

1 **Urban-rural gradients of POPs soil concentrations at a regional scale:**
2 **quantification and prediction.**

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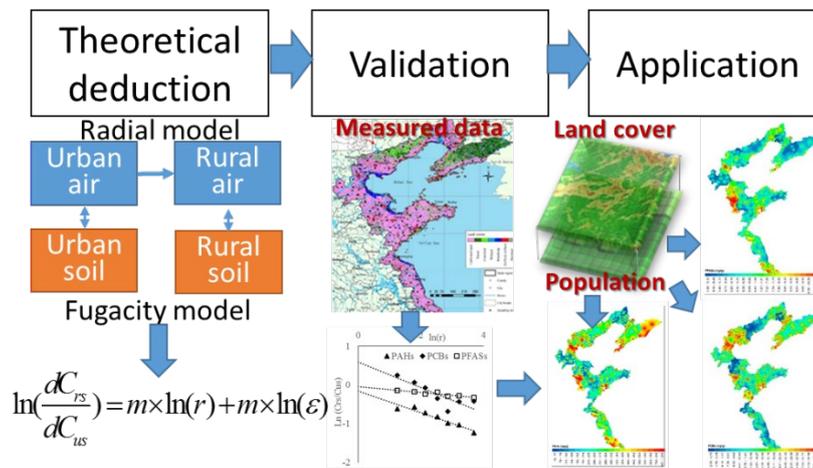
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12 **Abstracts**

13 Quantitative study of urban-rural gradients for persistent organic pollutants (POPs)
14 concentrations is extremely important for understanding POPs behavior and for
15 ecological risk assessment and management. However, little has been studied about
16 quantitative calculation for soil concentrations of POPs from urban to rural areas. In
17 this paper, 153 soil samples were collected in the coastal region of Bohai Sea and
18 Yellow Sea, and the distributions of polychlorinated biphenyls, Polyaromatic
19 hydrocarbons, Perfluoroalkyl and polyfluoroalkyl substances were examined. Urban-
20 rural gradient model (URGM) was derived by using atmospheric point source diffusion
21 model combined with fugacity model to prove potential mathematical relationships

22 among urban and rural soils, and the URGM was proved suitable for simulating urban-
 23 rural soil concentrations through validation of measured data. Significantly linear
 24 correlation was found between POPs amount in surface soil and city population,
 25 between POP concentrations and the artificial surface area. Urban-rural POPs
 26 concentrations were simulated by URGM model, and calibrated by urban population
 27 and land-cover data. The results showed that it was suitable for simulating urban-rural
 28 POPs concentrations at a regional scale. This study could provide a new method for
 29 quantifying urban-rural gradients for POPs concentrations, and promote quantitative
 30 research on coupling between land cover, socio-economic data and POPs
 31 concentrations.

32 Graphic Abstract:



33

34 **Key words:** urban-rural gradient; POPs; land use; regional scale; spatial prediction;

35 coastal region

36

37 **1. Introduction**

38 The fate of persistent organic pollutants (POPs) has brought great concern from

39 considerable government officials and scientists, particularly when local emission of
40 chemicals has resulted in dispersed contamination of large areas ¹. Assessment of POPs
41 state and impacts is very important for environmental management and human health
42 risk management. However, the differences of chemical properties and regional
43 emissions, heterogeneity of environmental media properties, spatially and temporally
44 limited large-scale monitoring data have led to difficulties in the assessment of
45 contaminants in regional scales. Based on measured data of POPs, quantitative research
46 of distributed regularity in environmental media is extremely important for an
47 understanding of what the fate of a chemical is at the regional scale.

48 Urban areas are generally regarded as major sources of some POPs to the
49 surrounding regions. Pollutant concentrations in environmental media could be directly
50 affected by huge differences of contaminant emission per unit area between urban and
51 rural regions. For example, emissions of PAHs, PCBs and PBDEs in urban areas may
52 be several times or even thousands of times higher than that in suburbs, which may
53 result in POPs concentrations in urban soils several times or even tens of times higher
54 than that in rural soils ^{2, 3}. However, little work was relevant to quantitative spatial
55 distribution of POPs soil concentrations between urban and rural regions at a regional
56 scale^{4, 5}. As a follow-up of our previous study², this paper is intended to study
57 differences in distribution and behavior of POPs between urban and rural areas.

58 At present, there are mainly two methods for describing POPs gradients from
59 urban to rural regions. One is that a few representative studies about POPs transport
60 simulation between urban and rural regions using fugacity model. For example,

61 Multimedia Urban Model (MUM) and Spatially Oriented MUM (SO-MUM) were
62 used to simulate the fate of POPs from urban to suburban in Canada^{4, 6, 7}; the
63 Berkeley-Trent-Urban-Rural Fate Model (BETR) was provided for Switzerland to
64 investigate the origin of PCBs in air and to investigate their long-term fate and mass
65 balance in the environment⁵; the spatially resolved Berkeley-Trent-Urban-Rural Fate
66 Model (BETR-UR) was designed by coupling land cover information to simulate the
67 transport of POPs between urban and rural areas². The other method is that the
68 analysis of numerous measured data through systematic description^{8, 9} or some classic
69 atmospheric transport models¹⁰. For example, ratios of urban to rural concentrations
70 was used to assess urban-rural trends for PCBs¹¹, OCPs¹¹ and PBDEs¹². Melymuk et
71 al.¹⁰ used a radial dilution model to simulate the urban-rural gradients in air and
72 proved the influence of the central city as a source of contaminants to the surrounding
73 environment. However, most of them were not able to provide quantitative
74 information about distribution or transport of POPs between urban and rural areas⁵,
75 and some research of POPs gradients were mainly focused on air concentrations¹⁰.
76 Soil as one of the important environmental media for POPs sink, quantitative study of
77 spatial distribution of chemical concentrations in soils should also be paid more
78 attention. According to our previous study, more than 95%, 91%, 53% of the total BaP
79 ¹³, Phen², PFOS¹⁴ concentrations would be found in soils in Bohai coastal region
80 which implied that soils would serve as the predominant sink of POPs. Similar results
81 for PCBs and PBDEs were also found in Toronto⁶, which could be attributed to the
82 physical and chemical properties of different chemicals¹⁴. Hence, a quantitative study

83 of urban-rural POPs gradients in soils should be conducted for better understanding of
84 the POPs behavior and for ecological risk assessment and management of POPs at a
85 regional scale.

86 In this paper, measured data of PAHs, PFASs and PCBs were used to study the
87 concentration gradients of POPs between urban and rural areas. We will (1) first outline
88 the potential factors affecting the spatial distribution and accumulation of POPs; (2)
89 quantitatively describe what are the urban-rural gradients in the study area through
90 theoretical analysis (formula deduction) and validation of measured data; (3) illustrate
91 what extent of POPs in soils is attributed to land cover and population distribution; and
92 (4) demonstrate that multi-driving factors and urban-rural gradients are applied to
93 calibrate and predict the distribution of concentrations of PFASs, PAHs and PCBs in
94 soils.

