

Multiphase chemistry experiment in Fogs and Aerosols in the North China Plain (McFAN): integrated analysis and intensive winter campaign 2018†

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Received 6th August 2020, Accepted 5th October 2020

DOI: 10.1039/d0fd00099j

Fine-particle pollution associated with winter haze threatens the health of more than 400 million people in the North China Plain. The Multiphase chemistry experiment in Fogs and Aerosols in the North China Plain (McFAN) investigated the physicochemical mechanisms leading to haze formation with a focus on the contributions of multiphase processes in aerosols and fogs. We integrated observations on multiple platforms with regional and

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† Electronic supplementary information (ESI) available. See DOI: 10.1039/d0fd00099j



box model simulations to identify and characterize the key oxidation processes producing sulfate, nitrate and secondary organic aerosols. An outdoor twin-chamber system was deployed to conduct kinetic experiments under real atmospheric conditions in comparison to literature kinetic data from laboratory studies. The experiments were spanning multiple years since 2017 and an intensive field campaign was performed in the winter of 2018. The location of the site minimizes fast transition between clean and polluted air masses, and regimes representative for the North China Plain were observed at the measurement location in Gucheng near Beijing. The consecutive multi-year experiments document recent trends of PM_{2.5} pollution and corresponding changes of aerosol physical and chemical properties, enabling in-depth investigations of established and newly proposed chemical mechanisms of haze formation. This study is mainly focusing on the data obtained from the winter campaign 2018. To investigate multiphase chemistry, the results are presented and discussed by means of three characteristic cases: low humidity, high humidity and fog. We find a strong relative humidity dependence of aerosol chemical compositions, suggesting an important role of multiphase chemistry. Compared with the low humidity period, both PM₁ and PM_{2.5} show higher mass fraction of secondary inorganic aerosols (SIA, mainly as nitrate, sulfate and ammonium) and secondary organic aerosols (SOA) during high humidity and fog episodes. The changes in aerosol composition further influence aerosol physical properties, e.g., with higher aerosol hygroscopicity parameter κ and single scattering albedo SSA under high humidity and fog cases. The campaign-averaged aerosol pH is 5.1 ± 0.9 , of which the variation is mainly driven by the aerosol water content (AWC) concentrations. Overall, the McFAN experiment provides new evidence of the key role of multiphase reactions in regulating aerosol chemical composition and physical properties in polluted regions.

1. Introduction

In the recent decade, frequently occurring severe haze events in the North China Plain (NCP) have triggered numerous studies on the underlying formation mechanisms, and the contribution of multiphase chemistry to haze formation has become one of the focal points.^{1–4} Besides directly emitted primary pollutants,^{5,6} high levels of secondary inorganic aerosols (SIA, mainly sulfate, nitrate and ammonium) and secondary organic aerosols (SOA) have been observed during haze episodes in many regions of China. As an important component of fine particles, sulfate shows rapid formation during haze events and this high growth rate could not be explained by current state-of-the-art models, suggesting the existence of yet unknown sources of sulfate.² Cheng *et al.*¹ discovered that high rates of sulfate production and large differences between observed and modeled sulfate concentrations were related to high aerosol water content (AWC), suggesting that aqueous-phase oxidation in aerosol water may play a key role, following different reaction pathways depending on aerosol pH and oxidant concentration levels: at pH > 4.5 multiphase reactions of NO₂ and O₃ dominate, while at pH < 4.5 reactions involving transition metal ions (TMI) and H₂O₂ may prevail.¹ Though the importance of multiphase reactions has been widely accepted, the exact formation pathway is still under debate.^{1,2,7–16} Besides sulfate, nitrate, ammonium and secondary organic aerosols also show distinct



characteristics during the haze events, and the increased contribution of multiphase reactions has been suggested as a potential explanation.

