

# Predicting Organic-Inorganic Aerosol Efflorescence Using Thermodynamically Modeled Viscosity

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**Abstract.** Atmospheric aerosols, especially internally mixed organic-inorganic aerosols, exhibit complex phase behaviors that affect their size evolution, optical properties, and chemical reactivity, ultimately impacting climate and human health. Although parameterizations for secondary organic aerosol phase state exist, predictive models based on primary predictors for efflorescence in organic-inorganic aerosols remain underdeveloped. In this study, we evaluated several chemical parameters, including equivalent O:C ratio, organic mass fractions, glass transition temperature ( $T_g$ ), and viscosity ( $\eta$ ), and identified aerosol viscosity as the primary predictor of efflorescence relative humidity (ERH) in internally mixed organic-inorganic aerosols. We developed a linear viscosity-ERH model based on ERH and  $\log_{10} \eta$ , which defines the boundary conditions for aerosol efflorescence when  $\eta < 4.76 \times 10^2 \text{ Pa}\cdot\text{s}$ . Additionally, we showed that efflorescence is inhibited when  $\eta > 4.76 \times 10^2 \text{ Pa}\cdot\text{s}$ . Validation using an independent dataset showed strong agreement between predicted and experimentally measured ERH ( $R^2 = 0.95$ ). A multivariate regression model incorporating  $\eta$  and  $T_g$  improved prediction accuracy but was limited by  $T_g$  parameterization for complex organic-inorganic mixtures. Our findings highlight the role of aerosol viscosity in controlling efflorescence and emphasize the need to develop improved aerosol viscosity measurement techniques to better constrain aerosol phase transitions in atmospheric models.

## 1 Introduction

30 Atmospheric aerosols significantly influence climate and human health through both direct and indirect effects. They directly influence the energy balance by absorbing and scattering solar radiation and indirectly affect the climate by acting as cloud condensation nuclei (CCN), thereby altering cloud microphysical properties such as droplet size, lifetime, and optical characteristics. (Boreddy et al., 2014; Fan et al., 2010; Knopf and Alpert, 2023; Novo et al., 2021; Pöschl, 2005; Shrivastava et al., 2017; Smith  
35 et al., 2021) Among aerosol physicochemical properties, phase state (liquid, solid, or semi-solid) is particularly important, as it governs hygroscopicity, optical properties, and heterogeneous reactivity. (Fard et al., 2018; Freedman et al., 2024; Novo et al., 2021) Moreover, aerosol phase transition significantly affects viral activity during transmission. (Oswin et al., 2022) Consequently, aerosol phase states substantially impact climate, air quality, and human health. (Reid et al., 2018)

40 Atmospheric aerosols, particularly internally mixed secondary aerosols composed of organic and inorganic components, exhibit diverse phase states, including liquid, liquid-liquid phase-separated, semi-solid, amorphous solid, and crystalline solid, which govern their physicochemical properties. (Novo et al., 2021) For instance, aerosols undergoing efflorescence (i.e., crystallization) and deliquescence exhibit hysteresis in particle size and water content in response to relative humidity (RH) variations. (Choi and  
45 Chan, 2002; Guo et al., 2020) In contrast, amorphous organic aerosols generally do not exhibit the pronounced discontinuous hysteresis associated with efflorescence and deliquescence of crystalline particles. However, semi-solid or glassy amorphous particles may still show gradual, kinetically driven hysteresis in water uptake and release because of slow diffusion and equilibration. (Koop et al., 2011) Thus, effloresced (or crystallized) aerosols display fundamentally different hygroscopic behaviors from  
50 amorphous ones, introducing uncertainties in predicting mixing aerosol water uptake and particle size under varying RH conditions because phase transitions and kinetic limitations can lead to different water contents and growth factors during hygroscopic cycles. (Marcolli and Krieger, 2006; Mikhailov et al., 2009)

Aerosol particles consisting of more than a single phase may have crystalline as well as amorphous  
55 phases. It is important to note that, in internally mixed organic–inorganic aerosols, the onset of phase transitions such as deliquescence and efflorescence typically occur only in specific components rather than the entire particle. For example, crystalline inorganic salts or hydrates may undergo deliquescence

or crystallization, while the remaining particle matrix can persist in an amorphous, semi-solid, or aqueous state. (Hodas et al., 2016)

60 However, internally-mixed organic-inorganic aerosols are chemically complex, often containing thousands of distinct organic species, of which only about 10% have been identified. (Hallquist et al., 2009) This limited chemical characterization constrains our ability to accurately predict aerosol phase states. Existing parameterization models based on organic aerosol properties, such as molecular weight and oxygen-to-carbon ratio (O:C ratio) (Shiraiwa et al., 2017), elemental composition (DeRieux et al., 65 2018), and volatility (Li et al., 2020; Zhang et al., 2019) to predict the glass transition temperature ( $T_g$ ) have been established to predict the viscosity and phase state of organic aerosols but these models typically neglect efflorescence. Consequently, current parameterizations fail to distinguish between crystalline and amorphous solid states of internally mixed organic-inorganic aerosols and overlook the hysteresis in particle size and water uptake during efflorescence-deliqescence cycles. This limitation 70 introduces significant uncertainty in predicting aerosol size evolution and leads to biases in describing atmospheric aerosol phase behavior, hygroscopicity, and optical properties, ultimately affecting our understanding of atmospheric particle aging, CCN activation, as well as climate and health impacts.

