



Urban aerosol chemistry at a land-water transition site during summer – Part 2: Aerosol pH and liquid water content

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1 Abstract

2 3 Particle acidity (aerosol pH) is an important driver of atmospheric chemical processes 4 and the resulting effects on human and environmental health. Understanding the factors that 5 control aerosol pH is critical when enacting control strategies targeting specific outcomes. This 6 study characterizes aerosol pH at a land-water transition site near Baltimore, MD during summer 7 2018 as part of the second Ozone Water-Land Environmental Transition Study (OWLETS-2) 8 field campaign. Inorganic fine mode aerosol composition, gas-phase NH₃ measurements, and all 9 relevant meteorological parameters were used to characterize the effects of temperature, aerosol 10 liquid water (ALW), and composition on predictions of aerosol pH. Temperature, the factor linked to the control of NH₃ partitioning, was found to have the most significant effect on aerosol 11 pH during OWLETS-2. Overall, pH varied with temperature at a rate of -0.047 K⁻¹ across all 12 13 observations, though the sensitivity was -0.085 K⁻¹ for temperatures > 293 K. ALW had a minor





14	effect on pH, except at the lowest ALW levels (< 1 μ g m ⁻³) which caused a significant increase
15	in aerosol acidity (decrease in pH). Aerosol pH was generally insensitive to composition (SO_4^{2-} ,
16	$SO_4^{2-}:NH_4^+$, Tot-NH ₃ = NH ₃ + NH ₄ ⁺), consistent with recent studies in other locations. In a
17	companion paper, the sources of episodic NH_3 events (95 th percentile concentrations, $NH_3 > 7.96$
18	μ g m ⁻³) during the study are analyzed; aerosol pH was higher by only ~0.1-0.2 pH units during
19	these events compared to the study mean. A case study was analyzed to characterize the
20	response of aerosol pH to nonvolatile cations (NVCs) during a period strongly influenced by
21	primary Chesapeake Bay emissions. Depending on the method used, aerosol pH was estimated
22	to be either weakly (~0.1 pH unit change based on NH ₃ partitioning calculation) or strongly
23	(~1.4 pH unit change based on ISORROPIA thermodynamic model predictions) affected by
24	NVCs. The case study suggests a strong pH gradient with size during the event and underscores
25	the need to evaluate assumptions of aerosol mixing state applied to pH calculations. Unique
26	features of this study, including the urban land-water transition site and the strong influence of
27	NH3 emissions from both agricultural and industrial sources, add to the understanding of aerosol
28	pH and its controlling factors in diverse environments.





29 30	1 Introduction
31	The acidity, or pH, of atmospheric aerosols affects the chemical and physical properties
32	of airborne particles, and thus, their impacts on climate and health (Pye et al., 2020). The gas-
33	particle partitioning of semi-volatile acidic and basic compounds - notably NH3, HNO3, HCl,
34	and organic acids – depends in part on aerosol pH, which directly affects the particulate matter
35	(PM) mass concentration (Nenes et al., 2020). The solubility of many particulate components is
36	pH-dependent, including metals and nutrients, with implications for particle toxicity and nutrient
37	deposition to ecosystems (Fang et al., 2017; Kanakidou et al., 2016). The optical properties of
38	light-absorbing organic compounds, known as brown carbon, can exhibit a strong pH-
39	dependence, which directly affects their climate impacts (Phillips et al., 2017). Given the
40	importance of aerosol pH for atmospheric processes and the limitation in estimating acidity with
41	proxies (e.g., ion balances), there has been increased effort in recent years to identify the factors
42	that affect pH and to characterize temporal and spatial variations in the atmosphere (Hennigan et
43	al., 2015).
44	Globally, aerosol pH is often quite acidic due to the ubiquity and abundance of strong
45	acids like H ₂ SO ₄ , HNO ₃ , and HCl (Pye et al., 2020). Ammonia (NH ₃) is unique among species
46	that affect pH because it is the most abundant basic compound in the atmosphere. NH_3
47	partitioning is controlled by the concentration of strong acids and by the ambient temperature
48	and relative humidity, hence, the dependence of pH on both composition and meteorology
49	(Zheng et al., 2020). This explains why NH ₃ can exist partially in the gas phase even when the
50	aerosol pH is highly acidic (Weber et al., 2016). Due to its abundance and semi-volatile
51	properties, NH3 was identified as the most important buffering agent in aerosols across locations
52	with diverse emissions, composition, and climatology (Zheng et al., 2020). In single-phase





