

Emission Inventory and Control Policy for Non-road Construction Machinery in Tianjin

qijun zhang (✉ zhangqijun@nankai.edu.cn)

Nankai University

ning wei

Nankai University

Lei Yang

Nankai University

Xi Feng

CCSI

yanjie zhang

Nankai University

lin Wu

Nankai University

hongjun Mao

Nankai University

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1 **Emission inventory and control policy for non-road construction machinery**
2 **in Tianjin**

3 *Qijun Zhang^{1*}, Ning Wei¹, Lei Yang¹, Xi, Feng², Yanjie Zhang¹, Lin Wu¹, Hongjun Mao^{1*}*

4 ¹ Tianjin Key Laboratory of Urban Transport Emission Research, College of Environmental
5 Science and Engineering, Nankai University Tianjin, 300071, China

6 ²China Classification Society Industrial CORP, Tianjin 300457, China.

7 * **Corresponding author, e-mail address: zhangqijun@nankai.edu.cn (Qijun Zhang)**

8 **hongjunm@nankai.edu.cn (Hongjun Mao)**

9

10 **Abstract**

11 The establishment of a non-road construction machinery emission inventory forms
12 the basis for the analysis of pollutant emission characteristics and for the formulation
13 of control policy. We analysed and investigated data on populations, emission factors
14 and activity levels for the construction machinery in Tianjin to estimate an emission
15 inventory. Finally, a variety of emission reduction scenarios were used to simulate
16 emission reductions and propose the most effective control policy. The results show
17 that total emissions of CO, HC, NO_x, PM₁₀ and PM_{2.5} from non-road construction
18 machinery in Tianjin of 2018 reached 4180.78, 951.44, 5833.85, 383.92 and 365.70 t,
19 respectively. Forklifts, excavators and loaders were the three most important emission
20 sources in Tianjin. There are clear differences in the emissions of different districts.
21 Large machinery emissions were mainly distributed across the Binhai New Area, which

22 includes high volumes of port machinery and tractors in Tianjin Port. Based on various
23 emission reduction scenarios, the effect of emission reductions is estimated. The IAD
24 affected the reduction of CO and HC emissions with RR values of 17.6% and 17.3%,
25 respectively, while EMO affected the mitigation of PM₁₀ and PM_{2.5} emissions and RR
26 values by 18.0% and 18.4%, respectively. The emission reduction control policy for
27 non-road construction machinery is proposed, including the accelerated updating of
28 non-road machinery emission standards; integrating diesel engine research and
29 development institutions to accelerate the development of vehicle after-treatment
30 technology; and establishing a cooperation mechanism for scientific research institutes,
31 government departments and enterprises in the control of non-road mobile machinery
32 emissions.

33

34 **Keywords:** Non-road construction machinery; Emission inventory; Spatial
35 distribution; Scenario analysis; Control policy

36

37 **1 Introduction**

38 With continuous improvements in the control of motor vehicle emissions, pollution
39 from non-road mobile machinery has become increasingly prominent. The carbon
40 monoxide(CO), hydrocarbon(HC), nitrogen oxides(NOx) and particulate matter(PM)
41 emitted by non-road mobile machinery account for 18%-29% of mobile sources in the
42 world (Yan et al. 2013). Due to the numerous modes of non-road mobile machinery

43 production, the wide range of applications involved, the large number of possessions
44 involved, and uncontrolled pollution emissions, the management and control of those
45 emissions pose significant challenges. According to the China Motor Vehicle
46 Environmental Management Annual Report (2019)(MEE 2019), the NO_x emission of
47 construction machinery, agricultural machinery, ships and railway locomotives
48 accounted for 37%, 32%, 26% and 3%, respectively. In 2009, the European
49 Environmental Protection Agency published the emission inventory, which showed that
50 non-road mobile source was an important pollution source in Europe (EEA 2019). In
51 Yan's research, non-road mobile source will become the main source of mobile
52 pollution after 2030 and become the largest source of CO and HC pollutants in Asia
53 after 2050(Yan et al. 2013).

54 With the acceleration of urbanization and industrialization, the use of non-road
55 construction machinery in China has grown rapidly. With this trend, the population of
56 excavators and loaders increased from 120,000 and 230,000 vehicles in 2001 to 1.6
57 million and 1.55 million vehicles in 2017, reflecting 13.3- and 6.7-fold increases,
58 respectively(CCMA 2017). As emission standards for Chinese construction machinery
59 are still lax, emissions of NO_x and PM are high. Most construction and road
60 construction sites are distributed in densely populated urban living areas, and human
61 health is highly vulnerable to such pollution. Further reducing people's confidence in
62 air quality management directly affects the reputation of major cities in China.

63 Research on the emission control measures of non-road mobile source and their
64 impacts on air quality must be supported with accurate source emission inventories.

65 Since 1991, the US EPA has conducted a large number of studies on non-road mobile
66 source emissions. The CO, HC, NO_x, and PM emission inventories of non-road
67 construction machinery have been established(Chow et al. 2011; Diazrobles et al. 2009;
68 Lin et al. 2005; Sarwar et al. 2005). At the end of the 20th century, the European Union
69 also developed an air pollution emission inventory of non-road mobile source. The
70 results showed that non-road construction machinery contributions to NO_x and PM
71 were only less than those of industrial sources(Desouza et al. 2020; Winther and Nielsen
72 2011).

73 In China, the research on emission source inventory of non-road mobile source is
74 still in the initial stage. The Pearl River Delta (PRD), Yangtze River Delta (YRD) and
75 Beijing-Tianjin-Hebei (BTH) regions have begun to research the emission inventory of
76 non-road mobile sources(Hou et al. 2019; Lang et al. 2018; Li et al. 2019; Zheng et al.
77 2009; Zhong et al. 2018). Zheng Junyu has constructed a data set of non-road mobile
78 machinery based on the activity level, comprehensive emission factors and space-time
79 allocation factors in PRD. A bottom-up calculation method was used to establish the
80 non-road mobile machinery emissions inventory of the PRD in 2014. Construction
81 machinery emissions in the PRD were dominated by those generated by construction
82 vehicles and excavators, which contribute 40.1% and 33.9%, respectively(Bian YH
83 2018). Huang Cheng has established the emission inventory of air pollution sources for
84 non-road mobile machinery in the YRD based on a field survey of non-road mobile
85 machinery in these cities of the YRD. The results showed that CO and VOCs emitted
86 from agricultural machinery accounted for 88% and 77%, respectively. NO_x and PM_{2.5}

87 emissions from construction and municipal engineering machinery contributed 49%
88 and 35%, respectively(HUANG Cheng 2018). Non-road machinery emission test
89 research on BTH was relatively limited and has only started recently. An emission
90 inventory of typical agricultural machinery for Beijing from 2006 to 2016 has been
91 established based on the actual emission factors, population and activity level(Hou et
92 al. 2019). The results showed that CO, NO_x, HC and PM from agricultural machinery
93 decreased by 63.11%, 72.07%, 62.93% and 74.67% from 2006 to 2016. In Liang's study,
94 a high temporal-spatial resolution emission inventory for agricultural machinery in
95 China was developed(Lang et al. 2018). Whereas the above non-road mobile source
96 emission studies on BTH focus on agricultural machinery, there have been no
97 evaluations of the effects of construction machinery emissions and control policies in
98 BTH.

99 To fill this gap, we measured the spatial distribution of non-road construction
100 machinery in Tianjin through surveys and compiled the non-road mobile construction
101 machinery emission inventory of Tianjin in accordance with the methods of the
102 emission inventory guidelines. The Monte Carlo model was used to calculate the
103 uncertainty of the emission inventory. Finally, emission reduction scenarios were tested
104 to assess the effects of different policies and to provide a basis from which government
105 managers can formulate pollution control policies in the future.

106 2 Data and methodology

107 2.1 Estimating emissions methods

108 In this study, the emission of the pollutant from non-road construction machinery
109 sources in Tianjin was estimated for the base year 2018.