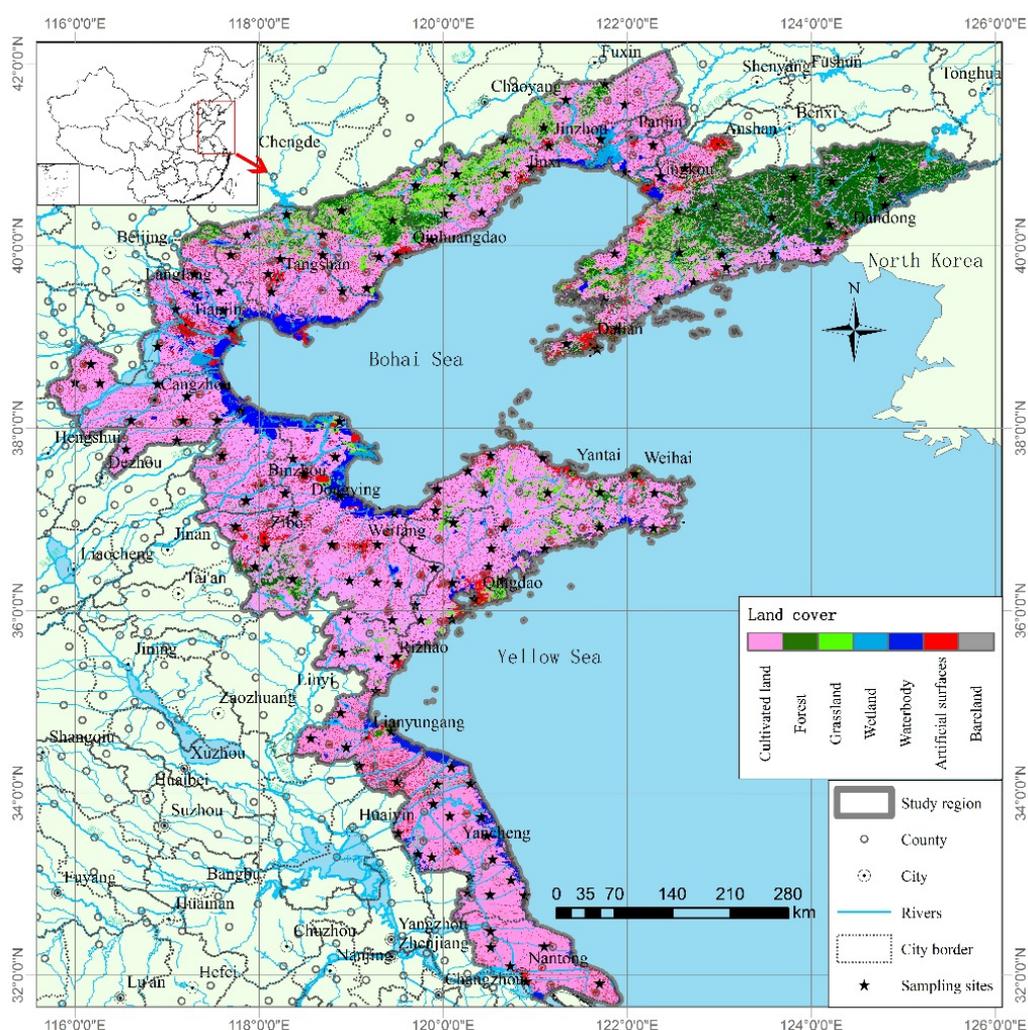
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96 **2. Material and Methods**

97 **2.1 Study area and sample collection**

98 The study area consists of four provinces (Liaoning, Hebei, Shandong and Jiangsu)
99 and one municipality (Tianjin) around the Bohai Sea and Yellow Sea. A total of 153
100 surface soils were collected from 22 coastal cities (Figure S1). The study area was
101 divided into 50×50 km² grids, and 2~3 sites were included in each grid by considering
102 site selection principle of representative land cover, spatial uniformity, regional
103 environmental representation, and balanced distribution of sites. Top surface soil
104 samples (0-10cm) were collected using a stainless steel trowel that had been rinsed with

105 methanol. Each sample was composed by five sub-samples that were collected from the
106 center and four corners of an area of 100×100 m². Information about sampling sites
107 including coordinates, land use and detailed descriptions were recorded in the field. Soil
108 samples were transferred and stored in clean polypropylene (PP) zip lock bags. All
109 samples were dried in air, homogenized with a porcelain mortar and pestle, sieved with
110 a 2 mm mesh for PFASs and a 100 mesh for PAHs and PCBs, and stored in 250 mL PP
111 bottles at room temperature until extraction.



112

113 Figure S1. Distribution of sampling sites and land cover information along Bohai Sea

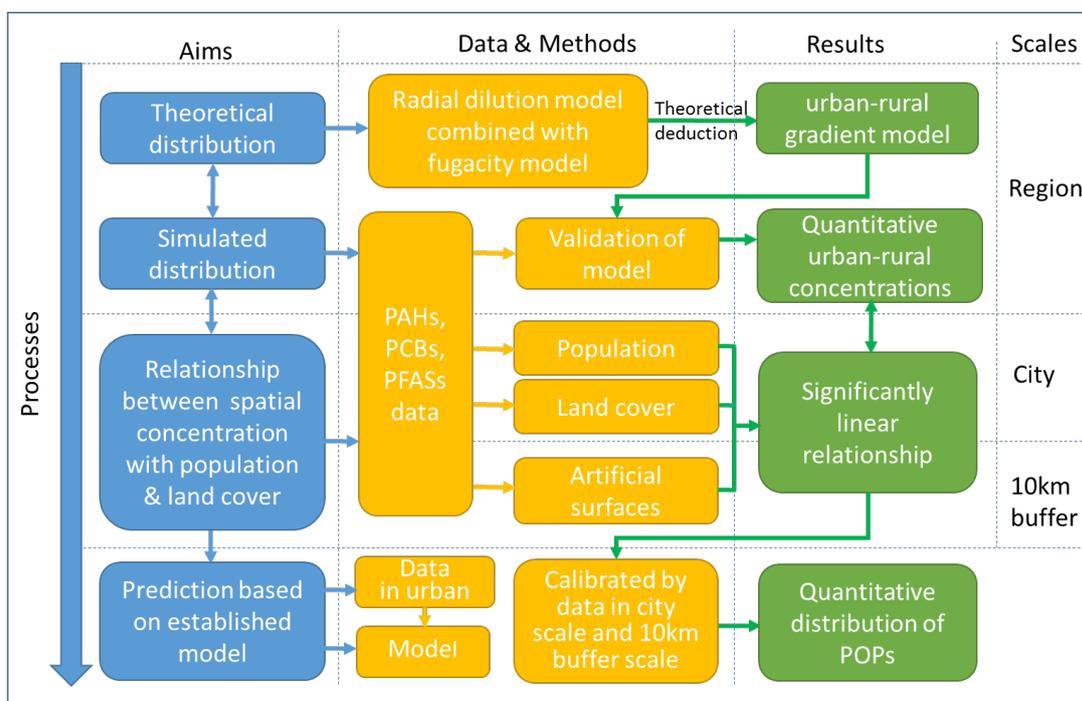
114 and Yellow Sea. **(Supporting materials)**

115

116 2.2 Analysis of samples and data

117 Analysis of PFASs, PAHs and PCBs, quality control and quality assurance were
118 using the same methods as those described previously by Meng et al.¹⁵, Jiao et al.¹⁶ and
119 Gao et al.¹⁷, respectively. Detail information are given in the SI.

120 Data analysis, including statistical analysis and t-test, was performed on SPSS 17.0.
121 Linear correlation and urban-rural model simulation were calculated on EXCEL 2013,
122 and maps of correlations were drawn on ORIGINLAB 2015. A confidence level of 95%
123 was used for the statistical tests and correlation analysis. Land cover data for year 2010
124 at 30 m resolution was from GlobeLand30 (<http://glc30.tianditu.com>). Spatial data
125 analysis and calculation, and spatial distribution map were all performed on ARCGIS
126 10.2. The flowchart of our study was shown as Figure 1.