53	aqueous particles, organic compounds have a minor effect on pH (Battaglia Jr et al., 2019),
54	though this is not the case for particles that have undergone liquid-liquid phase separation (Pye et
55	al., 2018). Non-volatile cations (NVC) typically contribute a minor fraction of PM mass but can
56	be critical for accurate predictions of pH, especially if NVC concentrations are overestimated
57	(Vasilakos et al., 2018). Globally, NVC are most important in regions heavily impacted by dust
58	emissions (Pye et al., 2020), but have minor effects on pH in other regions (Tao and Murphy,
59	2019; Zheng et al., 2020).
60	Aerosol pH is also strongly affected by meteorological factors. Equilibrium constants,
61	including those that determine the gas-particle partitioning and aqueous dissociation of semi-
62	volatile acids and bases, are temperature dependent. Temperature is a dominant factor driving
63	variability in the seasonal and diurnal cycling of pH (Guo et al., 2015; Tao and Murphy, 2019).
64	Temperature gradients, as occur in and around urban areas, can also drive large differences in
65	pH, even if the composition is uniform over the same scales (Battaglia et al., 2017). Relative
66	humidity (RH) regulates aerosol liquid water content (ALWC), which affects the partitioning of
67	soluble gases and aqueous phase solute concentrations. ALWC may be the most important
68	factor responsible for large pH differences observed between the southeast US (pH \sim 0.5-1.0) and
69	the heavily polluted North China Plain (pH ~4-5), the two regions where pH has been most
70	extensively studied to-date (Zheng et al., 2020). These effects can be complex, or even partially
71	offset. For example, an increase in temperature reduces pH owing to the shift in NH ₃
72	partitioning towards the gas phase, but NH ₃ emissions increase with temperature as well,
73	producing an increase in pH and partially offsetting the pH changes due to temperature (Tao,
74	2020). More research is needed to better understand how all these factors together affect pH, and
75	how this changes geographically and temporally.





76	In this study aerosol pH was characterized during the summertime (June 4 to July 5,
77	2018) at a land-water transition site near a large urban area (Baltimore, MD) as part of the
78	OWLETS-2 (second Ozone Water-Land Environmental Transition Study) field campaign. This
79	site is unique because meteorological phenomena, such as the bay breeze, affect pollution
80	dispersion and recirculation (Loughner et al., 2014). Baltimore is impacted by different regional
81	emission sources, as it is located in the populous and heavily trafficked I-95 corridor, and
82	downwind of the Ohio River Valley (He et al., 2013), and relatively close to regional agricultural
83	operations that emit large amounts of NH ₃ (Pinder et al., 2006). In the companion investigation,
84	the local sources of NH3 during the OWLETS-2 study were examined, including an analysis of
85	transient events with unexpectedly high NH3 concentrations (Balasus et al., 2021). In this study,
86	the effects of the observed NH3 concentrations (including transient concentrations of
87	unexpectedly high values), in combination with the unique meteorological phenomena associated
88	with the land-water transition, on ALW and aerosol pH were investigated.
89	
90	2 Methods
91	The OWLETS-2 study was conducted to characterize effects of meteorological
92	phenomena associated with the land-water transition on summertime air quality in Baltimore.
93	Hart-Miller Island (HMI, coordinates 39.2421°, -76.3627°), a site located on the Chesapeake Bay
94	~ 10 km east of downtown Baltimore, hosted many of the ground-based measurements during the
95	study (Fig. S1). Semi-continuous measurements of aerosol inorganic chemical composition and
96	gas-phase NH3 were conducted at HMI. The measurement details are provided in the companion
97	paper (Balasus et al., 2021). Briefly, the water-soluble ionic components of PM _{2.5} were
98	measured with a Particle-into-Liquid Sampler coupled to a dual Ion Chromatograph (PILS-IC,





- 99 Metrohm) operated according to Valerino et al. (2017). NH₃ was measured with an AiRRmonia
- 100 Analyzer (RR Mechatronics) (Norman et al., 2009). Meteorological parameters were measured
- 101 with a Vaisala MAWS201 Met Station at 1-minute resolution.
- 102 The 5-minute NH₃ measurements and 1-minute meteorological measurements were
- 103 averaged to the 20-mintue sampling time of the PILS-IC. Aerosol pH for each 20-min sample
- 104 was calculated according to Zheng et al. (2020). This method uses the relevant temperature-
- 105 dependent equilibrium constants and the measured concentrations of NH₃ and [NH₄⁺] to
- 106 calculate pH. The method of Zheng et al. (2020) is based upon a similar approach in prior
- 107 studies (Hennigan et al., 2015; Keene et al., 2004), with minor differences possibly due to
- 108 equilibrium constant values and/or the concentration basis (molality vs. molarity). Indeed, very
- 109 close agreement was observed (slope = 0.967, $R^2 = 0.977$, n = 872, Fig. S2) in calculated pH
- 110 between the methods of Zheng et al. (2020) and Hennigan et al. (2015). The aqueous phase
- 111 NH₄⁺ concentration is derived from the mass concentration of NH₄⁺ from the PILS-IC and the
- 112 ALWC. The ISORROPIA-II thermodynamic equilibrium model was used to calculate ALW
- using the PILS-IC, NH₃, and meteorological data as inputs (Fountoukis and Nenes, 2007). The
- 114 model was run in forward mode (NH₃ and aerosol NH₄⁺ were input at total NH₃) using the
- 115 metastable assumption according to the recommendation of Guo et al. (2015).
- 116 Although ISORROPIA can provide pH, the methods of Zheng et al. (2020) and Tao et al.
- 117 (2020) were used for pH calculations:

118
$$pH = pK_{a,NH_3}^* + \log_{10} \frac{[NH_3(aq)] + [NH_3(g)]}{[NH_4^+(aq)]}$$
(1)

