Supplemental Information

Roadside air pollution and secondary organic aerosol seasonal trends from an oxidation flow reactor in Seoul

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Table. S1 Seasonal roadside information during measurement

	Winter	Spring	Summer	Fall
Period	1/28/2019— 2/08/2019 *Korean New Year (KNY): 2/02/2019 00— 2/07 00	04/15/2019– 27/04/2019	07/09/2019– 07/16/2019	09/09/2019– 09/19/2019 *Korean Thanksgiving Day (KTD): 9/12/2019 00 – 9/16/2019 00
T (°C)	-3.9 ± 4.9	14.1 ± 4.7	24.8 ± 3.3	23 ± 3.3
RH (%)	47.3 ± 17	67.5 ± 21.6	67.7 ±15.7	64.3 ± 20.2
Traffic Volume (# day ⁻¹) ^a	67,279 ± 6,683	72,966 ± 1,975	70,028 ± 3,232	66,949 ± 5,554
% of gasoline b	54% 36,901 ± 4,480	46% 34,885± 1,137	43% 31,950 ± 1,736	46% 32,033 ± 1,254
% of diesel ^c	36% 25,868 ± 7,685	45% 33,212 ± 2,572	46% 32,976 ± 3,714	45% 30,471 ± 5,491
% of LPG ^d	$10\% \\ 4,510 \pm 1,234$	9% 4,868 ± 563	11% $5,102 \pm 486$	9% 4,446 ± 382
Traffic Speed (km hour ⁻¹) ^e	69.2 ± 11.9	69.3 ± 1.4	64.8 ± 6.9	63.9 ± 7.2

^{a,e} The total volume and average speed of traffic were measured by HD Digital Wave Radar(smart Sensor HD Model 126, Wavetronix, Provo, UT, USA), ^{b-d} the video watch analysis was applied to 6 min in an hour day the interval of ten minutes and calculated them using the statistical extraction method with traffic volume (# hour⁻¹). This approach was followed by Park et al. (2021a), who conducted the tunnel project in the same location with the same traffic instrument. ^{d)} for taxi vehicles.

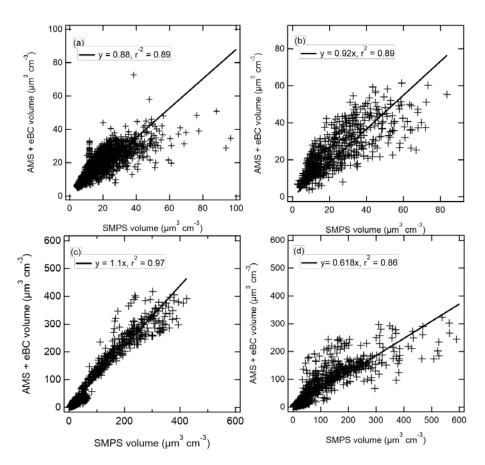


Fig. S1. Scatter plot of roadside and OFR aerosol volume result from AMS plus eBC vs. SMPS with regression line. (a), (b) and (C) were applied to the default CE = 1. (d) was calculated using composition-dependent collection efficiency (CDCE) as Middlebrook et al. (2012). Aerosol densities of 1.75 g cm⁻³ for nitrate, sulfate and ammonium, 1.52 g cm⁻³ for chloride, 1.77 g cm⁻³ for eBC (Ahlberg et al., 2019; Park et al., 2004; Palm et al., 2016). A density of organic (unit: cm⁻³) was estimated based on the improved elemental composition of [12 + 1*(H/C) + 16*(O/C)]/[7+5*(H/C) + 4.15*(O/C)] (Canagaratna et al., 2015; Kuwata et al., 2012).

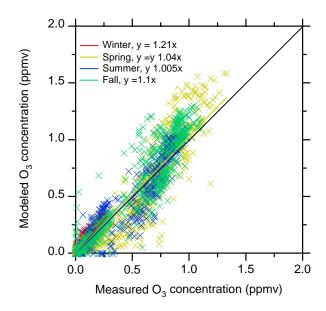


Fig. S2. Comparison of modeled O_3 concentration from photochemical model and measured O_3 concentration after OFR in seasons.

