Inventory of Atmospheric Pollutant Emissions from Burning of Crop Residues in China Based on Satellite-retrieved Farmland Data

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Abstract: The burning of crop residues emits large quantities of atmospheric aerosols. Published studies have developed inventories of emissions from crop residue burning based on statistical data. In contrast, this study used satellite-retrieved land-cover data (1 km \times 1 km) as activity data to compile an inventory of atmospheric pollutants emitted from the burning of crop residues in China in 2015. The emissions of PM₁₀, PM_{2.5}, VOCs, NO_x, SO₂, CO, and NH₃ from burning crop straw on nonirrigated farmland in China in 2015 were 610.5, 598.4, 584.4, 230.6, 35.4, 3329.3, and 36.1 Gg (1 Gg = 10^9 g), respectively; the corresponding emissions from burning paddy rice residues were 234.1, 229.7, 342.3, 57.5, 57.5, 1122.1, and 21.5 Gg, respectively. The emissions from crop residue burning showed large spatial and temporal variations. The emissions of particulate matter and gaseous pollutants from crop residue burning in nonirrigated farmland were highest in east China, particularly in Shandong, Henan, Anhui, and Sichuan provinces. Emissions from burning paddy rice residue were highest in east and central China, with particularly high levels in Shandong, Jiangsu, Zhejiang, and Hunan provinces. The monthly variations in atmospheric pollutant emissions were similar among different regions, with the highest levels observed in October in north, northeast, northwest, east, and southwest China and in June and July in central and south China. The developed inventory of emissions from crop residue burning is expected to help improve air quality models by providing high-resolution spatial and temporal data.

Keywords: crop residue burning; land-cover data; particular matter (PM); gaseous pollutants; emission inventory

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1 Introduction

China is one of the leading crop-producing countries, with a total grain production of $62\ 143.9 \times 10^4\ t$ in 2015 (NBSC, 2016a). This large crop production coupled with a lack of effective emission control strategies for

open biomass burning leads to the large-scale burning of crop straw in open fields during the harvest period. Previous studies have confirmed that crop residue burning contributes to the emission of large quantities of atmospheric pollutants (Zhang et al., 2006; 2009; Wang and Zhang, 2008; He et al., 2013; Zhou et al., 2017). An ac-

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curate inventory of atmospheric pollutants emitted by crop residue burning is necessary to understand, predict, and control the effects of agricultural emissions on air quality and human health.

Published studies have estimated the emissions of atmospheric pollutants from crop residue burning using both bottom-up and top-down methods. Several studies have developed inventories of atmospheric pollutants emitted by crop residue burning based on statistical yearbook data, which generally have province- or city-level resolution (Wang and Zhang., 2008; Peng et al., 2016). County-level pollutant emission inventories are difficult to develop because detailed data can not be obtained directly from statistical yearbooks (Zhou et al., 2017). Emission inventories developed using statistical data have lower resolutions than those based on fire count data from satellite observations. In the top-down method, fire count data are used to determine the temporal and spatial distributions of fire-related emissions. This method can reduce the uncertainty in the estimated emissions of atmospheric pollutants. However, inventories of emissions from crop residue burning tend to have large uncertainty due to the lack of localized vegetation data and/or the use of only province-level data of burned cropland area (Qiu et al., 2016; Wu et al., 2018). The emissions inventories that are based on bottom-up or top-down approaches are necessary, due to their performing cross-examinations of the atmospheric pollutant emissions between various agricultural inventories, it was an important method for verifying the accuracy of emission inventories (EEA, 2007).

In this study, we developed a high-resolution inventory of emissions from residual crop burning based on high-resolution land-cover data (1 km × 1 km) from satellite observations and county-level data burned cropland area data. Land-cover data derived from satellite remote sensing observations were used as activity data to quantify the emissions of atmospheric pollutants from crop residue burning in China for the year 2015 and characterize their spatial and temporal distributions. The resulting high-resolution inventory provides a detailed description of emissions from crop residue burning in China. The inventory is expected to improve the accuracy of atmospheric aerosol models and aid the design of national and local policies to control agricultural emissions.

2 Data and Methods

2.1 Study area

The study region, which includes the 27 provinces and four municipalities (Beijing, Tianiin, Shanghai, and Chongqing) of China (except for Hong Kong, Macao and Taiwan of China) ranges from 3°51'N to 53°33'N and 73°33′E to 135°05′E. Spatial variations in emissions are presented with a resolution of 1 km × 1 km in seven regions: north, northwest, east, central, south, southwest, and northwest China (Fig. 1 and Table 1). The spatial distributions of paddy and nonirrigated farmland in China indicate that paddy areas are primarily distributed in east China, south China, central China, Sichuan, and Heilongjiang, while nonirrigated farmland is prevalent in North China, Northwest China, Southwest China, Northeast China, Henan, and Shandong (http://www.resdc.cn/). We developed an inventory of atmospheric pollutant emissions from crop reside burning using a technology-based emission factor method. This emission inventory was developed for China in 2015 based on satellite-retrieved farmland data. The atmospheric pollutants considered were particulate matter (PM₁₀ and PM_{2.5}) and gaseous pollutants (SO₂, NO_x, CO, VOCs, and NH₃).