119
$$[NH_3(g)] = \frac{p_{NH_3}\rho_w}{RT AWC}$$
(2)

120
$$K_{a,NH_3}^* = \frac{[H^+(aq)]([NH_3(aq)] + [NH_3(g)])}{[NH_4^+(aq)]} = K_{a,NH_3}\left(1 + \frac{\rho_w}{H_{NH_3}RT AWC}\right)$$
(3)





where in Equation 1, $[NH_3(aq)]$ is the molality of NH₃ in solution (mol kg⁻¹, calculated by 121 122 multiplying H_{NH_3} and p_{NH_3}), $[NH_3(g)]$ is the equivalent molality of gaseous NH₃ in solution (mol kg⁻¹, given by Equation 2), and $[NH_4^+(aq)]$ is the molality of NH₄⁺ in solution (mol kg⁻¹). 123 124 Equation 2 calculates $[NH_3(g)]$, where p_{NH_2} is the partial pressure of NH₃ (atm), ρ_w is the density of water (µg m⁻³), R is the gas constant (atm L mol⁻¹ K⁻¹), T is temperature (K), and 125 AWC is aerosol water content ($\mu g m^{-3}$ air). Equation 3 calculates the last unknown term in 126 127 Equation 1, which is K_{a,NH_3}^* , the effective dissociation constant (where H_{NH_3} is the Henry's law 128 constant of NH₃ in mol kg⁻¹ atm⁻¹). 129 Direct measurements of aerosol pH are not available to test model predictions so the partitioning of semi-volatile species that depend on pH, most commonly NH₃/NH₄⁺ and 130 131 HNO_3/NO_3^{-} , is a key metric used to evaluate model performance (Pye et al., 2020). 132 ISORROPIA is used extensively for predictions of pH; however, in this study the measured and 133 ISORROPIA-predicted values of NH₃ partitioning ($\varepsilon_{NH3} = NH_3/(NH_3 + NH_4^+)$) did not agree well 134 (Fig. S3). There were systematic differences in pH between the two methods (mean pH 135 difference = 0.6 pH units), and they were not correlated ($R^2 = 0.097$, not shown). The source of the discrepancy in NH₃ partitioning (ISORROPIA modeled versus measurement-calculated) is 136 137 unknown, though it is not likely the result of an incorrect assumption of equilibrium (see the SI 138 and the companion paper Balasus et al. (2021) for more details). 139 140 **3** Results and Discussion

141 **3.1 Meteorological effect on pH**

As in many cities, the urban heat island effect in Baltimore evolves throughout the day,

143 with urban-rural temperature and RH gradients peaking at night (Battaglia et al., 2017).





144	Meteorological conditions at HMI demonstrate unique features associated with the land-water
145	transition. At night, the average temperature observed at HMI was close to conditions observed
146	in downtown Baltimore but transitioned to match the conditions at a nearby rural site during the
147	day (Fig. 1a). Overall, the range in average hourly temperatures at HMI was lower (4.3 °C; 22.5
148	-26.8 °C) than the averages observed at either the downtown site (range 6.2 °C; 22.8 – 29.0 °C)
149	or the rural site (range 8.8 °C; $17.5 - 26.3$ °C). Likewise, the average hourly RH profile at HMI
150	had a significantly smaller range $(13.7\%; 61.7 - 75.4\%)$ than the RH profiles at the downtown
151	(range 22.1%; 53.2 – 75.3%) or rural sites (range 30.3%; 55.9 – 86.2%) (Fig. 1b). The
152	differences shown in Fig. 1 were due to the proximity of HMI to the Chesapeake Bay. This has
153	strong implications for aerosol pH, which will be discussed in detail below.
154	Due to the meteorological conditions discussed above, the diurnal profile of ALWC at
155	HMI was unique (Fig. 2a). Typical profiles of ALWC in the eastern US closely follow RH, with
156	minima in the afternoon and maxima at night or in the pre-dawn morning hours (Guo et al.,
157	2015; Battaglia et al., 2017). During OWLETS-2, ALWC did not show a distinct diurnal profile
158	that was correlated with RH. Instead, the highest median ALWC at HMI occurred between
159	12:00 – 14:00, (local time, LT; during the study this is UTC-4) (Fig. 2a). This daily peak in
160	ALWC coincided with a pronounced enhancement in aerosol NO3 ⁻ , which is discussed in the
161	companion paper (Balasus et al., 2021). The partitioning of NH ₃ , ε_{NH3} , also showed a diurnal
162	profile that was unexpected (Fig. 2b). Due to the strong temperature dependence of vapor
163	pressure and equilibrium constants, NH3 partitioning typically shifts towards the gas-phase
164	during the daytime (ε_{NH3} increases) and shifts towards the aerosol phase at night (ε_{NH3} decreases)
165	(Guo et al., 2017). The increase in gas-phase NH ₃ emissions with increasing temperature can
166	also contribute to an elevated ε_{NH3} during the daytime. The diurnal profile of ε_{NH3} observed at