110 Non-road construction machinery emissions were calculated with the method
111 based on population and rated net power from the Non-road Mobile Source Emissions
112 Inventory Compiled Technical Guidelines (Guidelines)(MEE 2014). The calculation
113 formula is as follows:

$$114 \quad E = \sum_x \sum_y \sum_z (P_{x,y,z} \times G_{x,y,z} \times LF_{x,y,z} \times hr_{x,y,z} \times EF_{x,y,z}) \times 10^{-6}$$

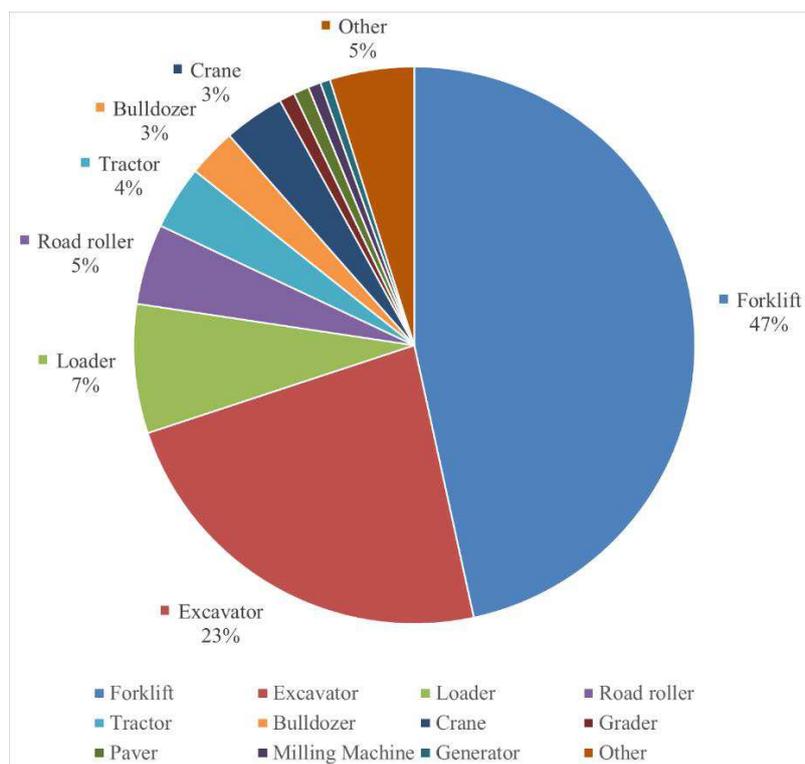
115 where E is annual emissions of CO, NO_x, HC, PM₁₀ and PM_{2.5} (t); x, y, and z are
116 the type of non-road mobile machinery, emission standard and power segment,
117 respectively; P is machinery population; G is the average rated power level, (kw/unit);
118 LF is load factor; hr is annual used hours of each non-road mobile machinery(h); and
119 EF is emission factor of each pollutant(g/kWh).

120 2.2 Construction machinery population

121 The population of non-road construction machinery in Tianjin were obtained from
122 government survey statistics and the China construction machinery industry yearbook.
123 Forklifts, loaders and excavators are the three main types of construction machinery
124 used in Tianjin. In addition, road rollers, bulldozers, and graders are often used in road
125 construction. In this study, construction machinery are classified as follows: forklifts,

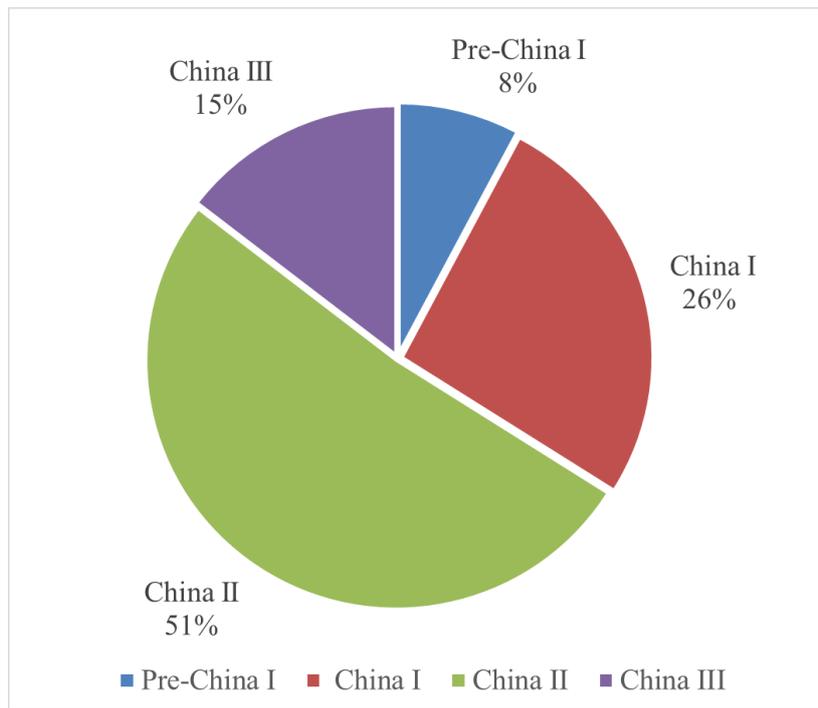
126 excavators, loaders, road rollers, bulldozers, tractors, cranes, graders, pavers, milling
127 machines, generators and other construction machinery.

128 According to surveys and estimates, the population of non-road construction
129 machinery in Tianjin was approximately 22481 by the end of 2018. The non-road
130 construction machinery population and type distribution in each district were detailed
131 in Figure 1 and Table S1. Forklifts, excavators and loaders constitute the largest
132 proportion of construction machinery, accounting for 47%, 23% and 7%, respectively
133 (Figure 1). Based on the classification of emission standards, the proportion of Pre-
134 China I, China I, China II and China III are 8%, 26%, 51% and 15%, respectively
135 (Figure 2). Table S2 shows that non-road construction machinery met Pre-China I and
136 China I emission standards before 2010. With the extension of use time, older vehicles
137 were scrapped and eliminated. China II emissions continued for nearly 6 years,
138 resulting in the China II population accounting for the largest proportion.



140

Figure 1 Type distribution of non-road construction machinery



141

142

Figure 2 Emission standard distribution of non-road construction machinery

143

144 2.3 Emission factors and activity data

145 Emission factor refers to the number of pollutants emitted per unit of activity level.

146 For non-road construction machinery, the pollutant emission concentration when

147 outputting a unit of power per unit of time was defined as the emission factor. In this

148 study, basic emission factors were obtained according to the Guidelines (Table S3).

149 The activity levels of non-road construction machinery are closely related to the

150 different operating regions. In this study, we conducted a large number of investigations

151 to make up for the lack of basic data on non-road machinery activity levels, which were

152 including the average rated power, load factor, and average annual operating hours. The

153 average rated power was obtained through dividing the total power by the population

154 in Tianjin. The load factor was 0.65, which was found in the relevant research. The
155 average annual operating hours of non-road construction machinery referred to the
156 values in the Guidelines, and the activity level of each non-road construction machinery
157 type was shown in Table S4.

158 **2.4 Uncertainty calculation**

159 Due to the lack of key data, errors in monitoring results and random errors, certain
160 uncertainties will be caused in establishing the emission inventory and, which including
161 errors resulting from the statistical process of determining activity levels and a lack of
162 localized emission factors. The Monte Carlo simulation model is used to quantify the
163 range of uncertainty(Kelliher et al. 2017; Ramirez et al. 2008; Sun et al. 2019).

164 Monte Carlo simulation is a commonly used random sampling simulation method.
165 Random sampling and simulation are performed to obtain the approximate solution to
166 a problem that characterizes the probability distribution characteristics of the problem.
167 The Monte Carlo simulation method involves three steps: 1) construct a probability
168 process and determine the distribution probability of each parameter used in a
169 calculation; 2) employ sampling from the known probability distribution and generate
170 random numbers satisfying this distribution by computer for mathematical experiments;
171 and 3) establish various estimators and investigate the estimators of random variables
172 after mathematical experiments to characterize the distribution characteristics of the
173 random distribution(Wu et al. 2009).

174 Crystal Ball software is used for Monte Carlo simulation in this study. We conduct

175 10,000 simulations of the uncertainty of the emission inventory of non-road
176 construction machinery at a 95% confidence level.

