

Supplementary information for

**Estimation of Secondary PM_{2.5} in China and the United States
using a Multi-Tracer Approach**

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Table S1. Impacts of primary sulfate/nitrate emission uncertainty on the estimated secondary proportion of PM_{2.5}¹ in China.

City	Change of sulfate/nitrate emission	
	10%	3%
Beijing	40.3	42.4
Tianjin	61.9	62.6
Shijiazhuang	44.8	45.2
Taiyuan	43.1	44.3
Hohhot	48.6	46.5
Shenyang	48.7	49.3
Changchun	47.9	46.7
Harbin	66.9	67.3
Shanghai	68.0	67.8
Nanjing	50.3	50.5
Hangzhou	45.6	45.3
Hefei	65.4	65.0
Fuzhou	64.8	62.1
Nanchang	62.5	62.7
Ji'nan	54.6	54.6
Zhengzhou	54.6	56.3
Wuhan	61.5	60.8
Changsha	65.9	64.9
Guangzhou	65.2	65.6
Nanning	65.2	65.2
Haikou	65.9	65.4
Chongqing	62.7	62.8
Chengdu	45.3	44.8
Guiyang	65.6	64.3
Kunming	70.4	70.5
Lhasa	56.1	55.1
Xi'an	52.6	53.0
Lanzhou	60.0	59.5
Xining	59.1	58.9
Yinchuan	59.5	60.1
Urumqi	72.1	70.3

¹ Based on the MEE observations in 2016.

Table S2. The comparison of two assumptions on weak correlation between secondary PM_{2.5} and X-tracer in the secondary proportions of PM_{2.5} (%) ¹.

City	$\delta 1$ ²	$\delta 2$ ³	$\delta 1 - \delta 2$
Beijing	40.3	38.2	2.1
Tianjin	61.9	61.4	0.5
Shijiazhuang	44.8	42.8	2.0
Taiyuan	43.1	40.9	2.2
Hohhot	48.6	46.8	1.7
Shenyang	48.7	47.0	1.7
Changchun	47.9	46.3	1.6
Harbin	66.9	68.3	-1.4
Shanghai	68.0	67.8	0.2
Nanjing	50.3	47.6	2.7
Hangzhou	45.6	43.4	2.2
Hefei	65.4	65.1	0.3
Fuzhou	64.8	63.9	0.8
Nanchang	62.5	61.2	1.4
Jinan	54.6	52.7	1.9
Zhengzhou	54.6	52.8	1.9
Wuhan	61.5	60.4	1.1
Changsha	65.9	65.7	0.2
Guangzhou	65.2	64.9	0.3
Nanning	65.2	64.8	0.5
Haikou	65.9	66.4	-0.4
Chongqing	62.7	61.9	0.7
Chengdu	45.3	43.8	1.4
Guiyang	65.6	66.1	-0.5
Kunming	70.4	70.9	-0.4
Lhasa	56.1	55.1	1.0
Xian	52.6	50.7	1.9
Lanzhou	60.0	59.6	0.4
Xining	59.1	58.3	0.8
Yinchuan	59.5	58.6	0.9
Urumqi	72.1	73.3	-1.2

¹ Based on the MEE observations in 2016.

² $\delta 1$ is mean value of the estimated interval.

³ $\delta 2$ is the specific value when $r=0$.

Table S3. List of 31 populous cities and 19 regional background cities and the corresponding averaged PM_{2.5} concentrations ($\mu\text{g m}^{-3}$) in China during the studying period.