167	HMI did not follow temperature, as the median ε_{NH3} peaked during the 06:00 – 08:00 LT, and
168	decreased slightly into the afternoon (Fig. 2b). This shows a shift of NH ₃ partitioning towards
169	the condensed phase as daily temperatures peaked. This was allowed by the ALWC remaining
170	steady throughout the afternoon. Overall, the median ε_{NH3} value for the entire OWLETS-2 study
171	was 0.915, showing NH ₃ partitioning was shifted towards the gas-phase.
172	The diurnal profile of aerosol pH computed using the method of Zheng et al. (2020)
173	followed a qualitatively similar pattern to prior studies (Battaglia et al. 2017), with maxima in
174	the early morning and minima in the afternoon; however, there was a much smaller amplitude in
175	the median hourly pH values (Fig. 3). The highest median pH value (1.97) was observed
176	between $07:00 - 08:00$ LT, while the lowest median pH (1.50) was observed between $16:00 - 08:00$ LT.
177	17:00 LT. For the entire study, there was only a \sim 1 pH unit difference between the 10 th and 90 th
178	percentile values (1.39 and 2.36, respectively). The relatively muted diurnal profile of aerosol
179	pH was due to the unique meteorology that resembled a nearby urban site at night and
180	transitioned to match the nearby rural site during the day. The companion paper shows the
181	diurnal profiles of aerosol inorganic composition (Balasus et al., 2021). Figures 2 and 3 are
182	consistent with recent studies that demonstrate the high sensitivity of pH to meteorological
183	factors (Battaglia et al., 2017; Tao and Murphy, 2019; Zheng et al., 2020). It is interesting to
184	note that the ISORROPIA predictions of aerosol pH yield a distinctly different diurnal profile
185	than the pH predicted by NH ₃ partitioning (Fig. S4). The difference in pH between the two
186	methods peaked in the afternoon between $12:00 - 14:00$ LT, when particles were most acidic,
187	and was a minimum in the early morning when pH was highest (Fig. S4). The reason for these
188	discrepancies are explored in the discussion below.





189	Results suggest that pH was most sensitive to temperature changes at HMI during
190	OWLETS-2 (Fig. 4a). Across the full temperature range, the observed pH sensitivity to
191	temperature was -0.047 K ⁻¹ . A recent study contrasting pH in the southeast US and the North
192	China Plain found that temperature affects pH linearly at a rate of approximately -0.055 K ⁻¹
193	(Zheng et al., 2020). A separate study from a Canadian observational network found that the
194	pH-temperature dependence is not linear, but changes with temperature (Tao and Murphy, 2019;
195	Tao, 2020). Over the range of conditions observed during OWLETS-2, Tao (2020) computed a
196	pH sensitivity to temperature of approximately -0.045 K ⁻¹ to -0.055 K ⁻¹ . In a previous study, the
197	pH sensitivity to temperature in Baltimore was -0.048 K ⁻¹ , though this was calculated for
198	conditions of constant atmospheric composition (Battaglia et al., 2017). While present results
199	share consistencies with these studies, the results in Fig. 4a suggest important differences, as
200	well. An increase in pH was observed with increasing temperature for conditions below 293 K
201	(n=156). In the companion paper, NH_3 concentrations dramatically increased with temperature <
202	293 K but exhibited a much weaker dependence on temperature for conditions > 293 K (Balasus
203	et al., 2021). Under the warmer conditions, the pH relationship with temperature was linear
204	during OWLETS-2, with a sensitivity of -0.085 K ⁻¹ (Fig. 4a). The results in Fig. 4a are
205	consistent with Tao (2020), as they demonstrate the offsetting responses of pH to temperature
206	through effects on NH ₃ emissions and partitioning.
207	These results contribute to the growing body of work demonstrating the importance of
208	temperature to aerosol pH. Collectively, these studies suggest that the sensitivity of pH to
209	temperature is constrained between -0.045 K ⁻¹ and -0.085 K ⁻¹ , although the present results
210	illustrate circumstances where a positive relationship between temperature and pH can exist, as

- 210 illustrate circumstances where a positive relationship between temperature and pH can exist, as
- 211 well. It is notable that similar $\Delta pH/\Delta T$ values are observed across a range of locations and for





- 212 variable data sets that include monthly averages of long-term observations (Tao and Murphy,
- 213 2019) and 20-min measurements made over a period of weeks (presented here).

214 Aerosol pH was not strongly affected by ALWC over the range of conditions observed 215 during OWLETS-2, except at ALWC below 1 µg m⁻³ (Fig. 4b). At the lowest ALWC levels, the 216 increase in pH occurs because of the diluting effect of water, however, at ALWC above 1 µg m⁻³, 217 other factors appear to be more important (e.g., temperature). ALWC was recently identified as 218 the most significant contributor to regional pH differences (Zheng et al., 2020). In that study, the 219 pH-ALWC relationship was highly non-linear, with the greatest sensitivity calculated at ALWC 220 at levels $< 25 \,\mu g \, m^{-3}$, conditions corresponding to all the OWLETS-2 observations. Tao (2020) 221 found that pH is extremely sensitive to RH, presumed to be a surrogate for ALWC, when RH 222 was < 20% or RH > 80%; however, pH was quite insensitive to RH variations in the region of 223 20% < RH < 80%. Approximately 25% (208 out of 875) of RH values during OWLETS-2 were 224 above 80%, yet no increase in pH was observed at the highest ALWC, suggesting that other 225 factors were offsetting the diluting effect of water as ALWC increased. It is interesting to note 226 that ISORROPIA predicts a stronger effect of ALWC on pH, with the diluting effect apparent as 227 pH increases with increasing ALWC (Fig. 4b). 228 The result shown in Fig. 4b is somewhat surprising because NH₃ partitioning was quite