While laboratory studies have provided substantial data on efflorescence in mixed organic-inorganic aerosols, a systematic understanding of the key factors that govern and predict efflorescence remains 75 limited. (Cai et al., 2017; Ma et al., 2021b; Wang et al., 2017) Efflorescence is a kinetically-controlled process that requires overcoming a nucleation energy barrier. This process is influenced by several factors. Component diffusion controls the mobility of ions and molecules and therefore affects the rate at which crystalline nuclei can form and grow; higher viscosity generally suppresses diffusion and delays efflorescence. (Ji et al., 2017; Mikhailov et al., 2009) Chemical composition, including solute 80 concentration, solubility, and vapor pressure, influences the degree of supersaturation and thus the thermodynamic driving force for crystallization. (Cohen et al., 1987; Gupta et al., 2015) In addition, molecular interactions within the particle phase can either facilitate or hinder the structural rearrangement required for nucleation, while interfacial properties such as surface tension affect the nucleation energy barrier by altering the energetic cost of forming a new phase interface. (Davis et al., 2015; Ma et al., 85 2021a; Mikhailov et al., 2009) Furthermore, thermodynamic constraints exists for efflorescence. Crystallization can only occur when the aqueous phase becomes supersaturated with respect to the relevant solid phase; otherwise, efflorescence is thermodynamically inhibited. Once supersaturation is

achieved, nucleation and crystal growth proceed through kinetically controlled processes. (Bouzidi et al., 2022; Marcolli and Krieger, 2006) The transition point, described by the efflorescence relative humidity (ERH), is therefore strongly influenced by these factors, which are modulated by the properties of organic components, and the hydration state of the inorganic salts. The O:C ratio is widely used as a parameter for organic compound polarity. (Yu et al., 2021) Previous studies have shown that increasing the mass fraction of organic compounds ( $\omega_{\text{org}}$ ), particularly with high-viscosity organics, can significantly alter mass transport processes, and thereby affect phase transition. (Parsons et al., 2004; Shiraiwa et al., 2013; Virtanen et al., 2010; Wang et al., 2017; Zobrist et al., 2011) Under such conditions, efflorescence may be kinetically suppressed because of limited molecular diffusion, causing aerosol particles to remain in an amorphous semisolid or glassy state; in this case, their phase-state evolution is more appropriately characterized with reference to the glass transition temperature ( $T_g$ ). (Berkemeier et al., 2016; Koop et al., 2011; Song et al., 2016b; Tong et al., 2011) However, direct and reliable measurements of the viscosity of atmospheric aerosols under atmospherically relevant conditions remain challenging (Fitzgerald et al., 2016; Hosny et al., 2016), particularly because water uptake can continuously modify particle composition, phase state, and molecular mobility. (Sheldon et al., 2023; Zobrist et al., 2008) This has resulted in limited observational data and motivated the use of thermodynamic models, such as AIOMFAC, to predict the viscosity of organic–inorganic mixed aerosols. (Lilek and Zuend, 2022)

In this study, we develop a parameterization for predicting the ERH of organic-inorganic aerosols by integrating literature-reported ERH data, investigating different aerosol physicochemical parameters, focusing on aerosol viscosity ( $\eta$ ) and the collective effect of O:C ratio,  $\omega_{\text{org}}$ , and  $T_g$ . The established framework enhances our understanding of aerosol phase behavior and size change during hygroscopic cycles, offering valuable insights into atmospheric processes, such as gas-particle partitioning, heterogeneous chemistry, and cloud formation.

