Fine particle pH and gas–particle phase partitioning of inorganic species in Pasadena, California, during the 2010 CalNex campaign

Hongyu Guo¹,1, Jiuneng Liu¹,a, Karl D. Froyd2,3, James M. Roberts2, Patrick R. Veres2,3, Patrick L. Hayes3,4,5, Jose L. Jimenez3,5, Athanasios Nenes1,6,7,8, and Rodney J. Weber1

¹School of Earth and Atmospheric Sciences, Georgia Institute of Technology, Atlanta, GA, USA
²Chemical Sciences Division, Earth System Research Laboratory, NOAA, Boulder, CO, USA
³Cooperative Institute for Research in Environmental Sciences (CIRES), Boulder, CO, USA
⁴Department of Chemistry, Université de Montréal, Montréal, Québec H3T 1J4, Canada
⁵Department of Chemistry and Biochemistry, University of Colorado Boulder, Boulder, CO, USA
⁶School of Chemical and Biomolecular Engineering, Georgia Institute of Technology, Atlanta, GA, USA
⁷Foundation for Research and Technology, Hellas, Greece
⁸National Observatory of Athens, Athens, Greece
¹now at: Atmospheric Sciences and Global Change Division, Pacific Northwest National Laboratory, Richland, WA, USA

Correspondence to: Rodney J. Weber (rweber@eas.gatech.edu) and Athanasios Nenes (athanasios.nenes@gatech.edu)

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Abstract. pH is a fundamental aerosol property that affects ambient particle concentration and composition, linking pH to all aerosol environmental impacts. Here, PM₁ and PM₂.₅ pH are calculated based on data from measurements during the California Research at the Nexus of Air Quality and Climate Change (CalNex) study from 15 May to 15 June 2010 in Pasadena, CA. Particle pH and water were predicted with the ISORROPIA-II thermodynamic model and validated by comparing predicted to measured gas–particle partitioning of inorganic nitrate, ammonium, and chloride. The study mean ± standard deviation PM₁ pH was 1.9 ± 0.5 for the SO₄²⁻–NO₃⁻–NH₄⁺–HNO₃–NH₃ system. For PM₂.₅, internal mixing of sea salt components (SO₄²⁻–NO₃⁻–NH₄⁺–Na⁺–Cl⁻–K⁺–HNO₃–NH₃–HCl system) raised the bulk pH to 2.7 ± 0.3 and improved predicted nitric acid partitioning with PM₂.₅ components. The results show little effect of sea salt on PM₁ pH, but significant effects on PM₂.₅ pH. A mean PM₁ pH of 1.9 at Pasadena was approximately one unit higher than what we have reported in the southeastern US, despite similar temperature, relative humidity, and sulfate ranges, and is due to higher total nitrate concentrations (nitric acid plus nitrate) relative to sulfate, a situation where particle water is affected by semi-volatile nitrate concentrations. Under these conditions nitric acid partitioning can further promote nitrate formation by increasing aerosol water, which raises pH by dilution, further increasing nitric acid partitioning and resulting in a significant increase in fine particle nitrate and pH. This study provides insights into the complex interactions between particle pH and nitrate in a summertime coastal environment and a contrast to recently reported pH in the eastern US in summer and winter and the eastern Mediterranean. All studies have consistently found highly acidic PM₁ with pH generally below 3.

1 Introduction

Ambient aerosol particles affect human health and climate (Lim et al., 2012; IPCC, 2013), and have many other environmental effects. Particle pH is linked to all of these by altering the fundamental aerosol properties of particle mass and chemical composition. For example, some important pathways leading to secondary organic aerosol (SOA) formation from biogenic volatile organic compounds (VOCs), such as isoprene and α-pinene, are catalyzed by H⁺ (Jang et al., 2002; Gao et al., 2004; Edney et al., 2005; Surratt et al.,...
directly affects particle mass and composition through altering the partitioning of both semi-volatile inorganic and organic acids between particle and gas phases (Guo et al., 2016). pH affects the nitrogen cycle through gas–particle partitioning of HNO$_3^-$–NO$_3^-$ and NH$_3$–NH$_4^+$, impacting deposition patterns due to large differences in gas versus particle dry deposition rates (Huebert and Robert, 1985; Duyzer, 1994; Schrader and Brummer, 2014).

Particle pH is linked to adverse health impacts, both directly and indirectly. Synergistic adverse health effects have been observed between ozone and acidic aerosols (Last, 1991; Enami et al., 2008) and epidemiological studies have reported adverse health outcomes associated with strong aerosol acidity (Koutrakis et al., 1988; Thurston et al., 1994; Dockery et al., 1996; Raizenne et al., 1996; Gwynn et al., 2000; Lelieveld et al., 2015). Low pH increases the solubility of transition metals, such as iron and copper (Meskhidze et al., 2003; Oakes et al., 2012; Longo et al., 2016; Fang et al., 2017), which have been linked to aerosol toxicity through aerosol oxidative effects (Ghio et al., 2012; Verma et al., 2014; Fang et al., 2015, 2017). Metal mobility also affects nutrient distributions with important impacts on photosynthesis productivity (Duce and Tindale, 1991; Meskhidze et al., 2003; Nenes et al., 2011; Ito and Xu, 2014; Myriokefalitakis et al., 2015; Li et al., 2017), carbon sequestration and ocean oxygen levels (Ito et al., 2016).

Due to limitations with direct particle pH measurement techniques, fine particle pH has often been indirectly inferred from aerosol composition based on ion balances or cation-anion molar balances, e.g., ammonium to sulfate (NH$_4^+$ / SO$_4^{2-}$) molar ratios. However, these are largely inaccurate pH proxies (Guo et al., 2015, 2016; Hennigan et al., 2015; Weber et al., 2016), which if used can provide misleading indications on the level of acidity present in the aerosol and pH-related properties. Alternatively, particle pH is more accurately calculated with a thermodynamic model, such as ISORROPIA-II (Nenes et al., 1998; Fountoukis and Nenes, 2007) or E-AIM (Clegg et al., 1998, 2003; Wexler et al., 2015; Weber et al., 2016), which if used can provide accurate pH proxies (Guo et al., 2015, 2016; Hennigan et al., 2015; Weber et al., 2016), if which if used can provide misleading indications on the level of acidity present in the aerosol and pH-related properties.

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This work adds to our investigation of particle pH in differing locations and under different emission characteristics. We have reported that ground-level pH in the southeastern US is 0.9 ± 0.6 (mean ± standard deviation, SD) in summer and 2.0 ± 1.0 in winter (Guo et al., 2015), and 0.8 ± 1.0 in the boundary layer and lower free troposphere (<5 km altitude) over broad regions of the eastern US in winter, based on aircraft data (Guo et al., 2016). A pH of 1.3 ± 1.1 has also been reported in various air masses advected to Crete in the eastern Mediterranean (Bougiatioti et al., 2016). This study focuses on particle pH in an urban coastal site, Pasadena, CA, and investigates the reasons for significantly higher nitrate mass loadings compared to those of the southeastern US (Zhang et al., 2007; Hand et al., 2012).