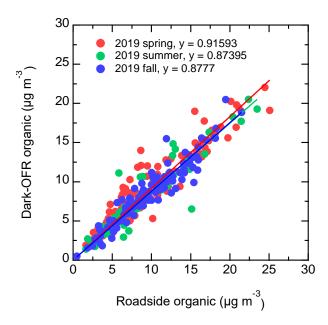


Fig. S3. Relationship between the organic after OFR inside 0% lamp intensity and roadside organic. The winter season wasn't conducted of dark-OFR, so we applied the average value 0.88919 of other seasons.

Section S1. Morning hour SOA estimation (Fig. S4, Equation S1–S3, Table S2–4)

The purpose of this method is to investigate the seasonal trend in the contribution levels of major component groups among the secondary organic aerosols (SOA) formed in the morning hours (9:30–11:30) on the roadside. Since the concentration of anthropogenic volatile organic compounds (AVOCs) was not measured in this study, the results of the tunnel project conducted previously in the same location were utilized. Since the measurements of the tunnel project were conducted in winter (January) and spring (April) at the same time as this project, the data show consistency. Likewise, for measurements in the fall (September), the SOA components were measured using a proton transfer reaction mass spectrometry (PTR-MS, Iconic), so the measurement hours are the same. However, in the case of summer (July), data from the summer (July) 2018 was used, and data were matched only for overlapping days of the week (Table S2). Fig. S4 shows the pattern of traffic volume during the study period. The average traffic volume from July 28 to August 6, 2018 was $70,328 \pm 856$ (vehicles day⁻¹), which does not show a significant difference from $70,025 \pm 3,232$ (vehicles day⁻¹), the traffic volume during the summer measurement period of this study. Thus, it is thought that there was no significant change in the pollutants from motor vehicle emissions.

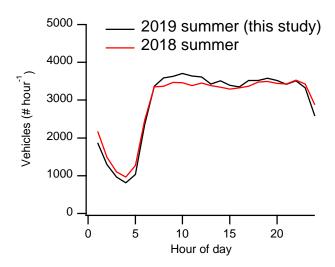


Fig. S4. Hourly pattern of average traffic volume (# hour⁻¹) in 2019 and 2018 summer projects. The traffic volumes were measured by HD Digital Wave Radar (smart Sensor HD Model 126, Wavetronix, Provo, UT, USA).

Table. S2. The result of anthropogenic volatile organic compounds (AVOCs): winter, spring, and fall were the same period, and the summer result was part of the 2018 summer project.

Unit: $\mu g m^{-3}$

Season	Benzene	Toluene	Ethylbenzene	<i>m,p</i> -xylene	o-xylene
Winter	4.3 ± 1.2	10.3 ± 8.2	1.4 ± 1-1	1.9 ± 1.3	0.7 ±0.5
Spring	3.1 ± 0.8	20.6 ± 11.9	5.7 ± 3.7	5.8 ± 3.5	2 1.2
Summer 2018	3.7 ± 2.0	16.3 ± 8.4	6.8 ± 3.7	5.9 ±2.9	4.8 ± 2.2
Fall	2.3 ± 0.7	2.1± 1.1		$1.8 \pm 0.9^{a)}$	

a) The sum of ethylbenzene and *m,p,o*-xylene

Biogenic VOCs (BVOCs) have been reported to be closely associated with temperature and photosynthetically active radiation (PAR) (Alves et al., 2016; Seco et al., 2015; Warneke et al., 2010). Therefore, in this study, referring to the method of Saha et al. (2018), using the result values of BVOCs measured using PTR-MS during the fall season, the formula proposed by Guenther et al. (1995) was applied to derive the seasonal concentration of BVOCs with respect to temperature and PAR. In the equations S1–S3, c_T is a function of temperature, and c_L is a function of PAR, and since radiation is not measured in this study, the value of PAR was assumed to be 1.

