



Supplement of

Marine-derived water-soluble organic nitrogen in coastal air: influence of ocean productivity on atmospheric nitrogen cycling

Jiao Tang et al.

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1 **Text S1. PMF model**

2 Data Input

3 The data input includes concentrations and uncertainties of various types of chemicals
4 in 84 TSP samples. The types of chemicals include WSON, OC, EC, metal elements
5 (Ni, V, Pb, Cu, Zn, Mn, Fe, Al), water-soluble ions (K^+ , Na^+ , Mg^{2+} , Ca^{2+} , NH_4^+ , Cl^- , SO_4^{2-} ,
6 NO_3^-), biomass burning tracers (levoglucosan, mannosan, galactosan), steranes,
7 hopanes, and SOA tracers (2-MGA, 2MGL, MBTCA, o/p-phthalic acid). The uncertainty
8 file provides species-specific parameters that EPA PMF 5.0 uses to calculate
9 uncertainties for each sample. If the concentration is less than or equal to the method
10 detection limit (MDL), the uncertainty (Unc) is calculated using a fixed fraction of the
11 MDL (1), and the concentration is replaced with half of the MDL.

$$12 \text{ Unc} = \frac{5}{6} \times \text{MDL} \quad (1)$$

13 If the concentration is higher than the MDL, the calculation is based on a user- provided
14 fraction of the concentration and MDL (Equation 2). Al element was set as bad,
15 resulting in solutions that didn't converge. Weak signals were found for species of
16 galactosan, mannosan, levoglucosan, hopanes and steranes, 2-MGA, and o-Phthalic
17 acid, which had a low signal-to-noise ratio ($S/N < 3$). Other species with an S/N greater
18 than 3 were set as strong.

$$19 \text{ Unc} = \sqrt{(\text{Error Fraction} \times \text{concentrations})^2 + (0.5 \times \text{MDL})^2} \quad (2)$$

20 Model run

21 The PMF model was executed using three to seven factors, 100 runs, and a random
22 starting point, resulting in converged solutions. Despite yielding the lowest $Q_{\text{robust}}/Q_{\text{true}}$
23 ratio (1.043, approaching unity), the six-factor solution exhibited mixing of the NH_4^+
24 and SO_4^{2-} with biomass burning tracers. This chemically implausible separation
25 rendered the six-factor interpretation problematic. Consequently, the seven-factor
26 solution was selected for source profile analysis as it provided physically meaningful
27 resolution.

28 *For WSON was set as the total variable:*

29 Uncertainty Estimation: To understand the uncertainty of the seven-factor solution, we
30 did a complementary analysis of displacement (DISP), BS, and bootstrapping with
31 displacement (BS-DISP). We used DISP intervals to include the effects of rotational
32 ambiguity, but not the effects of random errors in the data. BS intervals included the
33 effects of random errors and partially included the effects of rotational ambiguity. Finally,
34 BS-DISP intervals included the effects of both random errors and rotational ambiguity.
35 Tables S4–S5 show that the seven-factor solutions for WSON do not have significant

36 rotational ambiguity, and the base model and error estimates can be interpreted. With
37 the absence of any swaps, dQ_{\max} provides confidence that these solutions are well
38 constrained.

39 As shown in Figure S8, factor 1 was linked to high contribution of V, suggesting that it
40 came from a shipping emission source (Celo et al., 2015; Viana et al., 2009). Factor 2
41 was characterized by high levels of NH_4^+ and SO_4^{2-} , which was a secondary sulfate
42 source. Factor 3 (dust) showed a high contribution of Mn (69 %), Fe, Mg^{2+} , and Ca^{2+} .
43 Factor 4 was identified as SOA source because it was associated with high loadings
44 of 2-MGA, 2-MGL, MBTCA, and p-phthalic acid. Factor 5 was linked to biomass
45 burning, with high loadings of galactosan, mannosan, and levoglucosan, NH_4^+ , NO_3^- ,
46 and p-phthalic acid also had high loading for this source, as reported in the source
47 profiles of biomass burning (Li et al., 2007). Factor 6 showed high loadings with Pb,
48 Cu, and Zn and was identified as vehicle emissions and fossil fuel combustion. Factor
49 7 was associated with high levels of Na^+ , Cl^- , and Mg^{2+} , which was identified as sea
50 spray aerosols via the bubble-busting processes (Facchini et al., 2008).

51 **Text S2. Potential Source Contribution Function (PSCF) Model**

52 The PSCF model facilitates the identification of source region by partitioning the
53 potential source domain into a grid matrix of $i \times j$, as detailed in our previous study
54 (Geng et al., 2020; Tang et al., 2024). The PSCF analysis typically generates PSCF_{ij}
55 values ranging from 0 to 1, with elevated PSCF_{ij} suggesting an increased likelihood
56 that the ij_{th} cell being the source region. A notable constraint of PSCF-based
57 methodologies is the necessity to establish a weighing function to designed to diminish
58 the influences of cells with minimal residence time, a phenomenon often manifested
59 as “trailing effects” (Petit et al., 2017). To investigate the partitioning of aerosol sources
60 and to ascertain the contributions of various aerosol emission sources and
61 transportation mechanisms, we conducted a weighted Potential Source Contribution
62 Function (WPSCF) analysis. Given that PSCF represents a conditional probability, the
63 associated error escalates with increased distance between the grid and the sample
64 points (Tiwari et al., 2018). In this investigation, the experimental area was divided into
65 a $0.25^\circ \times 0.25^\circ$ grid. The threshold for the computation of m_{ij} was established at the 75th
66 percentile. To mitigate the impact of small n_{ij} values on PSCF, a weighting function was
67 applied (Tiwari et al., 2018):

$$68 \quad w_{ij} = \begin{cases} 1.00 & n_{ij} > 80 \\ 0.70 & 20 < n_{ij} \leq 80 \\ 0.42 & 10 < n_{ij} \leq 20 \\ 0.05 & n_{ij} \leq 10 \end{cases} \quad (3)$$

69 **Text S3. Predicting ALWC Using a Thermodynamic Model**

70 To predict the mass concentration of aerosol liquid water content (ALWC), we used a

71 thermodynamic model, ISORROPIA-II, <http://nenes.eas.gatech.edu/ISORROPIA/>,
72 which calculates the ALW concentration with particle-phase concentrations of Na^+ ,
73 SO_2^- , NH_4^+ , NO_3^- , Cl^- , Ca^{2+} , K^+ , and Mg^{2+} , as well as meteorological conditions (RH
74 and ambient temperature) as the input (Fountoukis and Nenes, 2007), which was
75 described in our previous studies (Xu et al., 2022). Briefly, the contributions of particle
76 water associated with organic fractions to ALW were estimated by calculating the
77 organic hygroscopicity parameter. ALW was calculated as the sum of water associated
78 with individual aerosol chemical components (i.e., the sum of ions and lumped organics)
79 based on the Zdanovskii–Stokes–Robinson (ZSR) relationship.