2 Methods

2.1 Sampling site

Aerosol and gas measurements were conducted on the California Institute of Technology campus in Pasadena, California (34.140582° N, 118.122455° W; altitude above sea level: 235 m), as part of the 2010 California Research at the Nexus of Air Quality and Climate Change (CalNex) campaign from 15 May to 15 June 2010 (Ryerson et al., 2013). The CalNex ground site was located within the Los Angeles Basin and approximately 16 km northeast of central Los Angeles (hereafter referred to as LA) and 5 km south of the San Gabriel Mountains. With the dominant wind from the southwest, the site was regularly impacted by emissions transported from the coast and central LA area (Washenfelder et al., 2011; Hayes et al., 2013). Meteorological data used in the following analysis are from the National Oceanic and Atmospheric Administration’s (NOAA) CalNex researchers meteorological station.

2.2 Instrumentation

2.2.1 Particles

- PILS-IC: PM$_{2.5}$ (particles with aerodynamic diameters <2.5 µm at ambient conditions) water-soluble ions were measured with a particle-into-liquid sampler coupled with ion chromatographs (PILS-IC), similar to that described elsewhere (Orsini et al., 2003; Hennigan et al., 2006; Sullivan et al., 2006; Peltier et al., 2007). The operation of PILS-IC during CalNex has been discussed by Liu et al. (2012) in detail. Ambient air was sampled through a URG (Chapel Hill, NC, USA) PM$_{2.5}$ cyclone and mixed with near 100°C water vapor generated from deionized water. After growth, droplets were collected by impaction, producing a continuous liquid sample for online ion chromatograph analysis. All ambient data were blank-corrected by periodically measuring filtered ambient air. PM$_{2.5}$ anion data were available throughout the CalNex study, whereas cation data were only available for the last week (8–14 June). Since particle acidity predictions require both anion and cation data, discussions on PM$_{2.5}$ pH will include only the last
2.2.2 Gases

- AMS: PM$_1$ non-refractory inorganic and organic components were measured by a high-resolution time-of-flight aerosol mass spectrometer (HR-ToF-AMS, Aerodyne Research Inc., hereafter referred to as “AMS”; DeCarlo et al., 2006). The operation procedure of the AMS during this study has been described in Hayes et al. (2013) and measurement uncertainty estimated at 30% (Middlebrook et al., 2012). The AMS detects non-refractory species in dried aerosols through an aerodynamic focusing lens, a detection chamber where aerosols are flash vaporized and ionized, followed by a time-of-flight mass spectrometer. The comparison of AMS PM$_1$ and PILS-IC PM$_{2.5}$ measurements was summarized in Hayes et al. (2013) and is discussed further below. In the following analysis PM$_1$ species are AMS data and PM$_{2.5}$ species are PILS-IC data.

- PALMS: single aerosol composition and size for diameters 0.15–3 μm were measured by an online particle analysis by laser mass spectrometer (PALMS) (Murphy et al., 2006; Froyd et al., 2010). In this method, individual aerosols scatter light from a continuous laser beam and trigger an excimer laser that ionizes the single particle. The resulting ions are analyzed by a time-of-flight mass spectrometer to generate a complete positive or negative mass spectrum per particle. In this study, PALMS data provide insights into the aerosol mixing state.

2.2.3 Gases

- QC-TILDAS: gas-phase NH$_3$ was quantified using a quantum cascade tunable infrared laser differential absorption spectrometer (QC-TILDAS), developed by Aerodyne Research Inc. and described in details in Ellis et al. (2010). QC-TILDAS uses a thermoelectrically cooled pulsed quantum cascade (QC) laser, which measures NH$_3$ at 967 cm$^{-1}$ in the infrared regime. The laser beam is directed into an astigmatic Herriot absorption cell, where it passes between two highly reflective mirrors before leaving the cell and arriving at a thermoelectrically cooled mercury cadmium telluride (HgCdTe) infrared detector. A 12 m heated (40 ± 2°C) and insulated 0.95 cm perfluoroalkoxy (PFA) line (Clayborn Lab, Truckee, CA, USA) connects a short (10 cm; 8 m above ground) custom-designed quartz inlet to the QC-TILDAS. The inlet includes two ports for the introduction of calibration and background gas designed so that the flows follow the same path through the inlet as the ambient sample. Zero-air measurements and ammonia calibrations were performed periodically during the CalNex campaign. The detection limit was 1.5 ppbv for 1 s data and decreased to 0.42 ppbv for 1 min averaging data. The overall NH$_3$ measurement uncertainty was 10%.

- NI-PT-CIMS: gas-phase HNO$_3$ and HCl measurements were made using a negative-ion proton-transfer chemical ionization mass spectrometer (NI-PT-CIMS; hereafter referred to as “CIMS”). Details of the NI-PT-CIMS design and operation during CalNex have been reported in Veres et al. (2008). Acidic molecules are ionized by proton transfer with acetate ions and detected as conjugate anions with a quadrupole mass spectrometer. The CIMS was placed on top of an instrument trailer at 3 m height with a heated (75°C) 0.32 cm Teflon inlet. 1.3 m in length, sampling at 5 m relative to ground level. (Possible biases in measured HNO$_3$ and HCl due to aerosol volatilization in the heated inlet is discussed in Sect. 4.1). In the field, instrument backgrounds were quantified every 190 min for a duration of 30 min. The detection limits were 0.080 ppbv for HNO$_3$ and 0.055 ppbv for HCl. Overall measurement uncertainty was 35%. All gas-phase concentrations reported in ppbv were converted to μg m$^{-3}$ at ambient conditions to be consistent with particle-phase measurements and for thermodynamic calculations. In all cases, gas and particle concentrations are reported and utilized in the model using ambient conditions (i.e., all volumetric concentrations are at ambient temperature and pressure).

2.3 pH and gas–particle partitioning modeling

pH is defined as the negative logarithm of the hydronium ion (H$_3$O$^+$) activity in an aqueous solution. Hereafter we denote H$_3$O$^+$ as H$^+$ for simplicity, while recognizing that the hydronated hydrogen ion is rare in aqueous solutions. pH is given by

$$\text{pH} = -\log_{10} \gamma_{H^+} H^+_\text{aq} = -\log_{10} \frac{1000 \gamma_{H^+} H^+_\text{air}}{W_i + W_o} \approx$$

$$- \log_{10} \frac{1000 \gamma_{H^+} H^+_\text{air}}{W_i},$$

(1)

where $\gamma_{H^+}$ is the hydronium ion activity coefficient (in this case assumed = 1), $H^+_\text{aq}$ (mole L$^{-1}$) the hydronium ion concentration in particle liquid water, $H^+_\text{air}$ (μg m$^{-3}$) the hydronium ion concentration per volume of air, and $W_i$ and $W_o$ (μg m$^{-3}$) the bulk particle water concentrations associated with inorganic and organic species, respectively. $W_o$ can be calculated by Eq. (5) in Guo et al. (2015). Both $H^+_\text{air}$ and $W_i$ are outputs of the thermodynamic model, ISORROPIA-II,
which was used to determine the composition and phase state of an NH$_4^+$−SO$_4^{2−}$−NO$_3^−$−water inorganic aerosol in thermodynamic equilibrium with its corresponding gases. (In some cases Cl$^−$–Na$^+$−K$^+$ were also included). A similar approach has been used in several studies for contrasting summer and winter conditions in the eastern US (Guo et al., 2015, 2016; Weber et al., 2016) and eastern Mediterranean (Bougiatioti et al., 2016, 2017).