2.2 Algorithms for estimating pollutant emissions from crop residue burning

In this section, we describe the method used to estimate emissions from crop residue burning. The method

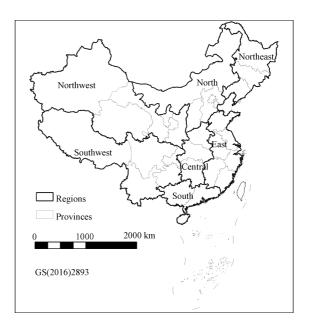


Fig. 1 Study region

employed was derived from 'Technical guidelines on emission inventory development of air pollutants from biomass burning (on trial)' (MEE, 2014) with modification based on the specific activity data used in this study. The emissions of atmospheric pollutants $E(\mathbf{g})$ are given by

$$E = \sum_{i=1}^{M} \sum_{j=1}^{N} \left(A \times P_{i,j} \times Q_{i,j} \times R_i \times B_{i,j} \times EF_i \times F_i \times 100 \right)$$
 (1)

where M is the number of crop types (i), N is the number of grids (j) that represent nonirrigated farmland or paddy regions (land-cover data, $1 \text{ km} \times 1 \text{ km}$), A is the area of the grid (km^2) , $P_{i,j}$ is the crop output per area of crop type i in grid j (kg/ha), $Q_{i,j}$ is the percentage of cropland area of crop type i out of all sown areas in grid j (statistical data and county-level data; %), R_i is the ratio of crop residue to production for crop type i (%), $B_{i,j}$ is the percentage of crop straw burned in fields in region j (%), F_i is the combustion efficiency of crop type i (0.9), and EF_i is the emission factor of crop type i (g/kg).

The statistical data for $P_{i,j}$ and $Q_{i,j}$ were collected at the county-level resolution from statistical yearbooks. In the case where county-level activity data $(P_{i,j})$ or $Q_{i,j}$ could not be directly obtained from the yearbook, we calculated county-level crop yield based on the county-level grain yield since grain yield and crop yield are strongly correlated at the prefecture resolution (R = 0.747) (Zhou et al., 2017). The R_i data were based on a questionnaire implemented by Peng et al. (2016). The

values of $B_{i,j}$, F_i , and EF_i were obtained according to MEE (2014).

2.3 Activity data

The $P_{i,j}$ data were obtained from the National Bureau of Statistics (http://data.stats.gov.cn/), China Rural Statistical Yearbook (NBSC, 2016a), and the Chinese Statistical Yearbook (NBSC, 2016b). The $Q_{i,j}$ data for each grid (the percentages of crop area out of the total sown area within each county) were obtained from provincial statistical yearbooks (ABS, 2016; BBS, 2016; CBS, 2016; FBS, 2016; GDBS, 2016; GXBS, 2016; GSBS, 2016; GZBS, 2016; HBS, 2016; HBBS, 2016; HNBS, 2016; HUBBS, 2016; HUNBS, 2016; HLJBS, 2016; IMBS, 2016; JSBS, 2016; JXBS, 2016; JLBS, 2016; LBS, 2016; NBS, 2016; QBS, 2016; SBS, 2016; SXBS, 2016; SDBS, 2016; SHBS, 2016; SCBS, 2016; TJBS, 2016; TBS, 2016a; 2016b; XBS, 2016; YBS, 2016; ZBS, 2016), including those of the 27 provinces and four municipalities.

The paddy and nonirrigated farmland areas were extracted from the data of land-cover change in China (remote sensing monitoring data; http://www.resdc.cn/default.aspx) with a resolution of 1 km × 1 km. The 'Raster to points' tool in ArcGIS was used to extract the latitudes and longitudes of the paddy and nonirrigated farmlands. The resulting paddy and nonirrigated farmland areas in 2015 in China are shown in Table 1.

Table 1 Extraction of paddy and nonirrigated farmland areas from the land-cover data of China in 2015

| Regions in China | Province | Paddy area (km²) | Nonirrigated farmland area (km²) | Regions in China | Province | Paddy area (km²) | Nonirrigated farmland area (km²) |
|------------------|----------------|---------------------|----------------------------------|------------------|-----------|---------------------|----------------------------------|
| North | Beijing | 131 | 4279 | Central | Hubei | 37744 | 29507 |
| | Hebei | 4538 | 91141 | | Hunan | 43538 | 16402 |
| | Inner Mongolia | 1209 | 112766 | South | Guangdong | 25106 | 16933 |
| | Shanxi | 48 | 59999 | | Guangxi | 24887 | 26073 |
| | Tianjin | 601 | 6052 | | Hainan | 3020 | 5644 |
| Northeast | Heilongjiang | 32288 | 131734 | Southwest | Chongqing | 11160 | 26071 |
| | Jilin | 11216 | 64328 | | Guizhou | 14247 | 34439 |
| | Liaoning | 9411 | 55105 | | Sichuan | 42540 | 76425 |
| East | Anhui | 42952 | 36177 | | Tibet | 186 | 4391 |
| | Fujian | 13950 | 6581 | | Yunnan | 17243 | 50670 |
| | Jiangsu | 41065 | 25004 | Northwest | Gansu | 186 | 65075 |
| | Jiangxi | 32365 | 12229 | | Ningxia | 4752 | 13137 |
| | Shandong | 1759 | 99375 | | Qinghai | 0 | 8224 |
| | Shanghai | 3396 | 241 | | Shaanxi | 8224 | 61423 |
| | Zhejiang | 21569 | 3122 | | Xinjiang | 562 | 76549 |
| Central | Henan | 7595 | 98985 | Total | Total | 457488 | 1318081 |