To elucidate the chemical mechanisms leading to severe haze formation in Beijing and the NCP, a number of field campaigns and laboratory studies have been carried out. These early studies have revealed several challenges. For example, the observed aerosol concentrations in Beijing were found to be strongly influenced by atmospheric transport processes, which may challenge proper analysis of the prevailing chemical reactions.² Likewise, simple laboratory experiments may differ from that in the real atmosphere, in which synergic effects or high ionic strengths may lead to orders of magnitude difference in reaction rates. Besides, many studies lack information about aerosol pH, which controls the rates of many atmospheric multiphase reactions.¹

Under this background, the Multiphase chemistry experiment in Fogs and Aerosols in the North China Plain (McFAN) was organized to advance our understanding of the physical-chemical mechanisms leading to severe haze formation, especially with a focus on the contributions of multiphase processes in aerosols and fogs. Instead of Beijing, we selected a site located in the central polluted region of the NCP, where the influence of transport was smaller. We made a comprehensive design to cover more parameters (such as aerosol pH) that are required for a closure study on the multiphase reactions. We also introduced an environmental chamber system and an automatic-shifting aerosol inlet system to perform kinetic experiments under real ambient air conditions. In this overview paper, the field site and the instrumentation deployed during the McFAN experiment are firstly described. Then new observation-based findings are presented.

2. Experiment design and criteria for site selection

One of the main scientific objectives of the McFAN experiment was to investigate the effects of multiphase processes on particle formation and evolution. To achieve this goal, the related research was schemed into three cases: (1) the formation and evolution of aerosols during fog conditions (*i.e.*, fog case); (2) the formation and transformation of aerosols during high relative humidity days and the respective impact of multiphase reactions (*i.e.*, high RH case); (3) the formation and transformation of aerosols during low humidity days (*i.e.* low RH case). Given the temporal and spatial variations of the air quality in the NCP, one intensive field campaign was performed from 11th November to 24th December 2018 at the Gucheng site, to capture fog and haze events. The observation was equipped with abundant online and offline instruments, allowing to record and analyze meteorological parameters as well as variations and properties of aerosol and gas species. The instruments and their measured parameters are listed in Table 1. Briefly, the trace gas instruments (O_3 , NO_x , CO, SO_2 , NH_3) from Chinese Academy of Meteorological Sciences (CAMS, see Table 1) were housed in an air-conditioned room in a two-story building in the southern part of the Gucheng station,¹⁷ and the other instruments were installed in two air-conditioned containers placed on the north side of the site. To investigate multiphase reaction kinetics, an environmental chamber system was also employed. The



Table 1 Instrumentation, measurement parameters, and partners involved in the McFAN experiment

	Instrument	Measured parameter	Organization ^a
Aerosols	Scanning Mobility Particle Sizer (SMPS)	Aerosol number size distribution (10–1000 nm)	JNU
	Aerodynamic Particle Sizer (APS)	Aerosol number size distribution (0.5–20 μm)	JNU
	Particle Size Magnifier (PSM)	Aerosol number concentration (1–4 nm)	MPIC
	Nephelometer	Aerosol optical properties	PKU
	Aethalometer AE33	Black carbon optical properties	JNU
	Aerosol photometer	Aerosol optical properties	JNU
	Capture Vaporizer Time-of-Flight Aerosol Chemical Speciation Monitor (CV-ToF-ACSM)	Aerosol composition	NJU, IAP-CAS
	OC/EC Aerosol Analyzer	Aerosol composition	JNU
	Four-channel high-volume aerosol sampler	Aerosol composition	JNU
	Cascade impactor	Aerosol composition	CAMS
	Cloud Condensation Nucleus Counter (CCNC)	Aerosol hygroscopicity (combined with DMA)	JNU
	Hygroscopic Tandem Differential Mobility Analyzer (HTDMA)	Aerosol hygroscopicity	JNU
	Humidified nephelometer	Aerosol hygroscopicity	PKU
	Single-Particle soot Photometer (SP2)	Black carbon	JNU
	Centrifugal Particle Mass Analyzer (CPMA)	Black carbon density (combined with SP2)	JNU
	Trace gases	Ground-based Counterflow Virtual Impactor (GCVI)	Fog
Proton-Transfer-Reaction Mass Spectrometry (PTR-MS)		VOCs	JNU
Chemical Ionization Mass Spectrometry (CIMS)		VOCs	JNU
Model 43i SO ₂ analyzer		SO ₂	MPIC, CAMS
Model 42i NO _x analyzer		NO _x	MPIC, CAMS
Model 49i O ₃ analyzer		O ₃	MPIC, CAMS
LOng Path Absorption Photometer (LOPAP)		HONO	JNU
Monitor for AeRosols and Gases in ambient Air (MARGA)		Trace gases: HCl, SO ₂ , NH ₃ , HNO ₃ , HONO	JNU
Picarro G2103 gas concentration analyzer		NH ₃	JNU
Economical ammonia analyzer (LGR DLT-100)		NH ₃	CAMS



Table 1 (Contd.)