177 **2.5 Scenario analysis for emission reduction**

178 Several policies have been introduced in many provinces and cities to control non-
179 road construction machinery and include the elimination of old machinery(EOM) and
180 the installation of after-treatment devices(IAD) (Table 1). According to a preliminary
181 investigation, it is more difficult to install exhaust after-treatment equipment onto Pre-
182 China I and China I emission standard construction machinery due to the poor emission
183 conditions of such machinery. It is possible to reduce emissions by eliminating old
184 machinery. The emissions of China II non-road mobile machinery can be significantly
185 reduced by installing after-treatment devices (DOC and DPF). The efficiency of
186 pollutant emission reduction after transformation can reach roughly 70%.

187 Table 1 Specific definitions of Non-road construction machinery reduction scenarios in Tianjin.

Scenarios	Specific definition
Normal situation (NS)	Take 2018 as the base year, emission control is maintained in 2018
Install after-treatment device (IAD)	50% of the China II emission standard construction machinery were equipped with after-treatment devices
Eliminate old machinery (EOM)	50% of Pre-China I and China I construction machinery were eliminated.

188 Reduction effects under different control scenarios were estimated by Reduction

189 Rates (RR) relative to the normal situation (NS) as given by the following equation:

$$190 \quad RR_{ERS} = \frac{E_{NS} - E_{ERS}}{E_{NS}} \times 100\%$$

191 Where ERS is the emission reduction scenario (including IAD, EOM); E is the
192 non-road construction machinery emissions.

193 **2.6 Quality Assurance and Quality Control**

194 The quality control of non-road construction machinery inventory data involved
195 the following three steps:

196 1) Data sources on activity levels were checked and calibrated to ensure the
197 accuracy of records used in the field investigation. 2) The consistency of the survey
198 space and of the time range of non-road construction machinery pollution sources was
199 checked. 3) The integrity of the survey data was checked to ensure that the data covered
200 all types of non-road construction machinery pollution sources.

201 **3 Results and discussion**

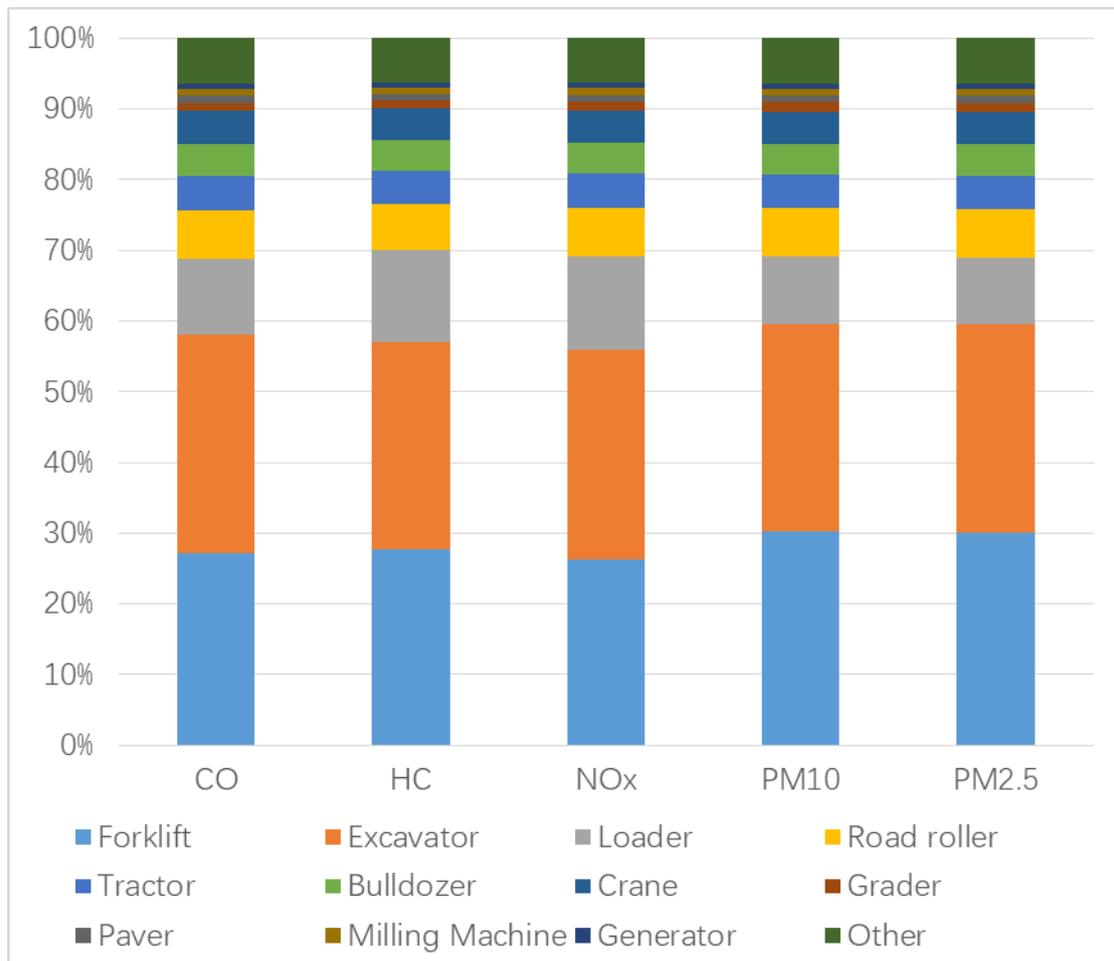
202 **3.1 Emission inventory of non-road construction machinery**

203 The emission inventory of non-road construction machinery in Tianjin for 2018
204 was estimated using the methods described in the prior section. Total atmospheric
205 pollutant emissions of CO, HC, NO_x, PM₁₀ and PM_{2.5} were 4180.78, 951.44, 5833.85,
206 383.92 and 365.70 t, respectively. These values were several times higher than those of
207 non-road mobile sources estimated in 2015(ZHANG Yi 2017). The average power
208 calculation method was used, as it is more accurate than the fuel calculation methods

209 used in previous similar studies. The population data used in this study were also more
210 reliable and based on actual surveys. As HC and NO_x from construction machinery
211 were the important precursors of atmospheric PM_{2.5} in megacities(Guo et al. 2020),
212 emissions of non-road mobile machinery cannot be ignored. The emission control of
213 non-road construction machinery in such cities must be improved.