In previous studies, the effect of $W_o$ on pH has been investigated and found to be minor (Guo et al., 2015). pH in this study based solely on $W_i$ was 0.15–0.23 units systematically lower than pH predicted with total particle water ($W_i + W_o$) and highly correlated ($R^2 = 0.97$). In this study, ΔpH values of +0.12 and +0.19 units were estimated when including $W_o$ based on average and maximum organic hygroscopic parameter $κ_{org}$ of 0.13 and 0.23, respectively (Mei et al., 2013). Sensitivity of pH to effects of $W_o$ are smaller in CalNex due to smaller fractions of $W_o$ to total particle water (21 %) compared to that found in SOAS (Southern Oxidant and Aerosol Study; 35 %). Given this relatively small deviation (on average 0.12 unit), we report pH only considering $W_i$.

ISORROPIA-II was run assuming particles were “metastable” with no solid precipitates (H$^+$ is not stable in an effloresced aerosol) – a reasonable assumption considering the high RH range observed during this study (mean ± SD of RH = 79 ± 17 %). In our previous pH studies, we only considered data for RH between 20 and 95 %. At low RH (e.g., <20 %), aerosols are less likely to be in a completely liquid state (Ansari and Pandis, 2000; Malm and Day, 2001; Fountoukis and Nenes, 2007; Bertram et al., 2011), and the “glassy” SOA may impede the partitioning of semi-volatile species due to decreased diffusion in the particles (Ye et al., 2016), and uncertainties in predicted pH are expected to be large due to uncertain activity coefficients associated with highly concentrated solutions (Fountoukis et al., 2009). At RH >95 %, large pH uncertainty is introduced due to the exponential growth in particle liquid water with RH and propagation of RH sensor uncertainties (Malm and Day, 2001; Guo et al., 2015). The CalNex RH ranged from 22 to 100 %; therefore periods for RH above 95 % were excluded.

The model was also run in “forward” mode, which calculates the gas–particle equilibrium partitioning concentrations based on the input of total concentration of a species (i.e., gas + particle). Use of total species as ISORROPIA-II inputs produces substantially better predictions compared with only particle-phase concentration inputs in either “forward” or “reverse” modes since in the former cases (“forward” mode with only particle-phase input), particle-phase semi-volatile species concentration is underpredicted due to some fraction repartitioned into the gas phase in the model, and in the latter cases (“reverse” mode), measurement errors often result in large model biases in pH (Hennigan et al., 2015).

The predicted gas- or particle-phase semi-volatile compounds can be compared to measurements for validating the thermodynamic calculations. Possible partitioning pairs for ISORROPIA-II are HNO$_3$–NO$_3^−$, NH$_3$–NH$_4^+$, and HCl–Cl$^−$ (discussed in Sect. 3.2 and 3.3). This method is effective when gas- and particle-phase components have substantial fractions in both the gas and particle phases. For example, in the southeastern US in summer, ammonia partitioning (gas/(gas + particle)) varied between 10 and 80 %, whereas nitric acid partitioning was mostly near 80 % and PM$_1$ nitrate level close to the detection limit (larger uncertainty), making ammonia partitioning much more useful than nitric acid for evaluating thermodynamic models (Guo et al., 2015). In contrast, nitric acid partitioning ranged from 0 to 100 % (average ~50 %) and so was used to evaluate pH predicted in the northeastern US in the cold season (Guo et al., 2016).

Using gas–particle phase partitioning to constrain particle pH can be complicated by the presence of multiple phases within the particle, which may distribute inorganic species amongst multiple phases, each with their own water activity and hence inorganic concentration. Lab studies show that liquid–liquid phase separations are always observed at O : C (organic aerosol atomic O to C ratio) ≤0.5, whereas no phase separations occur for O : C ≥0.8 (Bertram et al., 2011; Song et al., 2012; You et al., 2013, 2014). The likelihood for phase separation decreases at higher RH (0.5 < O : C < 0.8) and only has a weak dependence on T (Schill and Tolbert, 2013; You and Bertram, 2015). For conditions during SOAS (O : C = 0.75 ± 0.12, RH = 74 ± 16 %, T = 25 ± 3 °C), we found that thermodynamic calculations accurately predicted bulk particle water and ammonia partitioning over the complete T (18 to 33 °C) and RH (36 to 96 %) ranges (Guo et al., 2015). During the Wintertime Investigation of Transport, Emissions, and Reactivity (WINTER) aircraft study (O : C = 0.78 ± 0.11, T = 0 ± 8 °C), we found that HNO$_3$–NO$_3^−$ partitioning was accurately predicted for RH >60 % (Guo et al., 2016). Compared to SOAS and WINTER (both in the eastern US), in this study the smaller O : C (0.52 ± 0.10) (Hayes et al., 2015) means we cannot outright exclude the possibility of phase separation, but very high RH (79 ± 17 %) makes it less likely.

In running ISORROPIA-II to predict pH and semi-volatile species partitioning, it is also assumed that the particles are internally mixed, that pH does not vary with particle size (i.e., bulk properties represent the overall aerosol pH), and that the ambient aerosols and gases are in thermodynamic equilibrium. For the WINTER study, which included measurements over coastal and marine areas, we found that PM$_1$ pH was accurately predicted with only particle-phase SO$_4^{2−}$, NO$_3^−$, and NH$_4^+$ (and gas-phase HNO$_3$), whereas sea salt components had some, but generally small, effects on the prediction of particle pH (except on rare occasions when the mole fraction of NaCl to total soluble inorganic ions was greater than 50 %) (Guo et al., 2016). The mixing state of sea salts with SO$_4^{2−}$, NO$_3^−$, and NH$_4^+$ is a critical issue in predicting particle pH in LA and is discussed below.
3 Results

3.1 PM$_1$ and PM$_{2.5}$ inorganic composition and gas–particle partitioning

Time series for various measured parameters during CalNex are shown in Fig. 1. CalNex $T$ and RH were 18 ± 4°C and 79 ± 17% (mean ± SD). During the first half of the campaign, 15 to 29 May, daily maximum $T$ was below 26°C and PM$_1$ (AMS) and PM$_{2.5}$ (PILS-IC) SO$_4^{2-}$, NO$_3^-$, and NH$_4^+$ showed a general decreasing trend (PM$_{2.5}$ NH$_4^+$ data were not available in this period). The second half of the campaign started with a warmer period (30 May to 7 June), with the highest $T$ reaching 29°C. During this period, SO$_4^{2-}$, NO$_3^-$, NH$_4^+$, and HNO$_3$ were significantly higher than the first half and reached campaign maximum of 9.7 µg m$^{-3}$ PM$_1$ SO$_4^{2-}$, 20.1 µg m$^{-3}$ PM$_1$ NO$_3^-$, 9.6 µg m$^{-3}$ PM$_{2.5}$ NH$_4^+$, and 33.1 µg m$^{-3}$ HNO$_3$. The peak concentrations of combined PM$_1$ inorganics alone exceeded the National Ambient Air Quality Standards (NAAQS) PM$_{2.5}$ 24 h limit of 35 µg m$^{-3}$ (https://www.epa.gov/criteria-air-pollutants/naaqs-table), with the largest contribution from NO$_3^-$. Gaseous components other than HNO$_3$, such as NH$_3$ and HCl, showed different patterns than the above species, indicating different sources and sinks.