sensitive to ALWC (Fig. 5); the relatively invariant aerosol pH is unexpected given the increase in NH₃ uptake in the presence of ALW. NH₃ partitioning shifted towards the condensed phase (ε_{NH3} decreased) at increasing ALWC, consistent with the results of Nenes et al. (2020). ε_{NH3} was more sensitive to ALWC than it was to either temp or RH (Fig. S6). This result shows the importance that ALWC can have on PM_{2.5} mass concentrations, as water serves as an important medium enhancing the condensation of organic and inorganic water-soluble species (Carlton et





235	al., 2018; El-Sayed et al., 2016). Fig. 5 demonstrates the importance of ALWC in changing the
236	dry deposition of reactive nitrogen species. Nenes et al. (2021) predicts that NH ₃ and HNO ₃ dry
237	deposition rates will both be high under the conditions observed during OWLETS-2 (i.e., ALWC
238	$<10~\mu g$ m $^{-3}$ and pH \sim 1.5). This is also consistent with a modeling study showing increased dry
239	deposition of reactive nitrogen in coastal regions, including the OWLETS-2 study domain
240	(Loughner et al., 2016).
241	
242	3.2 Composition Effects on Aerosol pH
243	In contrast to meteorological factors, aerosol composition did not have a major effect on
244	pH variability during OWLETS-2. Consistent with prior studies, neither the sulfate
245	concentration nor the NH4 ⁺ :SO4 ²⁻ molar ratio contributed significantly to pH variability (Weber
246	et al., 2016; Hennigan et al., 2015). Fig. 4c shows that pH was also relatively insensitive to the
247	Tot-NH ₃ concentration, in agreement with the results from other locations (Zheng et al., 2020;
248	Tao, 2020; Weber et al., 2016). While the predictions of ISORROPIA generally did not capture
249	the trends between pH and the meteorological factors, it is interesting to note that ISORROPIA
250	predicts that pH is relatively insensitive to Tot-NH ₃ , as well, except at the highest concentrations
251	(Fig. S5).
252	Composition and concentration differences between HMI and the urban and rural sites
253	are analyzed in more detail in the companion paper (Balasus et al., 2021). In the companion
254	paper, episodic NH3 events that derived from dairy, poultry, and industrial sources were
255	characterized (Balasus et al., 2021). For the events with complete aerosol composition and
256	meteorology data (8 out of 11 NH ₃ events total), the average and medial aerosol pH values (2.00
257	and 1.96, respectively) were only moderately higher than the study average and median pH





258	values (1.85 and 1.83, respectively); this difference is not statistically significant at the 95%
259	confidence interval. This includes an aerosol pH of 1.92 at the peak NH3 concentration observed
260	during the entire study (19.3 μ g m ⁻³), an event influenced by industrial emissions near downtown
261	Baltimore (Balasus et al., 2021). Together, the results suggest that pH may be more affected by
262	the proximity of the HMI site to the Chesapeake Bay than it was to regional agricultural NH_3
263	emissions, or to episodic NH ₃ events from local industrial sources.
264	
265	3.3 Case Study: Effect of NVCs on pH
266	NVCs affect thermodynamic predictions of NH3 partitioning and thus, pH (Guo et al.,
267	2018; Vasilakos et al., 2018). Seawater is alkaline and primary marine emissions contain high
268	concentrations of NVCs (O'Dowd and De Leeuw, 2007). Marine aerosols rapidly acidify,
269	typically in seconds or minutes, though the timescale depends upon particle size (Angle et al.,
270	2021; Pszenny et al., 2004). Studies have examined the acidity of sea spray aerosols and their
271	evolution but none, to the authors' knowledge, have done so in a polluted urban environment.
272	The OWLETS-2 study offered a unique opportunity to analyze the pH of primary marine
273	particles emitted within several km of a large urban area. The Chesapeake Bay is brackish, with
274	increasing salinity moving down the bay towards the Atlantic Ocean (Pritchard, 1952). Near the
275	OWLETS-2 measurement site at HMI, salinity is variable but average conditions are \sim 5 g kg ⁻¹
276	(www.chesapeakebay.net, last accessed 20-November 2020). This is about a factor of seven
277	lower than typical seawater, but shows the potential for primary emissions to contribute salts that
278	could impact aerosol pH at HMI.
279	Elevated concentrations of Na^+ and Cl^- in $PM_{2.5}$ were infrequently observed during
280	OWLETS-2, with one notable 36-hour period showing evidence of primary marine impact. Na ⁺