127

128

Figure 1 Architecture map

129

130 **3. Results and discussion**

131 **3.1 Contaminant concentrations and distribution in soils**

132 A summary statistics of PCBs, PFASs and PAHs concentrations in soils is
133 presented in Table S1 for comparison. Coefficients of variation for PCBs, PFASs and
134 PAHs were 1.74, 0.80 and 2.11 (Table S1a), respectively, which indicated that spatial
135 variability of the POPs levels in soils are large due to different natural and
136 anthropogenic impacts, and the sampling sites from the highly contaminated areas
137 resulted in a wide range of POPs levels in soils. Since our focus was put on the
138 relationship between urban and rural soil levels but not on highly contaminated areas
139 in this paper, the outliers were not taken into consideration (Table S1 b, c, d do not
140 include 2 PCBs, 3 PFASs and 2 PAHs outliers). A reference value in Table S1 is the
141 limitation of toxic substances in soils^{18, 19}.

142 **3.1.1 Polychlorinated biphenyls**

143 The concentrations of PCBs ranged from 2.94 to 385.67 ng/g, with mean value of
144 19.89 ng/g (Table S1^a). The mean levels of PCBs were larger than the global
145 background concentration for 5.41 ng/g in soil²⁰, and also larger than mean soil
146 concentration for 0.52ng/g across China²¹ where both urban and rural soils were
147 sampled. Nevertheless, the values in the present study were comparable with other
148 studies (average 1.07 - 35.5 ng/g) in East China²², Northeast²³, South²⁴ and North
149 China²⁵. PCBs levels increased from barrenland and grassland to arable, forests and
150 orchard, and those in urban soils are largest. PCBs levels in the urban soils (24.78
151 ng/g) were three times higher than that in grassland (8.55 ng/g) and barrenland soils

152 (8.33 ng/g), which indicated that primary sources of PCBs in rural area might come
 153 from urban area through air transport. A high degree of land development with high
 154 population density in the study area implied that barrenland was relatively less
 155 affected by human activities, which may be the reason why the lowest PCBs levels
 156 were found in the barrenland soils. The mean PCBs concentration in the study area
 157 was close to the 20ng/g limitation¹⁸, while the median was far lower than this value.
 158 The mean PCBs concentrations in soils from Tangshan, Qingdao, Tianjin, Rizhao,
 159 Cangzhou, Jinzhou and Zibo were larger than the reference value (Table S1b), which
 160 indicated that the soils in this region were subject to intensive domestic and industrial
 161 activities in and around the sampling sites.

162 **Table S1 Descriptive statistics of PAHs, PCBs and PFASs in soils (Supporting materials)**

Values (ng/g) ^a	PCBs	PFASs	PAHs	City ^b	N*	PCBs	PFASs	PAHs	
Ref. value	20.00**	-	1000.00***			20.00	-	1000.00	
Mean	19.89	7.51	466.73	Tangshan	9	30.97	6.14	875.15	
Minimum	2.94	2.76	26.63	Dandong	9	14.98	6.42	818.70	
Maximum	385.67	63.97	11463.52	Zibo	3	20.93	9.76	713.45	
Std. Deviation	34.73	6.05	985.69	Tianjin	7	25.45	7.07	503.71	
Variance	1206.29	36.57	971579.28	Dalian	11	18.36	6.75	498.07	
Skewness	8.17	6.86	9.54	Binzhou	4	14.11	7.87	489.94	
Kurtosis	82.15	56.77	103.91	Nantong	6	10.93	8.27	478.31	
Percentiles	10	5.11	4.28	123.66	Jinzhou	5	21.65	7.45	406.59
	25	7.97	5.02	167.86	Qingdao	10	28.69	7.31	362.82
	50	12.09	6.54	273.02	Yingkou	4	18.84	6.16	325.04
	75	17.69	7.68	449.66	Lianyungang	6	9.04	6.95	323.35
	90	36.89	10.25	825.74	Dongying	5	14.42	6.73	320.52
Distance from Edge of City ^c	PCBs N	PFASs (151*)	PAHs (151*)	Huludao	10	10.09	6.22	293.70	
0-2km	18	20.70	8.37	763.15	Yancheng	16	10.60	6.59	281.83
2-5km	17	26.64	7.26	412.09	Qinhuangdao	5	6.91	6.59	258.94
5-7km	12	21.96	7.00	433.64	Cangzhou	11	22.77	5.78	245.56

Values (ng/g) ^a	PCBs	PFASs	PAHs	City ^b	N*	PCBs	PFASs	PAHs	
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	75	17.69	7.68	449.66	Lianyungang	6	9.04	6.95	323.35
	90	36.89	10.25	825.74	Dongying	5	14.42	6.73	320.52
Distance from Edge of City ^c	PCBs N (151*)	PFASs (150*)	PAHs (151*)	Huludao	10	10.09	6.22	293.70	
7-10km	13	19.18	6.59	372.65	Yantai	10	14.19	6.66	244.81
10-15km	23	14.60	6.95	333.10	Panjin	3	6.41	5.58	226.94
15-20km	13	10.37	6.25	285.56	Weifang	10	13.00	7.78	224.44
20-30km	40	13.50	6.28	272.85	Rizhao	5	23.72	6.12	193.91
30-50km	17	13.63	6.02	220.56	Weihai	4	11.60	6.00	170.78
Landuse ^d	N*	PCBs	PFASs	PAHs	Landuse ^d	N*	PCBs	PFASs	PAHs
Urban soils	19	24.78	8.47	570.80	Arable soils	89	16.22	6.63	367.68
Orchard soils	10	19.62	6.30	256.65	Grassland soils	2	8.55	5.67	716.33
Forest soils	22	16.30	6.44	340.97	Barrenland soils	11	8.33	6.59	144.08

163 *the number does not contain the abnormal value; ** 18 *** 19

164 3.1.2 Polyaromatic hydrocarbons

165 The mean concentration of PAHs in the study area was 466.73 ng/g that was
166 significantly lower than the reference value (1000 ng/g). However, the PAHs
167 concentrations of 12 soil samples are higher than the reference value, where 10
168 sampling points are located at or around the urban areas (<2 km), an outlier (11463.52
169 ng/g) at Cangzhou industrial zone, and an outlier (3803.45 ng/g) near the steel plant in

170 Tangshan. According to guideline values for soil contamination ²⁶, this reference value
171 is a threshold for human exposure risks. The mean PAHs concentrations in different
172 cities showed that Tangshan, Zibo, Tianjin, Dandong and Dalian are relatively high, and
173 those in Weihai and Rizhao are relatively low (Table S1b). The distribution of PAHs
174 concentrations in soils in different cities are similar with the distribution of PAHs
175 emissions reported by Shen et al. ²⁷. The mean value of PAH concentrations for urban
176 soils (570.80 ng/g) is large, and this level is about half order of magnitude higher in
177 forest (340.97 ng/g), arable (367.68 ng/g) and orchard soils (256.65 ng/g). The mean
178 PAHs concentration in grassland soils is the largest (716.33 ng/g), however, these two
179 sites may be more representative of the local soil concentrations. Higher levels of PAHs
180 in grassland were also found in other research¹⁹, which indicated PAHs in the grassland
181 may be easily accumulated in the grass soils. The mean PAHs concentration in
182 barrenland soils is also the lowest, similar with that of PCBs, which showed less impact
183 of human activities. However, all the different concentrations in different soils probably
184 reflect a combination of the local heating sources, traffic as well as long-range transport.
185 And also, we could speculate about the sources of elevated concentrations from rural
186 area to urban area. PAHs in soils surrounding steel plants, power plants and industrial
187 parks were found with high concentrations. Due to a number of heavily polluting
188 factories have been moved from cities to their suburbs over the past decade, soil
189 pollution in urban areas may be alleviated. However, since human population and heat
190 of combustions in urban areas are relatively concentrated, PAHs concentrations in urban
191 soils remain relatively higher, while rural areas may be mainly impacted by local straw

192 combustion and household coal burning in winter.