PM$_1$ (AMS) SO$_4^{2-}$, NO$_3^-$, and NH$_4^+$ were highly correlated with PM$_{2.5}$ (PILS-IC) measurements, with $R^2$ between 0.8 and 0.9 (Fig. S1). PM$_1$ / PM$_{2.5}$ mass ratios were 97.5 ± 5.4% for SO$_4^{2-}$, 63.5 ± 22.1% for NO$_3^-$, and 92.3 ± 9.9% for NH$_4^+$. Note that the mass fractions for SO$_4^{2-}$ and NO$_3^-$ are campaign averages, but NH$_4^+$ mass fraction is only for the last week when PM$_{2.5}$ NH$_4^+$ was available. Nearly 40% of the NO$_3^-$ was found between 1 and 2.5 µm, whereas most SO$_4^{2-}$ and NH$_4^+$ were associated with PM$_1$. (Ratios based on regression slopes shown in Fig. S1.) Hence, the NO$_3^-$ for the 1 to 2.5 µm size range was likely associated with some non-volatile cations, such as Na$^+$ and K$^+$ instead of NH$_4^+$. The PM$_{2.5}$ Cl$^-$ / Na$^+$ molar ratio was 0.5 ± 0.2. Given that these species are mainly transported from the coastal regions in the form of sea salts (NaCl), this indicates that roughly half of the Na$^+$ had reacted with HNO$_3$ resulting in depleted Cl$^-$ (Robbins et al., 1959; Langer et al., 1997). These results are consistent with the analysis of Hayes et al. (2013).

Measurements of semi-volatile particle (NO$_2^-$, NH$_4^+$, Cl$^-$) and corresponding inorganic gases (HNO$_3$, NH$_3$, HCl) are also investigated. The gas–particle phase partitioning is described as the particle-phase mass concentration divided by the total mass concentration (gas + particle), e.g., $\varepsilon$(NO$_3^-$) = NO$_3^-$ / (HNO$_3$+NO$_2^-$). The campaign average partitioning ratios were as follows: for PM$_1$ $\varepsilon$(NO$_3^-$) = 39 ± 16%, PM$_1$ $\varepsilon$(NH$_3^+$) = 56 ± 26%, PM$_{2.5}$ $\varepsilon$(NO$_3^-$) = 54 ± 10%, and PM$_{2.5}$ $\varepsilon$(Cl$^-$) = 39 ± 26%. All partitioning ratios are near 50%, making them useful for assessing pH predictions by comparing measured versus ISORROPIA-predicted ratios.

3.2 PM$_1$ pH prediction and verification

PM$_1$ pH was determined to be on average (±SD) 1.9 ± 0.5 for the complete study, for model inputs of PM$_1$ inorganic AMS-measured components SO$_4^{2-}$, NO$_3^-$, and NH$_4^+$, gases HNO$_3$ and NH$_3$. Although the CalNex ground site was influenced by sea salt components, Na$^+$ and Cl$^-$ were not included in the PM$_1$ pH since NaCl is typically found mainly at sizes above 1 µm and the mixing states of PM$_1$ NaCl with SO$_4^{2-}$, NO$_3^-$, and NH$_4^+$ remains to be investigated. PALMS single-particle data indicated that for the particle size range with D$_{\text{vac}}$ (dry volume-equivalent diameter) between 0.15 and 780 nm (~1 µm), by number 27% of PM$_1$ sea salt particles (Na$^+$-rich particles without crustal materials) had observable NO$_3^-$ signals and by mass only 12% of PM$_1$ are sea salt types.
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Table 1. Number fraction of sea salt particles with observable nitrate signals and mass fraction of sea salt particles to total mass in two size ranges, 0.15–1 and 1–2.5 μm. Sea salt particles are identified as Na\(^+\)-rich particles without crustal elements. The number fraction was determined directly from the PALMS data and the mass fraction was calculated based on the number fractions and size distributions, assuming dry particle densities.

<table>
<thead>
<tr>
<th>Particle size, μm</th>
<th>Number fraction of sea salt particles with observable nitrate signal</th>
<th>Sea salt particles mass fraction to total</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.15–1</td>
<td>27 %</td>
<td>12 %</td>
</tr>
<tr>
<td>1–2.5</td>
<td>85 %</td>
<td>63 %</td>
</tr>
</tbody>
</table>

(Table 1), suggesting external mixing of NaCl with NO\(_3^-\) is the main form. For the 1 to 2.5 μm size range the number and mass fractions were 85 and 63 %, respectively.

We assess predicted pH from the thermodynamic model by comparing predicted and measured gas–particle partitioning of NH\(_3^-\)–NH\(_4^+\) and HNO\(_3^-\)–NO\(_3^-\). Comparison of HNO\(_3^-\)–NO\(_3^-\) and NH\(_3^-\)–NH\(_4^+\) predictions to their measured values is shown as Fig. 2. Gas-phase HNO\(_3\) and NH\(_3\) and particle-phase NH\(_4^+\) are on average within 10 % and highly correlated, \(R^2 = 0.8\). Despite a high correlation \((R^2 = 0.76)\), ISORROPIA-II predicted particle-phase NO\(_3^-\) is systematically higher than measured, with a regression slope of 1.28. Two bands are observed that are related to RH or time of day (Fig. 2b); for mid-range RH (50–70 %) daytime data, ISORROPIA-II slightly underpredicts NO\(_3^-\) and for high RH (90 %) nighttime data, NO\(_3^-\) is overpredicted. This leads to a regression slope of 2 comparing predicted to measured ε(NO\(_3^-\)) and a large intercept (Fig. 2c, also see Fig. S5). In contrast, predicted versus measured ε(NH\(_4^+\)) is close to 1 : 1 and highly correlated \((R^2 = 0.8\) Fig. 2f), and there is a much weaker systematic variability related to RH or time of day.

3.3 PM\(_{2.5}\) pH prediction and verification

Predicting the bulk pH of PM\(_{2.5}\) is more complicated since the particles larger than 1 μm in the PM\(_{2.5}\) fraction are not necessarily in equilibrium due to increased timescale for equilibration (>20–30 min) (Fountoukis et al., 2009), especially in an environment with rapidly changing concentrations of key species (e.g., HNO\(_3\)). This leads to greater uncertainty when predicting partitioning of semi-volatile species. Furthermore, the pH analysis now needs to consider sea salt components \((\text{SO}_4^{2-}–\text{NO}_3^-–\text{NH}_4^+–\text{Na}^+–\text{Cl}^-–\text{K}^+–\text{HNO}_3–\text{NH}_3–\text{HCl} \text{ system})\). Whereas single-particle PALMS data suggested that only a small fraction (27 %) of the sea salt particles less than 1 μm were internally mixed with nitrate, for sizes between 1 and 2.5 μm, the majority (85 %) were. Overall we find bulk PM\(_{2.5}\) pH considering sea salt components higher than PM\(_1\) by 0.8 units on average. For data from the last week of the study (i.e., period of PILS data that in-
Includes measurements of sea salt components, PM$_1$ pH was 1.9 ± 0.4, similar to that of the complete study (1.9 ± 0.5), and in contrast to an average PM$_{2.5}$ pH of 2.7 ± 0.3. A comparison of the pH is shown in Fig. 3. Addition of non-volatile Na$^+$ and K$^+$ increases pH. Na$^+$ is the more important cation in this case, as the Na$^+$ levels were several times larger than K$^+$ (0.77 ± 0.39 vs. 0.20 ± 0.09 µg m$^{-3}$, or 0.33 vs. 0.05 mol m$^{-3}$).