Instrument	Measured parameter	Organization ^a
Multiphase Environmental chamber system reactions	Multiphase reaction kinetics, sulfate formation rate	MPIC

^a JNU: Jinan University; NJU: Nanjing University; PKU: Peking University; IAP-CAS: Institute of Atmospheric Physics Chinese Academy of Sciences; CAMS: Chinese Academy of Meteorological Sciences; MPIC: Max Planck Institute for Chemistry.

chambers were installed on the roof of one container and the associated analyzers were located inside the container. Data obtained from the chamber experiments are still under analysis, therefore are not presented in this study.

The Gucheng site was selected to carry out the McFAN campaign. The site is ~100 km southwest of Beijing and ~35 km north of Baoding city. It is a suburban site located at the Ecological and Agricultural Meteorology Station (39°09' N, 115°44' E) of the Chinese Academy of Meteorological Sciences.¹⁸ It is surrounded by agricultural fields (for cultivation of wheat and corn) and the closest residential town (Dingxing county) is ~1.5 km away. The location of the site minimizes fast transition between clean and polluted air masses (e.g., in Beijing), and helps to maintain a pollution regime representative for the NCP, making it an ideal position to investigate atmospheric chemical processes.

3. Overview of meteorology, gas and aerosol variation

Fig. 1 shows the observed time series of meteorological parameters, concentrations of inorganic gas pollutants and PM_{2.5} during the McFAN campaign. The average wind speed observed over the whole campaign was only ~0.5 m s⁻¹, but occasionally it could reach up to ~8.5 m s⁻¹. For most of the measurement time, the wind speed displayed a diel variation with higher values during daytime than at nighttime. The level of ambient air relative humidity (RH) for the whole campaign can be generally classified into two stages: from 11th November to 03rd December, the daily averaged RH covered a high range of 45–89%, deemed as high humidity period; from 04th to 24th December, the daily averaged RH was in a relatively low range of 23–69%, considered as low humidity period. Moreover, two typical fog events (with RH = 100%) were observed during the campaign. All these periods are indicated by the colored rectangles in Fig. 1B. The ambient temperature (*T*) ranged from -14 °C to 18 °C, with an average of ~1.3 °C. More information of the meteorological conditions can be found in the ESI.† The campaign-averaged concentrations of key gas pollutants were: ~10 ppb for SO₂, ~32 ppb for NO₂, ~28 ppb for NH₃ and ~7 ppb for O₃ (Fig. 1C and D). The mass concentration of PM_{2.5} covered a wide range from ~3 to ~462 μg m⁻³ (Fig. 1E) with an average of ~121 μg m⁻³. Additional information about volatile organic compounds (VOCs) can be found in Fig. S3.† Fig. 2 shows the variations of





Fig. 1 Time series of meteorological parameters (A and B), inorganic gas pollutants (C and D) and mass concentration of $PM_{2.5}$ (E) during the McFAN campaign. The colored rectangles in panel (B) indicate different characteristic periods: dark blue – fog, light blue – high humidity, orange – low humidity. The mass concentration of $PM_{2.5}$ is obtained by summing up elemental carbon (EC) and other non-refractory aerosol components measured by the OC/EC analyzer and the CV-ToF-ACSM (Table 1), respectively.



Fig. 2 Time series of submicron aerosol (PM_1) mass concentrations (A and B) and mass fractions of different components (C). In panel (B), the colored rectangles have the same meaning as in Fig. 1. Positive matrix factorization analysis is employed to classify organic aerosol (OA) into various primary OA (POA) factors including hydrocarbon-like OA (HOA), cook-related OA (COA), biomass burning OA (BBOA) and coal combustion OA (CCOA), and a secondary OA (SOA) factor: oxygenated OA (OOA). More details can be found in Fig. 5 and in the work of Sun *et al.*¹⁹

chemical compositions of submicron particles (PM_1). The mean mass concentration of PM_1 was $\sim 67 \mu\text{g m}^{-3}$ with a maximum of $\sim 250 \mu\text{g m}^{-3}$. In the low humidity period, the averaged contributions of inorganic components (mainly as



nitrate, sulfate and ammonium) and organics were 34% and 57%, respectively. While in the high humidity period, the contribution of inorganics increased to 50%, slightly higher than that of the organic aerosols (43%). During fog periods, the averaged mass fraction of inorganics reached 59% and the contribution of organics dropped to 34%. Among organics, the mass fraction of secondary organic aerosols (SOA) in total organic aerosols (ACSM-determined) increased from 23% in the low humidity period to 47% and 53% in high humidity and fog episodes, respectively.