Province	Populous city	PM _{2.5}	Regional background city	PM _{2.5}
Beijing	Beijing	69.4	-	-
Tianjin	Tianjin	69.5	-	-
Hebei	Shijiazhuang	92.0	Zhangjiakou	33.3
Shanxi	Taiyuan	62.8	-	-
InnerMogolia	Hohhot	42.0	Xilingol	16.2
Liaoning	Shenyang	57.5	-	-
Jilin	Changchun	51.6	Yanbian	31.6
Heilongjiang	Harbin	57.5	Daxinanling	20.4
Shanghai	Shanghai	45.6	-	-
Jiangsu	Nanjing	52.8	-	-
Zhejiang	Hangzhou	49.1	Zhoushan	26.4
Anhui	Hefei	61.6	Huangshan	28.3
Fujian	Fuzhou	27.8	Nanping	23.3
Jiangxi	Nanchang	40.9	-	-
Shandong	Jinan	75.3	Weihai	34.0
Henan	Zhengzhou	80.2	-	-
Hubei	Wuhan	62.0	-	-
Hunan	Changsha	57.9	-	-
Guangdong	Guangzhou	38.4	Shanwei	27.3
Guangxi	Nanning	39.0	Beihai	29.8
Hainan	Haikou	20.5	Sanya	16.0
Chongqing	Chongqing	50.8	-	-
Sichuan	Chengdu	59.4	Aba	15.1
Guizhou	Guiyang	36.7	Qianxinan	18.2
Yunnan	Kunming	28.8	Diqing	13.5
Xizang	Lhasa	22.5	Linzhi	10.5
Shannxi	Xi'an	68.8	-	-
Gansu	Lanzhou	50.1	Jiayuguan	30.9
Qinghai	Xining	49.0	Yushu	18.0
Ningxia	Yinchuan	44.7	Guyuan	35.6
Xinjiang	Urumqi	70.4	Altay	11.3

Table S4. List of PM_{2.5} component measurements ($\mu\text{g m}^{-3}$) of China in previous studies.

City	Period	PM _{2.5}	SO ₄ ²⁻	NO ₃ ⁻	NH ₄ ⁺	SOA ¹	SPM/PM _{2.5}	Mean SPM/PM _{2.5}	References
Beijing	2012 Summer	103	20.6	15.8	8.3	10.1	53.2%		Tian et al., 2015
	2012 - 2013	72	9.3	11.9	5.3	9.6	50.3%		Liu et al., 2018b
	2013 Winter	159	25.4	19.0	15.6	32.2 ^a	58.0%		Tao et al., 2015
	2013 Winter	143	23.9	20.2	16.5	15.4	53.1%		Huang et al., 2014b
	Jan 2014	153	9.6	12.1	6.7	33.8 ^c	40.6%		Gao et al., 2018
	Apr 2014	115	10.7	10.7	11.4	15.2 ^c	41.6%		Gao et al., 2018
	Jul 2014	96	25.6	25.6	14.1	11.1 ^c	79.7%	31% ~ 80%	Gao et al., 2018
	Oct 2014	139	21.1	45.5	13.9	23.0 ^c	74.5%		Gao et al., 2018
	2014 Winter	138	21.0	26	14.1	17.4	51.8%		Lin et al., 2016
	Jun 2014 - Apr 2015	100	14.3	17.1	11.5	12.4 ^b	55.6%		Huang et al., 2017
	May 2015 - Apr 2016	114	8.6	11.1	5.2	10.1	30.9%		Yu et al., 2019
	Jul 2015 - Apr 2016	81	9.6	12.4	8.6	7.7	47.1%		Xu et al., 2019
	Oct 2016	95	16.8	16.8	12.5	12.3 ^b	61.5%		Zhang et al., 2018
MTEA estimation ²							41%		This study
Tianjin	Jun 2014 - Apr 2015	106	16.6	16.2	13.7	10.4 ^b	53.8%		Huang et al., 2017
	Jul 2015 - Apr 2016	86	12.1	13.9	10.5	7.3	51.0%		Xu et al., 2019
	Jul 2014	113	12.2	16.2	9.3	11.0 ^c	43.0%	41% ~ 54%	Gao et al., 2018
	Oct 2014	101	12.8	9.9	8.2	11.1 ^c	41.4%		Gao et al., 2018
	2014 Winter	183	19.5	40.7	15.1	21.8 ^c	53.1%		Gao et al., 2018
	May 2015 - Apr 2016	120	18.1	20.3	8.5	10.0	47.4%		Liu et al., 2018a