To examine the effects of sea salt components on the thermodynamic predictions, we compared the observed to measured partitioning of PM$_{2.5}$ semi-volatile species in Fig. 4. ISORROPIA-II was run with two differing inputs, one with Na$^+$, Cl$^-$, and K$^+$ and the other run without these ions. In both cases only PILS-IC PM$_{2.5}$ data were used. All other input parameters, including NH$_4^+$, SO$_4^{2-}$, NO$_3^-$, RH, and $T$, were the same. Figure 4 shows that inclusion of Na$^+$, Cl$^-$, and K$^+$ improves the prediction of HNO$_3$–NO$_3^-$ partitioning. For HNO$_3$, NO$_3^-$, and ε(NO$_3^-$), predicted levels are somewhat closer to the measurements and the scatter in the data is reduced. However, like the PM$_1$ analysis above, the slope between predicted and measured ε(NO$_3^-$), 2.4, is significantly larger than 1. The deviation is again related to RH, resulting from a diurnal dependence. Unlike the HNO$_3$–NO$_3^-$ partitioning, NH$_3$–NH$_4^+$ partitioning is not as sensitive to inclusion of Na$^+$, Cl$^-$, and K$^+$. Overall, ε(NH$_4^+$) is on average underestimated by 17% compared to measured (average ratio). HCl–Cl$^-$ partitioning is well captured by ISORROPIA-II with regression slopes of 1.05, 0.95, and 1.14 and $R^2$ of 0.98, 0.84, and 0.81 for HCl, Cl$^-$, and ε(Cl$^-$), respectively.

An analytical calculation of HNO$_3$–NO$_3^-$ partitioning can also be used to assess whether the shift of one pH unit caused by Na$^+$, Cl$^-$, and K$^+$ is consistent with observed nitric acid partitioning for PM$_{2.5}$. The analytical calculation is based on Eq. (3) in Guo et al. (2016) and a detailed equation derivation can be found in Supplement Sect. S2. To minimize the effects of $T$ and $W_i$ variability on partitioning, and focus on the role of pH, data for a relatively small $T$ (17–23°C) and $W_i$ (5–15 µg m$^{-3}$) range were selected. Predicted activity coefficients, $\gamma_{H^+}$ and $\gamma_{NO_3^-}$, extracted from ISORROPIA-II, were input in the analytical calculation to account for solution non-ideality. The product of the activity coefficients, $\gamma_{H^+}\gamma_{NO_3^-}$, was on average 0.28 with Na$^+$, Cl$^-$, and K$^+$ in the model and 0.19 without Na$^+$, Cl$^-$, and K$^+$ ($\gamma_{H^+}\gamma_{NO_3^-}$ is smaller without Na$^+$, Cl$^-$, and K$^+$ due to less predicted $W_i$, thus overall larger ionic strength). The analytical calculated S curves are plotted with the measurements and ISORROPIA-II predictions in Fig. 5. As noted, including these components changes the activity coefficient $\gamma_{H^+}\gamma_{NO_3^-}$ (as can be seen by the difference in the two curves in Fig. 5) and also slightly increases the liquid water. But the most important effect is reducing H$^+$, resulting in a shift to higher pH. This analysis also shows that the measured ε(NO$_3^-$) comes into better agreement with its theoretical S curve for the SO$_4^{2-}$–NO$_3^-$–NH$_4^+$–Na$^+$–Cl$^-$–K$^+$–HNO$_3$–NH$_3$–HCl system (red points closer to red curve compared to blue points and blue curve). Similar to ε(NO$_3^-$), measured ε(Cl$^-$) is also found to be in good agreement with the S curve (Fig. S2).

### 3.4 Average diurnal trends

- **PM$_1$**: The diurnal variations in $T$, RH, pH, LWC, and HNO$_3$–NO$_3^-$, and NH$_4^+$–NH$_4^+$ partitioning are shown together in Fig. 6. Due to the inverse variation between $T$ and RH diurnal patterns, predicted particle water ($W_i$) reached a daily maximum before dawn and decreased rapidly with RH after sunrise. To be consistent with pH, particle water data are not plotted for RH above 95%. (For RH above 95%, particle water increased continuously at night until reaching the daily highest RH at 05:30.) Between 13:00 and 20:00 LT (local time), $W_i$ stayed consistently low (~5 µg m$^{-3}$). PM$_1$ pH generally tracked liquid water. pH was lower in the daytime due to less liquid water, reaching a minimum value of 1.6 at approximately 16:00. After that, pH continued to increase to its daily maximum of 2.4 at midnight, tracking the liquid water concentrations. This pH diurnal pattern is similar to that observed in the southeastern US (Guo et al., 2015). A very large peak in gaseous HNO$_3$ was observed during the day produced from rapid photochemical reactions of NO$_x$ with the hydroxyl radical in the LA outflows (Veres et al., 2011). In contrast, NO$_3^-$ peaked at dawn under conditions of low $T$ and high RH, which favored nitrate condensation (S curves shifted to lower pH; see Fig. S3 and S4), consistent with a previous study in Mexico City (Hennigan et al., 2008). There was also a small peak in NO$_3^-$ near midday when HNO$_3$ peaked, simply due to the large amount of fresh HNO$_3$, despite the trend of a continuous ε(NO$_3^-$) decrease. The diurnal variation in ε(NO$_3^-$) was similar to liquid water and pH. The PM$_1$ NH$_4^+$ trend followed NO$_3^-$ and all inorganic species (NH$_4^+$, SO$_4^{2-}$, NO$_3^-$) exhibited a peak during the day. ε(NH$_4^+$) had a similar diurnal pattern to ε(NO$_3^-$).

- **PM$_{2.5}$**: Fig. 7 compares the diurnal trends of PM$_1$ pH (AMS data, SO$_4^{2-}$, NO$_3^-$, NH$_4^+$) and PM$_{2.5}$ pH (PILS-IC SO$_4^{2-}$, NO$_3^-$, NH$_4^+$, Na$^+$, Cl$^-$, K$^+$) for the last week of the study. PM$_1$ pH had a similar profile to that for the complete study – lower in the afternoon and following the $W_i$ trend – whereas the PM$_{2.5}$ diurnal pH trend was nearly flat. The difference was caused by more non-volatile Na$^+$ than volatile Cl$^-$ (PM$_{2.5}$ Cl$^-$ / Na$^+$ molar ratio was 0.5 ± 0.2), which is related to the HNO$_3$ and HCl trends. Sea salt components in the 1 to 2.5 µm size range react with the daytime high HNO$_3$ forming NO$_3^-$ and gas-phase HCl (simplified as Cl$^-$→M→HNO$_3$→NO$_3^-$→HCl) (Robbins et al., 1959;
Figure 4. Intercomparisons of predicted and measured gas–particle phase partitioning for PM$_{2.5}$ particles for two scenarios: ISORROPIA-II input without (left) and with (right) Na$^+$, Cl$^-$ (and HCl), and K$^+$. The other input $\text{SO}_4^{2-}$, $\text{NO}_3^-$, $\text{NH}_4^+$, $\text{NH}_3$, $\text{HNO}_3$, RH, and $T$ are the same in the two cases. The PM$_{2.5}$ data for the last week during CalNex study are shown above. ODR fits are applied.

Figure 5. Comparison of measured $\varepsilon$($\text{NO}_3^-$) for PM$_{2.5}$ (data points) to S curves, which are predicted from theory and include activity coefficients from ISORROPIA-II. The product of the activity coefficients, $\gamma_{\text{H}^+}\gamma_{\text{NO}_3^-}$, was on average 0.28 with Na$^+$, Cl$^-$, and K$^+$ and 0.19 without Na$^+$, Cl$^-$, and K$^+$. In both cases pH (data points) is predicted by ISORROPIA-II.