## 2 Methods

### 2.1 The dataset for the parameterization

We compiled aerosol ERH data from our previous studies and peer-reviewed literature identified through Web of Science using keywords: aerosol, hygroscopic, and efflorescence. All studies were laboratory-based, with ERH measured using IR spectroscopy (Braban and Abbatt, 2004; Cai et al., 2017; Ghorai et

al., 2014; Ji et al., 2017; Ma et al., 2021b; Shi et al., 2017; Wang et al., 2017; Wu, 2017; Xu et al., 2022), Raman spectroscopy (Chang, 2020; Hu et al., 2025; Ushijima et al., 2021; Wang, 2018; Yu et al., 2012), optical microscopy (Bertram et al., 2011; Huang et al., 2024; Parsons et al., 2004) or Hygroscopicity Tandem Differential Mobility Analyzer (HTDMA). (Laskina et al., 2015) In multicomponent organic–  
 120 inorganic particles, multiple efflorescence relative humidities may exist due to different salts crystallizing at distinct RH levels, and efflorescence typically occurs over a RH range. In this study, “aerosol ERH” is defined as the mean value of this transition range, serving as a simplified but effective macroscopic parameter for the overall phase transition of the particle.

A total of 102 ERH data points for organic-inorganic mixture aerosols were collected: 66 ERH points for  
 125 the training set (Table S1) and the remaining 36 points for the validation set (Table S2). The training set includes aerosol compositions containing 14 organic compounds (composed of C, H, and O) and 4 inorganic salts (ammonium sulfate, ammonium nitrate, sodium chloride, and potassium chloride) mixed in varying ratios. All ERH data in the training were from our previous studies, and all corresponding viscosity data were derived using the AIOMFAC model. The particles investigated in our previous  
 130 studies typically range from 1 to 10  $\mu\text{m}$  in diameter. The validation set contains 36 ERH data from systems involving 12 organic compounds and 4 inorganic salts (Table S2), selected from studies that either reported both aerosol ERH and viscosity or observed no efflorescence. For dataset included in this study, experiments were conducted over several hours with slow RH change rate ( $< 0.1\%/s$ ) to ensure near-equilibrium conditions. Although these timescales are shorter than typical atmospheric aerosol  
 135 lifetimes (days to weeks), previous studies have shown that nucleation induction times in high-viscosity systems can be extremely long due to diffusion limitations. (Wang et al., 2017) Therefore, the suppression of efflorescence at high viscosity reported in the cited studies is likely a fundamental kinetic constraint rather than an artifact of limited experimental duration. Moreover, the empirical data utilized to derive the regression model in this study were all obtained within a temperature range of  $298 \pm$   
 140 10 K.

Key physicochemical parameters for predicting aerosol ERH include the equivalent O:C ratio of the mixture

$$\text{O:C}_{\text{mixture}} = \left( \frac{n_{\text{O,org}}}{n_{\text{C,org}}} \right) \cdot x_{\text{org}} \quad (1)$$

where  $x_{\text{org}}$  is the organic molar fraction.

145

$$x_{\text{org}} = \frac{n_{\text{org}}}{n_{\text{org}} + n_{\text{inorg}}} \quad (2)$$

and  $n_{\text{inorg}}=0$  mol for purely organic aerosols (making  $\text{O:C}_{\text{mixture}}=\text{O:C}$ ). Other parameters include  $\omega_{\text{org}}$ , fractional  $T_g$  at ERH ( $T_{g(\omega_{\text{org}})}$ ), and viscosity.

## 2.2 AIOMFAC model

Aerosol viscosity at the corresponding ERH and 298 K was estimated using the Aerosol Inorganic-  
 150 Organic Mixtures Functional Groups Activity Coefficients (AIOMFAC) model. (Zuend et al., 2008, 2011) AIOMFAC is a thermodynamics-based group-contribution framework developed for aqueous organic-inorganic mixtures. To predict the viscosity of complex organic-inorganic mixtures, the AIOMFAC-VISC framework conceptually divides the mixture into two distinct sub-systems. For the aqueous inorganic (electrolyte) sub-system, it employs a semi-empirical model based on Eyring's  
 155 absolute rate theory, which relates viscosity to the molar Gibbs energy of activation for viscous flow ( $\Delta g^*$ ) taking into account temperature, ion activities, and ionic strength. Conversely, for the aqueous organic sub-system, the model relies on a combinatorial-activity-weighted approach based on functional group contributions. (Gervasi et al., 2020) Adopting the UNIFAC concept, AIOMFAC segments organic molecules into functional subgroups, treating the liquid mixture as a 'solution of groups' to manage  
 160 chemical complexity. (Fredenslund et al., 1975; Hansen et al., 1991) In our study, the viscosity of mixed organic-inorganic phases was estimated using the default "aquelec" (electrolyte-aware water) approach in AIOMFAC-web. This method represents the effects of ions by first computing the viscosity of the aqueous electrolyte mixture (excluding organics), and then using that sub-system's viscosity to represent a modified viscosity of water in the calculation of the electrolyte-free organic sub-system. Validation  
 165 against experimental data (e.g., Figure 11 in Lilek and Zuend et al.) has demonstrated that the aquelec approach accurately characterizes the viscosity of internally mixed aerosols such as sucrose-nitrate systems, establishing it as a robust tool for our predictions. (Lilek and Zuend, 2022; Song et al., 2021; Zuend et al., 2011) Validation against experimental viscosity data for sucrose–nitrate systems demonstrated that the aquelec approach accurately characterizes the viscosity of internally mixed  
 170 organic–inorganic aerosols, establishing AIOMFAC-VISC as a robust tool for predicting atmospheric aerosol viscosity when experimental measurements are unavailable. (Lilek and Zuend, 2022)