Real BVOCs_iconc = base conc (fall)
$$\times$$
 $c_T \times c_L$ (S1)
Isoprene

$$c_T = \frac{e^{(37.711 - 0.398570815 \times dT)}}{1 + e^{(91.301 - dT)}}, \qquad dT = \frac{28668.514}{K}$$
 (S2)

Monoterpenes, sesquiterpene

$$c_T = e^{(0.09 \times (K-303))}$$
 (S3)

Among the roadside organic, for calculation of semi-volatile organic compounds (SVOCs) and intermediate organic compounds (IVOCs) production, hydrocarbon-like organic aerosol (HOA) factor of the positive matrix factorization (PMF) approach was used (Paatero and Tapper, 1994). The PMF in this study was applied the Multilinear Engine 2 solver (ME-2, Paatero, 2000) controlled within the Source Finder software (SoFi Pro 8.0.4) toolkit for Igor pro 8 (Canonaco et al., 2013, 2015, 2021; Crippa et al., 2014; Paatero, 2000). However, in this study, because source groups other than HOA were not considered, PMF was run through a comprehensive unconstrained run under the condition of a-value 0. The HOA profile extracted

in this way was compared with representative tracer ions, gases, and eBC in terms of the Pearson correlation coefficient (r). In addition, the HOA profile was also compared with the HOA profile of Crippa et al. (2013) provided from the high resolution AMS spectral database for final decision.

HOA is related to fossil fuel emissions such as motor vehicles, and typical mass-to-charge ratios (m/z) in AMS database are 41, 43, 55, and 57, showing strong signals for $C_3H_5^+$, $C_3H_7^+$, $C_4H_7^+$, and $C_4H_9^+$ (Crippa et al., 2013; Kim et al., 2018; Sun et al., 2011; Zhang et al., 2005). In addition, HOA shows a high correlation with carbon monoxide (CO), nitrogen oxide (NOx), and equivalent black carbon (eBC), which are representative products of combustion (Lee et al., 2017; Li et al., 2019; Shah et al., 2018). Table S3 shows the results of comparing Pearson's r between the components in this study and unconstrained HOA factor, and for AMS tracer ion, the Pearson's r was 0.5–0.99, confirming a high correlation. Also, the Pearson's r for CO, eBC, and NO $_x$ was 0.34 to 0.78, thus showing a high correlation with the major components of motor vehicle emissions. In the case of NH $_3$, a correlation of 0.16–0.67 was derived due to the short distance between the sample inlet and the lane and NH $_3$ emissions from gasoline and LPG vehicles.

Table. S3. Correlation coefficients (Pearson's r) of HOA factor in the roadside of four seasons with AMS tracer ions, gases, particle, and reference profile of HOA.

Compounds	Winter HOA	Spring HOA	Summer HOA	Fall HOA
$C_3H_5(m/z 41)$	0.9	0.62	0.5	0.8
$C3H_7(m/z 43)$	0.98	0.92	0.89	0.97
$C_4H_7(m/z 55)$	0.97	0.83	0.77	0.94
$C_4H_9(m/z 57)$	0.99	0.97	0.94	0.99
eBC	0.62	0.39	0.36	0.34
NO_x	0.67	0.78	0.82	0.53
CO	0.42	0.58	0.57	0.41
NH ₃	0.16	0.67	0.58	0.42
HOA profile ^{a)}	0.95	0.92	0.93	0.93

a) This profile was Crippa et al. (2013)'s result that uploaded to the AMS spectral database(http://cires1.colorado.edu/jimenez-group/HRAMSsd/)

For the final calculation of SOA estimation, Equation 1 in the manuscript was used, and the OH rate constant for each required component and the SOA yield under high NO_x conditions are summarized in Table S4.

Table. S4. Lists of organic groups OH rate constant (25 °C, molecules cm⁻³) and SOA yield under high NO_x condition.

