1	Spatiotemporal variations and factors of air quality in urban central
2	China during 2013–2015
3	
4	Supplementary data
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#### 12 Data and Methods

#### 13 **Data**

Designed as a mix of nine urban sites and one background sites for the point layout of ambient air 14 quality monitoring, presently there are ten national automated monitoring sites in Hefei. Monitoring 15 systems were installed and used to measure the pollutant concentrations according to China 16 Environmental Protection Standards (HJ655-2013). The mass concentrations of PM<sub>2.5</sub> and PM<sub>10</sub> were 17 measured using the Tapered Element Oscillating Microbalance analyzers, and levels of SO<sub>2</sub>, NO<sub>2</sub>, O<sub>3</sub> 18 19 and CO were measured using the using of the ultraviolet fluorescence method, the chemiluminescence 20 method, the UV-spectrophotometry method, and the non-dispersive infrared absorption method (or the 21 gas filter correlation infrared absorption method), respectively. The diurnal concentrations of each 22 pollutant were calculated only when there were more than 16 h valid data. To better study the temporal changes of pollutants, the corresponding data are divided into four seasons: spring (March-May), 23 summer (June-August), autumn (September-November) and winter (December-February). 24

## 25 Methodology

# 26 Hybrid Single Particle Lagrange Integrated Trajectory (HYSPLIT) model

HYSPLIT-4 model was developed in the Air Resources Laboratory (ARL) of the National Oceanic 27 and Atmospheric Administration (NOAA). Corresponding meteorological data with a 1° × 1° grid used 28 in the HYSPLIT-4 model was acquired from the Global Data Assimilation System (GDAS) 29 (ftp://arlftp.arlhq.noaa.gov/pub/archives/gdas1/). The backward trajectories were run four times per day 30 at a starting times of 0:00, 06:00, 12:00, and 18:00 UTC (8:00, 14:00, 20:00, and 02:00 LT, respectively) 31 32 with starting height of 500 m above ground level (agl). In order to divide the seasonal trajectories into 33 distinct transport groups, cluster analysis was carried out on the basis of the hierarchical clustering 34 method [1].

# 35 **Potential source contribution function (PSCF)**

The zone of concern is divided into small equal grid cells  $(i \times j)$ . The PSCF value in the  $ij^{\text{th}}$  cell represents the probability of air parcel residence time and is expressed as:

$$PSCF_{ij} = \frac{m_{ij}}{n_{ii}} \tag{1}$$

where *i* and *j* denote the latitude and longitude,  $n_{ij}$  is on behalf of the total number of endpoints that fall in the *ij*<sup>th</sup> cell, and  $m_{ij}$  represents the number of endpoints in the same cell associated with samples for which the monitored mass concentration of pollutant exceeds the criterion value in the *ij*<sup>th</sup> cell. The GB3095-2012 PM<sub>2.5</sub> guideline value for daily average of 75 µg/m<sup>3</sup> is used as the criterion-value. To better reflect the uncertainty in cells, the PSCF values should be multiplied by an arbitrary weighting function  $W(n_{ij})$  when  $n_{ij}$  is lower than three times of average number of trajectory endpoints ( $n_{ave}$ ) in each cell [2,3], and the weight PSCF (WPSCF) is described as follows:

$$WPSCF_{ij} = \frac{m_{ij}}{n_{ij}} \times W(n_{ij})$$
<sup>(2)</sup>

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$$W(n_{ij}) = \begin{cases} 1.00, \ 3 n_{ave} < n_{ij} \\ 0.70, \ 1.5 n_{ave} < n_{ij} \le 3n_{ave} \\ 0.40, \ n_{ave} < n_{ij} \le 1.5 n_{ave} \\ 0.20, \ n_{ij} \le n_{ave} \end{cases}$$
(3)

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# 49 Concentration-weighted trajectory (CWT) model