4. Impact of multiphase processes on aerosol composition and acidity

4.1. Aerosol chemical composition

Ambient RH strongly influences the reaction pathways and their magnitude.^{20,21} Under high RH conditions (*e.g.*, during fog and haze episodes), aerosols tend to become liquid with increased surface and volume concentrations due to hygroscopic growth while under low RH aerosols become solid or semi-solid particles. Thus, we first compared the aerosol composition between different RH conditions. Fig. 3 shows the diel variation of the composition of PM₁ between three cases (as marked in Fig. 1 and 2): low humidity (RH = 23–69%), high humidity (RH = 45–89%) and fog (RH = 100%) along with the averages of the entire campaign period. For the entire campaign (Fig. 3I), PM₁ was on average composed of ~9% elemental carbon (EC), ~11% sulfate, ~20% nitrate, ~15% ammonium, ~3% chloride, ~15% POA and ~27% SOA. The inorganic mass fraction substantially increased from low humidity to high humidity and to the fog case, and SOA also showed a similar but lesser increasing trend even with a decreasing contribution of the total OA. Such phenomenon suggests enhanced formation of secondary species with increasing RH, potentially due to the multiphase processes.²² Interestingly, while the mass fraction of sulfate follows the order of low humidity < high humidity < fog periods, for nitrate the order is low humidity <

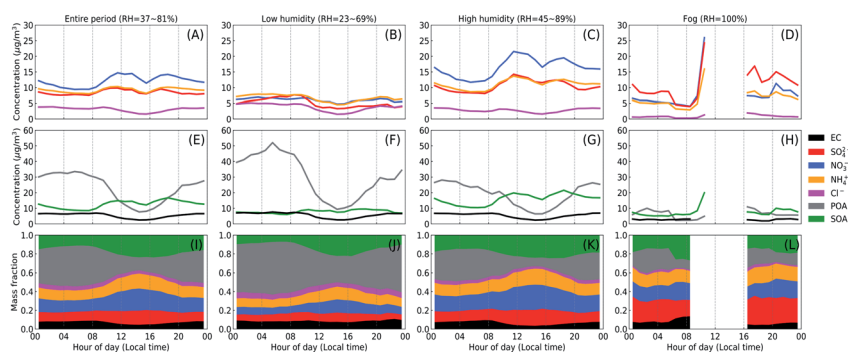


Fig. 3 Diel variations of the submicron aerosol (PM₁) mass concentrations (A–H) and mass fractions (I–L) of the major components of submicron aerosols (PM₁) averaged over the entire campaign and focus periods with different relative humidity (RH) conditions. The solid lines in (A–H) show the mean value of hourly averaged concentrations of each component. The displayed RH range relevant to each focus period is based on daily averaged data.



fog < high humidity (Fig. 3I–L). A clear RH effect is also found on NOR and SOR (molar ratio of nitrate or sulfate to the sum of nitrate and NO_2 or sulfate and SO_2), which represents the degree of secondary formation of nitrate and sulfate.^{2,23} As shown in Fig. 4, the variations of SOR and NOR follow the same trend as the mass fraction of sulfate and nitrate. These results indicate an important role of high RH in promoting the formation of SIA and SOA, probably through multiphase reactions. Information of gases (SO_2 , NO_2 , NH_3 , O_3 and VOCs) and $\text{PM}_{2.5}$ concentrations can be found in the ESI.†

4.2. Size dependence of aerosol chemical composition

Sun *et al.*¹⁹ characterized aerosol composition and sources of organic aerosol (OA) during the McFAN campaign in 2018, and also the compositional differences between PM_1 and $\text{PM}_{2.5}$ by using the CV-ToF-ACSM (Table 1). As depicted in Fig. 5, comparable contributions of OA and secondary inorganic aerosol (SIA) were found during the high humidity period ($\text{RH} = 73 \pm 24\%$), while the low humidity period ($\text{RH} = 48 \pm 18\%$) showed a reduced contribution of SIA. OA composition was also substantially different with a much higher contribution of secondary OA (46–47%) during the high humidity period than the low humidity phase (19–21%). In contrast, primary OA from biomass burning, coal combustion, and traffic emissions dominated OA (71–73%) during the low humidity period (Fig. 5B). These results highlight that meteorological conditions, in particular RH, play an important role in secondary aerosol formation in NCP, and hence change the contributions of primary and secondary aerosol.