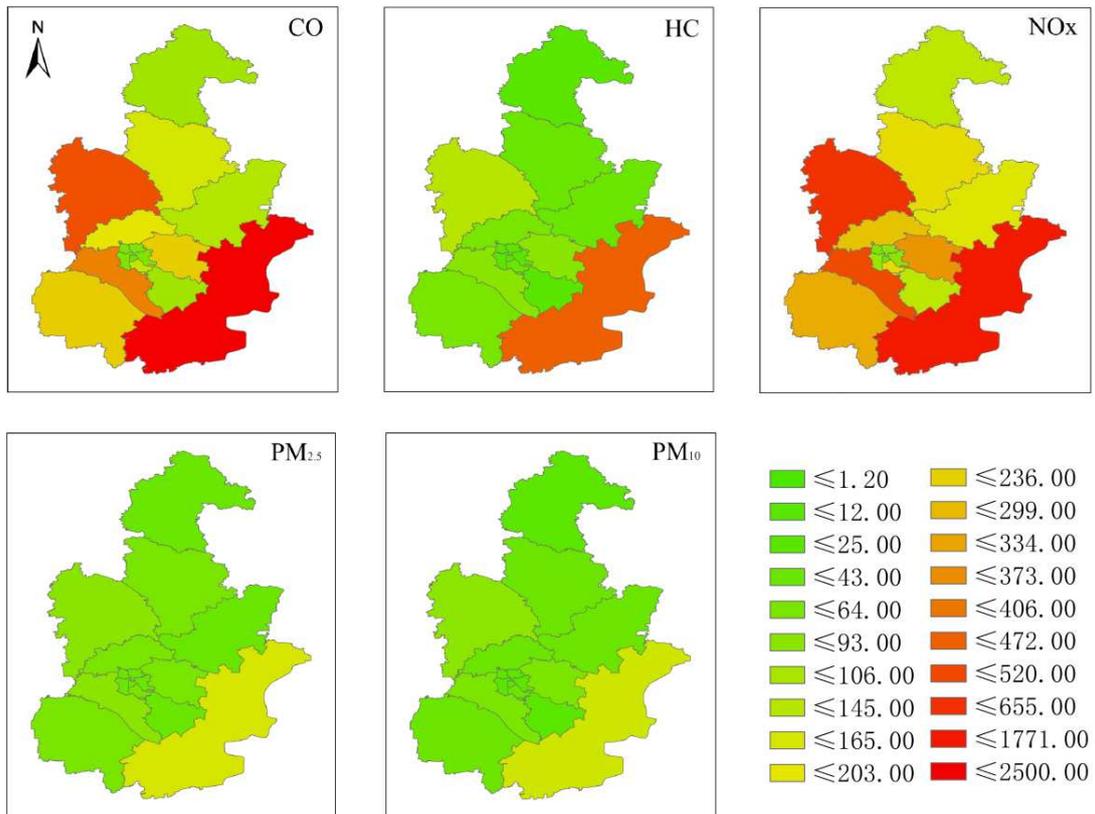
214 The contribution rate of emissions from different construction machines was
215 shown in Figure 3. Among the non-road construction machinery, forklifts, excavators
216 and loaders were the most important emission sources in Tianjin. While the emissions
217 of individual forklifts were not significant, they generated the largest proportion of
218 pollutants due to their frequent use. Excavators and loaders accounted for 30.8%,
219 contributing 40-60% of all pollutants. Therefore, control policies on non-road mobile
220 machinery should focus on large populations of construction machines. The spatial
221 distribution of pollutants from non-road construction machinery in Tianjin is illustrated
222 in Figure 4. There are obvious differences in the emissions of different districts. Large
223 machinery emissions were mainly distributed across the Binhai New Area, Wuqing
224 District, Xiqing District, Dongli District, Jinghai District and Beichen District.
225 Especially for the Binhai New Area, the proportion of total emissions reached 43% for
226 all pollutants. The Heping, Hongqiao, Hebei, Hedong and Nankai District were the
227 regions with the lowest emissions, which was due to the high rates of urbanization and
228 less building construction in these five regions. Emissions of non-road construction
229 machinery were significantly negatively correlated with the level of urbanization. The
230 contributions of non-road construction machinery emission sources in different districts

231 were shown in Figure 5. The results showed that forklifts, excavators, loaders and road
 232 rollers are construction machines with the greatest contributions in each district. Among
 233 them, the contributions of emission sources in Heping District and Binhai New Area
 234 were different from those of other districts. As Heping District has the highest level of
 235 urbanization and there are no large factories in the area, emission sources here mainly
 236 included excavators, loaders and road rollers. The Binhai New Area is the largest
 237 district in Tianjin with a large number of factories and the Tianjin Port(Chen et al. 2016;
 238 Na et al. 2017). On port terminals, a large number of tractors were operated. Therefore,
 239 in the Binhai New Area, forklifts, excavators, loaders, road rollers and tractors
 240 generated the most emissions.



242

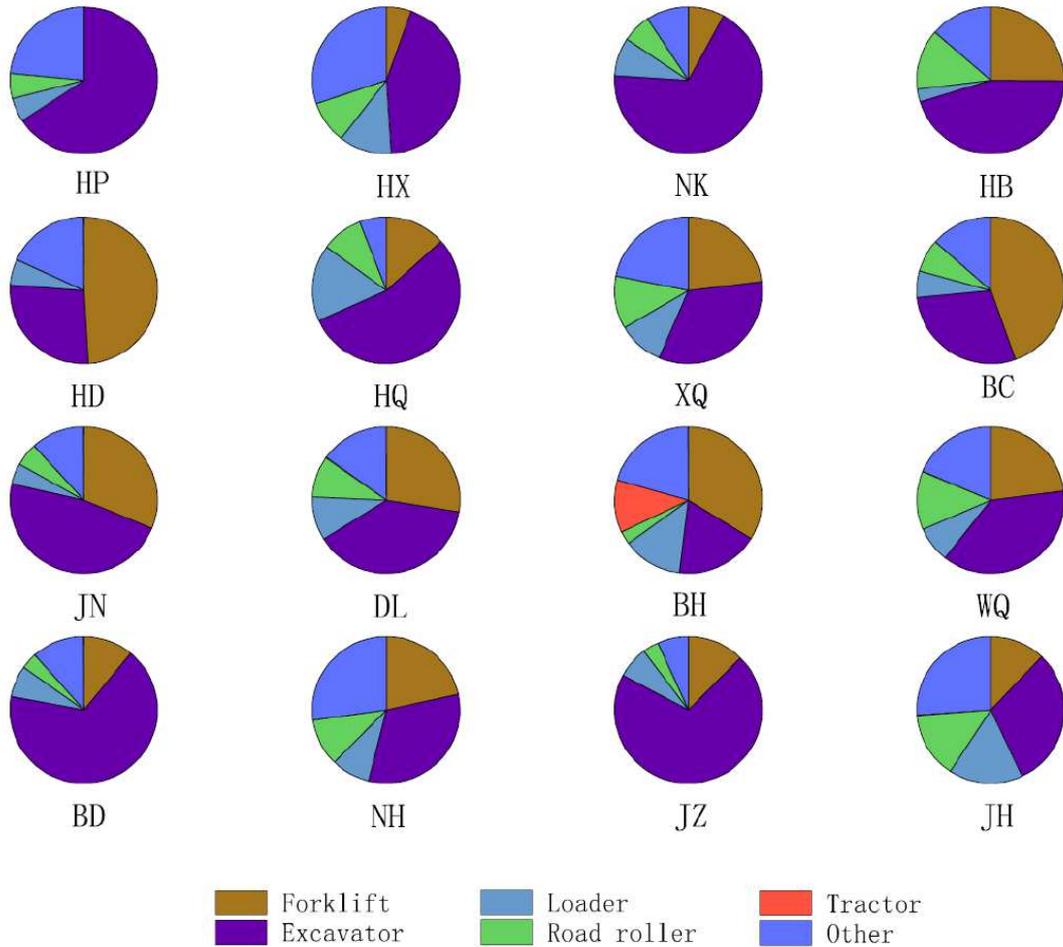
Figure 3 the contribution rates of emissions from different construction machines



243

244

Figure 4 Spatial distribution of total emissions from construction machinery (t)



245

246 Figure 5 The contributions of non-road construction machinery emission sources

247 in different districts (%)