193 **3.1.3 Perfluoroalkyl and polyfluoroalkyl substances**

194 A distribution with lower level and narrower range was observed for PFASs
195 compared with PAHs and PCBs. The concentrations of PFASs ranged from 2.76 to
196 63.97 ng/g, with a mean value of 7.51 ng/g (Table S1a). The levels of PFASs in the
197 study area were significantly lower than those in Shanghai (mean value 10.4 ng/g²⁸),
198 while significantly higher than those with mean value of 0.98 ng/g around Bohai Sea
199 ¹⁵, which indicated that land use, population density, industrial emission, historical
200 accumulation may lead to differences in the regional distribution. PFASs in urban
201 soils had generally higher mean values than the other soils, while for arable, orchard,
202 barrenland and forest soils the mean and median PFAS values stayed close. Similar
203 distribution trend was also reported by Meng et al. ¹⁵. It reflects continuous
204 accumulation of contaminants in soils in the study area has probably led to increasing
205 concentration, and intensive industrial and domestic processes in urban areas might
206 greatly affect the PFASs concentrations in the surrounding soils. The mean PFASs
207 concentration in soils from Zibo was the largest (Table S1b), which may be associated
208 with emissions of solid waste, sewage and waste gas from fluorine chemical related
209 industries. Besides, the mean PFASs concentrations in soils from Nantong, Tianjin,
210 Jinzhou, Qingdao and Weifang were relatively higher (from 7.07 to 8.27 ng/g), for
211 urbanization relevant indicators such as industrialization, urban population, and GDP
212 per capita are relatively larger. The level of urbanization and change of PFOS
213 concentrations are positively correlated ¹⁵ which may be the key factors for the

214 differences of spatial PFASs concentrations among the different cities.

215 **3.2 Theoretical derivation of urban-rural gradient model in soils**

216 Cities could be regarded as major sources of PAHs, PCBs and other POPs to the
217 rural regions, which was reflected in huge differences in contaminant emissions
218 between urban and rural areas. Lots of studies on POPs concentrations in different
219 media were reported, but little was studied about POPs behavior between urban and
220 rural areas at a regional scale². Quantitative studies on urban and rural concentrations
221 of pollutants mainly focus on those in atmospheric media using ratios of urban to rural
222 concentrations^{12, 29, 30}. The use of these ratios were proved to be suitable for simulation
223 of concentrations in atmosphere¹⁰. However, rather than simple ratios in air, in our
224 study, the dependent distance functions adapted to soil concentration gradients between
225 urban and rural areas were clearly demonstrated by using formula deduction and
226 validation of a large number of measured data.

227 To better understand the mechanism of chemical transport between urban and
228 suburban areas, we adopted the fugacity equations to describe the transport process
229 from atmosphere to soil (Figure S2). In the model, the bulk Z values (mol/Pa/m³) which
230 are specific to the capacity of a phase for a chemical, and D values (mol/Pa/h) which
231 are used to quantify the intermedia transport and transformation processes of chemicals
232 between the two compartments, were used to calculate POPs transport from air to soil
233 (details of D and Z values were described by Mackay³¹).

234 We assume that the pollutants in soils are mainly from the interactions between the
235 atmosphere and soil while neglect other processes such as sewage irrigation and solid

236 waste pollution. When the chemical concentrations in atmosphere and soil reach
 237 equilibrium in urban or rural area, the following relationship exists for soils:

$$238 \quad f_a (D_1 + D_2 + D_3 + D_4) = f_s (D_5 + D_6) \quad (1)$$

239 Concentrations of chemicals in air and soil were as follows:

$$240 \quad C_a = f_a \times Z_a \quad (2)$$

$$241 \quad C_s = f_s \times Z_s \quad (3)$$

242 Where, f is fugacity (mol/Pa/m³); C is concentration; subscript a is air; w , water; s ,
 243 soil; Q , aerosol; $D_1, D_2, D_3, D_4, D_5, D_6$ are transportation process of pollutant diffusion
 244 from air to soil, wet deposition of gas, dry deposition, wet deposition, diffusion from
 245 soil to air, and degradation, respectively. According to the equations (1-3), relations of
 246 concentrations between soil and air when reaching a steady state could be expressed by
 247 the equation (4). The parameters in equation (4) could be calculated by the equations
 248 (4-14), using which we could analyze the potential factors that impact the chemical
 249 transport in the different areas or media ².

$$250 \quad C_a = C_s \frac{(D_5 + D_6)Z_a}{(D_1 + D_2 + D_3 + D_4)Z_s} \quad (4)$$

$$251 \quad Z_a = 1/(RT) \quad (5)$$

$$252 \quad Z_w = 1/H = Z_a/K_{aw} \quad (6)$$

$$253 \quad Z_Q = 10^{(\log K_{oa} + \log f_o - 11.91)} \times Z_a \times \rho_Q \times 10^9 \quad (7)$$

$$254 \quad Z_s = 0.41 \times K_{ow} \times Z_w \times f_{oc} \times \rho_p \quad (8)$$

$$255 \quad D_1 = A / (1 / (k_{va} \times Z_a) + 1 / k_{vw} \times Z_w) \quad (9)$$

$$256 \quad D_2 = A \times U_r \times Z_w \times (1 - f_w) \quad (10)$$

$$257 \quad D_3 = A \times v_Q \times Z_Q \times U_p \times (1 - f_w) \quad (11)$$

258 $D_4 = A \times S_r \times v_Q \times Z_Q \times U_r \times (1 - f_w)$ (12)

259 $D_5 = A / (1 / (k_{va} \times Z_a) + 1 / k_{vw} \times Z_w) = D_1$ (13)

260 $D_6 = 0.693 \times V_s \times Z_s / \tau_s$ (14)

261 R is the gas constant (8.314 Pa m³/mol K); T , absolute temperature (K); f_{oc} , the
 262 fraction for organic carbon; ρ_Q , and ρ_p , the density of aerosols and soil particles (kg/m³);
 263 A , the media interfacial area (m²); K , Partition coefficient; K_{Qa} , aerosol/ air partition
 264 coefficient; K_{aw} , aerosol/ water partition coefficient; K_{ow} , octanol/water partition
 265 coefficient; k , the mass transfer coefficient (MTC), (m/s); k_{va} , the air side MTC (m/s);
 266 k_{vw} , the water side MTC (m/s); v , the volume fraction; v_Q , aerosols volume fraction; v_a ,
 267 air volume fraction; v_w , water volume fraction; v_s , solid volume fraction; S_r , the
 268 scavenging ratio, U_r , the rain rate (m/h), U_p , the dry deposition rate (m/h), f_w , the canopy
 269 wet interception fraction, τ_s , half-life of concentrations in soils(h); V_s , volume (m³).

270 The gradient of chemical atmospheric concentrations caused by the point source
 271 could be generalized by using a radial dilution model ³²,

272 $C_r = C_0 \times r^m$ (15)

273 where C_r is air concentration from a given radius r to the point source with concentration
 274 C_0 , and similarly identification of point source origin of atmospheric emission could
 275 also use this type of model to calculate with spatially-distributed air concentration data¹⁰.

276 At a regional scale, urban areas are generally regarded as major sources of POPs to the
 277 neighboring rural areas ⁶, and also urban areas are relatively small compared with the
 278 rural area, so atmospheric emission from urban areas could be generalized as point
 279 source emission in a large area. Some studies showed that similar atmospheric

280 concentration gradients from urban to rural areas, e.g., ratios of urban-rural air
281 concentrations³⁰, urban-rural air comparisons³³, and simulation results of spatially-
282 distributed POPs concentrations in air^{4, 6}. Melymuk et al¹⁰ demonstrated that PAHs,
283 PCBs and PBDEs concentrations in air could be simulated using logarithmic form of
284 the equation (15) with measured spatially distributed air concentrations.