80

81 **Table S1.** The concentrations of TSP, OC, EC, WSON, NH₄⁺-N, NO₃⁻-N, and the ratios of OC/EC and WSON/WSON. In addition, the
82 meteorological data, including relative humidity (RH), wind speed (WS), and temperature (Temp.) were also presented. Precipitation and radiation
83 data are obtained from historical reanalysis datasets from European Centre for Medium-Range Weather Forecasts (ECMWF). Cluster indicates
84 the air masses analyzed by HYSPLIT in Figures 1b and S1–S2.

Date	TSP ($\mu\text{g m}^{-3}$)	OC ($\mu\text{g m}^{-3}$)	EC ($\mu\text{g m}^{-3}$)	OC/EC	WSON ($\mu\text{gN m}^{-3}$)	NH ₄ ⁺ -N ($\mu\text{gN m}^{-3}$)	NO ₃ ⁻ -N ($\mu\text{gN m}^{-3}$)	RH (%)	WS (mph)	Temp. (°C)	Precipitation (mm)	Radiation (W m ⁻²)	Cluster
2016/1/18	42	10	1.7	6.1	0.31	0.11	0.55	80	5.6	29	0.453	2845.62	1
2016/1/25	66	19	1.4	13	0.73	0.32	0.73	65	11	17	0.049	456.26	3
2016/2/1	97	18	2.4	7.7	1.2	0.34	1.6	75	4.3	30	0.264	4340.11	1
2016/2/8	161	38	2.8	14	1.1	2.2	2.7	43	7.8	21	0.000	8144.62	5
2016/2/15	88	15	1.9	8.0	1.3	0.28	1.1	76	5.8	30	0.115	3605.48	1
2016/2/22	60	11	1.8	5.9	0.89	0.58	1.3	74	6.7	29	0.017	3355.82	2
2016/2/29	118	23	1.9	12	1.3	1.1	0.69	48	5.0	29	0.000	4671.43	3
2016/3/4	101	13	1.6	8.1	1.3	0.40	0.75	71	6.8	30	0.004	4642.70	1
2016/3/8	47	8.7	1.3	6.7	0.35	0.51	0.98	74	9.2	30	0.094	4934.61	2
2016/3/10	37	9.7	1.6	6.0	0.50	0.34	0.58	78	7.5	30	0.000	4929.85	2
2016/3/14	33	8.0	1.3	6.2	0.49	0.10	0.45	73	7.8	31	0.000	5116.68	2
2016/3/16	40	9.9	1.4	7.3	0.41	0.22	0.56	69	9.7	31	0.005	5223.13	2
2016/3/18	59	10	1.4	7.5	0.71	0.52	0.50	74	9.4	31	0.001	5809.82	2
2016/3/20	51	8.0	1.0	7.9	0.60	0.48	0.44	74	11	31	0.000	5953.51	2
2016/3/22	80	13	1.6	8.1	0.91	0.75	0.34	74	9.0	31	0.000	7301.90	2
2016/3/24	111	15	1.7	8.9	0.65	1.26	0.55	64	7.3	32	0.000	6984.91	1
2016/3/26	54	11	1.2	9.3	0.80	0.71	0.78	59	8.3	30	0.184	3269.41	3

2016/3/28	111	24	2.1	12	1.2	2.10	2.7	68	7.1	29	0.016	2870.25	4
2016/3/30	57	11	1.4	7.6	0.84	0.54	0.94	77	6.2	30	1.276	3838.34	2
2016/4/1	43	8.3	1.3	6.2	0.54	0.39	0.82	71	8.7	31	0.278	4539.01	2
2016/4/3	18	6.1	0.83	7.3	0.57	0.35	0.39	68	8.7	31	0.012	4226.13	2
2016/4/5	28	6.9	1.1	6.2	0.64	0.13	0.39	69	8.8	32	0.168	5370.84	2
2016/4/7	35	8.7	1.2	7.2	0.57	0.31	0.32	72	10	32	0.064	3968.69	2
2016/4/9	25	5.4	0.95	5.7	0.53	0.09	0.24	67	10	32	0.012	5640.03	2
2016/4/13	63	8.7	1.3	6.6	0.59	1.8	0.28	55	9.8	33	0.000	6572.45	1
2016/4/15	37	7.1	0.95	7.5	0.41	1.1	0.12	65	9.8	33	0.224	5262.17	2
2016/4/17	32	7.9	0.16	49	0.27	1.1	0.10	71	9.7	33	0.544	5817.36	2
2016/4/19	26	6.2	1.1	5.9	0.41	0.57	0.12	68	9.7	32	0.448	5535.08	2
2016/4/21	34	6.6	0.91	7.2	0.57	0.07	0.41	63	9.9	32	0.372	4912.19	2
2016/4/23	38	8.4	0.99	8.5	0.92	0.08	0.35	65	10	33	0.311	5393.97	2
2016/4/25	40	9.8	1.1	8.6	0.98	0.35	0.26	69	9.7	33	0.218	5708.40	2
2016/4/27	47	11	1.2	8.5	1.2	0.53	0.44	69	9.7	33	0.438	4995.66	2
2016/4/29	29	7.1	1.0	7.1	0.84	0.36	0.16	72	9.1	33	0.152	4496.61	2
2016/5/1	20	5.4	0.96	5.6	0.84	0.23	0.20	74	6.4	30	0.580	4786.53	1
2016/5/3	23	5.7	1.3	4.3	0.77	0.31	0.18	63	8.5	33	0.443	5948.07	2
2016/5/5	21	5.4	0.91	5.9	0.88	0.09	BDL	66	7.9	33	0.003	5921.03	2
2016/5/9	20	3.7	0.63	5.9	0.66	0.06	0.20	61	10	34	0.058	4461.39	2
2016/5/11	36	8.7	1.0	8.5	1.1	0.19	0.34	66	9.5	34	0.647	5327.31	2
2016/5/15	44	11	1.1	9.5	1.1	0.50	0.53	73	7.7	31	2.188	3869.72	1
2016/5/17	41	9.9	1.1	8.8	1.2	0.67	0.64	76	6.8	30	2.089	2627.67	1
2016/5/19	36	7.7	1.1	7.0	0.93	0.17	0.29	66	7.9	32	0.792	3161.86	1
2016/5/21	25	5.8	0.94	6.2	0.92	0.02	0	65	9.3	32	1.402	2336.95	6