In published studies, the emissions of atmospheric pollutants from crop residue burning were estimated based on statistical yearbook data or fire count data derived from satellite observations. Alternatively, in this study, we used land-cover data as activity data to calculate the pollutant emissions from crop residue burning. We therefore analyzed the correlation between the total area of sown crops determined from statistical data (China Bureau of Statistics) and farmland area determined from land-cover data. The results indicated a good correlation ($R^2 = 0.8044$; Fig. 2), confirming that our method was valid. Compared to other methods, the use of high-resolution activity data (paddy and nonirrigated farmland areas determined from land-cover data) improved the resolution of the inventory of pollutant emissions from crop residue burning.

2.4 Emission factors

Table 2 summarizes the emission factors used in this study for the burning of different types of crop straws. The emission factors were taken from MEE (2014). These emission factors are lower than those of Wang and Zhang (2008) and Qiu et al. (2016) but higher than

those of Zhou et al. (2017). It would be useful to collect more local data in future studies.

2.5 Uncertainty analysis of atmospheric pollutant emissions

The Monte Carlo method (Cao et al., 2006; Zhang et al., 2009) was used to analyze the uncertainties in pollutant emissions from crop residue burning. These uncertainties were mainly introduced during the calculation process and originated from the activity data, emission factors, and other key parameters (e.g., the ratio of crop residue to production) (Cao et al., 2006; 2011; Zhang et al., 2009; Li et al., 2019). The uncertainty ranges in the emission estimates for different pollutants were obtained in the four following steps. First, we assumed that the uncertainties in the activity data, emission factors, and other key parameters could be described by normal or lognormal distributions. In this study, activity data were assumed to follow uniform distributions (Oiu et al., 2016), and emission factors were assumed to follow lognormal distributions (IPCC, 2011; Peng et al., 2016; Zhou et al., 2017). Second, we extracted the values of analogue activity data and emission factors data that fit

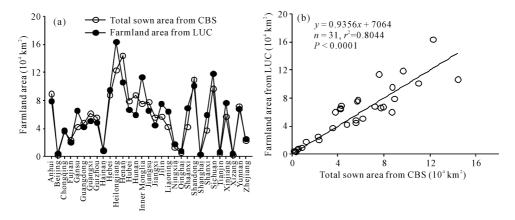


Fig. 2 Comparison of farmland area from LUC (land-cover change data) and CBS (China Bureau of Statistics) in different provinces in China, respectively (a) and regression analysis between total sown areas of crops from the CBS and farmland area from LUC (b) in 27 provinces and four municipalities in China in 2015

Table 2 Emission factors of atmospheric pollutants from crop residue burning

| | | | | | | | | _ |
|--------------|----------------------------|------------------|-------------|--------------|--------------|--------------|-----------|---|
| Type of crop | $PM_{10}\left(g/kg\right)$ | $PM_{2.5}(g/kg)$ | VOCs (g/kg) | $NH_3(g/kg)$ | $NO_x(g/kg)$ | $SO_2(g/kg)$ | CO (g/kg) | - |
| Wheat | 11.95 | 11.71 | 10.4 | 0.68 | 4.3 | 0.44 | 53.0 | - |
| Maize | 7.73 | 7.58 | 7.48 | 0.37 | 3.31 | 0.85 | 59.6 | |
| Rice | 5.78 | 5.67 | 8.45 | 0.53 | 1.42 | 0.53 | 27.7 | |
| Others | 6.93 | 6.79 | 8.45 | 0.53 | 2.92 | 0.53 | 49.9 | |
| | | | | | | | | |

the normal or lognormal distribution. The numerical values of activity data, emission factors, or other key parameters were used to calculate the pollutant emissions based on the algorithms discussed in section 2.1. Third, we extracted the values of key parameters (e.g., activity data and emission factors) over 200 000 times to repeatedly calculate the pollutant emissions. Finally, the standard deviations of the resulting distributions were estimated with a 95% confidence interval.