248

249 3.2 Emissions under different reduction scenarios

250 Emissions from non-road construction machinery and reduction rates under

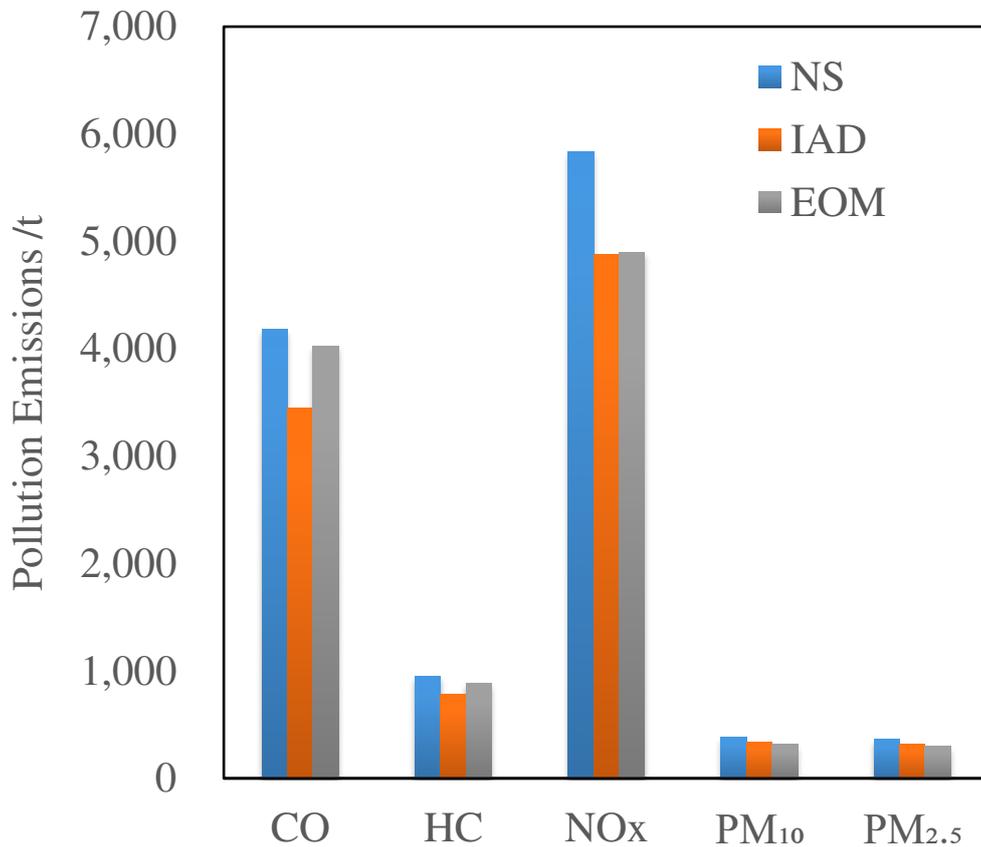
251 different reduction scenarios were shown in Figure 6 and Table 2. The emission will be

252 controlled to different levels under different emission reduction scenarios. The IAD has

253 an overwhelming effect on reductions of CO and HC emissions with RR values of 17.6%

254 and 17.3%, respectively, while EOM has limited effects on the reduction of CO and

255 HC emissions with RR values of 3.7% and 6.9%, respectively. The results indicate that
256 the IAD and EOM have similar effects on NO_x emissions control. RR values are 16.4%
257 and 16.0% for the IAD and EOM, respectively. For PM₁₀ and PM_{2.5}, the RR of EOM
258 is higher than the IAD, indicating that the elevation of emission standards has a more
259 obvious effect on PM emission reduction. At present, research on the emissions of non-
260 road mobile machinery is in its infancy, and there are gaps in relevant non-road
261 construction machinery policy measures. The simulation results of the emission
262 reduction scenarios in this study can thus guide future control policy development.



263
264 Figure 6 the emissions from non-road construction machinery under different
265 reduction scenarios

266

267

Table 2 Reduction rates of emissions under different scenarios

reduction rate	CO	HC	NO _x	PM ₁₀	PM _{2.5}
IAD	17.6%	17.3%	16.4%	12.4%	12.5%
EOM	3.7%	6.9%	16.0%	18.0%	18.4%

268

269 **3.3 Uncertainty analysis**

270 In this study, non-road mobile construction machinery population, activity level
271 and emission factor were the three forms of emission inventory uncertainty considered.

272 At present, several cities in China have not established a registration and filing
273 system for non-road mobile construction machinery. Non-road mobile construction
274 machinery is widely used, and its working environments tend to be very harsh. It is
275 difficult to determine the population, power level and age of construction machinery.
276 To mitigate uncertainty as much as possible, we have done our best to carry out detailed
277 and accurate survey classifications. However, the types of sites surveyed, construction
278 stages considered and the representativeness of mechanical samples may introduce
279 uncertainty.

280 The activity level of construction machinery was obtained through a large sample
281 survey of more than 20,000 machines. However, some uncertainty and large relative
282 deviations in activity level information for some machinery may remain due to the
283 variety of machinery considered.

284 Emission factors are the weakest facet of research on the emission inventory of

285 non-road construction machinery. The research results of non-road construction
286 machinery emission factors are still very limited. It is impossible to accurately assess
287 the emission levels of various machinery types, engine power and emission standards.
288 It is thus necessary to supplement actual measurement research on the localized
289 emission factors of various non-road mobile construction machinery to mitigate
290 emission inventory uncertainty.

291 According to the Monte Carlo model, the uncertainty of the emission inventory is
292 calculated based on the logarithm distribution with parameters such as emission factor,
293 activity level, and population with an error level of 10%. The average uncertainty
294 ranges (at a 95% confidence level) for CO, HC, NO_x, PM₁₀ and PM_{2.5} were measured
295 as -13.64%~18.12%, -15.22%~15.79%, -13.32%~15.13%, -13.58%~15.37%, and
296 -13.27%~15.58%, respectively. While the uncertainty of the emission inventory is
297 inevitable, the comprehensive prediction of the emissions of non-road mobile
298 machinery can still guide the formulation of environmental policies.

299 **3.4 Control policy**

300 The emission standard of non-road mobile machinery must be improved. In China,
301 most non-road mobile machinery only adheres to China II emission standard. Many
302 vehicles can meet the China II emission standard by installing DOC, EGR and other
303 after-treatment devices, which has hindered the development of advanced control
304 technologies such as DPF and SCR. Many companies cannot afford to make significant
305 technology and capital investments to adopt advanced control technology, hindering

306 technological research and development.

307 Therefore, the short-term goal of pollution control can be achieved by improving
308 emissions management and accelerating the installation of DPF and SCR advanced
309 exhaust treatment devices for non-road mobile machinery. The ultimate goal of the
310 long-term control of non-road mobile machinery exhaust pollution can be achieved by
311 providing preferential policies for research and development enterprises, accelerating
312 the development of DPF and SCR advanced technology, and reducing costs of
313 transformation.

314 In terms of the integration and development of non-road diesel engine
315 manufacturers, due to the small scale and insufficient technical reserves of many diesel
316 engine manufacturers in China, their manufacturing levels are limited, and the precision
317 of key components of developed engines is low. As a result, the combustion of domestic
318 engines cannot achieve the best efficiency or the requirements of higher emission
319 standards. It is necessary to integrate decentralized capital, introduce technology,
320 improve the production of non-road engines, and ultimately achieve emission reduction
321 and quality improvements to non-road mobile machinery.

322 Pollution problems related to non-road mobile machinery have been attracted to
323 the attention of the Chinese government. It is necessary to establish a cooperative
324 mechanism for scientific research institutes, government departments, enterprises and
325 other relevant parties in the control of non-road mobile machinery emissions. First,
326 scientific research institutes must establish preliminary emission standards based on the
327 data, and the government must then promulgate a detailed plan of action according to

328 these standards. Enterprises must provide prompt feedback on the implementation of
329 such standards and finally achieve the precise control of non-road machinery emissions.

330 **Conclusion**

331 This study provides comprehensive estimations of non-road construction
332 machinery emissions of CO, HC, NO_x, PM₁₀, and PM_{2.5} for Tianjin. Total atmospheric
333 pollutant emissions of CO, HC, NO_x, PM₁₀ and PM_{2.5} were measured at 4180.78,
334 951.44, 5833.85, 383.92 and 365.70 t, respectively. Among non-road construction
335 machinery, forklifts, excavators and loaders were found to be the three most important
336 sources of emissions in Tianjin, contributing 70% of all pollutants. Regarding spatial
337 distributions, there are clear differences in the emissions of different districts. The
338 contribution of emission sources in Heping District and the Binhai New Area is
339 different from those of other districts.

340 Analysis results of emission reduction scenarios show that the IAD has an
341 overwhelming effect on the reduction of CO and HC emissions with RR values of 17.6%
342 and 17.3%, respectively. The EMO has a more obvious effect on the reduction of PM
343 emissions. For NO_x, the IAD and EMO have similar effects on NO_x emissions control
344 with RR values of 16.4% and 16.0% for IAD and EMO, respectively.

345 Finally, the emission reduction control policy for non-road construction machinery
346 is proposed to provide basic support for the future management of non-road mobile
347 machinery. The emission standard for non-road mobile machinery must be improved.
348 The development of vehicle after-treatment technology should be accelerated to reach

349 international standards, and cooperation between scientific research institutes,
350 government departments and enterprises in the control of non-road mobile machinery
351 emissions must be established.