281	and Cl^{-} were well-correlated ($R^2 = 0.78$) from 11-June to 13-June, a period that coincided with
282	the highest concentrations of both species (Fig. 6 and Fig. 7). It is noteworthy that wind speeds
283	were not elevated during this time (average winds = 3.1 m s^{-1} compared to campaign-average
284	wind speeds of 2.9 m s ⁻¹); longer-term measurements would be needed to characterize factors
285	driving the primary bay emissions. During this event, the pH calculation using NH ₃ partitioning
286	suggests that the primary marine emissions had a minimal effect on aerosol pH. The average pH
287	during this period was 1.98, which was only slightly higher than the average for the entire study
288	(1.85). Further, as the total $Na^+ + Cl^-$ concentration increased by more than an order of
289	magnitude, from 0.05 $\mu g~m^{\text{-3}}$ around 18:00 LT on 11 June to 0.9 $\mu g~m^{\text{-3}}$ at 09:00 LT on 12 June,
290	the pH only increased by 0.1 pH unit during the same period (Fig. 6). Gas-phase NH ₃ data were
291	not available for the entire NaCl event, so pH calculations were limited to the first ~20 hours.
292	The pH predictions from ISORROPIA during this period display a more significant effect
293	on aerosol pH. The average pH predicted by ISORROPIA is 3.08, which is 0.7 pH units higher
294	than the study average from ISORROPIA. Further, while the pH calculated using NH_3
295	partitioning is insensitive to Na ⁺ and Cl ⁻ , ISORROPIA predicts a rise of 1.4 pH units as Na ⁺ and
296	Cl ⁻ increase (Figure 6). The NH ₃ partitioning predicted by ISORROPIA deviates from the
297	observations during this event (r = -0.19, Fig. S7). The most likely explanation for this behavior
298	is different chemical compositions of the coarse- and fine-mode particles. Fresh marine
299	emissions acidify quickly, and evidence was found for chemical processing of NaCl. HNO3
300	displacement of HCl is a well-known phenomenon in sea salt particles (Brimblecombe and
301	Clegg, 1988). Chloride:sodium ratios slightly above unity were observed when aerosol nitrate
302	concentrations were low, and below unity when nitrate concentrations were elevated (Fig. 7).
303	Nitrate formation, including HNO3 uptake to sea salt, is highly sensitive to pH (Kakavas et al.,





- 304 2021; Vasilakos et al., 2018). Nitrate (along with Na⁺ and Cl⁻) are direct inputs to ISORROPIA,
- 305 suggesting that the pH trend in Fig. 6 is due to the increased influence of coarse-mode particles.
- 306 The PILS inlet was equipped with a 2.5 µm cut-point cyclone (URG-2000-30-EH, URG Corp.),
- 307 which allows some penetration of particles with $d_p < 4.5 \,\mu\text{m}$
- 308 (http://www.urgcorp.com/products/inlets/teflon-coated-aluminum-cyclones/urg-2000-30eh, last
- 309 accessed 29 January 2021). Likewise, primary sea salt emissions often exhibit a size distribution
- 310 tail that extends below $2.5 \,\mu m$ (Feng et al., 2017).
- 311 The above analysis identifies limitations computing aerosol pH with both approaches and 312 highlights opportunities to use complementary information from each to inform factors driving 313 aerosol pH. ISORROPIA's assumption of an internally mixed aerosol distribution cannot be 314 applied to predict pH in this system (Fountoukis and Nenes, 2007). However, it does provide 315 insight into the likely presence of coarse mode particles with pH significantly higher than the 316 fine mode. Conversely, Na⁺, Cl⁻ and NO₃⁻ are not direct inputs into the pH calculation using 317 NH₃ partitioning, though they are used to compute aerosol liquid water, which is an input in the 318 pH calculation (Zheng et al., 2020). These NVCs have been shown to have sometimes 319 significant effects on the ratios of ammonium-sulfate, which could lead to inconsistencies in the 320 calculation of aerosol pH when considered in the ALWC calculation only (Guo et al. 2015). 321 NH₄⁺ resides predominantly in the fine mode (Seinfeld and Pandis, 2016), so the partitioning 322 approach is unlikely to capture the acidity of the coarse mode or fine-mode particles in the tail of 323 the distribution of primary emissions that may be externally mixed with secondary particles, such 324 as dust or sea salt. This may lead to underestimates of NVC effects on aerosol pH using the partitioning approach. The combined information from both methods suggests that aerosols 325 326 sampled at HMI during the NaCl event were characterized by a strong size-dependent pH





327	gradient, with fine-mode particles more acidic (pH \sim 2) than the coarse mode (pH up to 4.5).
328	Estimates of pH derived from size-segregated aerosol composition measurements have observed
329	the same phenomenon (Angle et al., 2021; Fang et al., 2017; Kakavas et al., 2021; Keene et al.,
330	2002).
331	
332	4 Conclusions
333	There is growing recognition of the importance of aerosol pH affecting atmospheric
334	processes relevant to public health and ecosystems (Fang et al., 2017; Nenes et al., 2021).
335	Observations of spatial and temporal variations in pH are needed so that the factors that control
336	pH and contribute to variability in different environments can be fully understood. This study is
337	unique as it represents the first characterization of aerosol pH at a land-water transition site near
338	a large urban area. Baltimore, MD, is impacted by regional agricultural emissions and by
339	industrial point-source emissions of NH ₃ . The companion paper examines the sources of
340	episodic NH ₃ events and the associated effects on aerosol composition (Balasus et al., 2021).
341	Although average and peak NH3 concentrations during this study were significantly higher than a