Sun *et al.*¹⁹ further analyzed the chemical differences between PM_1 and $\text{PM}_{2.5}$ during both high and low humidity periods. As indicated in Fig. 5, aerosol composition and OA composition of PM_1 and $\text{PM}_{2.5}$ were relatively similar during both focus periods, despite the concentration differences by up to 30%. However, the decreases in $\text{PM}_1/\text{PM}_{2.5}$ ratios as a function of RH were also observed for secondary organic and inorganic aerosol species, which was likely due to the changes in aerosol hygroscopicity and phase states. In contrast, primary aerosol

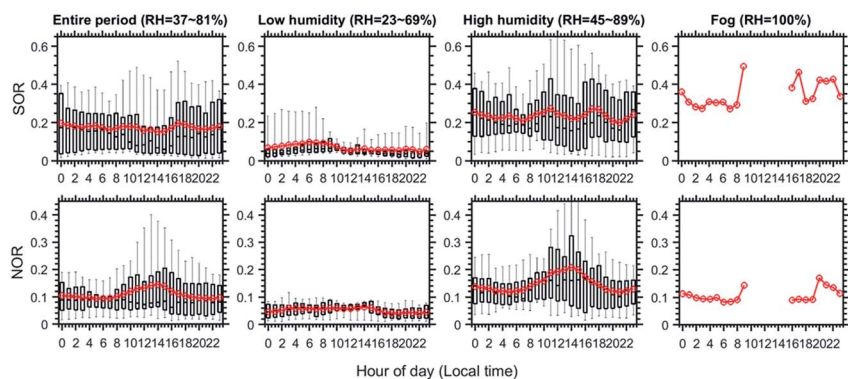


Fig. 4 Diel variations of the ratios of nitrate to nitrate plus NO_2 (NOR) and sulfate to sulfate plus SO_2 (SOR) averaged over the entire campaign and focus periods with different relative humidity (RH) conditions. The boxes and whiskers indicate percentiles (90th, 75th, 50th (median), 25th and 10th).



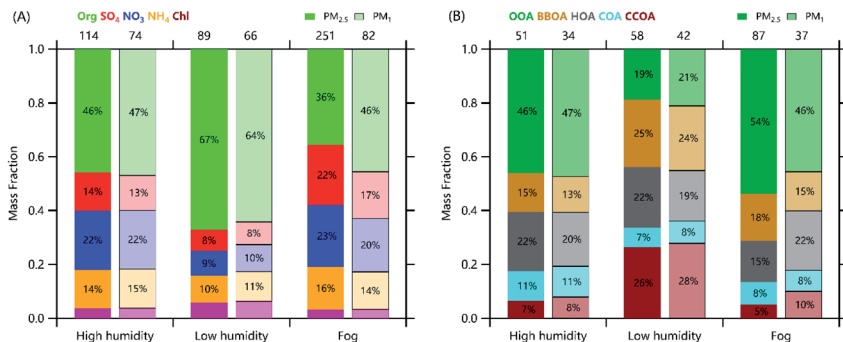


Fig. 5 Average (A) chemical composition of non-refractory PM_{2.5} and PM₁, and (B) OA composition during the focus periods with different relative humidity (RH) conditions. The numbers on the top of the figures are the average mass concentrations ($\mu\text{g m}^{-3}$) for each period. Adopted and modified from Sun *et al.*¹⁹

species with low hygroscopicity did not show clear RH dependence of PM₁/PM_{2.5} ratios. Large differences in both mass concentrations and composition were observed during fog events. For instance, PM₁ on average accounted for 33% of PM_{2.5} due to the rapid hygroscopic growth of aerosol particles under high RH levels, and PM_{2.5} showed largely elevated contributions of SIA (61% *vs.* 51%) and SOA (54% *vs.* 46%) compared with PM₁. Further analysis showed that the chemical differences between PM₁ and PM_{2.5} had negligible impacts on predictions of particle acidity, while they could affect aerosol liquid water content by up to 50–70%.