Compound	OH rate constant (cm ³ molec. ⁻¹ s ⁻¹)	Reference SOA Yield		Reference			
AVOCs	AVOCs						
Benzene	1.2×10^{-12}	Atkinson et al. (2006)	0.281	Ng et al. (2007a)			
Toluene	6.05×10^{-12}	Atkinson et al. (2006)	0.08	Ng et al. (2007b)			
Ethylbenzene	7 × 10 ⁻¹²	Atkinson and Arey (2003)	0.072	Yao et al. (2016)			
m+p-xylene	1.87×10^{-11}	Atkinson and Arey (2003)	0.08	Ng et al. (2007a)			
o-xylene	1.36 × 10 ⁻¹¹	Atkinson and Arey (2003)	1.37E-4	Zhang et al. (2020)			
BVOCs							
Isoprene	1×10^{-10}	Atkinson (1997)	0.02	Yao et al. (2016)			
Monoterpenes (α-, β-pinene)	6.63 × 10 ⁻¹¹	Atkinson (1997)	0.11	-α -pinene Ng et al. (2007b) β-pinene Sarrafzadeh et al. (2016)			
Sesquiterpene (longifolene)	4.8×10^{-11}	Atkinson and Arey (2003)	0.89	Ng et al. (2007b)			
S/IVOCs vehicl	les exhaust						
Gasoline	2.0×10^{-11}	Nault et al. (2018)	0.24	Shah et al. (2020)			
Diesel	2.0×10^{-11}	Nault et al. (2018)	0.38	Shah et al. (2020)			

Table. S5. Registered vehicles in South Korea in 2019. This result was in supporting information of Park et al. (2021b)

No.	Evol	South K	orea	Seoul		
110.	Fuel	Count	Portion	Count	Portion	
1	Gasoline	21,921,558	46.3%	1,597,621	51.1%	
2	Diesel	19,915,086	42.1%	1,140,532	36.5%	
3	Liquefied petroleum gas (LPG)	4,009,460	8.5%	279,713	9.0%	
4	Kerosene	2	0.0%	-	-	
5	Electric	179,836	0.4%	10,175	0.3%	
6	Alcohol	2	0.0%	-	-	
7	Compressed natural gas (CNG)	76,294	0.2%	9,607	0.3%	
8	Liquefied natural Gas (LNG)	10	0.0%	-	-	
9	Hybrid (gasoline + electric)	976,256	2.1%	78,778	2.5%	
10	Hybrid (diesel + electric)	1,204	0.0%	31	0.0%	
11	Hybrid (LPG + electric)	33,980	0.1%	1,939	0.1%	
12	Hybrid (CNG + electric)	654	0.0%	51	0.0%	
13	Hybrid (LNG + electric)	-	0.0%	-	-	
14	Hydrogen	10,166	0.0%	143	0.0%	
15	Others	230,224	0.5%	5,699	0.2%	
	Total	47,354,732	100%	3,124,289	100%	

Table. S6. Average ($\pm 1\sigma$ standard deviation) concentration of gaseous and aerosols on the roadside with comparison data of the monitoring station from the distance of 3.2 km in this site.

	Winter	Spring	Summer	Fall
CO (ppm)	1.49 (± 0.36)	1.83 (±0.63)	$1.26 (\pm 0.6)$	1.11 (± 0.38)
CO ₂ (ppm)	589 (± 112)	640 (± 143)	605 (± 143)	569 (±113)
NH ₃ (ppb)	81.5 (± 28.4)	141.7 (± 50)	94.3 (± 41.3)	62.4 (± 27)
NO _x (ppb)	686 (± 442)	798 (± 499)	407 (± 325)	308 (± 267)
NO / NOx	0.7	0.8	0.7	0.8
SO ₂ (ppb)	7.39 (± 5.66)	9.66 (± 5.71)	6.61 (± 1.63)	1.41 (± 1.04)
O ₃ (ppb)	4.35 (± 4.84)	3.45 (± 2.8)	6.26 (± 3.82)	9.63 (± 4.02)
PM _{AMS+eBC} (μg m ⁻³)	30.4 (± 13.1)	38.0 (± 19.8)	31.1 (± 18)	25.3 (± 12.3)
eBC	9.0 (± 6.43)	18.7 (± 12.31)	11.6 (± 8.23)	10.0 (± 7.83)
Organic	10.9 (± 4.8)	9.1 (± 5.2)	$8.7(\pm 4.9)$	10.1 (± 4.8)
Nitrate	4.7 (± 3.4)	4.4 (± 3.9)	3.4 (± 3.7)	1.4 (± 2.3)
Sulfate	2.9 (± 2.0)	3.3 (± 2.0	5.0 (± 4.2)	2.5 (± 1.7)
Ammonium	2.5 (± 1.5)	2.4 (± 1.6)	2.4 (± 2.1)	1.2 (± 1.1)
Chloride	$0.5 (\pm 0.3)$	$0.2 (\pm 0.2)$	0.1 (± 0.1)	0.1 (± 0.1)
Near monitoring sta	ation far from 3.2	km classified urba	nn air pollutants in	Seoul
$PM_{10} (\mu g m^{-3})$	59.4	41.3	7.7 ^{a)}	8.1
$PM_{2.5} (\mu g m^{-3})$	24.9	19.5	8.1 b)	11.3
O ₃ (ppb)	16.8	30.1	35.2 °)	27.7
NO ₂ (ppb)	35.5	35.6	9.3 ^{d)}	16.8
CO (ppm)	0.6	0.5	0.31 ^{e)}	0.43
SO ₂ (ppb)	3.9	3.2	4 ^{f)}	4