2016/5/23	39	7.1	0.93	7.6	0.98	0.39	0	63	10	31	12.116	2166.24	6
2016/5/25	39	8.1	1.4	5.7	0.69	0.11	0.31	58	11	32	1.066	1400.26	6
2016/5/27	39	6.6	1.0	6.6	0.78	0.16	0.39	62	8.3	32	1.198	2902.14	6
2016/5/29	46	6.0	0.88	6.8	0.89	0.23	0.52	61	8.8	32	2.901	902.48	1
2016/5/31	42	7.9	1.2	6.8	0.84	0.05	0.41	82	7.5	29	2.012	1927.43	1
2016/6/2	41	7.3	1.3	5.7	0.92	0.05	0.40	80	6.3	30	1.046	2302.63	2
2016/6/14	37	7.9	1.3	5.9	0.68	0.00	0.26	68	8.1	31	0.363	2907.11	6
2016/10/22	18	5.4	1.2	4.6	0.49	0.26	0.07	73	8.9	30	3.326	2077.57	6
2016/10/24	26	6.2	1.1	5.7	0.72	0.70	0.25	73	6.7	29	1.569	3366.86	1
2016/10/26	17	5.8	1.2	4.9	0.61	0.12	0.09	89	5.4	28	6.510	2059.29	4
2016/10/28	30	7.4	1.5	5.1	0.71	0.19	0.25	84	5.9	29	1.616	3684.34	4
2016/10/30	18	5.8	1.0	5.6	0.47	0.09	0.08	89	5.0	27	5.517	2707.15	5
2016/11/1	26	8.9	1.3	7.1	0.73	0.12	BDL	82	6.1	29	0.203	2615.73	3
2016/11/3	40	10	1.6	6.5	0.92	0.35	0.39	75	5.2	28	1.372	3752.32	3
2016/11/5	75	17	1.9	9.1	1.3	1.1	1.4	70	4.3	29	0.001	3672.66	4
2016/11/7	60	12	2.3	5.4	1.2	1.2	1.4	80	3.2	29	9.042	787.82	4
2016/11/9	43	10	1.7	6.1	1.2	0.62	0.98	89	5.3	28	6.884	317.76	4
2016/11/13	49	14	1.7	8.6	1.2	0.19	0.57	75	6.3	29	3.125	4701.36	5
2016/11/15	44	12	1.3	9.1	1.5	BDL	BDL	72	5.8	30	0.028	4410.31	5
2016/11/17	62	12	1.3	9.4	0.99	0.12	0.45	67	7.4	29	0.000	7075.15	5
2016/11/19	46	11	1.1	9.6	1.2	0.42	0.61	74	6.3	29	1.836	2255.39	4
2016/11/21	93	24	1.6	15	1.5	0.46	0.90	73	5.8	30	0.581	3279.00	4
2016/11/23	57	13	1.2	11	1.1	0.25	0.48	75	6.4	29	10.938	1452.02	4
2016/11/25	64	15	2.3	6.5	0.95	0.25	0.57	83	4.9	29	0.156	3046.12	5
2016/11/27	41	9.3	0.96	9.6	0.75	0.15	0.23	69	5.6	29	0.130	1766.73	5

2016/11/29	78	15	1.5	10	0.92	0.44	0.82	71	4.6	27	0.000	5736.78	5
2016/12/1	66	12	1.4	8.5	0.97	0.49	0.48	68	6.0	27	0.002	5181.22	5
2016/12/3	64	12	1.3	9.2	0.88	0.45	0.56	67	8.2	28	0.012	2908.52	5
2016/12/5	51	14	1.3	11	1.3	0.41	0.33	72	5.1	28	0.614	2746.71	5
2016/12/7	73	14	1.5	9.4	1.7	0.67	0.56	65	6.5	27	0.931	2198.62	5
2016/12/9	103	21	2.1	10	1.4	0.91	1.1	65	4.2	27	0.000	5908.49	5
2016/12/11	106	24	1.8	13	1.7	1.7	1.2	62	6.2	27	0.021	3175.09	4
2016/12/13	98	25	1.9	13	1.3	0.86	1.2	60	7.3	27	0.005	4615.21	4
2016/12/15	46	9.2	0.85	11	0.75	0.09	0.22	68	8.8	28	5.802	1251.42	3
2016/12/17	81	17	1.3	13	1.5	0.43	0.93	62	7.5	25	0.179	929.87	5
2016/12/20	124	29	2.0	15	2.1	0.75	2.1	65	5.8	29	0.003	3904.70	5
2016/12/27	50	13	0.92	14	1.0	0.17	0.40	58	5.9	29	0.000	5351.00	3
2017/1/3	57	12	1.3	9.8	1.1	0.20	0.54	70	7.6	27	0.114	2825.05	3
2017/1/7	110	25	2.0	12	2.3	0.59	1.3	71	7.0	28	2.041	330.58	4
2017/1/14	90	20	1.9	10	1.7	0.50	0.80	75	4.7	29	0.599	3377.36	5
2017/1/21	66	14	1.1	13	1.0	0.30	0.87	62	6.1	29	0.000	3822.33	3
2017/1/28	98	22	1.8	12	1.6	0.50	1.2	55	5.1	27	0.000	5900.06	4

85 BDL: below detection limits.

86

87

88 **Table S2.** The concentration of TSP, carbonaceous aerosol, WSON, WSIN, and
 89 molecular markers in marine-, mixed-, and continental-influenced days.

