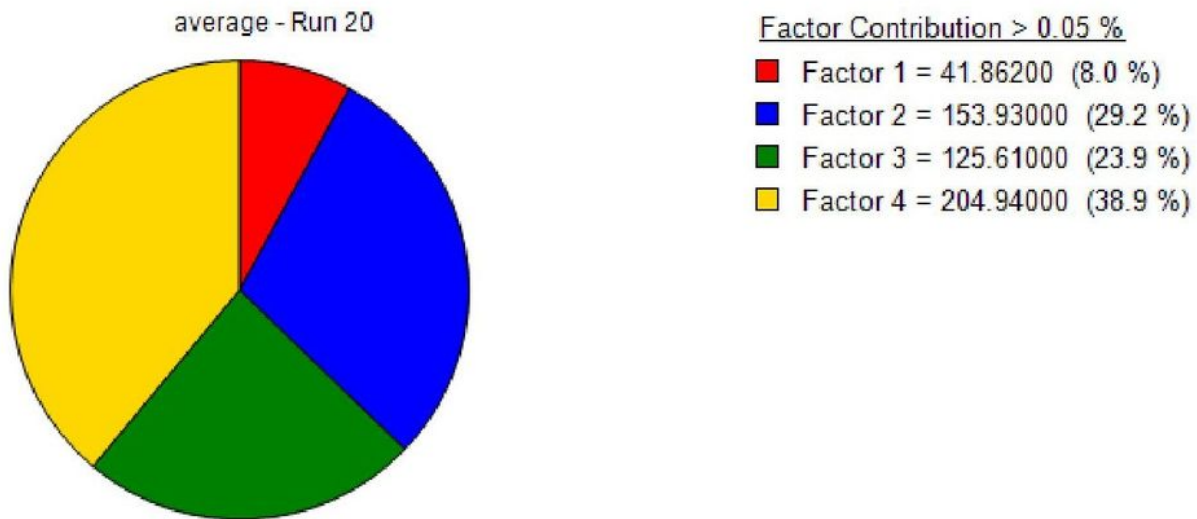
**b) Sediment**

**Figure 5**

Results of PMF source apportionment modeling for heavy metals in (a) water and (b) sediment in the Fenghe River, Shannxi.



**a) Water**



**b) Sediment**

**Figure 6**

The average factor contribution ration for heavy metals in (a) water and (b) sediment in the Fenghe River, Shannxi.