4.3. Aerosol acidity

To examine aerosol acidity during the campaign, we modelled the aerosol pH based on the ISORROPIA model²⁴ and observation data. The concentrations of sulfate, nitrate, ammonium and chloride were taken from the CV-ToF-ACSM measurement, and the ammonia concentrations were taken from observations using the Picarro G2103 gas analyzer. Only the data points with RH > 40% are used here. Sensitivity studies show that the uncertainty in chloride measurements by the ACSM is expected to result in a pH variation of -0.15 to 0.35 , while the missing measurements of non-volatile species (*i.e.*, Na⁺, Ca²⁺, K⁺ and Mg²⁺) are expected to result in a pH underestimation of 0.07 to 0.35 . In general, simulated gas–particle partitioning of ammonia agrees well with the observations. As shown in Fig. 6, the aerosol pH averaged 5.1 ± 0.9 over the whole observation period. A clear diel variation is observed for the entire and high humidity periods, with an obvious drop in pH during the daytime, and bottomed around 15:00. This pH drop is mainly driven by the diel variation in RH and therefore the aerosol water content (AWC) concentrations.²⁵ The diel variation is not complete for low humidity and fog periods, as there is no available afternoon data. The absence of afternoon data during the low humidity period is due to its low RH (<40%) and the aerosols were mostly dry. Similarly, due to stronger solar radiation and lower RH, no foggy events appeared during the afternoons. For these two periods, the nighttime pH was relatively stable, both of which fluctuating around 5.5. A more



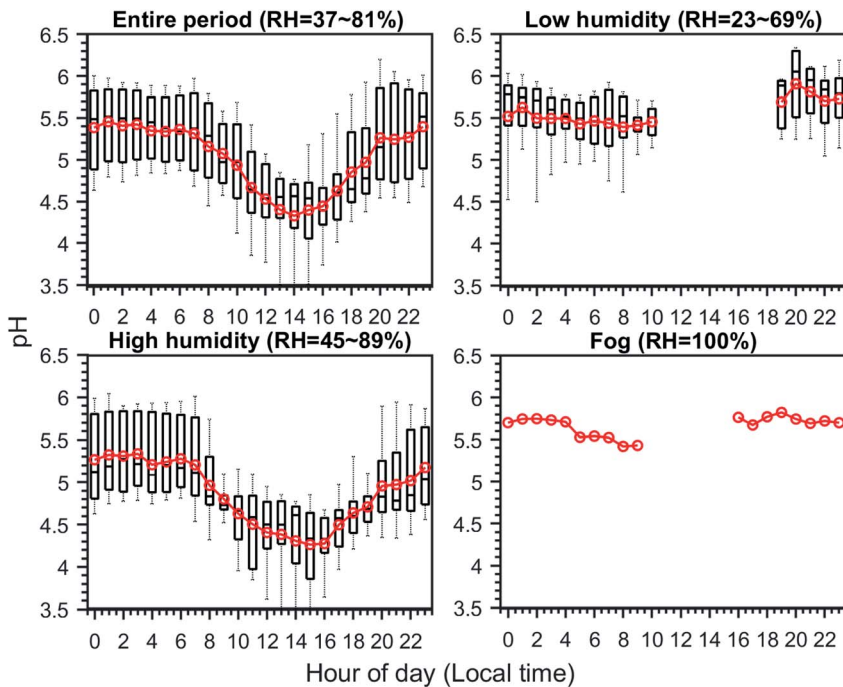


Fig. 6 Diel variation of aerosol pH averaged over the entire campaign and focus periods with different relative humidity (RH) conditions. The boxes and whiskers indicate percentiles (90th, 75th, 50th (median), 25th and 10th), and the red circles represent arithmetic mean values. Note that the displayed RH ranges for each focus period represent daily averages. Due to limited data points at each hour, only mean values are shown for the fog period.

detailed analysis on the drivers of aerosol pH can be found elsewhere (Zheng *et al.*, Drivers of aerosol pH during the multiphase chemistry experiment in fogs and aerosols (McFAN) experiment in the North China Plain, in preparation).