	Marine (n = 36)		Mixed (n = 21)		Continental (n = 27)		All (n = 84)	
	Ave	SD	Ave	SD	Ave	SD	Ave	SD
TSP ($\mu\text{g m}^{-3}$)	42	21	54	24	73	34	55	30
OC ($\mu\text{g m}^{-3}$)	8.4	2.4	11	4.7	17	7.7	12	6.3
EC ($\mu\text{g m}^{-3}$)	1.2	0.32	1.4	0.48	1.6	0.40	1.4	0.43
WSOC ($\mu\text{g m}^{-3}$)	4.6	1.8	5.2	2.5	8.5	4.7	6.0	3.6
WSON ($\mu\text{gN m}^{-3}$)	0.72	0.25	1.0	0.34	1.2	0.43	0.95	0.40
NH ₄ ⁺ -N ($\mu\text{gN m}^{-3}$)	0.42	0.38	0.27	0.25	0.68	0.56	0.47	0.44
NO ₃ ⁻ -N ($\mu\text{gN m}^{-3}$)	0.40	0.28	0.53	0.39	0.93	0.68	0.60	0.52
Na ⁺ ($\mu\text{g m}^{-3}$)	1.4	0.74	0.57	0.46	0.34	0.28	0.86	0.74
Cl ⁻ ($\mu\text{g m}^{-3}$)	0.27	0.16	0.20	0.14	0.21	0.10	0.23	0.14
*nss-K ⁺	0.49	0.16	0.49	0.25	0.77	0.45	0.58	0.33
*nss-SO ₄ ²⁻	5.5	2.9	2.9	2.0	4.0	2.6	4.0	2.6
Na ⁺ / Σ ions	0.11	0.040	0.060	0.038	0.028	0.023	0.028	0.023
OC/EC	8.1	7.2	8.4	2.4	10.1	3.0	8.9	5.2
WSOC/OC	0.54	0.11	0.44	0.08	0.49	0.08	0.50	0.10
WSON/WSOC	0.17	0.070	0.22	0.070	0.17	0.069	0.18	0.07
Molecular tracers (ng m^{-3})								
Levoglucosan	70	54	140	112	337	282	174	207
Galactosan	2.3	1.4	4.9	3.6	11	8.2	5.6	6.17
Mannosan	5.8	3.8	9.2	6.3	17	12	10.3	9.29
Hopanes and steranes	1.2	0.4	1.3	0.7	1.3	0.7	1.3	0.60
2-MGA	3.2	1.5	5.0	2.6	6.6	4.1	4.8	3.2
2-MGL	36	22	29	17	38	23	35	21
MBTCA	11	7.2	31	22	29	12	22	17
o-Phthalic acid	7.9	4.4	4.1	3.4	7.5	7.3	6.8	5.5
p-Phthalic acid	10	7.9	31	25	43	26	26	25

90 nss: non-sea-salt

91

92 **Table S3.** The monthly averaged WSON concentration in the selected coastal cities
 93 that are impacted by marine air masses on a global scale and the corresponding AEC
 94 (air mass exposure to Chl a). The Chl a concentration was from the MODIS-Aqua
 95 (EARTHDATA, <https://giovanni.gsfc.nasa.gov/giovanni/>).

Sites	Data	Monthly averaged WSON (nmol m ⁻³)	AEC (mg m ⁻³)	References
Keelung City	2006/07	27.2	0.136	Chen and Chen (2010)
	2006/08	22.9	0.190	
South China Sea	2005/04	65	0.163	Shi et al. (2010)
Jiaozhou Bay, China	2015/07	190	0.860	Xing et al. (2018)
Hong Kong	2011/08	95	0.524	Ho et al. (2019)
	2011/09	95	0.319	
Hong Kong	2017/01	130.5	0.391	Leung et al. (2024)
	2017/03	114.7	0.306	
	2017/04	119.8	0.151	
	2017/05	117.0	0.230	
	2017/06	82.6	0.181	
	2017/07	114.8	0.250	
	2017/08	90.6	0.499	
	2017/09	93.3	0.445	
Eastern Mediterranean	2011/01	11.6	0.081	Tsagkaraki et al. (2021)
	2011/02	11.1	0.081	
	2011/04	9.45	0.118	
	2011/05	12.3	0.085	
	2011/06	22.2	0.071	
	2011/07	14.7	0.028	
	2011/08	16.4	0.082	
	2011/09	13.9	0.097	
	2011/10	8.12	0.073	
	2011/11	11.7	0.062	
Okinawa Island	2010/06	34.6	0.068	Kunwar and Kawamura (2014)
	2010/07	2.88	0.030	
	2010/08	2.88	0.035	
	2010/09	5.77	0.048	

	2010/10	4.32	0.141	
Barbados	2008/08	3.5	0.038	Zamora et al. (2011)
Huaniao island	2019/07	231.30	0.505	Tian et al. (2023)
	2019/08	211.56	0.474	
Sapporo	2010/06	6.56	0.159	Pavuluri et al. (2015)
	2010/07	4.63	0.135	
	2010/08	1.54	0.125	
Bangkok, Thailand	2016/01	36.96	0.197	This study
	2016/02	81.96	0.377	
	2016/03	52.03	0.199	
	2016/04	46.30	0.238	
	2016/05	63.93	0.311	
	2016/06	57.39	0.321	

96

97

98 **Table S4.** The mapping of bootstrap (BS) factors and Fpeak BS factors to base factors
 99 for seven-factor solutions. (Number of bootstrap runs: 100)

		Factor 1	Factor 2	Factor 3	Factor 4	Factor 5	Factor 6	Factor 7	Unmapped
BS Mapping	Factor 1	100	0	0	0	0	0	0	0
	Factor 2	0	91	0	3	2	4	0	0
	Factor 3	0	1	96	1	0	1	1	0
	Factor 4	0	1	0	99	0	0	0	0
	Factor 5	0	0	2	0	98	0	0	0
	Factor 6	0	0	0	1	0	99	0	0
	Factor 7	1	0	1	2	1	0	95	0
Fpeak BC Mapping (Fpeak value - 0.5)	Factor 1	100	0	0	0	0	0	0	0
	Factor 2	0	92	2	3	1	2	0	0
	Factor 3	0	0	96	3	1	0	0	0
	Factor 4	0	0	1	97	2	0	0	0
	Factor 5	0	0	1	2	96	1	0	0
	Factor 6	0	0	0	2	0	98	0	0
	Factor 7	1	1	1	3	0	1	93	0