Langer et al., 1997). This “chloride depletion” is a result of the higher volatility of HCl versus HNO$_3$ in the deliquesced sea salt aerosol (Nenes et al., 1998; Fountoukis and Nenes, 2007). The process can partly account for the large HCl peak and stronger Cl$^-$ depletion (wider gap between Na$^+$ and Cl$^-$) during the day, coinciding with high HNO$_3$ and low $W_i$ (evaporation of HCl occurs during evaporation of droplets). The slightly higher Na$^+$ generally increased pH due to the added non-volatile cations. These data are consistent with the discussion above indicating that bulk PM$_{2.5}$ pH is higher due to the contributions of sea salt aerosol components solely in the 1 to 2.5 µm range. For smaller particles (PM$_1$), these components do not significantly affect the $\text{SO}_4^{2-}$–$\text{NO}_3^-$–$\text{NH}_4^+$–$\text{HNO}_3$–$\text{NH}_3$ system resulting in a lower bulk PM$_1$ pH. Size-resolved particle pH and solubility of metals, reported in another study, are consistent with these findings (Fang et al., 2017). (Note that an attempt to calculate pH in the PM$_1$ to PM$_{2.5}$ size range was not successful due to highly uncertain data resulting from particle concentrations determined by difference from two separate measurements.)

4 Discussion

4.1 Cause for bias in $\varepsilon$($\text{NO}_3^-$)

The bias between ISORROPIA-predicted and observed nitrate partitioning may be a result of several causes. Since the $\varepsilon$($\text{NO}_3^-$) bias is seen for both PM$_1$ (AMS data) and PM$_{2.5}$ (PILS-IC data), the cause is apparently not associated with a specific aerosol measurement method. For the aerosol measurements, sampling artifacts associated with differences in
indoor–outdoor temperatures that varied with time of day could be one cause for the biases. Sample heating is most likely to occur at night (indoor \( T \) > ambient \( T \)) and can cause semi-volatile NO\(_3^-\) loss, whereas sample cooling during the day (indoor \( T \) < ambient \( T \)) can lead to vapor condensation and higher NO\(_3^-\). Differences in observed versus predicted NO\(_3^-\) are consistent with these trends (Fig. S5); measured NO\(_3^-\) is lower than predicted at night (negative artifact) and higher than predicted during the day (positive artifact). The same will apply to NH\(_3\)-NH\(_4^+\) partitioning, but to a lesser degree due to the addition of non-volatile sulfate. Indoor temperatures were recorded to be fairly constant at \( \sim 25 \, ^\circ\text{C} \) for the AMS trailer (Fig. S6, PILS trailer indoor temperatures are expected to be similar). Aerosol samples were heated by \( \sim 10 \, ^\circ\text{C} \) at night and negligibly during the middle of the day (Table S1). Possible biases due to effects of the altered RH and \( T \) on aerosol measurements were examined by comparing measurements to model results for partitioning of HNO\(_3^-\), NO\(_3^-\), NH\(_3\), and NH\(_4^+\) using sample line versus ambient conditions (Figs. S6 and S7), with all other model inputs the same. We find that ambient RH and \( T \) result in better agreement for NO\(_3^-\), \( \varepsilon(\text{NO}_3^-) \), NH\(_4^+\), and \( \varepsilon(\text{NH}_4^+) \), although some minor effect may be possible. We note that more extreme ambient to sample line temperature differences were experienced during the WINTER aircraft campaign (\( \Delta T \) on average \( +24 \, ^\circ\text{C} \)). In that case, a similar analysis also found no evidence for nitrate loss due to sample line heating (Guo et al., 2016). (Sample-line residence times of 0.5 to 2 s are in the same range for both studies.)

The sampling systems for HNO\(_3\) and NH\(_3\) involved heated sample inlets, which could also lead to bias due to evaporation of aerosol components. Here we only focus on the CIMS sampling line heating (75 \( ^\circ\text{C} \)) since it was more extreme than the NH\(_3\) (40 \( ^\circ\text{C} \)). For the CIMS system, a 75 \( ^\circ\text{C} \) inlet may cause particle nitrate and chloride evaporation (residence time \( = 0.32 \, \text{s} \), Table S1), resulting in over-measurement of
HNO$_3$ and HCl (the species measured by this instrument). To examine whether this is the cause for the differences in predicted and observed $\epsilon$(NO$_3^-$), we compared the partitioning of HNO$_3$–NO$_3^-$ and NH$_3$–NH$_4^+$ based on “corrected” HNO$_3$ assuming 10, 20 and 30% of the measured NO$_3^-$ evaporated in the CIMS inlet (the “corrected” HNO$_3$ is lower by subtracting the various fractions of PM$_1$ NO$_3^-$ from measured HNO$_3$). Comparing the orthogonal regression fitting lines in the Fig. S8, only slight improvements in predicted to measured $\epsilon$(NO$_3^-$) are found for increased NO$_3^-$ loss (Fig. S8c), but all lines converge at the same intercept. At 30% evaporation the comparison becomes worse for both $\epsilon$(NO$_3^-$) and $\epsilon$(NH$_4^+$). We conclude that potential inlet artifacts associated with the aerosol- or gas-phase sampling systems cannot explain the predicted versus measured $\epsilon$(NO$_3^-$) bias. This is consistent with the limited inlet heating and residence times in this study compared to some thermodenuder studies utilizing more extreme conditions (Huffman et al., 2009; Riipinen et al., 2010).

Another possible cause of the $\epsilon$(NO$_3^-$) bias is the effect of a large sea salt coarse mode on fine-mode semi-volatile species. For example, during nighttime, when HNO$_3$ concentrations are much lower relative to daytime, ISORROPIA-II may have overestimated NO$_3^-$ due to ignoring the presence of a larger more neutral coarse mode when calculating fine-mode aerosol concentrations assuming equilibrium. Various studies show the typical time for the fine mode to reach thermodynamic equilibrium is on the order of 20 min (Dassios and Pandis, 1999; Fouliou et al., 2009), whereas timescales for the coarse mode to equilibrate are much larger. The mixing of fine particles with high concentrations of coarse mode, mainly sea salt, particles could disrupt fine-mode equilibrium by the mass transfer of volatile fine-mode NO$_3^-$, via HNO$_3$, to the coarse mode, forming non-volatile salts, such as NaNO$_3$. This mechanism is consistent with the presence of high levels of sea salt in the LA region and the anti-correlation of a NO$_3^-$ prediction bias with HNO$_3$ concentration (Fig. S5). During the daytime, when HNO$_3$ concentrations are high, this bias would be minimal, but at night, when HNO$_3$ concentrations are low, the model, which ignores the presence of the coarse mode, would overpredict NO$_3^-$ concentrations by not considering fine-mode nitrate loss to the coarse mode. Again, partitioning of NH$_3$–NH$_4^+$ would not be affected as much by this process since a fraction of the NH$_4^+$ would be associated with SO$_4^{2-}$. Of the various locations where we have investigated pH, this study has the highest coarse-mode inorganic aerosol concentrations (Sardar et al., 2005, report levels of ~20 µg m$^{-3}$ year round) and is the only one where we have observed this bias (not observed in the eastern US reported by Guo et al., 2016).