a-f) There were loss data period of 7/11/2019 12:00 to 7/16/2019 17:00 in summer. So these data were not comparable with ours result.

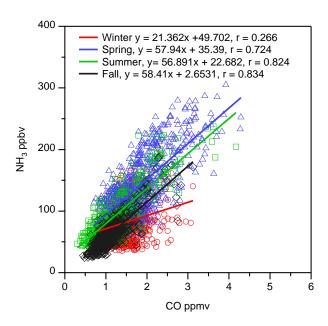


Fig. S5. The Pearson's correlation coefficients (r) between NH₃ (ppb) and CO (ppm) during real-time measurement period.

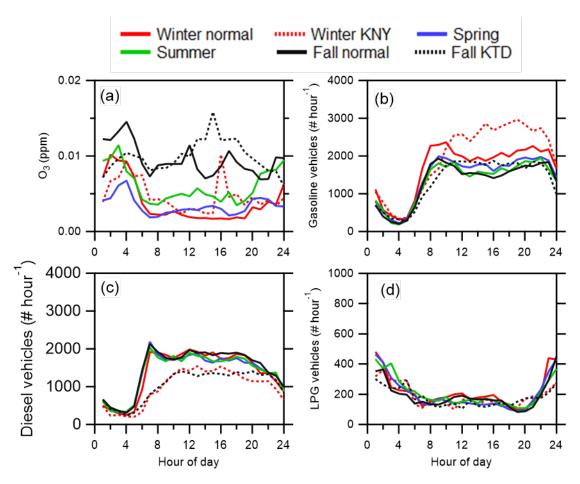


Fig. S6. Diurnal pattern of (a) O₃ and (b–c) average traffic volume of each fuel type (gasoline, diesel, and LPG) in seasons.

Table. S7. Comparison result (\pm 1 σ standard deviation) of normal days and holidays in Korea on the roadside.