3 Results

3.1 Levels of pollutant emissions from crop residue burning

Based on the land-cover change data, crop residue burning resulted in substantial emissions of atmospheric pollutants in 2015 (Table 3). The emissions of PM₁₀, PM_{2.5}, VOCs, NOx, SO₂, CO, and NH₃ from crop residue burning on nonirrigated farmland in 2015 were

Table 3 Particulate matter and gaseous emissions from nonirrigated farmland crop residue burning in China in 2015

| Regions in China | Province | $SO_{2}(t)$ | $NO_{x}(t)$ | $NH_3(t)$ | CO(t) | VOCs (t) | $PM_{10}(t)$ | PM _{2.5} (t |
|---------------------|----------------|-------------|-------------|-----------|---------|----------|--------------|----------------------|
| North | Beijing | 39 | 294 | 44 | 3972 | 704 | 789 | 774 |
| | Hebei | 1074 | 7457 | 1175 | 106604 | 18775 | 19548 | 19156 |
| | Inner Mongolia | 614 | 5109 | 824 | 67961 | 12775 | 13785 | 13508 |
| | Shanxi | 1266 | 9001 | 1436 | 125983 | 22834 | 24074 | 23593 |
| | Tianjin | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| Northeast | Heilongjiang | 1480 | 8428 | 1479 | 139582 | 23659 | 20395 | 19984 |
| | Jilin | 1756 | 12067 | 1836 | 170735 | 29658 | 31826 | 31190 |
| | Liaoning | 763 | 4298 | 609 | 65876 | 10370 | 10918 | 10701 |
| East | Anhui | 1844 | 15238 | 2417 | 201843 | 37653 | 41196 | 40369 |
| | Fujian | 129 | 712 | 120 | 11764 | 1945 | 1730 | 1695 |
| | Jiangsu | 1882 | 10555 | 1528 | 163268 | 25885 | 26702 | 26172 |
| | Jiangxi | 342 | 2131 | 321 | 31563 | 5276 | 5506 | 5396 |
| | Shandong | 6528 | 40926 | 5906 | 593396 | 98241 | 107043 | 104915 |
| | Shanghai | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | Zhejiang | 229 | 1457 | 215 | 20942 | 3550 | 3851 | 3775 |
| Central | Henan | 2721 | 15454 | 2391 | 242252 | 39648 | 38912 | 38136 |
| | Hubei | 950 | 5850 | 879 | 87025 | 14491 | 15098 | 14797 |
| | Hunan | 709 | 5220 | 830 | 71604 | 13153 | 14104 | 13822 |
| South | Guangdong | 1024 | 7625 | 1158 | 103839 | 18496 | 20418 | 20009 |
| | Guangxi | 895 | 5802 | 854 | 83450 | 14046 | 15200 | 14897 |
| | Hainan | 4 | 22 | 4 | 375 | 64 | 52 | 51 |
| Southwest | Chongqing | 774 | 4064 | 851 | 61375 | 13498 | 12230 | 11989 |
| | Guizhou | 199 | 1342 | 236 | 18629 | 3715 | 3795 | 3720 |
| | Sichuan | 3150 | 17416 | 3017 | 264191 | 49256 | 47949 | 47001 |
| | Tibet | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | Yunnan | 2549 | 16767 | 2620 | 236987 | 42466 | 45348 | 44446 |
| Northwest | Gansu | 683 | 5351 | 871 | 71924 | 13606 | 14585 | 14293 |
| | Ningxia | 313 | 2408 | 391 | 32704 | 6120 | 6519 | 6388 |
| | Qinghai | 144 | 928 | 132 | 13282 | 2200 | 2434 | 2385 |
| | Shaanxi | 719 | 5049 | 798 | 71062 | 12742 | 13442 | 13174 |
| | Xinjiang | 2590 | 19621 | 3152 | 267120 | 49607 | 53098 | 52036 |
| Total | | 35372 | 230592 | 36092 | 3329309 | 584432 | 610545 | 598372 |

estimated to be 610.5, 598.4, 584.4, 230.6, 35.4, 3329.3, and 36.1 Gg, respectively. The emissions of particulate matter and gaseous pollutants were highest in east China, particularly in the provinces of Shandong, Henan, Anhui, Zhejiang, and Hunan. The emissions of PM₁₀, PM_{2.5}, VOCs, SO₂, NH₃, CO, and NO_x in east China accounted for 30.5%, 30.5%, 29.5%, 31.0%, 29.1%, 30.7%, and 30.8% of nationwide pollutant emissions from crop residue burning, respectively. In contrast, the emissions were lowest in south China, which contributed 5.8%, 5.8%, 5.6%, 5.4%, 5.6%, 5.6%, and 5.8% of total national pollutant emissions, respectively. Beijing, Tianjin, and Shanghai had relatively low levels of atmospheric pollutant emissions from crop residue

burning on nonirrigated farmland due to their low levels of agricultural crop production and strict control policies for crop straw burning. The small percentage of crop straw burned in the fields of Inner Mongolia, where crop straws are mainly used in feed production for animal husbandry, led to low levels of atmospheric pollutant emissions from crop residue burning. The spatial patterns of emissions were similar for different atmospheric pollutants, with emission levels decreasing in the following order among the seven regions: East > Southwest > Northwest > Central > Northeast > North > South China.