MTEA estimation							63%		This study	
Shijiazhuang	Jun 2014 - Apr 2015	155	25.5	23.4	18.8	17.7 ^b	55.0%	51% ~ 55%		
	Jul 2015 - Apr 2016	105	16.8	14.9	12.3	9.6	51.0%			
MTEA estimation							49%		This study	
Shanghai	2012 Spring	70	15.3	8.6	6.4	5.7	51.4%	Zhao et al., 2015		
	2012 Summer	51	9.7	5.6	3.6	3.7	44.3%	Zhao et al., 2015		
	2012 Fall	82	17.9	20.2	7.8	7.7	65.4%	Zhao et al., 2015		
	2012 Winter	70	11.6	13.2	5.6	8.5	55.6%	Zhao et al., 2015		
	2012 Spring	64	12.0	10.8	4.3	4.9	50.0%	Huang et al., 2014a		
	2011- 2013 Spring	49	11.0	11.0	6.9	5.9	71.0%	Wang et al., 2016a		
	2011- 2013 Summer	31	8.1	5.2	4.2	4.7	67.3%	Wang et al., 2016a		
	2011- 2013 Fall	41	8.8	7.4	4.8	5.2	63.9%	Wang et al., 2016a		
	2011- 2013 Winter	65	13.0	13.2	8.3	6.7	63.4%	26% ~ 71%		
	2012 - 2013	68	13.6	11.9	5.8	8.6	58.7%	Liu et al., 2018b		
	Oct - Nov 2013	75	12.9	15.0	6.6	4.2	51.6%	Ming et al., 2017		
	Dec 2013 – Jan 2014	138	19.5	29.1	12.6	10.3	51.8%	Ming et al., 2017		
	Mar 2014 – Apr 2014	96	12.3	10.4	5.5	4.5	34.1%	Ming et al., 2017		
	Jun 2014 – Jul 2014	56	6.7	2.8	2.1	2.9	25.9%	Ming et al., 2017		
	2013 Winter	91	10.8	12.4	7.5	21.8 ^b	57.7%	Huang et al., 2014b		
Dec 2014 – Jan 2015	103	18.3	25.4	14.4	14.1 ^b	70.1%	Du et al., 2017			
Mar 2015 – Apr 2015	74	8.7	11.2	5.7	9.2 ^b	47.0%	Du et al., 2017			
MTEA estimation							67%		This study	
Nanjing	Apr – May 2013	110	23.1	11.7	6.4	17.7 ^a	53.5%	52% ~ 79%		

	Aug 2013	86	18.4	8.1	5.4	14.2 ^a	53.6%	Li et al., 2016
	Oct 2013	77	12.6	7.3	3.8	36.8 ^a	78.6%	Li et al., 2016
	Dec 2014 – Jan 2015	100	11.7	16.4	12.3	11.8 ^b	52.2%	Du et al., 2017
	Mar 2015 – Apr 2015	83	21.4	16.1	7.9	9.1 ^b	65.6%	
	MTEA estimation						53%	This study
Hangzhou	Oct 2013	36	9.7	5.3	6.0	6.5	76.4%	Wu et al., 2016
	MTEA estimation						53%	This study
	Dec 2012 – Jan 2013	75	10.6	5.8	5.1	6.5	37.3%	Liu et al., 2014
	2012 - 2013	75	13.1	7.2	4.8	8.4	44.6%	Liu et al., 2018b
	Nov 2012 – Dec 2013	61	9.3	5.7	4.2	11.9 ^b	51.0%	Chen et al., 2016
	2013 Summer	51	8.9	4.9	4.0	6.6	47.8%	Cui et al., 2015
	2013 Fall/Winter	68	9.8	7.3	4.5	9.4	45.6%	Cui et al., 2015
	2013 Winter	69	12.7	8.9	6.9	11.4 ^b	57.8%	Huang et al., 2014b
	2014 Spring	44	8.2	2.4	3.6	4.5	42.5%	Tao et al., 2017
	2014 Summer	37	7.6	0.3	2.6	3.7	38.4%	Tao et al., 2017
	2014 Fall	48	11.4	1.0	4.4	4.7	44.8%	Tao et al., 2017
	2014 Winter	63	9.8	5.5	4.8	7.0	43.0%	Tao et al., 2017
	MTEA estimation						66%	This study
	Dec 2012	137	13.5	9.8	6.6	21.6 ^b	37.6%	Zhang et al., 2015
	2012 Spring	164	17.8	15.2	6.5	13.9	32.6%	Niu et al., 2016
	2012 Summer	109	25.0	10.1	6.6	8.8	46.3%	Niu et al., 2016
	2012 Fall	155	18.7	16.5	8.2	18.4	39.9%	Niu et al., 2016
	Nov 2012 – Feb 2013	244	32.1	29.3	16.8	39.7	48.3%	Niu et al., 2016
Xi'an							33% ~ 55%	