100

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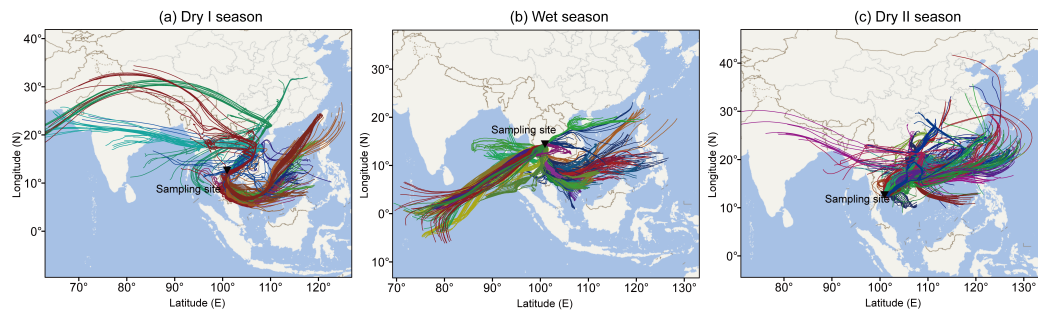
103

104 **Table S5.** Displacement (DISP) diagnostics of seven-factor solutions. Bootstrapping
 105 with displacement (BS-DISP) diagnostics of seven-factor solutions.

Constrained DISP Diagnostics:							
Error Code:	0						
%dQ:	-0.016758291						
Swaps by Factor:	0	0	0	0	0	0	0
Constrained BS-DISP Diagnostics:							
# of Cases:	81						
%dQ:	-0.106123702						
Swaps by Factor:	0	5	1	2	4	3	3

106

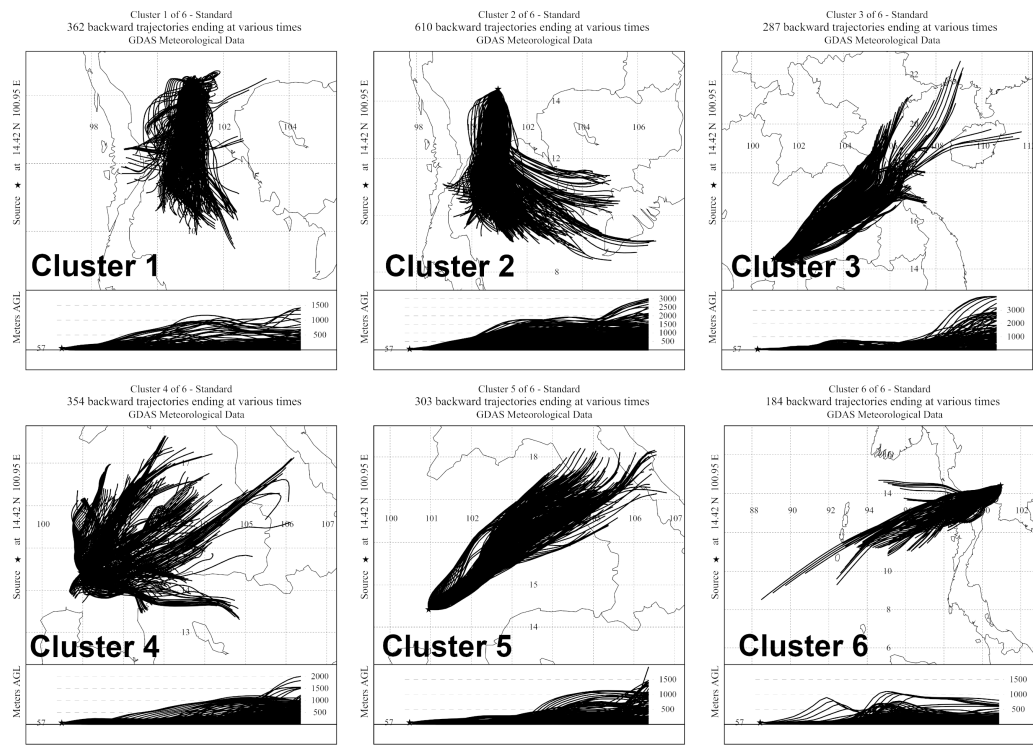
107



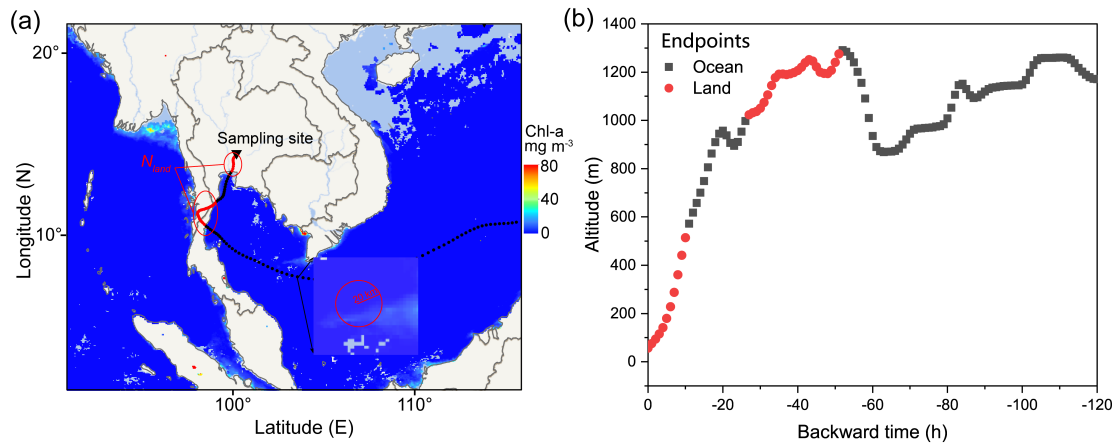
108

109 **Figure S1.** The 120 h back air-mass trajectories at Bangkok, Thailand, in the Dry I (a),
 110 Wet (b), and Dry II seasons (c). The air mass trajectories were analyzed using the
 111 HYSPLIT model. The map was created using ArcGIS software with the basemap from
 112 the National Platform for Common Geospatial Information Services. Map source and
 113 attribution: National Platform for Common Geospatial Information Services
 114 (www.webmap.cn).