Next, we estimated the emissions of particulate matter and gaseous pollutants from the burning of paddy rice straw in China in 2015 (Table 4). The burning of

Table 4 Particulate matter and gaseous emissions from paddy rice residue burning in China

| Regions in China | Province | $SO_{2}(t)$ | $NO_{x}(t)$ | NH ₃ (t) | CO(t) | VOCs (t) | $PM_{10}(t)$ | $PM_{2.5}(t)$ |
|------------------|----------------|-------------|-------------|---------------------|---------|----------|--------------|---------------|
| North | Beijing | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | Hebei | 19 | 52 | 19 | 1007 | 307 | 210 | 206 |
| | Inner Mongolia | 0 | 0 | 0 | 6 | 2 | 1 | 1 |
| | Shanxi | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | Tianjin | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| Northeast | Heilongjiang | 896 | 2399 | 896 | 46806 | 14278 | 9767 | 9581 |
| | Jilin | 217 | 583 | 217 | 11367 | 3467 | 2372 | 2327 |
| | Liaoning | 40 | 107 | 40 | 2079 | 634 | 434 | 426 |
| East | Anhui | 43 | 115 | 43 | 2252 | 687 | 470 | 461 |
| | Fujian | 386 | 1034 | 386 | 20161 | 6150 | 4207 | 4127 |
| | Jiangsu | 2711 | 7264 | 2711 | 141689 | 43223 | 29566 | 29003 |
| | Jiangxi | 2264 | 6067 | 2264 | 118345 | 36102 | 24694 | 24224 |
| | Shandong | 10 | 27 | 10 | 530 | 162 | 111 | 108 |
| | Shanghai | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | Zhejiang | 1016 | 2723 | 1016 | 53114 | 16203 | 11083 | 10872 |
| Central | Henan | 369 | 988 | 369 | 19273 | 5879 | 4021 | 3945 |
| | Hubei | 1886 | 5054 | 1886 | 98583 | 30073 | 20571 | 20179 |
| | Hunan | 4605 | 12339 | 4605 | 240688 | 73423 | 50223 | 49267 |
| South | Guangdong | 1512 | 4050 | 1512 | 78998 | 24099 | 16484 | 16170 |
| | Guangxi | 1383 | 3707 | 1383 | 72307 | 22058 | 15088 | 14801 |
| | Hainan | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| Southwest | Chongqing | 446 | 1195 | 446 | 23310 | 7111 | 4864 | 4771 |
| | Guizhou | 108 | 290 | 108 | 5665 | 1728 | 1182 | 1159 |
| | Sichuan | 2480 | 6646 | 2480 | 129640 | 39547 | 27051 | 26536 |
| | Tibet | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | Yunnan | 900 | 2412 | 900 | 47058 | 14355 | 9819 | 9632 |
| Northwest | Gansu | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | Ningxia | 24 | 65 | 24 | 1262 | 385 | 263 | 258 |
| | Qinghai | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| | Shaanxi | 149 | 399 | 149 | 7777 | 2372 | 1623 | 1592 |
| | Xinjiang | 3 | 8 | 3 | 150 | 46 | 31 | 31 |
| Total | | 21469 | 57521 | 21469 | 1122068 | 342291 | 234135 | 229680 |

paddy rice residue generated the emission of 234.1 Gg PM₁₀, 229.7 Gg PM_{2.5}, 342.3 Gg VOCs, 57.5 Gg NO_x, 57.5 Gg SO₂, 1122.1 Gg CO, and 21.5 Gg NH₃ in 2015. In contrast to the results obtained for nonirrigated farmland, the emissions of atmospheric pollutants from paddy rice straw burning were highest in east and central China (e.g., Hunan, Guangdong, Guangxi, Yunnan, and Jiangsu) and lowest in northwest and north China (e.g., Gansu and Tibet). This spatial distribution of emissions is related to the spatial distribution of paddy rice area sown in 2015. The emissions of atmospheric pollutants from paddy rice straw burning in central and east China accounted for 32.0% and 30.0% of all emissions from paddy rice straw burning in China, respectively; in contrast, northwest and north China accounted for only 0.8% and 0.1% of total emissions, respectively. Among the seven regions, the level of pollutant emissions from paddy rice straw burning decreased in the following order: central > east > southwest > south > northeast > northwest > north China.

3.2 Spatial distributions of pollutant emissions from crop straw burning

The spatial distributions of pollutant emissions (annual gridded emissions at a spatial resolution of $0.1^{\circ} \times 0.1^{\circ}$) from crop straw burning (including paddy rice and dryland crops) in China are shown in Fig. 3. The spatial emission patterns of the different atmospheric pollutants from crop residue burning were similar in 2015. Taking PM_{2.5} as an example, the emissions from crop residue burning were highest over east and central China, particularly in Shandong, Jiangsu, Zhejiang, and Hunan provinces, which have large areas of nonirrigated and paddy farmland relative to other regions in China. In Shandong, Jiangsu, Zhejiang, and Hunan provinces, the annual maximum emission of PM_{2.5} from crop residue burning exceeded 16 930 kg in each grid. Lower emission levels of less than 1000 kg/yr in each grid were observed in Beijing, Tianjin, Guizhou, Qinghai, and Tibet. These lower emissions can be attributed to the severe restrictions on open straw burning, high utilization rates of crop residues, and/or relatively small areas of nonirrigated farmland and paddy land in these regions. In most provinces, the average emissions of PM_{2.5} from crop residue burning ranged from 2000-26 000 kg/grid. The mean annual emissions of PM_{2.5} were 454.0 kg/km² for the burning of dryland crop (e.g., rice, wheat, and

maize) residues and 502 kg/km² for paddy rice straw burning.