342 nearby inland site, the effects on aerosol pH appear relatively insignificant, as pH during the

343 peak events was only ~0.1 pH unit higher than non-event periods. This finding is consistent with

344 studies at other locations that show aerosol pH is often insensitive to Tot-NH₃ and to the aerosol

345 NH₄:SO₄ ratio (Weber et al., 2016; Zheng et al., 2020; Tao and Murphy, 2019).

The unique characteristics of the OWLETS-2 study and measurement locations also offered insight into the composition and meteorological influences on aerosol pH. In the companion paper, the composition effects were shown to be muted in comparison to the

349 meteorological effects (Balasus et al., 2021). It was shown that the unique diurnal profiles,





350	particularly in ALW (which did not correlate with RH) and ε_{NH3} (which did not correlate with T)
351	resulted in meteorological factors, notably temperature, having the most important influence on
352	aerosol pH. Across the full temperature range of the study, the observed pH sensitivity to
353	temperature was -0.047 K ⁻¹ , with increases in sensitivity up to -0.085 K ⁻¹ when the temperature
354	was > 293 K. The sensitivity of aerosol pH shown here is in good agreement with previous
355	studies in the Baltimore region and beyond, e.g. Toronto (Tao and Murphy, 2019; Battaglia et al.
356	2017). Conversely, aerosol pH was not strongly affected by ALWC during the OWLETS-2
357	study, except when ALWC was below 1 μ g m ⁻³ , in contrast to the results of Zheng et al. (2020).
358	These results of Zheng et al. (2020) in identifying ALW as the most important factor driving
359	aerosol pH variability were derived from the results of studies in multiple locations, including
360	bulk aqueous solution. However, the analysis suggests that the factors that drive aerosol pH
361	variability may exhibit important site-to-site differences that must be considered before
362	generalizations are applied.
363	A case study of the NVC effect on aerosol pH had significantly different outcomes
364	depending on the method for calculating pH. ISORROPIA predicted a pH increase of $\sim 1.4~\text{pH}$
365	units during an event with primary aerosol emissions from the Chesapeake Bay, while the pH
366	calculation using NH ₃ partitioning predicted a much less significant effect (~0.1 pH unit). This
367	difference is attributed to the likely presence of externally mixed particles during the events,
368	which may include primary marine emissions elevated in NVCs. Bougiatioti et al. (2016)
369	evaluated aerosol pH at a remote site in the Mediterranean, where samples with a marine origin
370	demonstrated vastly different pH between fine (avg. $pH = 0.4$) and coarse mode (average $pH = 0.4$)
371	7.3) particles. Similarly, Keene et al. (2002) demonstrated the effect of marine aerosol size
372	distribution on aerosol pH, with fine mode particles predicted to reside in the range of 1-2, with





373	super- μ m particles to reside in the range of 3-4, consistent with the current results. Hence, a
374	limitation of this study is the lack of size-resolved aerosol composition measurements. This
375	study underscores the need to evaluate assumptions of internally mixed aerosols when applying
376	pH calculations, which may be a critical factor in overestimating the effects of NVCs on pH.
377	Models with size-resolved aerosol composition may be required to capture this effect across
378	scales in future studies (Kakavas et al., 2021).
379	
380	Data Availability
381	Data are available at https://www-air.larc.nasa.gov/cgi-bin/ArcView/owlets.2018.
382	
383	Supplement
384	The supplement related to this article is available online at:
385	
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393	
394	Author Contributions





- 395 CH, AC, and RD conceived the analysis and study participation. MB, NB, and KB collected and
- analyzed the PILS-IC and NH₃ data. NB, MB, and CH conducted the thermodynamic modeling
- analyses. VC and AC provided analytical input and interpretation. NB, CH, KB, and MB wrote
- 398 the manuscript. All authors provided feedback and revisions to the manuscript.
- 399

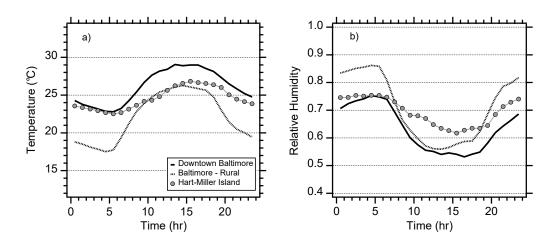
400 **Competing Interests**

401 The authors declare they have no conflict of interest.







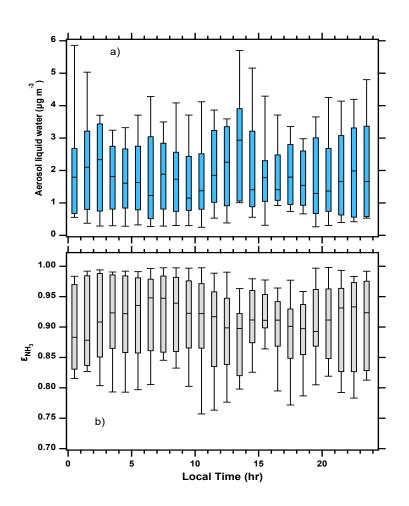


404

Figure 1: Diurnal profiles of (a) temperature and (b) relative humidity at three sites during the OWLETS-2 study. Hart-Miller Island (HMI) is a land-water transition site and was the location of aerosol composition and gas-phase measurements during the campaign. Temperatures at HMI resembled the downtown location during the night but showed characteristics of the rural site during the day. RH at HMI was between the urban and rural sites at night but was elevated during the daytime due to the Chesapeake Bay.