285 If C_o is the pollutant concentration in the urban center, the parameter m should have
286 significantly spatial difference for different sizes of cities. In simple term, each city
287 should calibrate the model parameters using the surrounding measured data, e.g. the
288 data from urban center to rural area, which might need more measured data to support.
289 If we assume all the chemical concentrations in the urban are the same, and then C_o is
290 the mean value, r should be the distance from the city edge. We use the constant ε value
291 to correct the distance for chemical concentration differences caused by different areas
292 of cities. The equation (15) is corrected for air gradient in a regional scale using
293 logarithmic transformation:

$$294 \quad \ln\left(\frac{C_r}{C_o}\right) = m \times \ln(\varepsilon r) \quad (16)$$

$$295 \quad \ln\left(\frac{C_r}{C_o}\right) = m \times \ln(r) + m \times \ln(\varepsilon) \quad (17)$$

296 where ε , a constant close to 1, denotes the degree of influence of urban area differences
297 on concentration gradient. ε value = 1 represents no significant effect of area
298 differences in cities on chemical gradient; ε value < 1 means that highest
299 concentration was in urban area, and it has some influences of area differences in cities
300 on chemical gradient; ε value >1 represents chemical concentrations in suburban
301 larger than in urban and rural regions probably due to industrial transfer from urban to

302 suburban areas. The parameter m is the slope of the linear regression between
 303 $\ln(C_r/C_0)$ and $\ln(\varepsilon r)$.

304 According to the equation (4, 15 and 16), the chemical concentrations in soil
 305 between urban and rural areas could be represented as follows:

$$306 \quad dC_{rs} \frac{(dD_{r5} + dD_{r6})Z_{ra}}{(dD_{r1} + dD_{r2} + dD_{r3} + dD_{r4})Z_{rs}} = dC_{us} \frac{(dD_{u5} + dD_{u6})Z_{ua}}{(dD_{u1} + dD_{u2} + dD_{u3} + dD_{u4})Z_{us}} \times (\varepsilon r)^m \quad (18)$$

307 where dC and dD denote chemical soil concentration per unit area and transportation
 308 process per unit area among the urban-rural transects. The subscript 'r' of variable dC
 309 and dD represents rural region from a given radius r to the urban center, and urban
 310 region; and 'u' represents urban area. Usually, Z values were related to the capacity of
 311 a phase for a chemical, so Z values for urban or rural media could be considered as the
 312 same, and then we can get equation (19).

$$313 \quad \ln\left(\frac{dC_{rs}}{dC_{us}}\right) = \ln\left(\frac{(dD_{u5} + dD_{u6})(dD_{r1} + dD_{r2} + dD_{r3} + dD_{r4})}{(dD_{u1} + dD_{u2} + dD_{u3} + dD_{u4})(dD_{r5} + dD_{r6})}\right) + m \times \ln(r) + m \times \ln(\varepsilon) \quad (19)$$

314 However, Song et al.² simulated the different transportation processes between
 315 urban and rural areas, which proved that parameters such as vegetation cover, soil
 316 properties, and atmospheric aerosol concentration significantly affect the chemical fate.
 317 If we actually consider the most potential parameter differences ² resulting from the
 318 given radius r to the urban edge, parameters in equation (19) and equation (9-14) except
 319 for v_Q , f_w , and Z_s all could be summarized as constants. Then we get equation (20), in
 320 which a , b , c , e , f and g could be got by deducing equations 9-14 and 19. v_Q in lower air
 321 actually is the function of r , but the distributions of f_w and Z_s values depend on
 322 vegetation cover and soil property. So, in equation 20, if anyone wants to study the
 323 effect of land cover and soil property on chemical distribution in soil, this model is a

324 simple method. Value m could be got from the concentration gradient in air, and then
325 the distributions of land cover and soil property would be considered to calculate all the
326 soil concentrations. Of course, if we assume the f_w and Z_s were spatially homogeneous,
327 and then C_{rs}/C_{us} was the function of single variable r (equation 20).

$$328 \quad \ln\left(\frac{dC_{rs}}{dC_{us}}\right) = \ln\left(a \frac{b + (c + ev_{\rho} + fv_{\rho})(1 - fw)}{b + gZ_s} dA\right) + m \times \ln(r) + m \times \ln(\varepsilon) \quad (20)$$

$$329 \quad \ln\left(\frac{dC_{rs}}{dC_{us}}\right) = m \times \ln(r) + m \times \ln(\varepsilon) \quad (21)$$

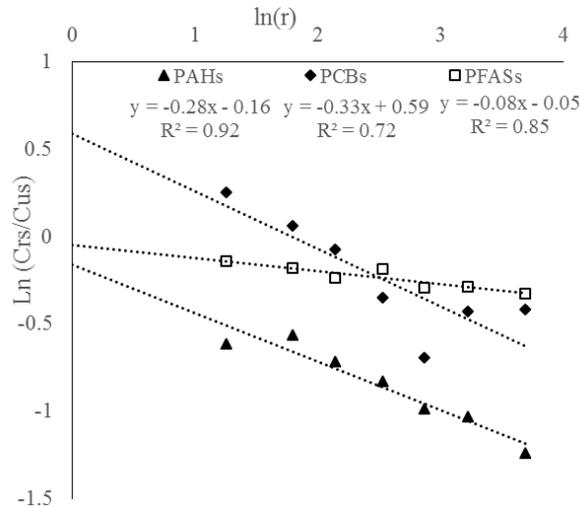
330 According to equation (19), if we assumed that the transportation process (D values)
331 in urban areas had no difference with that in rural areas, then equation (21) and (17)
332 were similar. As can be seen from equation (20), slope m is unrelated to the
333 transportation process such as atmospheric deposition, and in theory the slope m for
334 chemical concentration in soil and air is the same. We think it is important finding for
335 quickly estimating atmospheric and soil concentrations with lower cost and less time
336 due to mutual conversion relationship at a large scale. However, we were lack of the
337 measured data to prove the reliability, and in this paper we want to confirm the large-
338 scale adaptability of equation (21) and the possible affecting factors.

339 **3.3 Urban-rural gradient and validation of model**

340 In our study, PCB, PFAS, and PAH concentrations in the urban soils were
341 significantly larger than those in rural areas (Table S1d), where the urban soil
342 information of sampling sites was supported by the fieldwork including small down
343 town, built-up area, and green space. Generally, the concentrations of POPs gradually
344 decrease from urban to rural soils. Urban concentrations (0-2 km sites) were 3.5, 1.5
345 and 1.4 times higher than the rural concentrations (30-50 km sites) for Σ PAHs, Σ PCBs

346 and ΣPFASs, respectively (Table S1d). Mean PFAS and PAH concentrations in urban
347 soils were the largest, which indicated that industrial and domestic emissions of PFASs
348 and PAHs per unit area in cities were the largest. PAH concentrations in 5-7 km from
349 the urban center were slightly greater than those in 2-5km, with relatively higher
350 standard deviation of 432 ng/g. It implied relatively greater suburban industrial
351 emissions in 5-7 km from the urban center, and the aggregation and non-uniformity of
352 industrial distribution might be the main reason for the greater standard deviation of
353 PAH concentration. Urban–rural variation in PFAS gradients showed no large PFAS
354 emissions to air in suburbs, and PFASs in urban areas were the main sources at the
355 regional scale. PCB concentrations in soils with 2-5km away from the urban center
356 were the largest, and the possible reason was the industrial transfer from urban to
357 suburban areas.