For PM₁₀, the national average emissions in 2015 were 463.2 and 511.8 kg/km² for the burning of dryland crop straw and paddy rice straw, respectively; the corresponding values for CO were 2525 and 2453 kg/km², respectively, while those for VOCs were 443.4 and 748.2 kg/km², respectively. The annual emissions of SO₂, NH₃, and NO_x were smaller than those of PM₁₀, PM_{2.5}, and VOCs, which was attributed to their smaller emission factors. The average annual emissions of SO₂, NH₃, and NO_x for paddy rice straw were 46.9, 46.9, and 125.7 kg/km², respectively. The annual emissions of SO₂, NH₃, and NO_x for dryland crop straw burning were 26.8, 27.4, and 174.9 kg/km², respectively.

3.3 Temporal distributions of pollutant emissions from crop straw burning

The temporal distributions of the emissions of different pollutants from crop residue burning were similar in China in 2015. Taking PM_{2.5} as an example, the monthly contributions of PM_{2.5} emissions to annual PM_{2.5} emissions are shown in Fig. 3 to demonstrate the temporal distributions of atmospheric pollutant emissions from crop residue burning. The emissions of atmospheric pollutants exhibited similar monthly variations in different regions, with maximum values observed in October in North, Northeast, Northwest, East, and Southwest China and in June and July in other regions (central and South China). In North China, the emission peaks of PM_{2.5} observed in June (16.9% of total annual PM_{2.5} emissions) and October (80.6%) were related to the burning of both paddy rice and dryland crop straw. These peaks correspond to the harvest season, and winter wheat straw was mainly burned in June. In northeast China, the highest PM_{2.5} emissions were observed in September (43.2%) and October (56.3%), when the crop straw was cleared. Mainly maize straw and soybean straw were burned in October, while rice straw was burned in September after harvesting. In Northwest China, June, September, and October accounted for 10.5%, 30.9%, and 42.6% of annual PM_{2.5} emissions, respectively. The crop of winter wheat was burned in June; maize, rice, and soybean straws were burned in September and October; and cotton straw was burned in November. These crops straw contributions highlight the outdoor burning patterns in northwest China. Secondary

peaks in PM_{2.5} emissions were found in May (15.1%) and October (72.8%) and primary peak of the annual from crop residue burning in southwest China. May is the most important month for harvesting wheat and maize, and rice straw was mainly burned in October. In central and south China, the largest PM_{2.5} emissions were observed in June and July, which accounted for 43.5% and 54.2% of annual PM_{2.5} emissions, respectively. A secondary peak was observed in October, when high emissions occurred in regions with double-season crop systems. While east China also has a double-season

crop system, the $PM_{2.5}$ emissions peaked in October (42.0%) with a second peak in June (38.2%). The $PM_{2.5}$ emissions in June were dominated by the burning of winter-wheat straw, while those in October were dominated by the burning of maize, rice, and soybean straw.

Fig. 4b shows strong seasonal variations in the emissions of atmospheric pollutants from crop residue burning. Emissions were highest in October followed by June. The monthly variability in $PM_{2.5}$ emissions from crop residue burning is controlled by the seasonal operations involved in harvesting different crops.

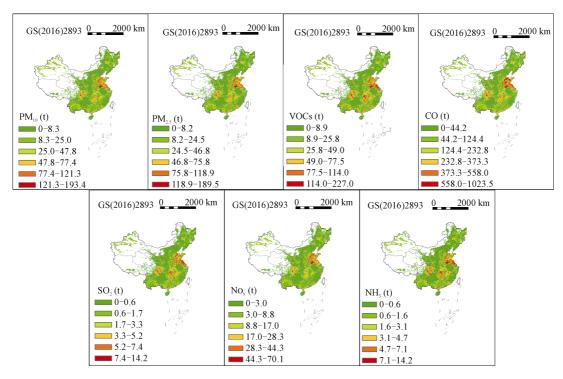


Fig. 3 The spatial distribution of (a) PM_{10} , (b) $PM_{2.5}$, (c) VOCs, (d) CO, (e) SO_2 , (f) NO_x , and (g) NH_3 of crop residue burning in China $(0.1^{\circ} \times 0.1^{\circ})$

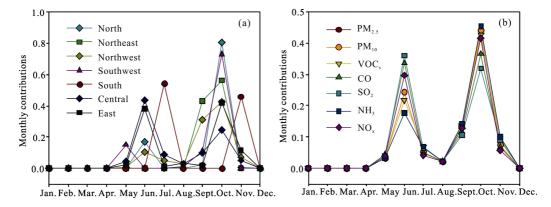


Fig. 4 Monthly contributions of atmospheric pollutant emissions in seven regions (a) and whole China (b)