Dec 2014 – Nov 2015		113	15.2	16.6	8.4	21.3	54.7%	Dai et al., 2018
MTEA estimation								
Chengdu	Oct – Nov 2014	62	10.5	9.3	6.9	8.3 ^b	56.5%	Wang et al., 2018
	Jan – Feb 2015	114	16.4	17.5	12.7	15.8 ^b	54.7%	Wang et al., 2018
	Apr 2015	48	8.3	5.9	5.1	5.0 ^b	44% ~ 57%	Wang et al., 2018
	Jul 2015	45	9.7	3.9	4.2	5.9 ^b	52.6%	Wang et al., 2018
	Jan 2015	48	6.1	3.7	2.4	8.7	43.5%	Li et al., 2017a
MTEA estimation								
2012 - 2013		74	19.7	6.5	6.1	8.6	55.3%	Liu et al., 2018b
Oct – Nov 2014		56	9.9	7.8	5.7	7.8 ^b	55.7%	Wang et al., 2018
Jan – Feb 2015		115	17.5	15.8	11.3	19.4 ^b	44% ~ 56%	Wang et al., 2018
Apr 2015		58	10.4	5.9	5.2	8.0 ^b	50.1%	Wang et al., 2018
Jul 2015		54	11.1	1.6	4.0	6.8 ^b	43.5%	Wang et al., 2018
MTEA estimation								
Dec 2012		120	11.8	7.2	6.7	21.2	39.1%	Tan et al., 2016
Jun – Jul 2013		34	4.3	1.9	1.9	5.8	40.9%	Tan et al., 2016
Apr – May 2014		83	4.0	1.7	0.8	8.0	18% ~ 41%	Wang et al., 2016b
Aug 2014		38	4.8	2.0	1.3	3.5	30.5%	Wang et al., 2016b
Oct 2014		93	5.8	7.1	3.6	12.7	31.4%	Wang et al., 2016b
Jan, Dec 2014		141	7.6	10.1	6.0	18.2	29.7%	Wang et al., 2016b
MTEA estimation								
Sep – Oct 2013		102	19.4	2.6	8.7	12.5 ^b	42.4%	Tang et al., 2017
Dec 2013 – Jan 2014		145	19.3	9.7	14.3	20.5 ^b	41% ~ 44%	Tang et al., 2017

	Apr – May 2014	97	17.0	1.4	7.5	14.0 ^b	41.1%	Tang et al., 2017
	Jul – Aug 2014	78	13.9	2.9	7.4	9.4 ^b	43.1%	Tang et al., 2017
	MTEA estimation						67%	This study
Haikou	Jan 2015	17	3.1	0.5	1.0	2.3	40.1%	Liu et al., 2017
	Mar 2015	9	1.6	0.2	0.5	1.2	38.8%	Liu et al., 2017
	Jul 2015	23	3.8	0.3	0.8	2.4	31.7%	Liu et al., 2017
	Sep 2015	47	7.9	3.1	3.0	2.8	35.7%	Liu et al., 2017
	MTEA estimation						61%	This study
Zhengzhou	Oct 2014	143	19.6	17.9	9.2	12.0	41.0%	Jiang et al., 2017
	Dec 2014 – Jan 2015	191	23.5	26.5	19.8	22.6	48.4%	Jiang et al., 2017
	Apr 2015	138	19.7	20.3	14.4	11.3	47.6%	Jiang et al., 2017
	Jul 2015	110	24.2	14.3	13.9	7.3	54.3%	Jiang et al., 2017
	MTEA estimation						60%	This study
Shenyang	2013 – 2014	82	13.2	4.6	4.5	11.7	41.5%	Liu et al., 2018b
	MTEA estimation						51%	This study
Lhasa	2013 – 2014	36	0.8	0.5	0.4	7.6	25.8%	Liu et al., 2018b
	MTEA estimation						64%	This study

¹ SOA = 0.5*OM, OM = f*OC. Default f is 1.2. In case of a, b and c, the f is 1.8, 1.6 and 1.4 respectively.