4.2 Why is nitrate much higher in LA compared to other regions investigated?

A comparison of pH and related statistics in five field studies is summarized in Table 2. The campaigns are CalNex, SOAS, SENEX, and WINTER, all conducted in the continental US. Also included are results from a study in the eastern Mediterranean (the campaign acronyms are given in Table 2). The SOAS (ground-based) and SENEX (aircraft-based) studies provide an interesting contrast with CalNex, that is, between the southeastern and southwestern US in summertime. WINTER aircraft data add the dimension of seasonal variation (summer versus winter). The eastern Mediterranean data provide a remote European (Crete) and urban (Athens) perspective, and a case where air masses were known to be impacted by biomass burning (BB). All pH values in Table 2 were calculated by ISORROPIA-II in forward metastable mode, but only the US data (SOAS, WINTER, CalNex) used gas–particle phase partitioning to constrain and verify the pH prediction for all the data analyzed. Lack of NH$_3$ or HNO$_3$ data for the eastern Mediterranean means that pH was likely underestimated by ~0.5 units (Bougiatioti et al., 2016). This was verified with a limited set of data when both aerosol and gas-phase data were available. It is noteworthy that, in all studies, pH was low and on average below 3. The eastern US regions are characterized with the lowest pH (~1) throughout the year, from ground level up to 5 km aloft (Guo et al., 2015, 2016: Xu et al., 2016). The highest pH of 2.8 ± 0.6 was consistently observed in biomass-burning-impacted air masses, regardless of season in both the remote and urban eastern Mediterranean and attributed to an abundance of NH$_3$ and fine particle K$^+$ in biomass burning emissions, which raised pH and NO$_3^-$ concentrations significantly (Bougiatioti et al., 2017). Biomass burning also accounted for the high pH, which approaches 3, in Athens during winter.

Comparing LA (CalNex) to the other summertime measurements in the eastern US (SOAS, SENEX), Table 2 shows that the LA ground site had uniquely higher NO$_3^-$ and HNO$_3$ mass concentrations and NO$_3^-$ was the most abundant (by mass) inorganic ion for PM$_1$ or PM$_2.5$. In contrast, in the southeastern US in summertime, SO$_4^{2-}$ was the dominant ion, NO$_3^-$ was only 5–14% of the SO$_4^{2-}$ mass. The higher total NO$_3^-$ (NO$_3^-$ + HNO$_3$) in LA indicates high local NO$_x$ relative to SO$_2$ sources. LA also had roughly 1 unit higher PM$_1$ pH, and much higher NO$_3^-$ concentrations relative to SO$_4^{2-}$. (Total ammonia (NH$_3$ + NH$_4^+$) was also higher, Table 2.) The ratio of total NO$_3^-$ and SO$_4^{2-}$ (sulfate is non-volatile, so total sulfate is equivalent to sulfate) was ~4 for CalNex, compared to 0.3 for SOAS and 0.8 for SENEX. The higher ratio of total NO$_3^-$ to SO$_4^{2-}$ can explain the higher LA pH through coupling of particle composition, hygroscopicity, and acidity. Consider the situation where there is initially a high HNO$_3$ concentration. Some HNO$_3$ will condense onto very acidic particles (e.g., even at low $\epsilon$(NO$_3^-$), with very
Table 2. Comparisons between different studies for particle pH, major inorganic ions and gases and meteorological conditions. All pH values are from ISORROPIA-II run in forward mode. The campaign acronyms other than CalNex stand for Southern Oxidant and Aerosol Study (SOAS), Southeastern Nexus of Air Quality and Climate (SENEX), and Wintertime Investigation of Transport, Emissions, and Reactivity (WINTER).

<table>
<thead>
<tr>
<th>Campaign</th>
<th>CalNex</th>
<th>SOAS</th>
<th>SENEX</th>
<th>WINTER</th>
<th>Studies in the eastern Mediterranean</th>
</tr>
</thead>
<tbody>
<tr>
<td>Type</td>
<td>Ground</td>
<td>Ground</td>
<td>Aircraft</td>
<td>Aircraft</td>
<td>Ground</td>
</tr>
<tr>
<td>PM cut size</td>
<td>PM$_{1}$</td>
<td>PM$_{2.5}$</td>
<td>PM$_{1}$</td>
<td>PM$_{1}$</td>
<td>PM$_{1}$</td>
</tr>
<tr>
<td>Season</td>
<td>(Early) Summer</td>
<td>Summer</td>
<td>Summer</td>
<td>Winter</td>
<td>Summer–autumn</td>
</tr>
<tr>
<td>Region/location</td>
<td>SW US</td>
<td>SE US</td>
<td>SE US</td>
<td>NE US</td>
<td>Crete, Greece</td>
</tr>
<tr>
<td>SO$_{4}^{2-}$, µg m$^{-3}$</td>
<td>2.86 ± 1.70</td>
<td>1.88 ± 0.69</td>
<td>1.73 ± 1.21</td>
<td>2.05 ± 0.80</td>
<td>1.02 ± 0.77</td>
</tr>
<tr>
<td>NO$_{3}^{-}$, µg m$^{-3}$</td>
<td>3.58 ± 3.65</td>
<td>3.74 ± 1.53</td>
<td>0.08 ± 0.08</td>
<td>0.28 ± 0.09</td>
<td>0.80 ± 1.03</td>
</tr>
<tr>
<td>HNO$_{3}$, µg m$^{-3}$</td>
<td>6.65 ± 7.03</td>
<td>4.45 ± 3.59</td>
<td>0.36 ± 0.14</td>
<td>1.35 ± 0.66</td>
<td>1.41 ± 1.83</td>
</tr>
<tr>
<td>$\varepsilon$(NO$_{3}^{-}$)</td>
<td>39 ± 16%</td>
<td>51 ± 18%</td>
<td>22 ± 16%</td>
<td>18 ± 6%</td>
<td>37 ± 26%</td>
</tr>
<tr>
<td>Total NO$_{3}^{-}$, µg m$^{-3}$</td>
<td>10.22 ± 9.74</td>
<td>8.19 ± 3.89</td>
<td>0.45 ± 0.26</td>
<td>1.63 ± 0.70</td>
<td>0.22 ± 0.21</td>
</tr>
<tr>
<td>NH$_{4}^{+}$, µg m$^{-3}$</td>
<td>2.06 ± 1.67</td>
<td>1.79 ± 0.65</td>
<td>0.46 ± 0.34</td>
<td>1.06 ± 0.25</td>
<td>0.50 ± 0.43</td>
</tr>
<tr>
<td>NH$_{3}$, µg m$^{-3}$</td>
<td>1.37 ± 0.90</td>
<td>0.75 ± 0.61</td>
<td>0.39 ± 0.25</td>
<td>0.12 ± 0.19</td>
<td>–</td>
</tr>
<tr>
<td>$\varepsilon$(NH$_{4}^{+}$)</td>
<td>45 ± 25%</td>
<td>71 ± 19%</td>
<td>50 ± 25%</td>
<td>92 ± 11%</td>
<td>–</td>
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<tr>
<td>Total NH$_{4}^{+}$, µg m$^{-3}$</td>
<td>3.44 ± 1.81</td>
<td>2.54 ± 0.89</td>
<td>0.78 ± 0.50</td>
<td>1.17 ± 0.81</td>
<td>–</td>
</tr>
<tr>
<td>Na$^{+}$, µg m$^{-3}$</td>
<td>–</td>
<td>0.77 ± 0.39</td>
<td>0.03 ± 0.07</td>
<td>–</td>
<td>0.23 ± 0.46$^d$</td>
</tr>
<tr>
<td>Cl$^{-}$, µg m$^{-3}$</td>
<td>–</td>
<td>0.64 ± 0.48</td>
<td>0.02 ± 0.03</td>
<td>–</td>
<td>0.34 ± 0.38$^d$</td>
</tr>
<tr>
<td>RH, %</td>
<td>79 ± 17</td>
<td>87 ± 9</td>
<td>74 ± 16</td>
<td>72 ± 9</td>
<td>58 ± 19</td>
</tr>
<tr>
<td>$T$, °C</td>
<td>18 ± 4</td>
<td>18 ± 3</td>
<td>25 ± 3</td>
<td>22 ± 3</td>
<td>0 ± 8</td>
</tr>
<tr>
<td>pH</td>
<td>1.9 ± 0.5</td>
<td>2.7 ± 0.3</td>
<td>0.9 ± 0.6</td>
<td>1.1 ± 0.4</td>
<td>0.8 ± 1.0</td>
</tr>
</tbody>
</table>