5. Impact of multiphase processes on aerosol physical properties

5.1. Aerosol hygroscopicity and phase state

Fig. 7 shows the diel variations of the measured cloud condensation nuclei (CCN) number concentration at two different supersaturation (SS) levels, under the three cases. For the entire period, N_{CCN} was lower during daytime ($\sim 09:00\text{--}20:00$) than at night and this difference became more pronounced at the higher SS value. Similar phenomenon has been reported in another study at the same site during winter time.²⁶ At larger SS values, particle size tends to be more crucial in affecting N_{CCN} .²⁷ The lower N_{CCN} during daytime was likely caused by the varied particle number size distribution (see Fig. S8†) and the decreased particle number concentration (see Fig. 8A–D). Compared to the low humidity case, N_{CCN} in the high humidity case was slightly higher at the lower SS value, which might be due to (1) the larger mass fraction of SIA (nitrate, sulfate and ammonium) and SOA



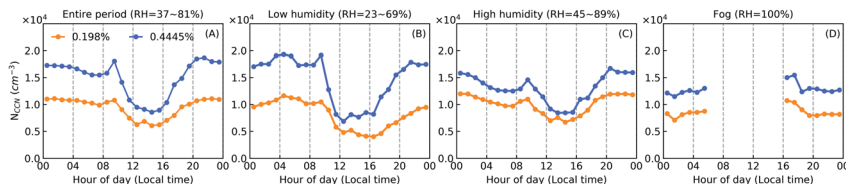


Fig. 7 Diel variations of CCN number concentration at two different supersaturation levels averaged over the entire campaign and focus periods with different relative humidity (RH) conditions. The symbols represent the mean value of hourly averaged concentrations.

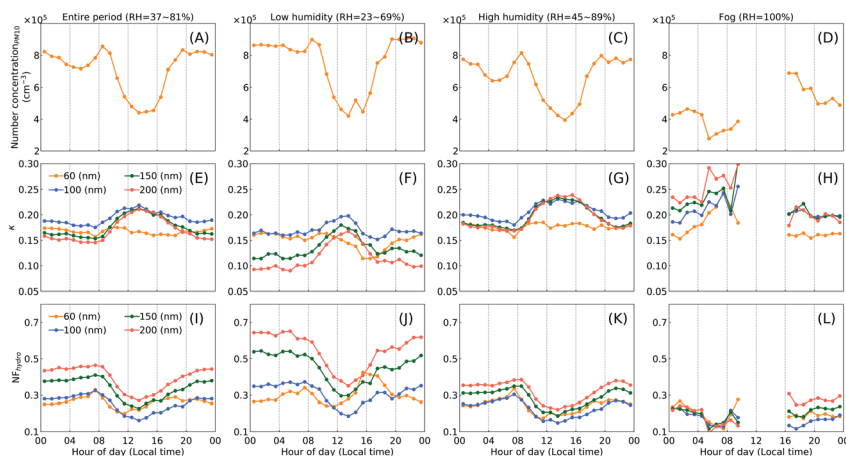


Fig. 8 Diel variations of particle number concentration in PM_{10} size range (top row), hygroscopicity parameter κ (second row) and number fraction of hydrophobic particles $N_{Fr_{hydro}}$ for selected particle sizes averaged over the entire campaign and focus periods with different relative humidity (RH) conditions.

found for the high humidity case in Fig. 3; (2) the increased concentration of larger-size particles under high RH. Increasing the fraction of hydrophilic inorganics and organics in aerosols will enhance their ability to activate into CCN, leading to a higher N_{CCN} . However, for the fog period, N_{CCN} was even lower than that of the low humidity case. Since fog droplets can efficiently scavenge particles with smaller sizes, the observed particle number concentration within fog events was much lower than during non-fog periods (Fig. 8). Thus, the low aerosol number concentration during fog events might result in the much less N_{CCN} .