² For period of 2014-2018.

Table S5. Impacts of anthropogenic emission uncertainty on the estimated secondary proportion of PM_{2.5}¹ in China.

City	Secondary proportion of PM _{2.5}	Change of secondary proportion of PM _{2.5}	
		$a - 0.1$	$a + 0.1$
Beijing	40.3	-2.9	2.1
Tianjin	61.9	-3.0	0.7
Shijiazhuang	44.8	-1.7	0.4
Taiyuan	43.1	-2.8	1.1
Hohhot	48.6	3.2	-2.1
Shenyang	48.7	-1.7	0.7
Changchun	47.9	2.2	-1.1
Harbin	66.9	-0.5	0.4
Shanghai	68.0	-6.7	-0.2
Nanjing	50.3	0.6	0.2
Hangzhou	45.6	0.0	-0.3
Hefei	65.4	-4.2	-0.5
Fuzhou	64.8	3.9	-2.7
Nanchang	62.5	-1.0	0.2
Ji'nan	54.6	0.4	0.1
Zhengzhou	54.6	-3.6	1.7
Wuhan	61.5	-2.4	-0.8
Changsha	65.9	-2.5	-1.0
Guangzhou	65.2	-1.5	0.5
Nanning	65.2	-1.8	-0.1
Haikou	65.9	1.2	-0.5
Chongqing	62.7	-1.2	0.2
Chengdu	45.3	2.8	-0.5
Guiyang	65.6	1.7	-1.2
Kunming	70.4	-2.5	0.0
Lhasa	56.1	1.9	-1.0
Xi'an	52.6	-0.5	0.4
Lanzhou	60.0	1.5	-0.5
Xining	59.1	-1.0	-0.2
Yinchuan	59.5	-0.7	0.7
Urumqi	72.1	-6.8	-1.8

¹ Based on the MEE observations in 2016.

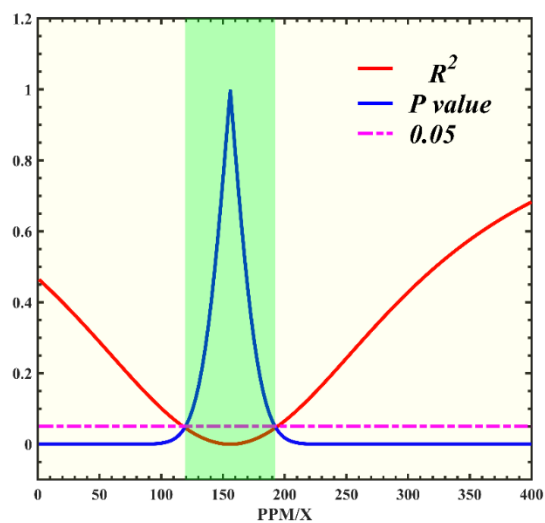


Figure S1. An illustration for scanning the aim interval (green shadow) in the MTEA approach.