358 In order to study urban-rural gradient rule, we count the distance of each point from
359 the nearest edge of city in ARCGIS software using 2010 land cover data, and then PCBs,
360 PFASs and PAHs concentrations in different segmented distances were analyzed.
361 Although the given heterogeneity of concentrations within an urban area was found in
362 our study, we hardly explain a generalized characteristic of cities using only 6 urban
363 sampling sites at a large region scale. Besides, urban areas accounted for a small
364 proportion of the entire study area. Therefore, the urban-rural gradients were mainly
365 concerned and each POP was considered the same in the urban area. Artificial urban
366 surface and the surroundings with 2 km distance were all regarded as urban areas in this
367 study due to industrial transfer from urban to suburban areas ².



368

369 **Figure 2. A logarithmic relationship exists between distance from edge of**
 370 **urban areas and the other sampling soil concentrations. Slops were significant at**
 371 **p<0.05 based on a linear regression t-test.**

372

373 Due to the elevated concentrations in the vicinity of the urban areas, we chose
 374 equation (21) to test the hypothesis that the urban areas were the main sources of
 375 chemical concentrations to the study region. We used the measured data to validate the
 376 urban-rural gradient model with a known source region (0-2km) and compare the
 377 degree to which that region influences concentrations of PFASs, PCBs and PFASs.
 378 According to the analysis results from more than 150 sampling sites, a significant
 379 relationship was found between $\ln(C_{rs}/C_{us})$ and $\ln(r)$ for PFASs, PCBs and PAHs
 380 (Figure 2), where r is the distance in km from the edge of urban areas and C_{us} is the
 381 average concentrations in 0-2km urban areas, which proves that urban-rural gradient
 382 model (equation 21) is applicable for simulating POPs concentrations in soils at the
 383 regional scale. PAHs decreased most rapidly with increasing distance from the urban
 384 area, dropping off by 71% at 30-50 km urban areas, while PCBs and PFASs decreased

385 by 34% and 28%, respectively. The different slopes of the regression for all compounds
386 indicated that the degree of declining concentrations along the urban–rural gradient
387 (Figure 2) may be affected by urban emissions, land cover, physicochemical properties,
388 and background emissions in suburban and rural areas.

389

390 **3.4 Relationship of soil concentration with population combined by land cover** 391 **data**

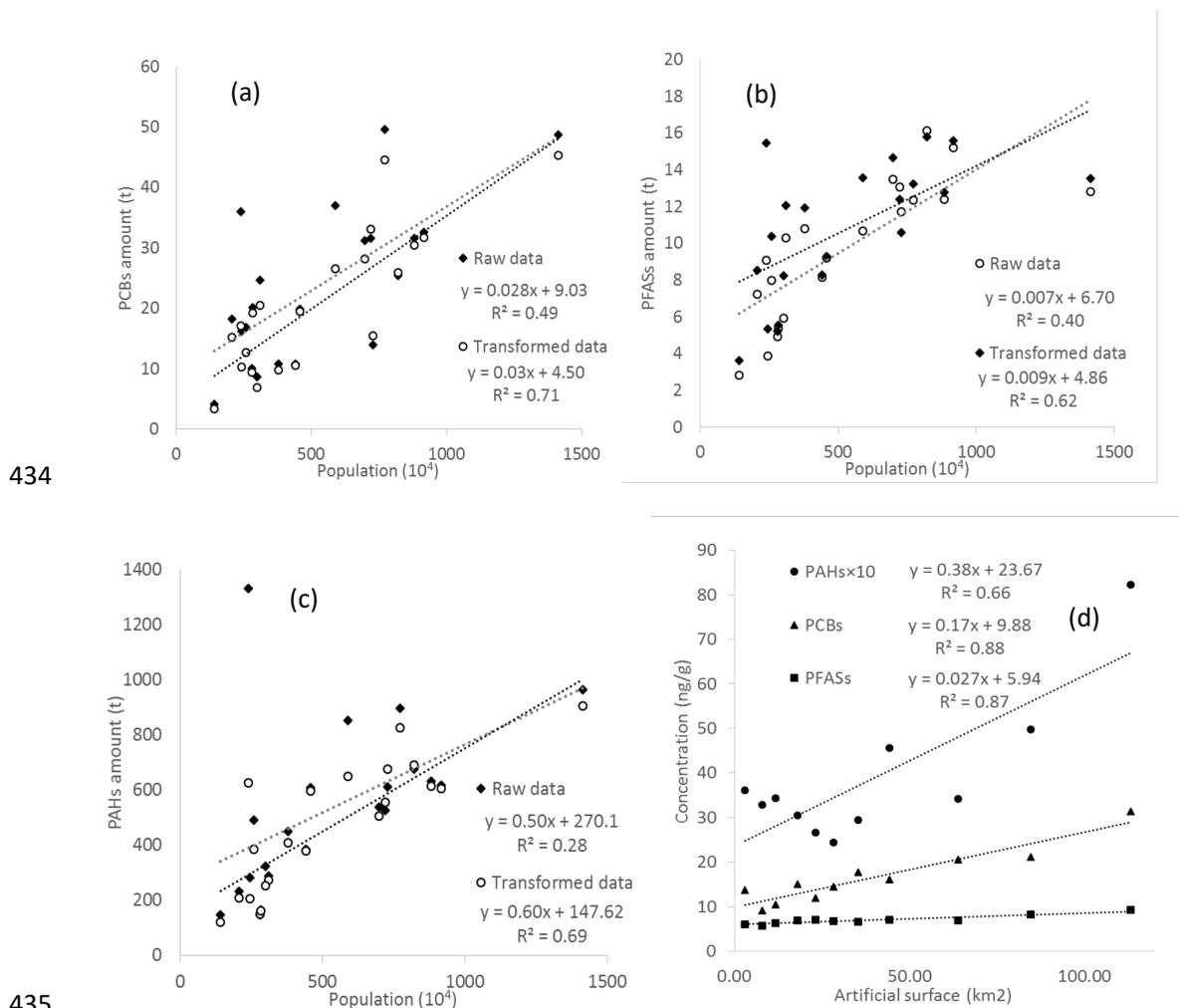
392 As mentioned above, soil concentration of PCBs, PFASs and PAHs generally
393 decreased from urban to rural areas in the study region. Meanwhile, variability of POPs
394 spatial distribution was very large (Table S1b) even at the same distance from the city
395 center. There may be a lot of reasons resulting in such a huge difference in the
396 concentrations of pollutants in the soil, for example, structure and intensity of POPs
397 emissions, wind speed and direction, topography, population distribution, wastewater
398 discharge, the amount of coal, land cover types, soil physical properties, soil organic
399 carbon³⁴, POPs properties and so on. Among so many factors, population is regarded as
400 one of the most important factors affecting POPs emission, e.g estimation of PAH
401 emissions^{13, 27}, PFAS domestic emissions^{35, 36} and PCB emissions^{37, 38}. At the same
402 time, many articles have demonstrated that the population and concentrations for POPs
403 in different environmental media showed a positive correlation^{15, 39, 40}. Land cover is
404 also one of the important factors that influence the distribution of POPs, and different
405 types of land use have a very significant impact on the distribution of POPs⁴¹. Therefore,
406 in this study, we try to analyze the relationship between spatial concentration of

407 pollutants and anthropogenic activities by using representative parameters of
408 population and land cover data.