The hygroscopicity parameter κ (ref. 28 and 29) showed higher values in daytime than at night (Fig. 8E–G), indicating the potential effect of photochemistry in enhancing particles' hygroscopicity. Based on hygroscopicity and aerosol composition measurements during the McFAN campaign, Kuang *et al.*¹⁷ found that the hygroscopicity parameter of organic aerosols, κ_{OA} , showed a prominent diel variation with a peak value present near 14:30 local time. And this diel variation was highly and positively correlated with the mass fraction of oxygenated organic aerosols (OOA), pointing to the important role of photochemical



processing in enhancing κ_{OA} . Similar photochemical impact was also reported by Wang *et al.*²⁷ at a suburban site in the central NCP. Moreover, an increase of κ could be found with increasing ambient RH, *i.e.*, from low humidity to fog case (Fig. 8F–H). Meanwhile, an opposite trend was observed for NF_{hydro} , meaning that the number fraction of hydrophobic particles is getting less. Along with the results shown in Fig. 3, we conclude that the enhanced hygroscopicity of aerosols is due to the increased hydrophilic fraction which is most likely caused by either high-RH favored multiphase reactions or photochemistry, and the coupling of both effects.

5.2 Aerosol optical properties

Aerosol optical properties have been found to show a strong RH dependence, which may result in a positive feedback and influence the planetary boundary layer (PBL) meteorology and the haze formation.^{30–32} Fig. 9 shows the optical properties of aerosols under the three cases. The diel variation of the Absorption Ångström Exponent (AAE) showed higher values during nighttime than during daytime with a much higher $\text{AAE}_{370-470}$ (average \pm std, 1.8 ± 0.35) than $\text{AAE}_{880-950}$ (0.8 ± 0.1) (Fig. S9†). This implies that, besides black carbon (BC), there existed some brown carbon (BrC), which had comparable light-absorbing capacities as BC at short wavelengths. Thus, the feature of higher nighttime $\text{AAE}_{370-950}$ in Fig. 9A–C might be due to increased levels of BrC generated by domestic heating and biomass burning during the night. Significant differences among different RH cases can be found for AAE and the Single Scattering Albedo (SSA). With increased RH from the low humidity to the high humidity case, AAE decreased from ~ 1.7 to ~ 1.55 and even down to ~ 1.4 under fog conditions. The decreased AAE suggests that the fraction of light-absorbing components such as BC and BrC is reduced, due to either decreased mass fraction or changed mixing state.^{33,34} For the SSA, however, its values under high humidity were remarkably higher than at low humidity, indicating an enhanced fraction of light-scattering components. This phenomenon could be due to the boosted formation of SIA (*i.e.*, nitrate,

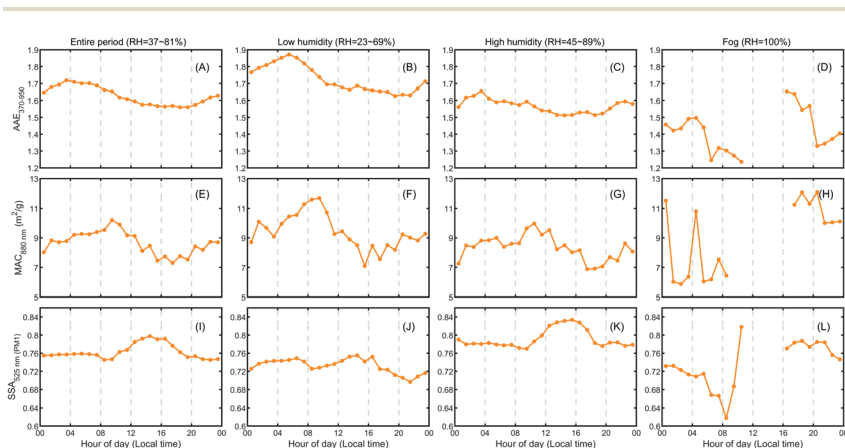


Fig. 9 Diel variations of Absorption Ångström Exponent (top row), Mass Absorption Cross section (second row), and Single Scattering Albedo (bottom row) averaged over the entire campaign and focus periods with different relative humidity (RH) conditions.



P.F. and Y.S. provided measurement data. G.L. and H.S. wrote the manuscript with inputs from Y.C., G.Z., U.P. and U.K.

Conflicts of interest

There are no conflicts to declare.

Acknowledgements

We acknowledge the National Key Research and Development Program of China (grant no. 2017YFC0210104), the National Natural Science Foundation of China (grant no. 91644218) and the support from the Max Planck Society. G. L. acknowledges the financial support from the China Scholarship Council (CSC). Open Access funding provided by the Max Planck Society.

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