Since PSCF just reflects the proportion of polluted trajectories in a cell grid, which is impossible to distinguish the contribution to the pollution levels of the target object for those who have the same PSCF value, CWT model is applied to weight trajectories with related PM<sub>2.5</sub> concentrations for better unraveling above limitation and calculating the relative contribution of different sources-areas. The geographical field is divided into grid cells representing an area of  $0.5^{\circ} \times 0.5^{\circ}$ . The CWT is defined as follows:

$$C_{ij} = \frac{\sum_{h=1}^{M} C_h \times \tau_{ijh}}{\sum_{h=1}^{M} \tau_{ijh}} \times W(n_{ij})$$
(4)

57 Where  $C_{ij}$  is designed as the average weight concentration of the trajectory *h* in the *ij*<sup>th</sup> cell, *h* 58 represents the index of the trajectory, *M* represents the total number of trajectories, *C<sub>h</sub>* denotes PM<sub>2.5</sub> 59 concentrations in the trajectory *h* through *ij*<sup>th</sup> cell,  $\tau_{ijh}$  is the time that trajectory *h* resides in the *ij*<sup>th</sup> cell (*i*, 60 *j*). The  $W(n_{ij})$  used in PSCF also is utilized to CWT method for reducing the uncertainty in cells.

The study domain is in the range of 20–60 °N and 90–130 °E, which includes more than 95% of areas covered by all the paths. The total number of endpoints in each season is about 51840, and the area covered by the trajectories is divided into 6400 grid cells ( $0.5^{\circ} \times 0.5^{\circ}$ ). Hence, each cell owns an average of 9 trajectory endpoints, that is, the *n*<sub>ave</sub> is equal to 9. Both PSCF and CWT analysis are run by the MeteoInfo soltware-TrajStat Plugin.

#### 66 Integrated exposure-response (IER) model

67 This model is based on the assumption that the cause-specific mortality rates are independent of 68 PM<sub>2.5</sub> concentrations. The relative risk (RR) for a specific disease due to PM<sub>2.5</sub> exposure is the ratio of 69 the probability of an event occurring in an exposed group to the probability of the event occurring in a 70 comparison, nonexposed group [4,5]. Adult STR (stroke, including ischemic and hemorrhagic stroke), Ischemic heart disease (IHD), lung cancer (LC) and chronic obstructive pulmonary disease (COPD), 71 ranking top ten in the international classification, are selected as the causes of mortality. The mortality 72 burdens attributable to ambient PM2.5 exposure in 2014 and 2015 are calculated using the 73 disease-specific mortality and Hefei population data. The RR of several causes of premature mortality 74 (AMort) among adults including STR, IHD, LC, and COPD, are estimated according to the IER 75 functions. 76

77 The RR for each disease is calculated based on equation (5):

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$$RR(C) = \begin{cases} 1 + \alpha (1 - e^{-\gamma (C - C_0)\delta}), & C \ge C_0 \\ 1 & , & C < C_0 \end{cases}$$
(5)

where *C* is the annual PM<sub>2.5</sub> exposure concentration ( $\mu$ g/m<sup>3</sup>); *C*<sub>0</sub> is the endpoint-specific theoretical minimum-risk concentration of PM<sub>2.5</sub>;  $\alpha$ ,  $\delta$  and  $\gamma$  are parameters used to describe the shape of C–RR curves. The population attributable fraction (AF), representing the fraction of mortality for a specific disease due to PM<sub>2.5</sub> exposure, is calculated by RR factors through the formula (6):

 $AF = (RR - 1)/RR \tag{6}$ 

The AFs are estimated with the annual average  $PM_{2.5}$  concentrations in 2014 and 2015 for four diseases. Formula (7) is used to calculate the  $\Delta$ Mort for a specific disease attributable to  $PM_{2.5}$  exposure 86 [6,7]. y<sub>0</sub> is the baseline mortality rate caused by a specific disease, and Pop is the total population

- 87 exposed to PM<sub>2.5</sub> in Hefei. The population data of Hefei in 2014 and 2015 are obtained from Hefei
- 88 Statistics Yearbook.