$^a$Only the last week of CalNex. $^b$PM$_{2.5}$ was sampled in the first half and PM$_{1}$ sampled in the second half of the study; various parameters were similar in both cases, crustal components were higher, but overall generally low so differences had minor effects, e.g., PM$_{2.5}$ Na$^{+}$ was 0.06 ± 0.09 and PM$_{1}$ Na$^{+}$ was 0.01 ± 0.01 µg m$^{-3}$. $^c$Only one flight (16 June 2013) statistic from the reference is shown. $^d$Externally mixed, thus not included in pH calculation. $^e$Estimated from offline measurement. $^f$A. Bougiatioti, personal communication, 2016. $^g$Averaged from identified biomass burning (BB) plumes from Crete and Athens studies due to the similar pH. $^h$K$^{+}$ was 0.36 ± 0.38 µg m$^{-3}$. $^i$Estimated by the S curve shift.
to SOAS and SENEX due to higher total NO$_3^-$ relative to SO$_4^{2-}$, which leads to more nitrate formation through feedbacks involving particle water and pH, as discussed above. This can also be seen in Fig. 8b.

$\varepsilon$(NH$_4^+$) S curves and data from the field studies are shown in Fig. 8c and d. Note that the $\varepsilon$(NO$_3^-$) and $\varepsilon$(NH$_4^+$) face opposite directions since it is acid versus base. NH$_3$ data were not available during WINTER so no measured $\varepsilon$(NH$_4^+$) points were plotted at 0°C and no comparisons as a function of T can be made. At 20°C, the $\varepsilon$(NH$_4^+$) of CalNex, SOAS, and SENEX all converge around one S curve, which was calculated assuming $\gamma_{H^+-NO_3^-} = 1$. The lower pH of SOAS and SENEX relative to CalNex resulted in generally higher $\varepsilon$(NH$_4^+$), and more ammonia partitioned to the particle phase. The data and predicted $\varepsilon$(NH$_4^+$) are in fairly good agreement.

S curves have significant utility for understanding how T, RH, and pH affect partitioning (e.g., see Fig. S3 and S4), but we also note that they can be used to estimate activity coefficients based on partitioning data, which may be particularly useful in situations where the data are not available from the literature (e.g., organic acids above a mixture of inorganic and organics).

5 Summary

pH of PM$_1$ and PM$_{2.5}$ and the semi-volatile gas–particle partitioning of HNO$_3$–NO$_3^-$, NH$_3$–NH$_4^+$, and HCl–Cl$^-$ in the LA Basin during the 2010 CalNex study were investigated. As a coastal urban site impacted by high sea salt aerosol components and high total nitrate (HNO$_3$+NO$_3^-$) and ammonia levels, this study provided a contrast to pH we have reported in the eastern US and eastern Mediterranean.

PM$_1$ single-particle analysis showed that 73% (by number) of PM$_1$ sea salt particles did not contain nitrate and sea salt type particles only contributed to 12% to PM$_1$ mass. Therefore, PM$_1$ pH was predicted solely on the SO$_4^{2-}$–NO$_3^-$–NH$_4^+$–HNO$_3$–NH$_3$ system (ISORROPIA-II inputs). This approach provided good agreement between observed and model-predicted partitioning of NH$_3$–NH$_4^+$ and HNO$_3$–NO$_3^-$, although a bias in $\varepsilon$(NO$_3^-$) was observed that depended on RH (day vs. night). Altering gas and particle temperatures from ambient due to the sampling configurations did not explain the bias. A likely cause is the loss of fine-mode NO$_3^-$ to coarse-mode sea salt, mainly at night, when HNO$_3$ concentrations are low, which is not considered by the thermodynamic model used to predict fine-mode concentrations.

The study mean (±SD) PM$_1$ pH in the LA Basin was 1.9 ± 0.5, roughly one unit higher than the average pH observed in summer in the southeastern US (0.9 ± 0.6 and 1.1 ± 0.4), despite similar RH and T ranges and both calculated for the SO$_4^{2-}$–NO$_3^-$–NH$_4^+$–HNO$_3$–NH$_3$ system. The cause may be much higher total nitrate concentrations (high NO$_3$ and proximity to locations of HNO$_3$ formation) relative to sulfate in LA. For example, high levels of HNO$_3$ can lead to at least some particulate NO$_3^-$, even if conditions are such that NO$_3^-$ partitioning is low (e.g., $\varepsilon$(NO$_3^-$) is small). NO$_3^-$ increases the particle water, which raises pH by dilution of H$^+$, allowing more partitioning (i.e., increases $\varepsilon$(NO$_3^-$)). When non-volatile sulfate dominates over total nitrate, sulfate controls liquid water and this effect is not observed. The
complex interactions between pH, LWC, \( T \), and \( \text{NO}_3^- \) are clearly illustrated by analytical (“S” curve) analyses.

Single-particle analysis showed that 85\% by number of sea salt particles in the 1 to 2.5\( \mu \)m nominal range contained nitrate and that the model-predicted partitioning of \( \text{HNO}_3^-\text{NO}_3^- \) for \( \text{PM}_{2.5} \) agreed better with the observed partitioning when sea salt components were included \((\text{SO}_4^{2-}\text{NO}_3^-\text{NH}_4^+\text{Na}^+\text{Cl}^-\text{K}^+\text{HNO}_3^-\text{NH}_3=\text{HCl} \) system). Bulk \( \text{PM}_{2.5} \) pH was 2.7 \( \pm \) 0.3, whereas for the same time period \( \text{PM}_1 \) pH was 1.9 \( \pm \) 0.4 and the diurnal pH profiles of \( \text{PM}_1 \) and \( \text{PM}_{2.5} \) also differed, all apparently due to the influence of sea salt aerosols.

The CalNex data provide unique contrast to pH reported in other regions and demonstrates the complex interactions between pH and emissions. They also support the general application of \( \text{SO}_4^{2-}\text{NO}_3^-\text{H}_2\text{O}=\text{NH}_3=\text{HNO}_3^-\text{NH}_4^+\text{Na}^+\text{Cl}^-\text{K}^+\text{HNO}_3^-\text{NH}_3=\text{HCl} \) system for predictions of \( \text{PM}_1 \) pH and gas–particle phase partitioning without considering sea salts or crustal elements, useful for regional or global modeling. They further illustrate that fine particles have surprisingly low pH in many locations, which has significant effects on the many environmental impacts of fine particles.

Data availability. The data were collected during the 2010 CalNex ground campaign in LA and open to the public at the NOAA website, https://esrl.noaa.gov/csd/groups/csd7/measurements/2010calnex/Ground/DataDownload/ (NOAA, 2016). The particle pH data are accessible by contacting the corresponding authors.

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Competing interests. The authors declare that they have no conflict of interest.

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