	Wi	nter	Fall		
compound	KNY	Winter-normal	KTD	Fall-normal	
$AMS + eBC$ $(\mu g m^{-3})$	32.16 (± 11.58)	28.95 (± 14.1)	23.64 (± 10.14)	26.54 (± 13.48)	
eBC	$8.2~(\pm~5.34)$	9.71 (± 7.17)	6.53 (± 3.62)	12.58 (± 9)	
Organic	11.38 (± 4.81)	10.52 (± 4.83)	10.81 (± 4.4)	9.62 (± 5.01)	
Nitrate	5.46 (± 3.27)	3.99 (± 3.27)	2.25 (± 3.25)	$0.8 (\pm 0.78)$	
Sulfate	3.63 (± 2)	2.24 (± 1.68)	2.57 (± 1.74)	2.51 (± 1.61)	
Ammonium	2.96 (± 1.55)	2.06 (± 1.4)	1.41 (± 1.42)	$0.98 (\pm 0.67)$	
Chloride	$0.53 (\pm 0.34)$	0.43 (± 0.28)	$0.08~(\pm~0.09)$	$0.05 (\pm 0.12)$	
O ₃ (ppb)	4.8 (± 6)	4 (± 3.4)	9.7 (± 3.5)	9.5 (± 4.4)	
CO (ppm)	$1.47 (\pm 0.33)$	$1.51 (\pm 0.38)$	$1.01~(\pm~0.27)$	$1.19 (\pm 0.44)$	
CO ₂ (ppm)	575.5 (± 91.5)	600.2 (± 125.7)	531.5 (± 75.3)	596.4 (± 127.8)	
SO ₂ (ppb)	4.97 (± 2.82)	9.5 (± 6.58)	$0.9 (\pm 0.5)$	1.8 (± 1.2)	
NO (ppb)	362.3 (± 198.6)	660.1 (± 433.8)	143.8 (± 103)	313.1 (± 254)	
NO ₂ (ppb)	145.4 (± 55.9)	203.1 (± 94.5)	48.3 (± 34.9)	85.9 (± 56)	
NO_x (ppb)	501.3 (± 239.7)	845.7 (± 510.1)	190.2 (± 134.8)	395.3 (± 304.7)	
NO/ NO _x	$0.69 (\pm 0.09)$	$0.75 (\pm 0.08)$	$0.75 (\pm 0.1)$	$0.77 (\pm 0.1)$	
NH ₃ (ppb)	91 (± 31.3)	73.3 (± 22.6)	54.8 (± 20)	68 (± 29.9)	
Ratio of the truck in diesel vehicles	0.11	0.3	0.09	0.28	

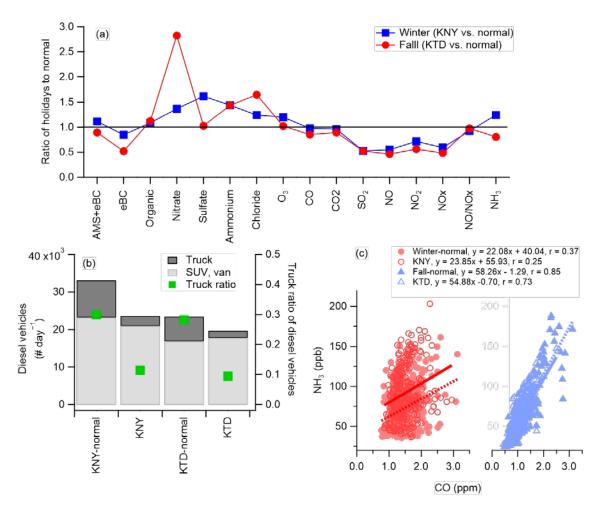


Fig. S7. (a) the ratio of holidays and normal days in gaseous compounds and aerosols. (b) The total volume of diesel vehicle and contribution of categories in diesel vehicles. (c) the comparison of NH₃ and CO in normal days and holidays on winter and fall.

Table. S8. The range of organic and inorganic potential enhancement and photochemical OH \cdot concentration modified OFR in seasons. (The result unit of compound in AMS is $\mu g \ m^{-3}$)

Compound	Winter normal (Winter KNY)	Spring	Summer	Fall normal (Fall KTD)
OA _{PE}	0.4-1.5 (0.8-1.6)	28.4-52.3	40.5-57.7	18.1–31.1 (12.1–21.9)
Nitrate _{PE}	0.4-1.5 (0.9-1.5)	82.2-141.2	158.6-251.4	165.3–290.8 (126.2–214.9)
Sulfate _{PE}	0.2-0.5 (0.1-0.6)	-0.2-0.01	-1.00.2	-1.60.9 (-1.60.7)
Ammonium _{PE}	-0.3-0.1 (-0.2-0.1)	25.6-43.7	49.9–78.9	42-73.3 (32.2-56.1)
Chloride _{PE}	-0.4-0.2 (-0.4-0.2)	0.06-0.1	0.22-0.24	0.1-0.3 (0.03-0.1)
$OA_{PE} / CO_{bkg-cor}$ (µg m ⁻³ ppm ⁻¹)	1.1-3.4 (1.9-8)	41.2-157.4	89.5–414.7	41.6–151.5 (34.9–77.3)
Photochemical OH· day	0.5—8.4 (0.6—8.5)	0.4-3.3	0.2-4.0	0.2-4.3 (0.2-4.3)

^{*} Negative values meant time difference between direct measurement and OFR measurement