409 The relationship of different city population and amount of POPs in surface soil
410 was analyzed, which showed a significant linear correlation (raw data in Figure 3). The
411 amount of POPs in each city was calculated according to mean concentration multiplied
412 by the urban area, soil depth (10cm), and soil density that were assumed all the same in
413 the study area. However, we found many outliers in Figure 3, such as the relatively high
414 POPs amount in Dandong, Jinzhou, Dalian, Tangshan and Huludao, which may be
415 caused by high emission per person, long-distance transportation, and long-term
416 accumulation. We compared land use and sampling sites in these cities, and found that
417 woodland and grassland areas in these cities were relatively large with a small degree
418 of landscape fragmentation (Table S1 and Figure S1). According to the principles
419 shown in Figure S2, chemical emissions and atmospheric deposition were more
420 susceptible to the surrounding anthropogenic activities. Population densities around a
421 large area of meadows and forests in the mountains were small, but those around arable
422 land were higher because of widely distributed villages (Figure S1). If we did not
423 consider the impact of wind speed, easy accumulation in meadows and woodlands with
424 high soil organic matter, and long-distance transport, the arable land may accept more
425 atmospheric deposition of POPs produced from the surrounding towns and villages. In
426 this regard, we use the coupling land use data to make a conversion for the total amount
427 of POPs, as shown in the following equation:

$$428 \quad TA = \sum_{k=0}^n C_{mean} * \rho * CA_k * h + \sum_{l=0}^n C_{bg} * \rho * CA_l * h \quad (22)$$

429 where, TA , translated data of total amount (ng); C_{mean} , mean concentration for each city
 430 in urban and arable soils, ng/g; C_{bg} , background value of different chemicals, ng/g; k ,
 431 land cover including artificial surface and arable land; l , land cover including grassland,
 432 woodland and barren land; ρ , soil density, 1.6g/cm^3 ; CA , area of land cover, cm^2 ; h ,
 433 depth, 10cm.



436 **Figure 3. The relations of POPs among soil, population and landscape patterns in**
 437 **different cities (a-c), and the relations between artificial surface area in 10km**
 438 **buffer area and concentration of POPs (d).**

439

440 It could be seen from Figure 3, linear fitting between transformed data and

441 population in different cities got better, compared with the raw data. R^2 of PCBs, PFASs
442 and PAHs were increased from 0.49, 0.40 and 0.28 to 0.71, 0.62 and 0.69, respectively.
443 Compared the transformed data with raw data (Figure 3), all the chemicals in Dandong,
444 Dalian and Tangshan were still above the trend line, indicating that the emissions per
445 capita in those regions were higher than other regions.

446 In order to study whether the chemical concentrations and the surrounding land
447 cover are interrelated, 2km, 5km, 10km and 20km buffer zones for each sampling site
448 were made using ARCGIS software, and the area of artificial surface in each buffer
449 zone was counted. We used artificial surface area to reflect cities and towns. Through
450 statistical analysis of the relationship between soil concentration and the surrounding
451 artificial surface area, we found that the artificial surface area within 10km buffer zone
452 and pollutant concentrations had a significant linear relationship (Figure 3), and
453 detailed statistical data were shown in Table S2. The linear correlation between artificial
454 surface in the other buffer zones and soil concentration was not significant. The range
455 of artificial surface area within 10km buffer zone was from 0 to 141 km². The soil
456 concentrations of PAHs, PCBs and PFASs with 113km² artificial surface surrounding
457 the sampling site were 2.3, 2.3 and 1.5 times than those with 3km² artificial surface.
458 Compared with the linear relationship between PAHs and artificial surface, R^2 of PCB
459 and PFAS were 0.88 and 0.87, respectively. This may be affected by the diversification
460 of PAHs sources, capacity of long-distance transportation, concentrated discharge of
461 industrial parks, and so on. The relationship between artificial surface area with 10km
462 scale and the chemical concentration may be a good approach to quickly estimate the

463 spatial distribution of POPs at the regional scale.

464 **3.5 Predicting the spatial distribution of POPs concentrations in soils**

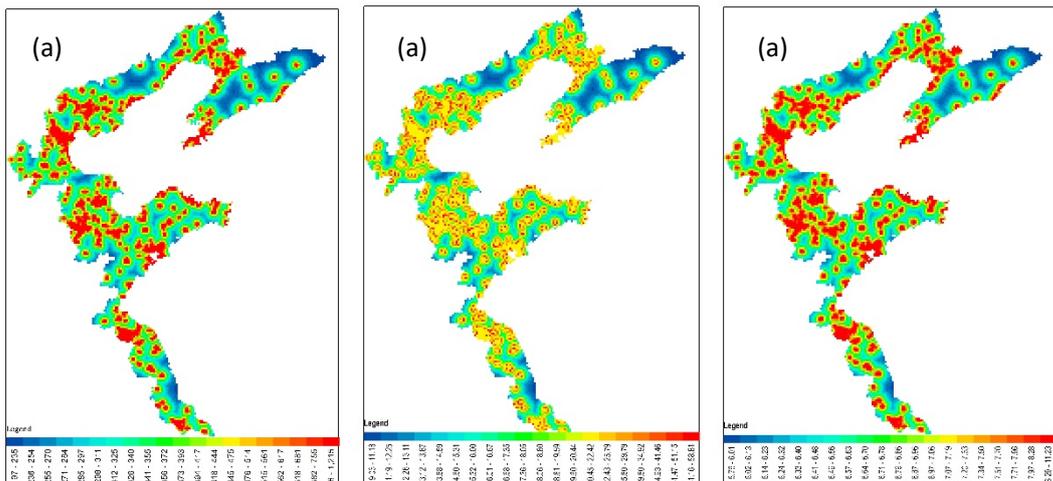
465 If PAH, PCB and PFAS concentrations in urban soil were known, Equation 21
466 could be used to calculate the concentration gradient in all suburbs. In this study, we
467 first generated 5×5km grids (Total 8692 grids) using ARCGIS, and then the distance
468 from the central point of each grid to the closest city edge was calculated. Not all of the
469 22 cities have monitoring data of POPs in urban soils, and a little measured data may
470 not be representative for the POPs concentrations in the urban soils. So, the first step,
471 the mean concentrations in this region were used as single parameter for the established
472 urban–rural gradient model to calculate soil concentrations for 5×5km grid points.
473 Spatial distribution of urban-rural POPs simulation results is shown in **Figure 4a**, and
474 parameters for calculation were shown in SI-Step1. Second step, the mean
475 concentration (parameter C_{us}) in each urban soil was calibrated according to the
476 relationship between the population and the chemical amount in soils (Figure 3, a-c),
477 while maintaining the total amount of pollutants unchanged in each city. The calibrated
478 results for parameter C_{us} were shown in Table S3. The spatial distribution of simulated
479 concentration in urban and suburban areas for the second step was shown in **Figure 4b**.
480 Finally, according to the artificial surface data in 10km buffer zones, chemical
481 concentrations in each site were calibrated again, and then regional distribution map of
482 contaminants was generated (Figure 4c). Calibrated process was shown in SI-Step3 and
483 Table S4.