115



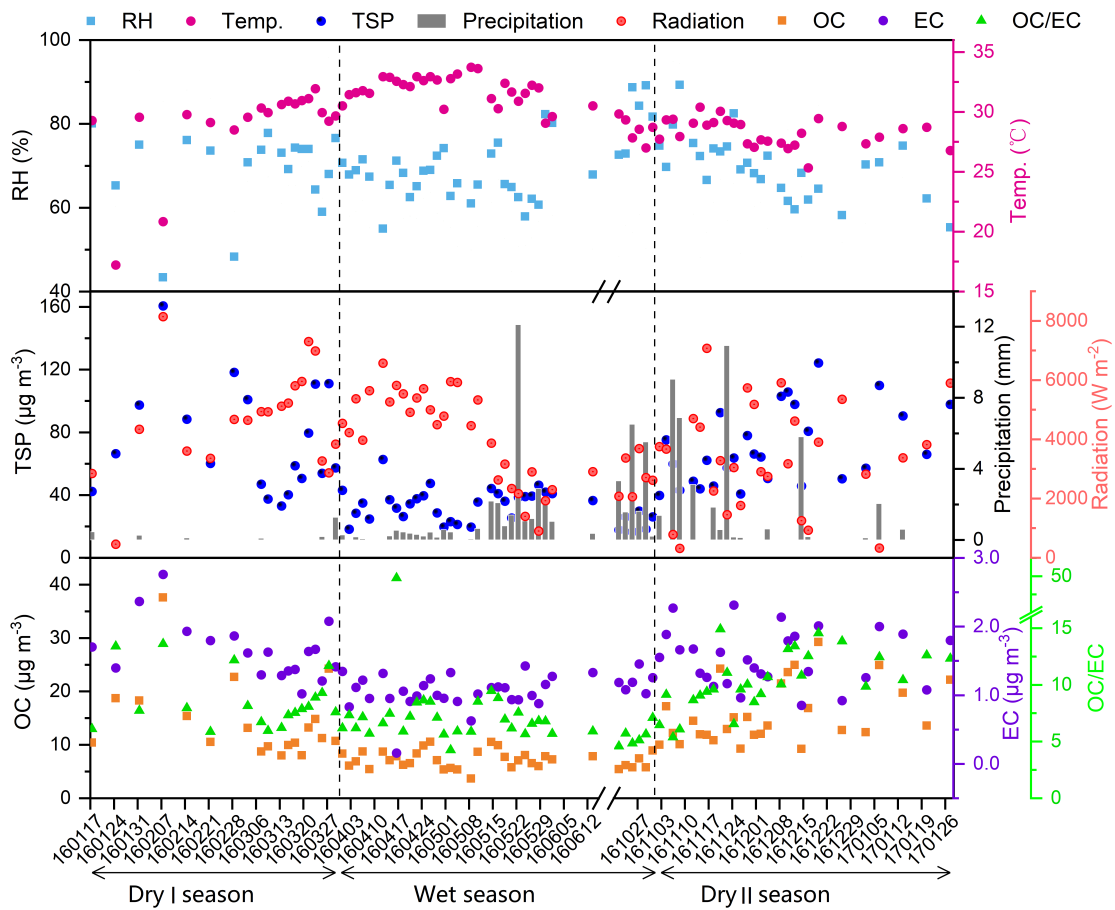
116
 117 **Figure S2.** The 120 h back air-mass trajectories at Bangkok, from Thailand, during
 118 January, 2016 to January, 2017. The backward trajectories were clustered into six
 119 groups. The air mass trajectories were analyzed using the HYSPLIT model.
 120



121

122 **Figure S3.** Schematic diagram of the calculations of (a) R_{land} and AEC, and (b) altitude
 123 of each air-mass endpoints using 120-hour trajectories. The red points refer to the air-
 124 mass endpoints along trajectories located over the land. This 20 km radius was
 125 representative of the area used as the search radius to compute the mean Chl-a
 126 concentration at the air-mass endpoints.

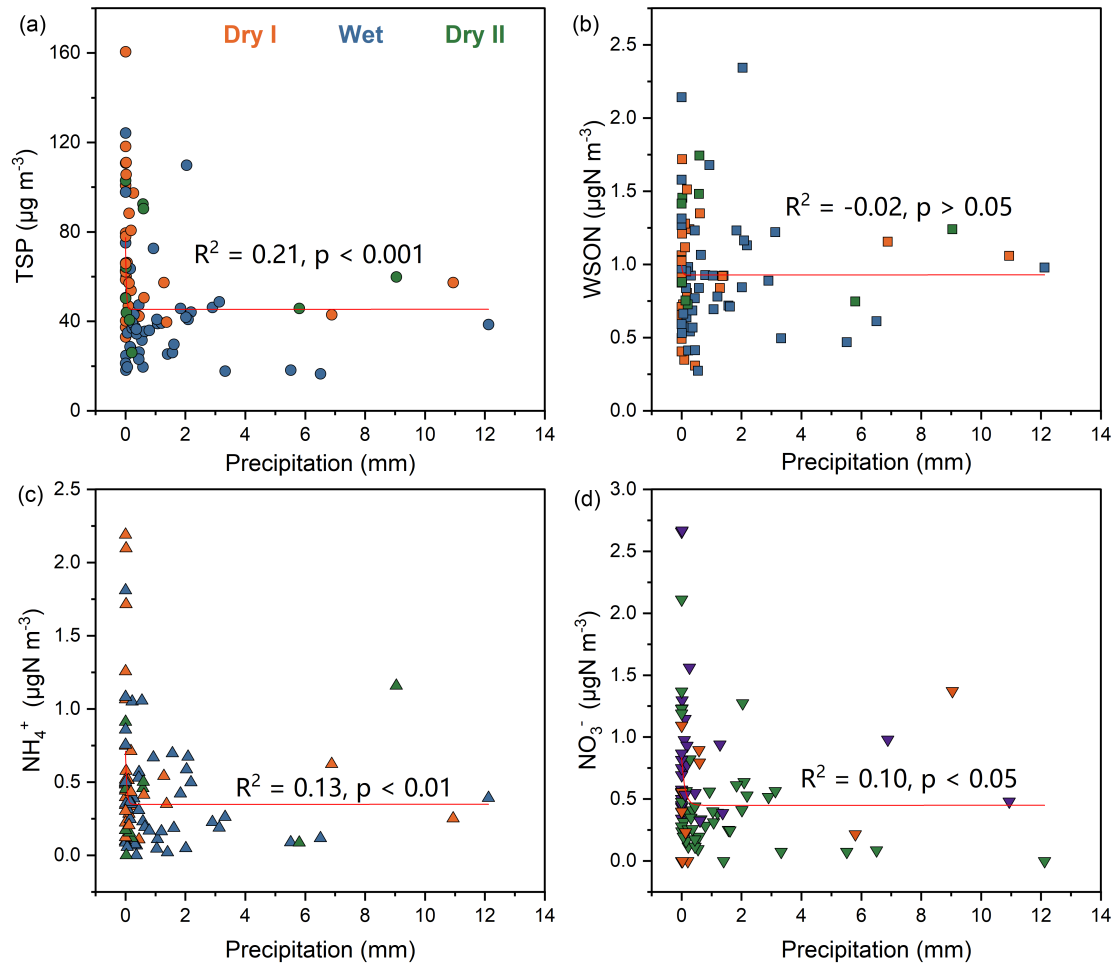
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129 **Figure S4.** Meteorological data including relative humidity (RH), temperature (Temp.),
 130 precipitation, and radiation, along with time-series of TSP, OC, EC, and the ratio of
 131 OC/EC.