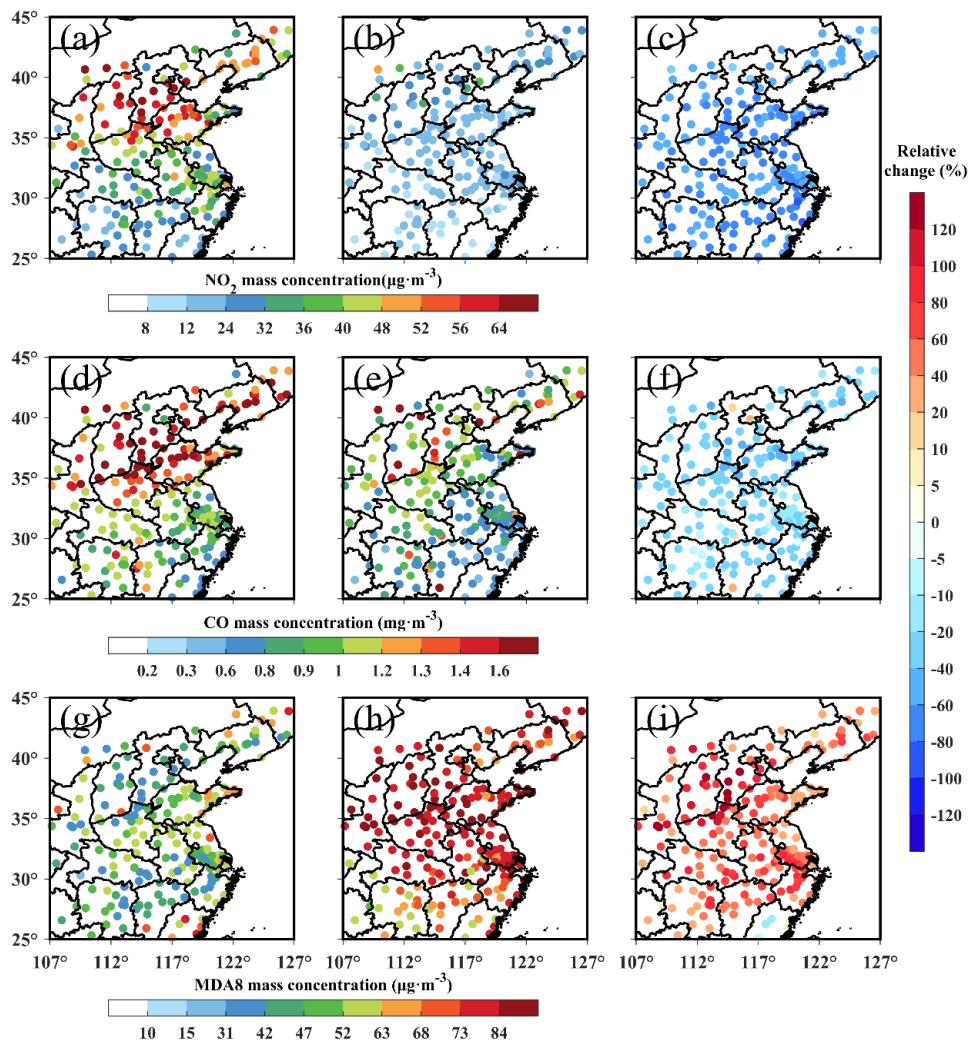


Figure S2. The spatial distribution of (a-b) NO₂, (d-e) CO and (g-h) maximum daily 8-h average O₃ (MDA8) mass concentrations before (01~23 Jan 2020) and during (23-Jan ~ 17-Feb 20) COVID-19 national lockdown. The right panel (c, f, and g) indicates relative change, i.e. (post-lockdown – pre-lockdown)/pre-lockdown.

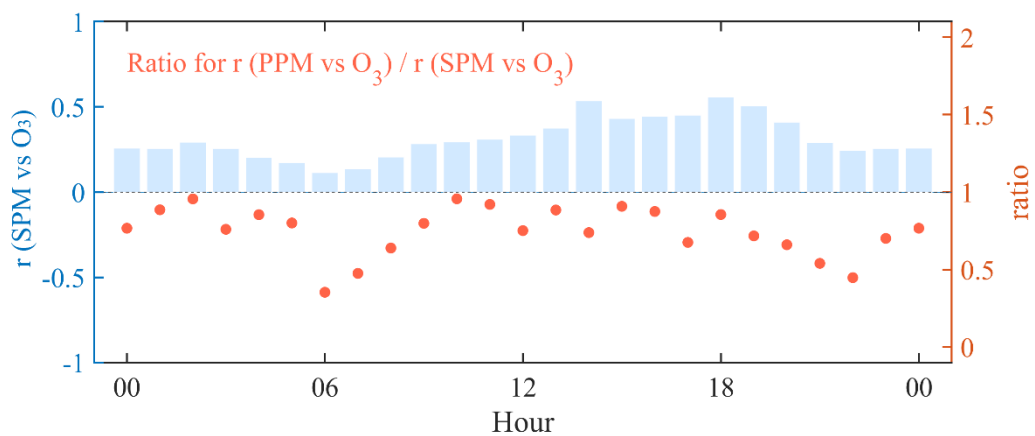


Figure S3. The nationwide correlation between PM versus O₃. Blue bars denote the correlation coefficients between SPM and O₃. The red dots indicate the ratios of the correlation coefficient of PPM vs O₃ to that of SPM vs O₃.