352 **Ethics approval and consent to participate**

353 Not applicable

354 **Consent for publication**

355 Not applicable

356 **Availability of data and materials**

357 The datasets used and/or analysed during the current study are available from the
358 corresponding author on reasonable request.

359 **Competing interests**

360 The authors declare that they have no competing interests.

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365 **Authors' contributions**

366 *Qijun Zhang*: Conceptualization, Methodology, Writing-Original draft preparation,
367 Writing-Reviewing and Editing.

368 *Ning Wei*: Data curation, Software.

369 *Lei Yang*: Data curation.

370 *Xi Feng*: Data curation.

371 *Yanjie Zhang*: Validation.

372 *Lin Wu*: Supervision.

373 *Hongjun Mao*: Funding acquisition, Methodology, Writing-Reviewing and Editing.

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Figures

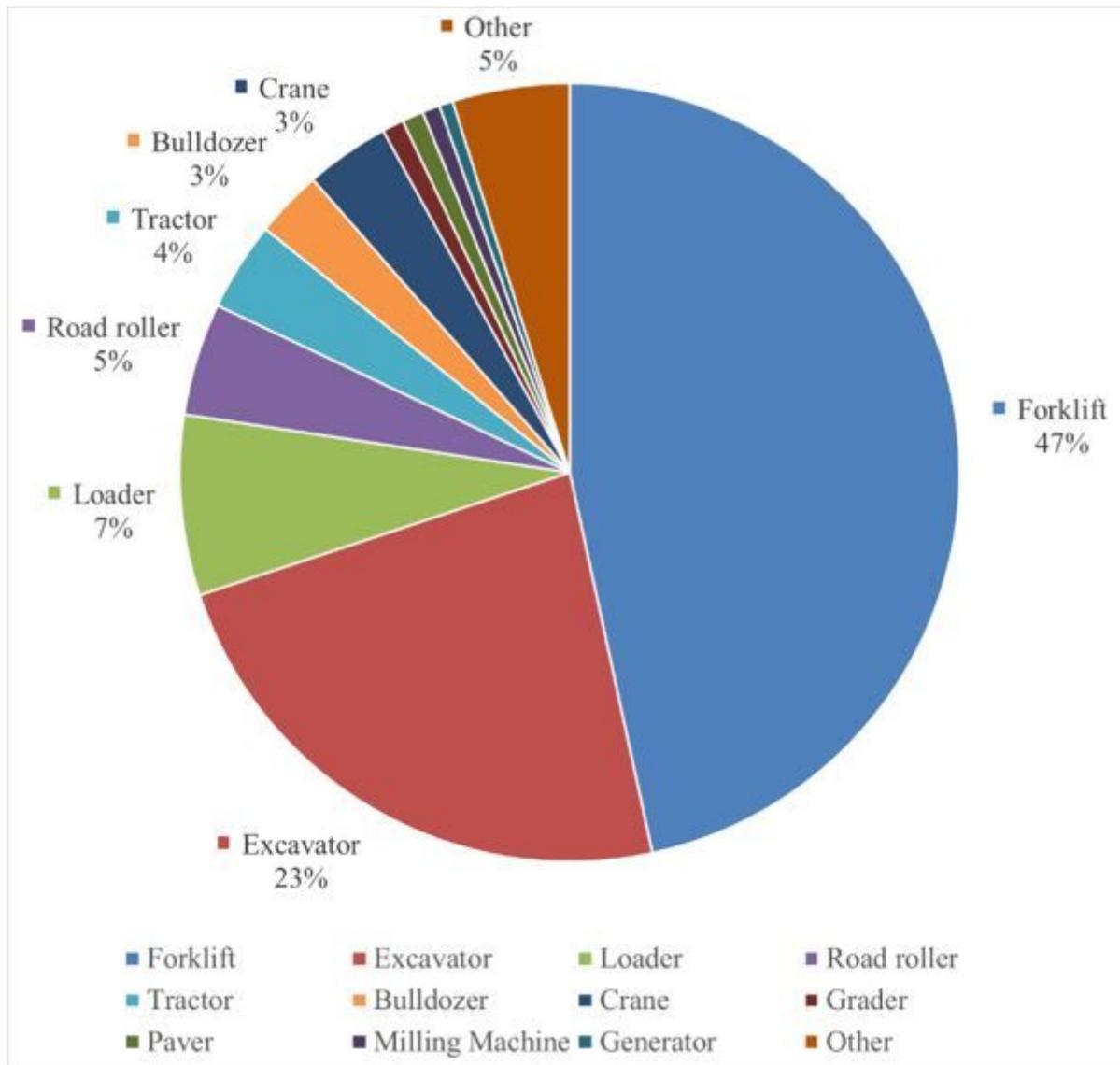


Figure 1

Type distribution of non-road construction machinery

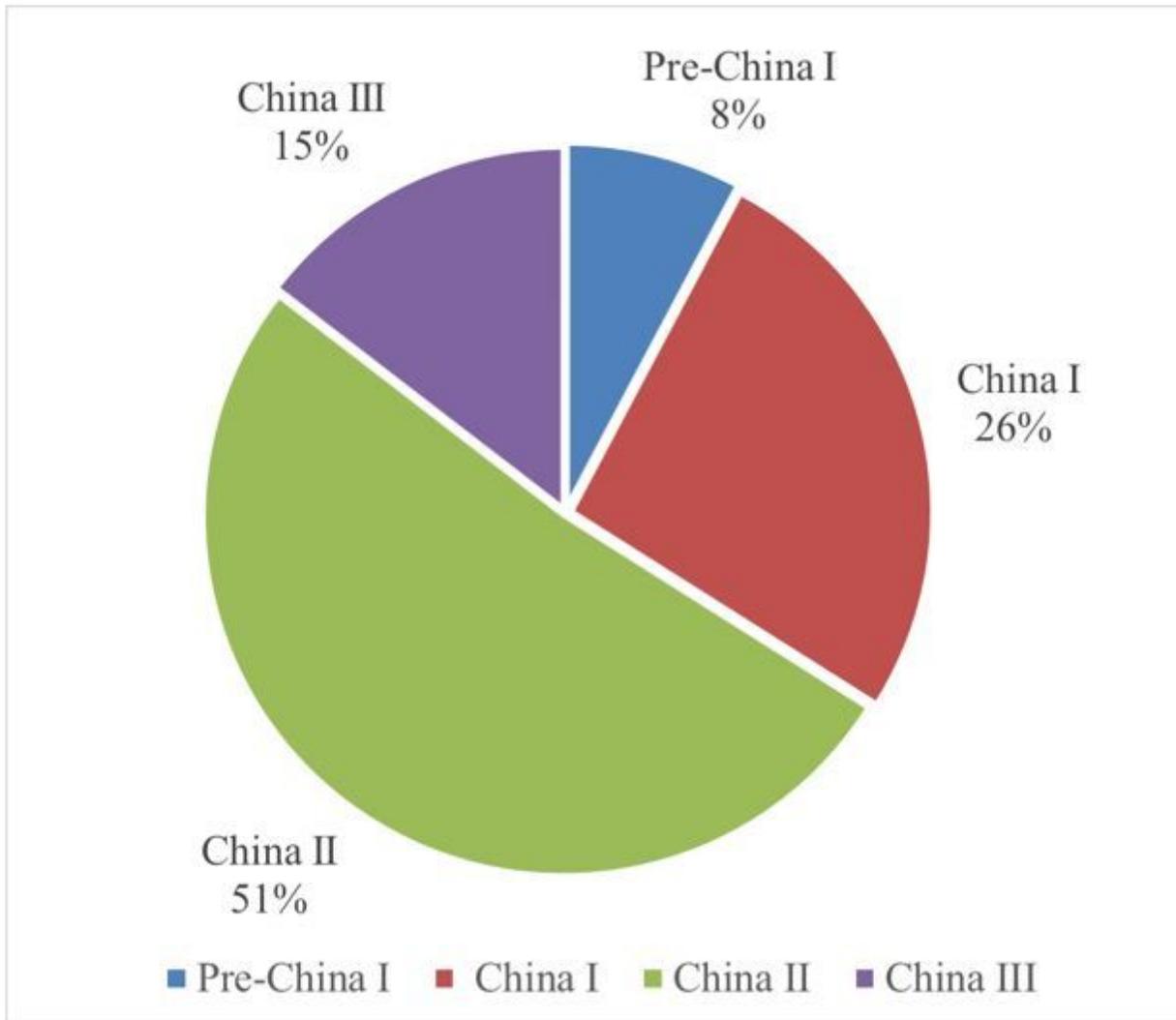


Figure 2

Emission standard distribution of non-road construction machinery

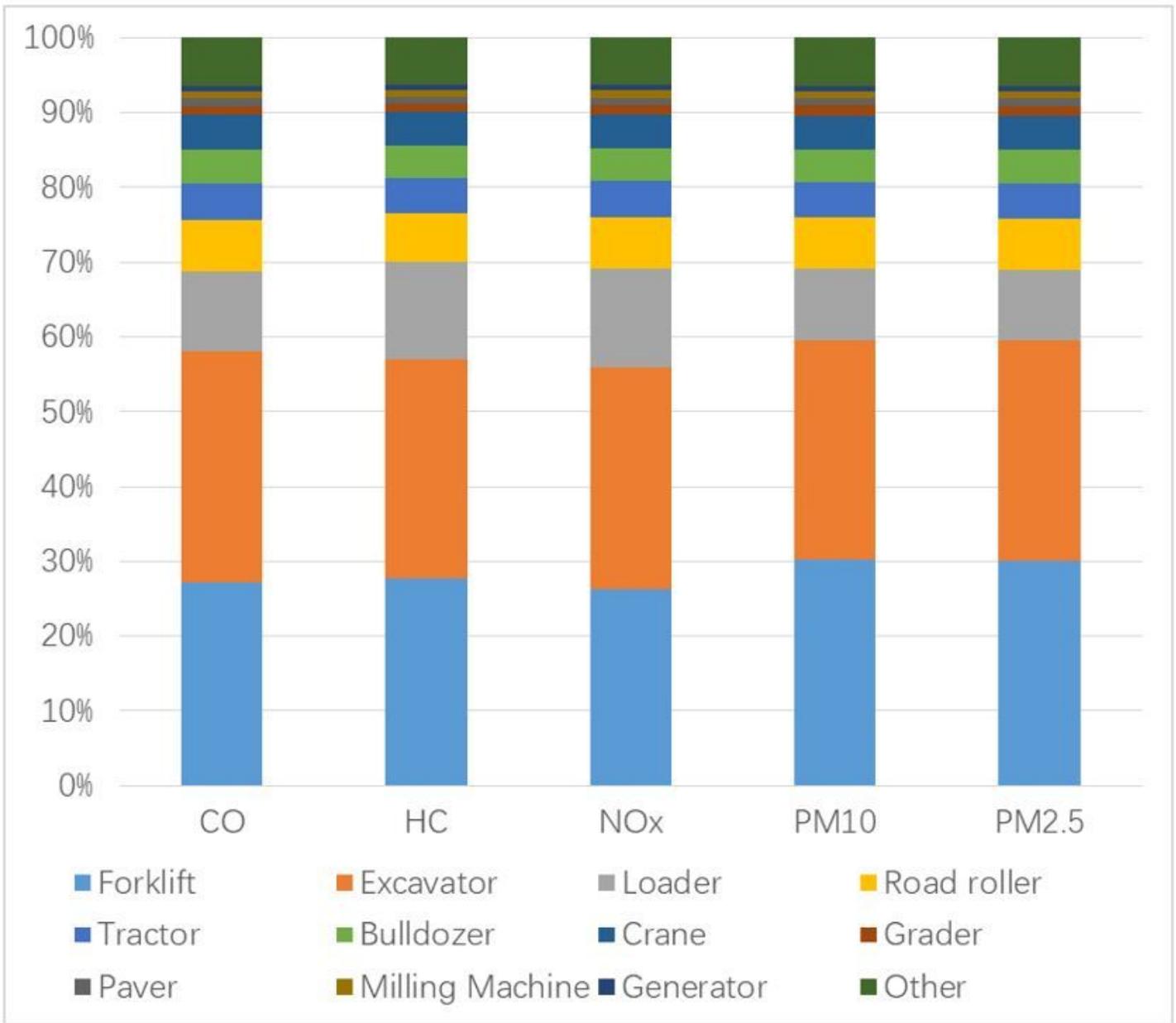


Figure 3

the contribution rates of emissions from different construction machines

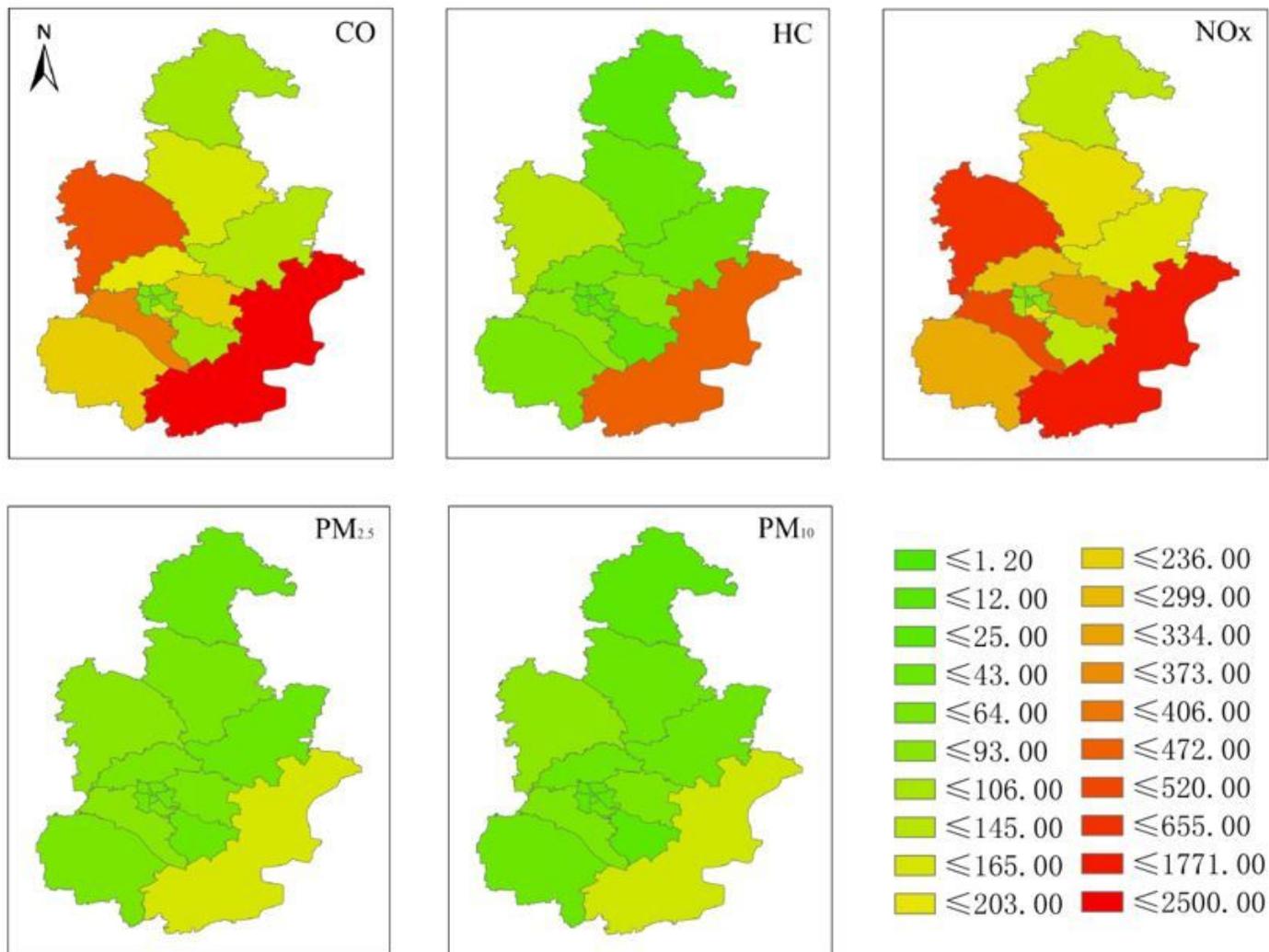


Figure 4

Spatial distribution of total emissions from construction machinery (t) 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.