484 Predicted and measured data of PAHs, PCBs and PFASs from urban to rural areas

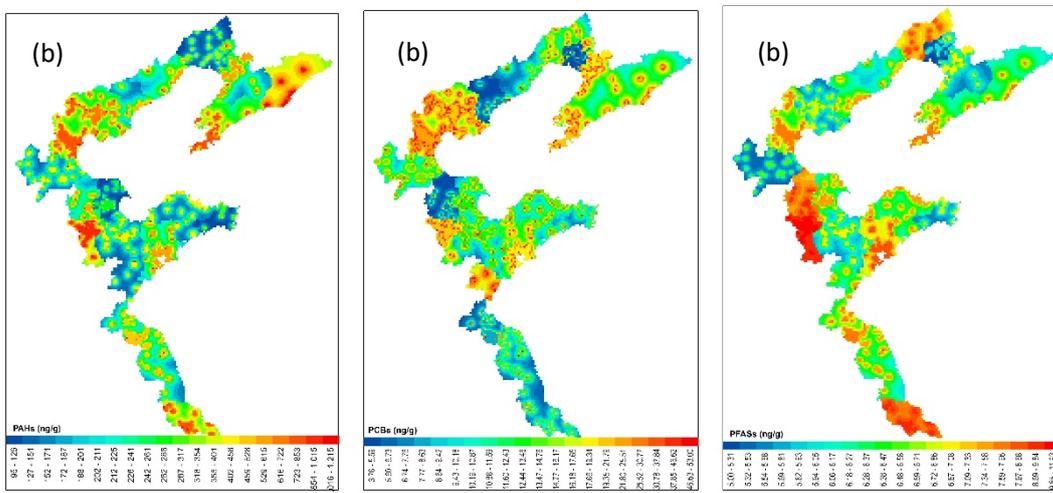
485 were shown in Figure 5, with the average relative errors of 12.84%, 17.77% and 2.81%,
 486 respectively. The statistical errors demonstrated that the precision of spatial simulation
 487 results for PAHs, PCBs and PFASs was relatively high using urban-rural gradient model
 488 with coupled population and land cover data. Compared this method with our previous
 489 BETR model^{13, 14}, it requires less parameters, easier and faster computation, but it can
 490 better characterize the gradients between rural and urban areas, with the accuracy of the
 491 simulation results comparable with that from BETR model^{2, 5, 14}.

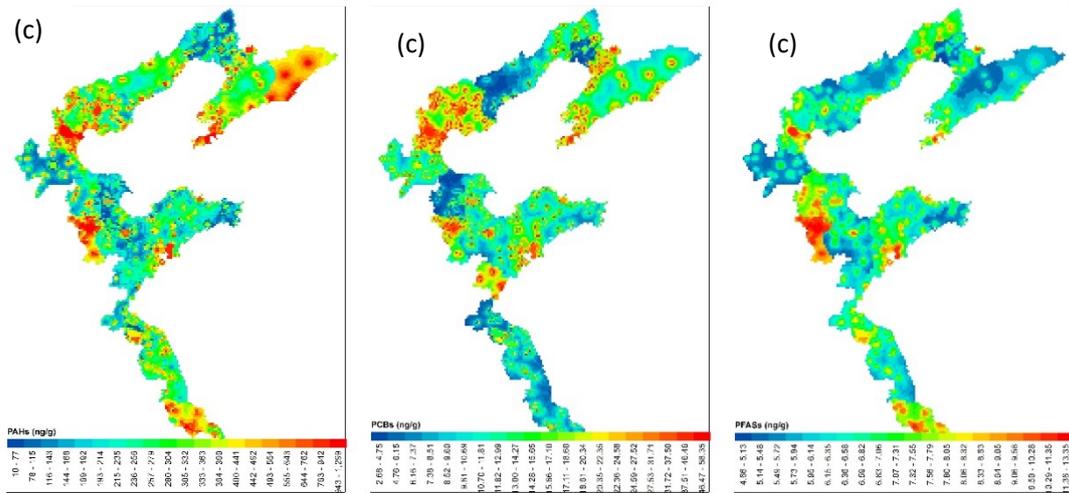
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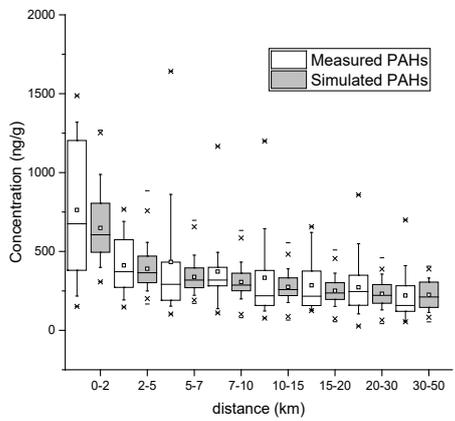




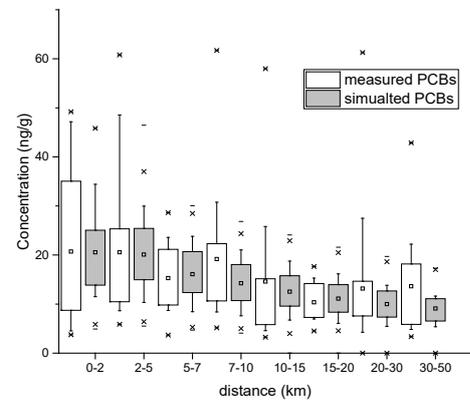
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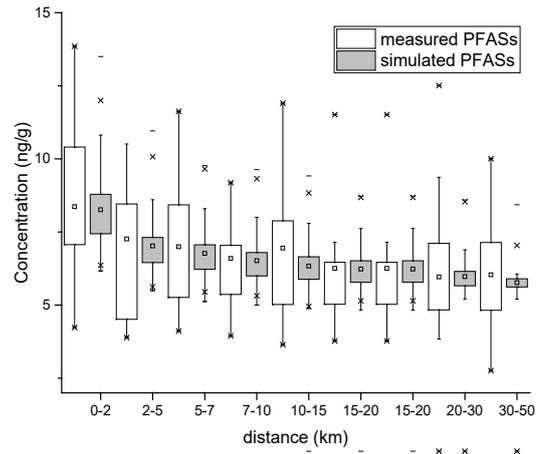
497 **Figure 4** Predicted distribution of PAHs, PCBs and PFASs. (a) simulated results using
 498 urban-rural model at a regional scale. (b) simulated results calibrated by population
 499 combined with land cover data at a city scale. (c) simulated results calibrated by
 500 artificial surface at a 10km scale.

501



502





503

504 Figure 5 Simulated and measured PAH, PCB and PFAS concentrations at different

505 distances from urban areas. **(Supporting materials)**

506

507 In this study, uncertainty of the simulation results may be affected by various
 508 assumptions when the prediction of soil POPs concentrations distribution was
 509 conducted, as some factors were not considered, such as wind speed and direction,
 510 centralized emission from industries, long-distance transmission, long-term
 511 accumulation, and seasonal change. These factors may result in prediction errors of
 512 concentrations in some sites. For example, the POPs concentrations in the surroundings
 513 of an industrial park with fluorine chemical plants, power plants and waste recycling
 514 plants may be higher than that in the other regions. In addition, there are higher organic
 515 matter contents (OMC) in the grassland and woodland soils, but whether the
 516 corresponding POPs concentrations are significantly related to OMC or not needs
 517 further verification.

518

519 Urban-rural gradient model was developed to simulate spatial distribution of POPs

520 in soils combined with urban population and land-cover data in this paper. We found
521 that the slope m for chemical concentrations in soil and air should be the same under
522 certain assumptions, which may provide a method for predicting atmospheric – soil
523 concentrations at a regional scale. Meanwhile, the model could be used either as a single
524 method for prediction or being combined with multi-parameters such as distribution of
525 emission for POPs, chemical properties, soil properties, industrial information, multi-
526 media properties, GDP and other social parameters. For example, total organic carbon
527 may be used to improve the accuracy of POPs concentrations in woodland or grassland
528 with high humus. Simpler parameters, faster deployment capability, and high accuracy
529 are the major features for this study, but more local and regional monitoring data are
530 needed to simulate the mechanism of distribution, seasonal and annual changes of the
531 urban-rural gradients with higher precision . This paper presented a new method
532 coupled land cover, socio-economic data, and POPs concentration for quantitative study
533 of POPs distribution at a regional scale

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541 **Appendix A. Supplementary data**

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