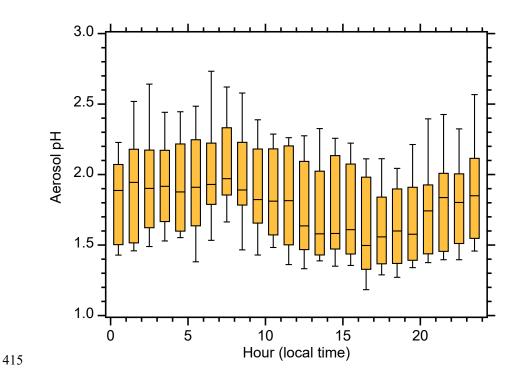
412 **Figure 2:** Box plots of (a) aerosol liquid water and (b) ε_{NH3} (ε_{NH3} = NH₃(g)/(NH₃(g) + NH₄⁺(aq)))



⁴¹⁴ percentiles.







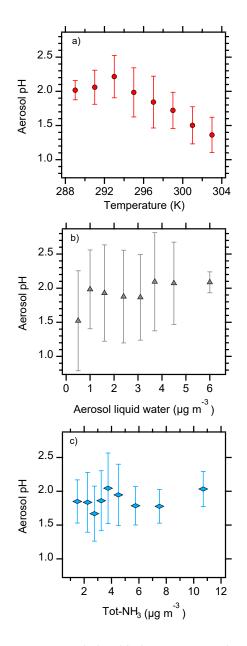
416 **Figure 3:** Box plot of the aerosol pH diurnal profile calculated using the method of Zheng et al.

417 (2020) at Hart-Miller Island during OWLETS-2. The statistics shown are the median, quartiles,

418 10^{th} and 90^{th} percentiles.







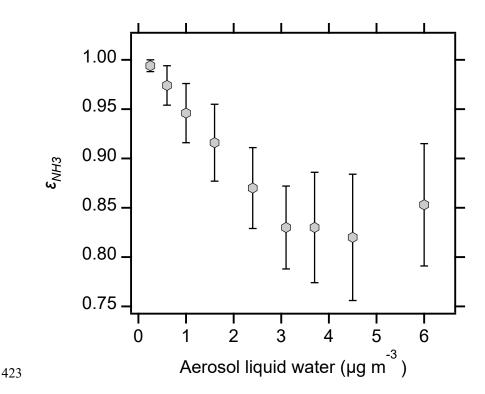
419

Figure 4: Relationship between aerosol pH and (a) temperature, (b) aerosol liquid water, and (c) total NH₃ (Tot-NH₃ = NH₃ + NH₄⁺). Symbols represent mean values while error bars represent

422 standard deviations.







424 **Figure 5:** Relationship between NH₃ partitioning ($\varepsilon_{NH3} = NH_3/(NH_3 + NH_4^+)$) and aerosol liquid

425 water. Symbols represent mean values while error bars represent standard deviations.





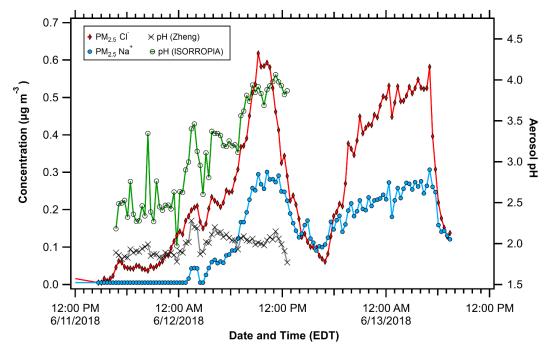




Figure 6: Concentrations of PM_{2.5} Na⁺ and Cl⁻ during a period with primary Chesapeake Bay

429 emissions. Aerosol pH calculated by NH₃ partitioning (Zheng et al. method) and ISORROPIA

430 show different behaviors, suggesting the aerosol distribution was externally mixed during this







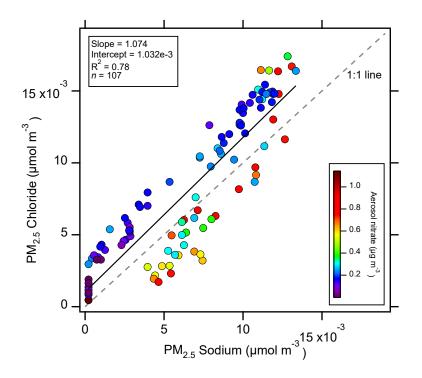


Figure 7: Correlation between $PM_{2.5}$ Na⁺ and Cl⁻ during the event shown in Figure 6. Lower

434 chloride:sodium ratios were observed at higher NO₃⁻ concentrations, suggesting HNO₃ uptake

435 displaced HCl.





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