3.4 Uncertainties in estimates of emissions from crop residue burning

We ran 200 000 Monte Carlo simulations to estimate the range of emissions. The resulting uncertainty ranges are shown in Table 5. The uncertainties ranges of atmospheric pollutants in the estimated emissions (min, max) (95% confidence intervals) of PM_{2.5}, PM₁₀, VOCs, NH₃, NO_x, SO₂, and CO were (21.9%, 106.4%), (2.5%, 71.1%), (3.7%, 73.2%), (32.8%, 121.9%), (-10.1%, 49.9%), (-12.8%, 45.6%), and (-10.9%, 48.6%), respectively. The main source of these uncertainties was the emission factors rather than the agricultural activity data from statistical yearbooks.

The uncertainty ranges for $PM_{2.5}$, PM_{10} , VOCs, and NH_3 emissions were similar (minimum and maximum uncertainty were both larger than 0%), while the uncertainty ranges of NO_x , SO_2 , and CO were similar (minimum < 0 and maximum > 0). This can be explained by the emission factors used in the estimations, which were much lower than $(PM_{2.5}, PM_{10}, VOCs, NH_3)$ or similar to $(NO_x, SO_2, and CO)$ the average values of emission factors used in published studies.

4 Discussion

Currently, the large amount of crop production coupled with a lack of effective emission control strategies for open biomass burning in China has led to the large-scale burning of crop straw in open fields during harvest periods. An accurate inventory of atmospheric pollutants emitted by crop residue burning is necessary to understand, predict, and control the effects of agricultural emissions on air quality and human health. However, past estimates of atmospheric pollutant emissions from crop residue burning were developed based on statistical yearbook data, which generally have low resolution (province- or city-level resolution). In this study, we

developed a high-resolution inventory of emissions from crop residue burning based on high-resolution land-cover data (1 km × 1 km) from satellite observations and county-level burned cropland area data. Data on land-cover change in China can generally be obtained easier than fire count data and are more accurate than statistical data. The emission inventory developed in this study based on land-cover change data has a higher resolution than previously reported inventories.

The spatial variations in atmospheric pollutant emissions from crop residue burning in 2015 were related to the spatial distributions of farmland area in China, including paddy and nonirrigated farmland area. The highest emissions from burning paddy rice straw were found in central and east China (e.g., Hunan, Jiangsu, Sichuan, and Jiangxi provinces), where large areas of sown paddy land were found in 2015. The highest emissions from burning dryland straw were found in east and southwest China (e.g., Shandong and Sichuan provinces) due to the large areas of nonirrigated farmland and the production of abundant straw resources in these regions. Control policies for crop straw burning also affected the spatial variations in the emissions of atmospheric pollutants from crop residue burning. Such policies resulted in lower emissions from crop residue burning in nonirrigated farmland in Beijing, Tianjin, and Shanghai. The monthly variability in pollutant emissions was controlled by the seasonal agricultural operations of different crop types and cropping systems in different regions. The uncertainties in the estimated emissions of different pollutants from crop residue burning were also calculated in this study. Uncertainty is a critical factor in verifying the accuracy of inventories of agricultural emissions. The range of uncertainty in our emissions estimates was as high as -12.8% to 106.4%; however, this range is smaller than those reported in many previous studies. The estimates in this

Table 5 Uncertainty ranges of atmospheric pollutant in emission estimates (min, max) in this study and other references (95% confidence intervals) (%)

| SO_2 | PM _{2.5} | PM_{10} | VOCs | NH ₃ | NO_x | СО | Reference |
|---------------|-------------------|-------------|-------------|-----------------|---------------|---------------|-------------------|
| (-71, 304) | (-60, 83) | - | (-67, 94) | - | (-49, 78) | (-91, 155) | Peng et al., 2016 |
| (-76, 75) | (-54, 34) | (-37, 80) | (-55, 21) | (-78, 76) | (-62, 24) | (-41, 47) | Zhou et al., 2017 |
| (-47, 121) | (-77, 213) | (-77, 213) | (-72, 213) | (-24, 78) | (-36, 92) | (-61, 152) | Qiu et al., 2016 |
| - | (-62.5, 105.1) | - | - | - | - | - | Li et al., 2019 |
| (-12.8, 45.6) | (21.9, 106.4) | (2.5, 71.1) | (3.7, 73.2) | (32.8, 121.9) | (-10.1, 49.9) | (-10.9, 48.6) | This study |

study were improved by the use of higher-quality activity data. The uncertainty was primarily introduced by the nonlocally derived emission factors.

The cross-examination of atmospheric pollutant emissions between different inventories is important to verify the accuracy of emission inventories. We compared the atmospheric pollutant emissions and their spatiotemporal distributions compiled in this study with those reported by Peng et al. (2016) for the year 2009 and Qiu et al. (2016) for the year 2013 along with the EDGAR emission inventories (Emissions Database for Global Atmospheric Research) (Fig. 5).