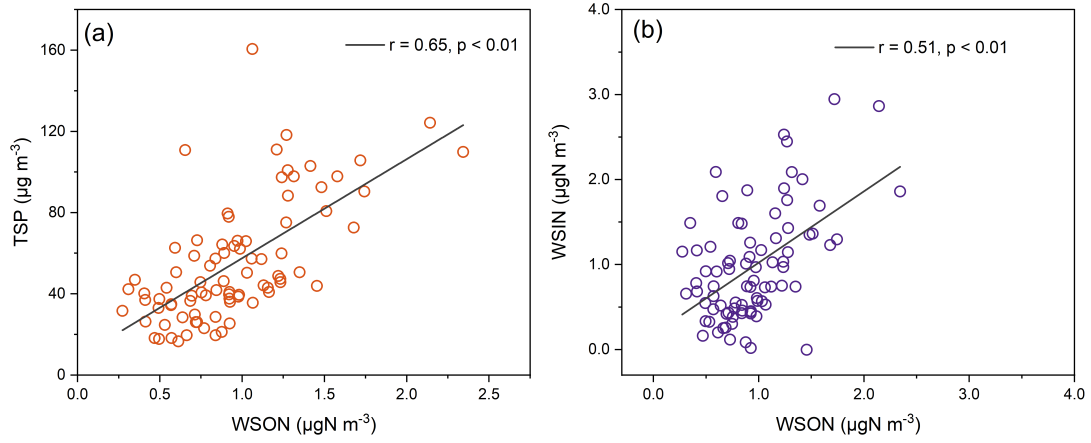
132



133

134 **Figure S5.** Exponential relationships between TSP (a), WSON (b), WSIN (c, d) and
 135 precipitation.

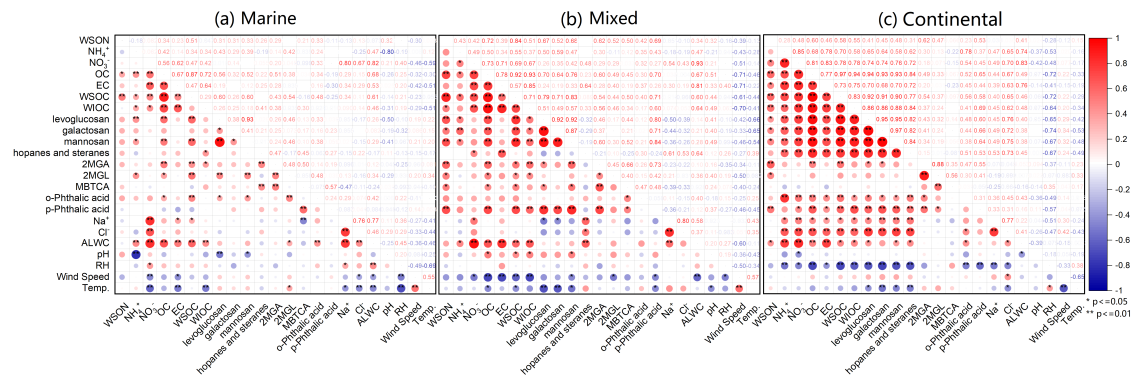
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138 **Figure S6.** Linear relations of WSON to WSIN (sum of NO_3^- -N and NH_4^+ -N).