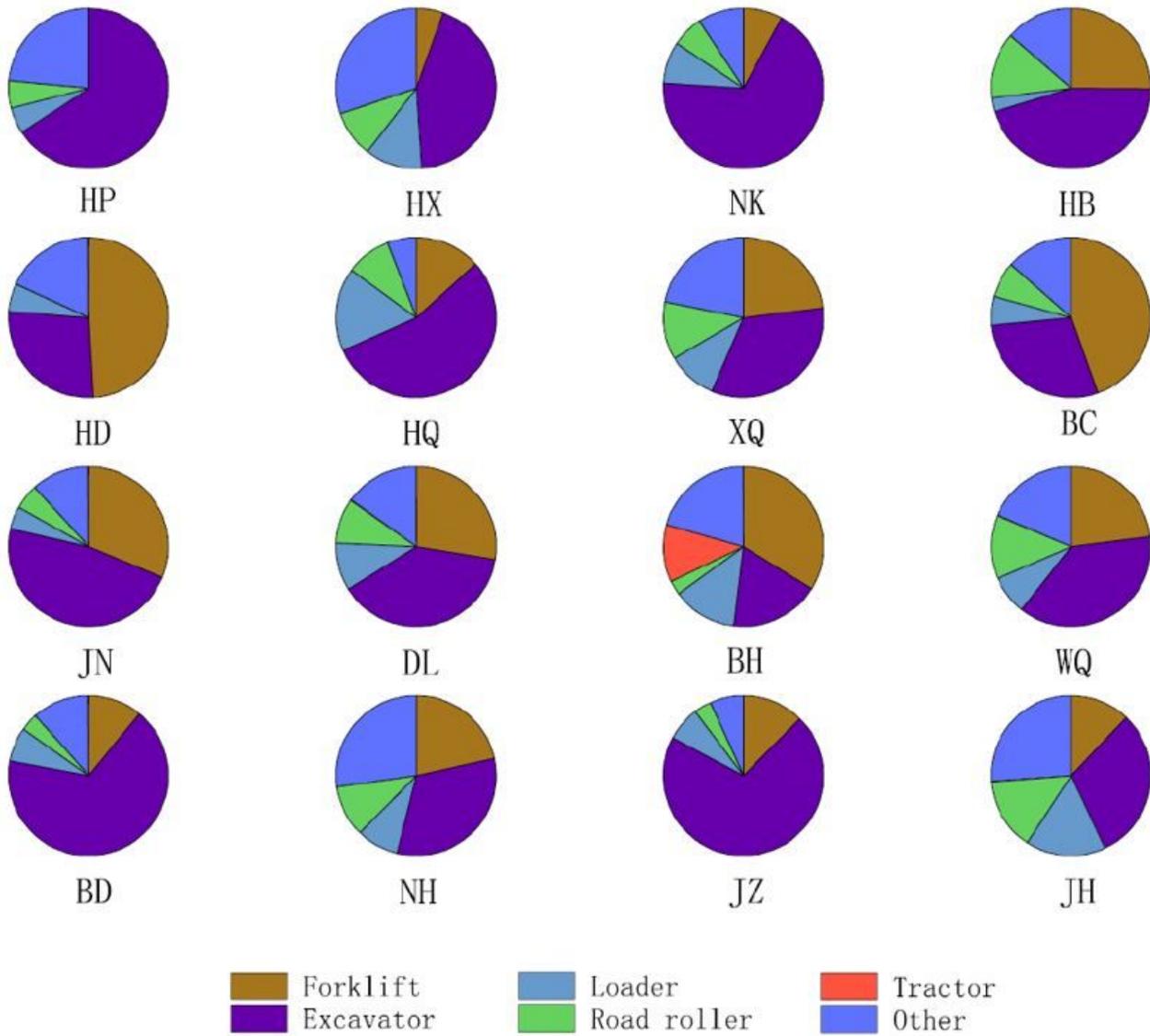


Figure 5

The contributions of non-road construction machinery emission sources in different districts (%)

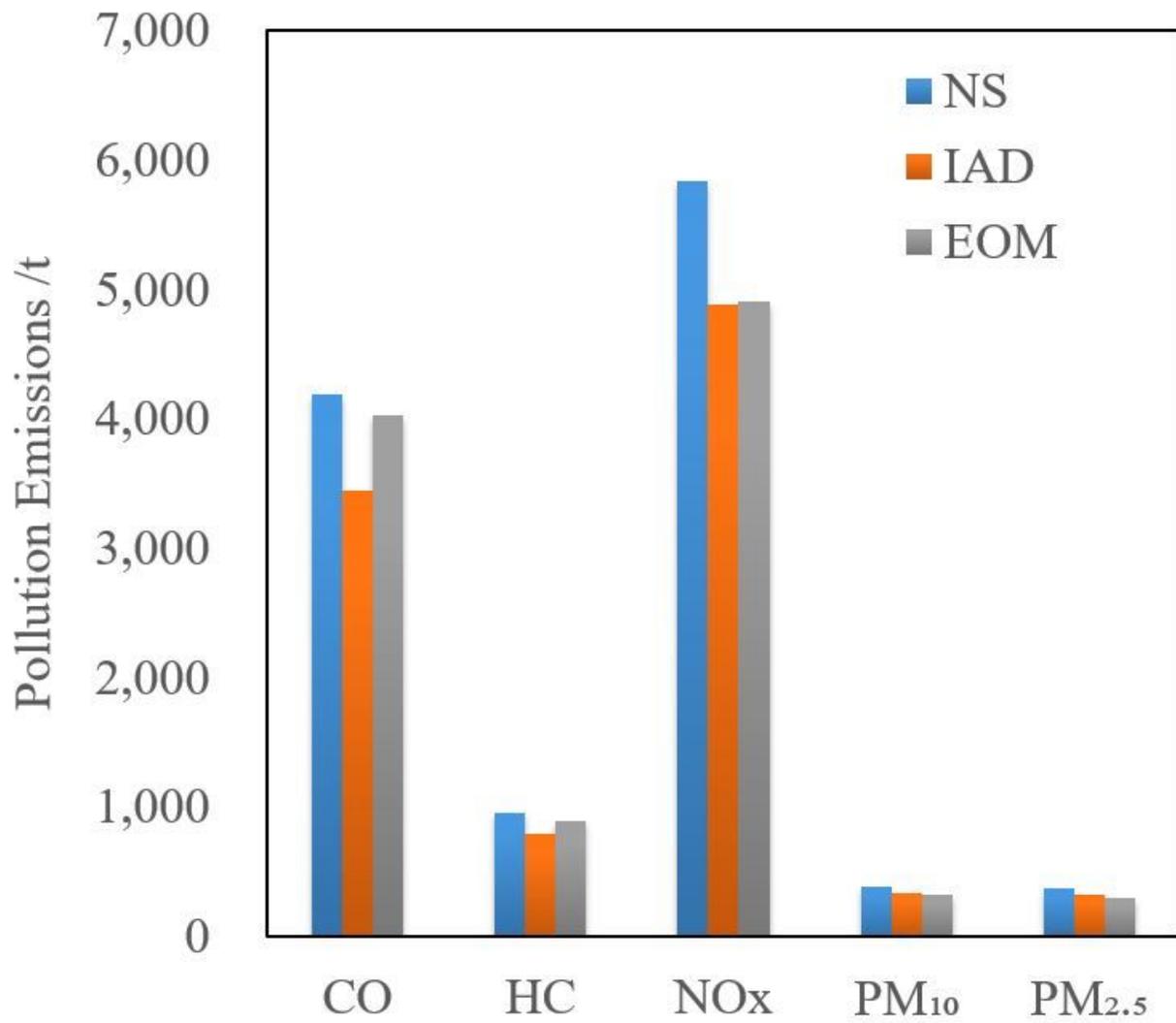


Figure 6

the emissions from non-road construction machinery under different reduction scenarios

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