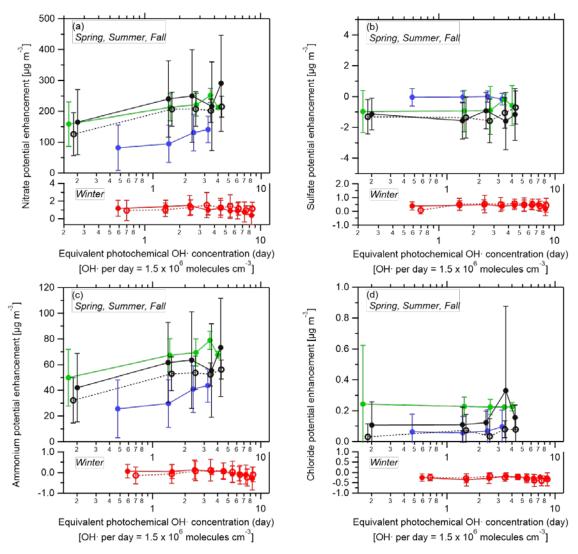


Fig. S8. Seasonal inorganics potential enhancement (inorganic $_{PE}$ = secondary inorganic aerosol formed and corrected in oxidation flow reactor – inorganic corrected in the roadside) by increasing OH· concentration inside the OFR. (a) nitrate. (b) sulfate. (c) ammonium. (d) chloride.

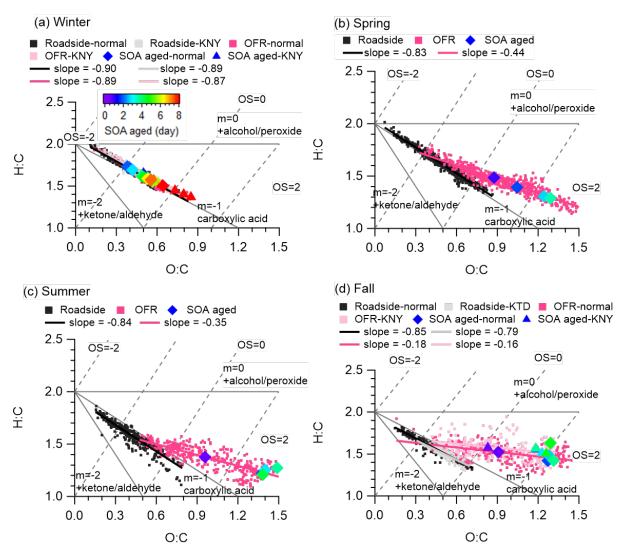


Fig. S9. Van Krevelen diagrams of H/C vs. O/C ratios of organic from roadside and after oxidation by OH•.

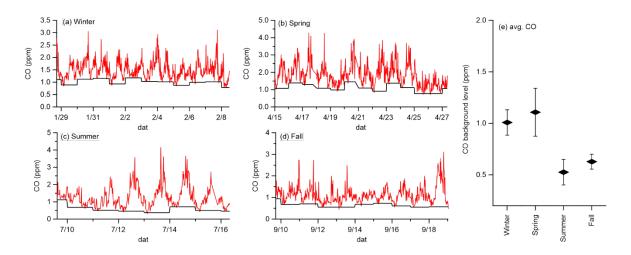


Fig. S10. (a–d) the timeline of CO (red line) and calculated background CO (black line) in seasons. (e) the average concentration of the background CO level on the roadside.

Table. S9. The range of organic, eBC, organic potential enhancement (OA_{PE}), and proportion of the OA_{PE} under ~4.5 OH \cdot day. (The resulting unit of compound is $\mu g \ m^{-3}$)

Commound	Winter		Coming	Cummon	Fall	
Compound	KNY	Normal	Spring	Summer	KTD	Normal
eBC	8.2	9.71	18.7	11.6	6.53	12.58
Organic	11.38	10.52	9.1	8.7	10.81	9.62
OA _{PE}	1.0	1.2	41	47.9	17.9	26.4
Ratio of OA _{PE} to total(organic, eBC, OA _{PE})	5.1%	5.5%	59.6%	70.2%	50.9%	54.3%

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