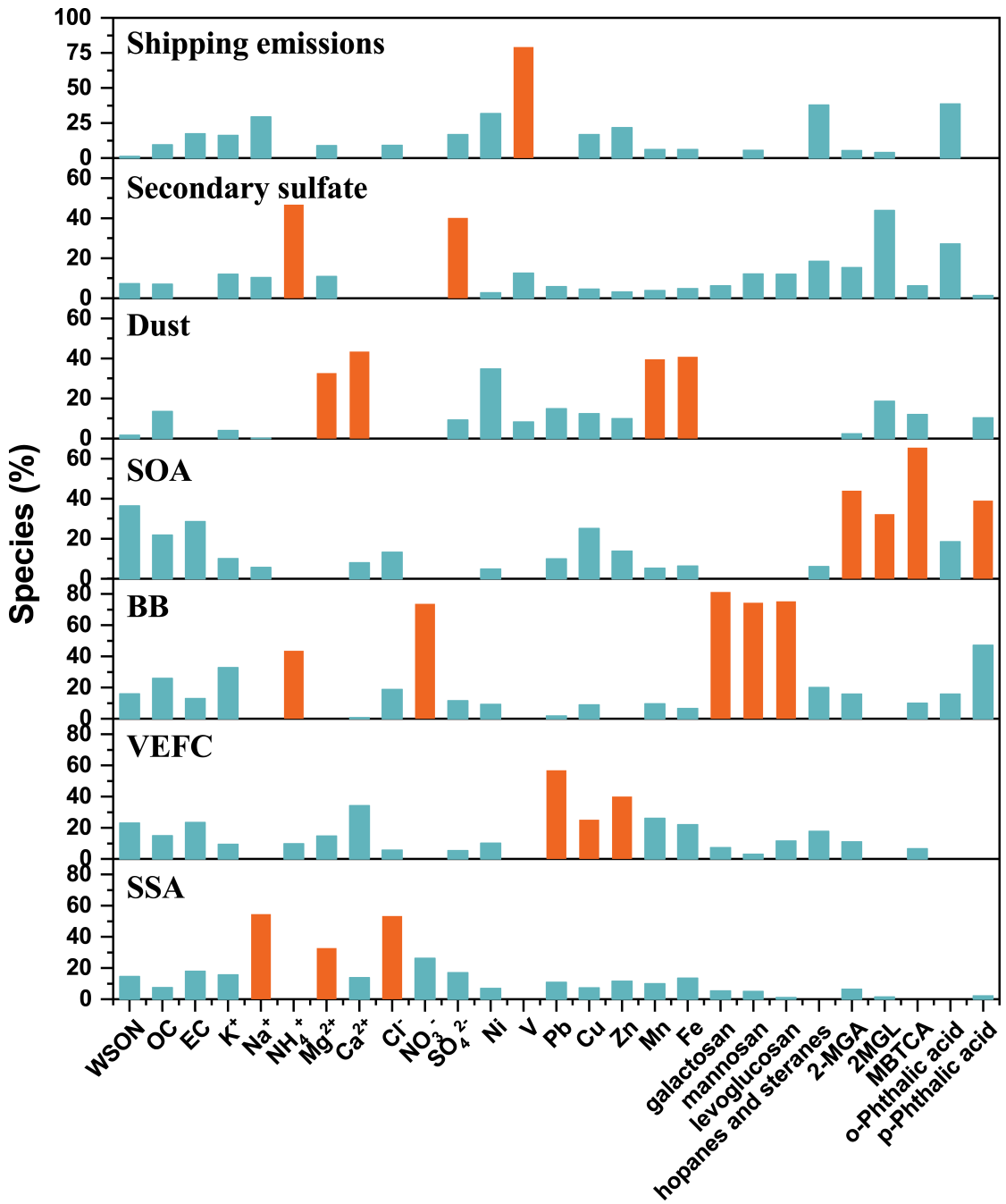
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141 **Figure S7.** The correlation hotspots in marine- (a), mixed- (b), and continental-
 142 influenced days (c).

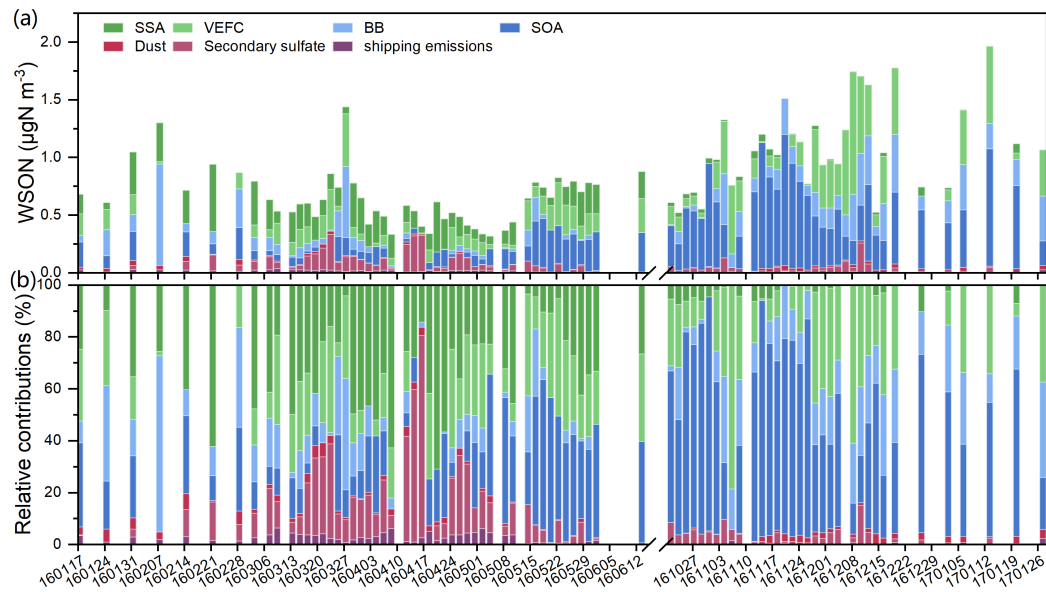
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145 **Figure S8.** Seven factor profiles derived from the PMF solution. The percentage of
 146 chemical species in each factor is shown.

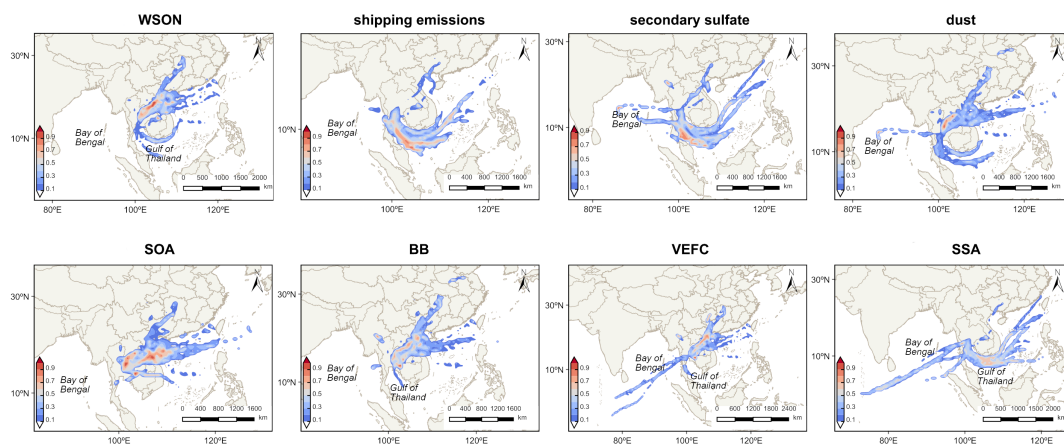
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148

149 **Figure S9.** The time series of the seven PMF source factors and their relative
 150 contributions to WSON (a, b).

151



152

153 **Figure S10.** PSCF results of WSON and the seven-factor derived WSON resolved by
 154 PMF model. Threshold values were set as the 75th percentile of the corresponding
 155 values. The 120-hr backward air mass trajectory was analyzed using the HYSPLIT
 156 model.

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