$$\Delta Mort = y_0 \times AF \times Pop \tag{7}$$

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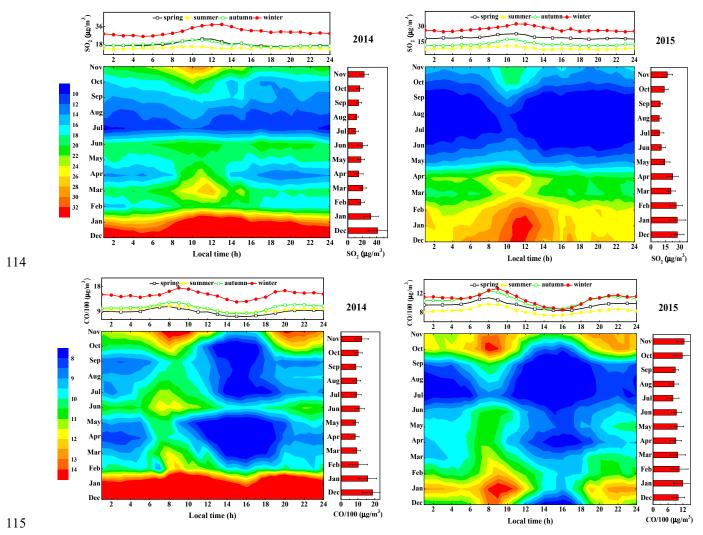
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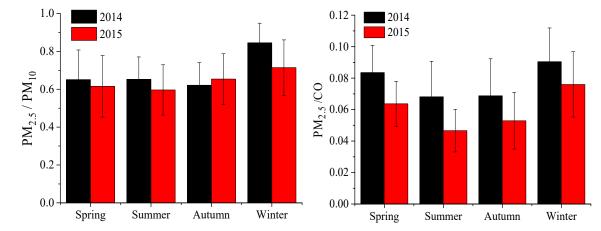
	PM <sub>2.5</sub> (μg/m <sup>3</sup> )	Attributable Fractions (AFs) (%)			
		STR	IHD	LC	COPD
2014	89	48	32	32	25
2015	61	44	29	26	21
CAAQS grade II	35	34	23	18	15
WHO IT-2	25	25	20	14	11
WHO IT-3	15	11	15	8	7
WHO AQG	10	4	10	4	3

**Table S1** The attributable fractions due to PM<sub>2.5</sub> for STR, IHD, LC and COPD.

**S**6



116 Fig. S1. Diurnal variations of SO<sub>2</sub> (upper) and CO (down).



119 Fig. S2. The seasonal variations of  $PM_{2.5}/PM_{10}$  ratios and  $PM_{2.5}/CO$  ratios.

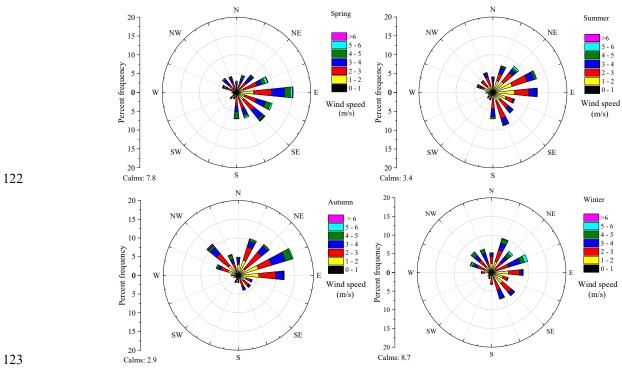
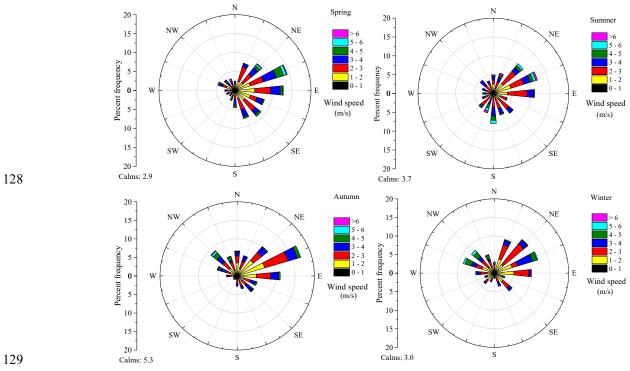


Fig. S3. The relationship between percent frequency and wind speed, direction for seasonal distribution in 2014 at
 Hefei.



130 Fig. S4. The relationship between percent frequency and wind speed, direction for seasonal distribution in 2015 at

131 Hefei.