Peng et al. (2016) used a bottom-up approach, while Qiu et al. (2016) use a top-down approach to compile the inventories. The emissions of pollutants from crop burning estimated in this study generally were lower than the estimates reported by Peng et al. (2016) and Qiu et al. (2016), and the spatial distributions were similar. The differences can be attributed to the different estimation methods, emission factors, and activity data. The emission factors for burned rice, maize, and cotton straw used in Peng et al. (2016) were higher than the values adopted in our study. The lower ratios of crop residue to production and crop yields used in Peng et al.

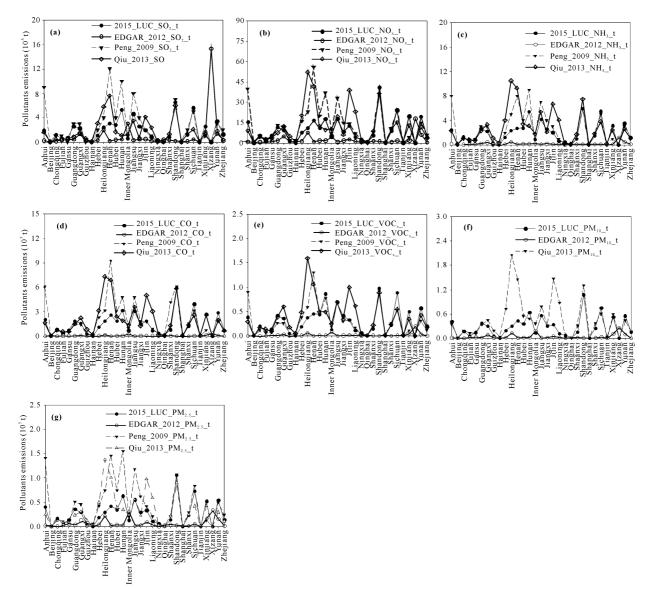


Fig. 5 Comparison of atmospheric pollutant emissions and their spatial distributions in this study with published emission inventories: (a) SO₂, (b) NO_x, (c) NH₃, (d) CO, (e) VOCs, (f) PM₁₀, and (g) PM_{2.5}. 2015_LUC: data from this study; EDGAR_2012: data from EDGAR; Peng_2009: data from Peng et al. (2016); Qiu_2013: data from Qiu et al. (2016)

(2016) did not contribute significantly to the difference between the findings of this study and their results. Comparing the estimated atmospheric pollutant emissions and their spatiotemporal distributions obtained in our study and by Qiu et al. (2016) confirmed the validity of our results. The agricultural activity data used in Qiu et al. were fire count data based on satellite observations, while land-cover data from remote sensing served as the activity data in our study. The emissions of atmospheric pollutants from crop residue burning in EDGAR (Crippa et al., 2018) are lower than those in 2015 LUC (the emission inventory of crop residue burning in China in 2015 that used satellite-retrieved land-cover data as activity data to be compiled). The emissions of particulate and gaseous pollutants from straw combustion between EDGAR and 2015 LUC exhibited significant differences in Inner Mongolia and Tibet, with EDGAR predicting higher pollutant emissions in those regions.

5 Conclusions

The burning of crop residues contributes to the emissions of large quantities of atmospheric aerosols. However, existing inventories of atmospheric pollutant emissions from crop residue burning based on statistical data have low resolution because county-level activity data cannot be obtained directly from statistical yearbooks. Emission inventories based on fire count data derived from satellite observations generally have high uncertainty because of their limited resolution, missing data for small fires, and the lack of localized vegetation data. To improve the resolution and accuracy of inventories of agricultural emissions, we extracted paddy and nonirrigated farmland areas from remote sensing land-cover data (http://www.resdc.cn/data.aspx?DATAID=184) and used them to develop a high- resolution inventory of emissions from crop residue burning.

The emissions of PM₁₀, PM_{2.5}, VOC_s, NO_x, SO₂, CO, and NH₃ from crop residue burning in nonirrigated farmland in China in 2015 were 610.5, 598.4, 584.4, 230.6, 35.4, 3329.3, and 36.1 Gg, respectively; the corresponding values for paddy rice residue burning were 234.1, 229.7, 342.3, 57.5, 57.5, 1122.1, and 21.5 Gg, respectively. The atmospheric pollutant emissions resulting from the burning of dryland crop residues were highest in east and central China, particularly in Shan-

dong, Jiangsu, Zhejiang, and Hunan provinces. For paddy rice straw burning, the atmospheric pollutant emissions were highest in south and central China, including Hunan, Guangdong, Guangxi, Yunnan, and Jiangsu provinces. The emissions of different atmospheric pollutants showed similar monthly variations among different regions, with maximum emissions observed in October in north, northeast, northwest, east, and southwest China and in June and July in other regions (central and south China).

The findings indicate that crop residue burning emits a large quantity of atmospheric pollutants during the harvesting season. Thus, it is necessary to manage crop residue burning more effectively during severe haze events. Further research is needed to improve agricultural inventories through improvements in key data inputs. For example, the percentage of crop straw burned in the field, which may vary over time and space, needs further study.

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