Accordingly, viscosity calculations were performed by specifying the functional groups of the organic

components, inorganic salt composition, organic-to-inorganic ratio (OIR, molar ratio of organic compounds to inorganic salts, excluding water), and a series of organic molar fractions in the solute mixture,  $x_{\text{org}}$ , where the solute consists of organic compounds and inorganic salts, ranging from 0.01 to 0.99 in increments of 0.01. For each system, the viscosity corresponding to the water activity closest to the literature reported ERH (within  $\pm 0.5\%$  RH) was selected. For example, in a 1,2,6-hexanetriol and ammonium sulfate (1:1) mixture, we defined 1,2,6-hexanetriol as 2  $\text{CH}_2(\text{OH})$ , 1  $\text{CH}(\text{OH})$ , 1 C and 3 OH. The viscosity of the mixture system was evaluated at a water activity of 0.47, which closely matches the ERH value in the literature (47.4%). Functional group definitions for all organic components are listed in Table S3. To assess the uncertainty associated with the thermodynamic state predictions, the sensitivity of viscosity ( $\pm \log_{10}(\eta/[\text{Pa.s}])$ ) was calculated for each mixture. This metric quantifies the variation in predicted viscosity resulting from a  $\pm 2\%$  perturbation in aerosol water mass fraction and serves as a proxy for uncertainty related to compositional variability. The corresponding values are listed in Table S1.

### 2.3 Calculation of the fractional glass transition temperature ( $T_{g(\omega_{\text{org}})}$ )

Multiple parameterizations exist for estimating the glass transition temperature ( $T_g$ ) of organic compounds, utilizing predictors such as chemical structure, volatility, or melting point ( $T_m$ ) (Armeli et al., 2023; Galeazzo and Shiraiwa, 2022; Koop et al., 2011; Li et al., 2020). Recent machine-learning models based on chemical structure do not strictly require  $T_m$  information, though including it as an optional parameter can further improve predictive accuracy. (Armeli et al., 2023) The framework proposed by Armeli et al. comprises four distinct modes, contingent upon the specific input parameters available for a given compound: the Functional Group Mode (FGM) and the SMILES Mode (SM). Each mode features a variant that includes melting temperature as an additional feature (FGM with  $T_m$ ; SM with  $T_m$ ) and a version that operates without it (FGM no  $T_m$ ; SM no  $T_m$ ). The  $T_g$  values and parameters calculated according to the methodology of Armeli et al. are summarized in Tables S4 – S7.

The volatility-based parameterization developed by Li et al. (Eq. 3) predicts  $T_g$  from the saturation mass concentration ( $C^0$ ).

$$T_g = 288.7 - 15.33 \times \log_{10}(C^0) - 0.33 \times [\log_{10}(C^0)]^2 \quad (3)$$

$C^0$  was estimated as a function of C, O, N, S numbers of the organic compound ( $\log_{10}C^0 = f(n_C, n_O, n_N, n_S)$ , Eq. 4) (Donahue et al., 2011; Li et al., 2016) :

$$\log_{10} C^0 = (n_C^0 - n_C)b_C - n_O b_O - 2 \frac{n_C n_O}{n_C + n_O} b_{CO} - n_N b_N - n_S b_S. \quad (4)$$

Here,  $n_C^0$  is the reference carbon number;  $n_C$ ,  $n_O$ ,  $n_N$ , and  $n_S$  denote the number of carbon, oxygen, nitrogen, and sulfur atoms, respectively;  $b$  coefficients ( $b_C$ ,  $b_O$ ,  $b_N$ , and  $b_S$ ) denote atomic contributions  
 205 fitted via multi-linear least squares analysis from 30,000 compounds across multiple classes (CH, CHO, CHN, CHON, CHOS, and CHONS). (Li et al., 2016)  $b_{(CO)}$  represents the carbon-oxygen nonideality.

A comparison between the experimentally measured  $T_g$  values of the organic compounds and the predictions derived from the methods of both Li et al. (Li et al., 2020) and Armeli et al. (Armeli et al., 2023) is presented in Table S7.

210 The presence of inorganic salts (e.g., sodium nitrate) has been demonstrated to modulate the  $T_g$  of organic–inorganic mixtures; specifically, depending on the effective  $T_g$  of inorganic salt, the  $T_g$  of the mixture can be either elevated or depressed. (Dette and Koop, 2015) Typically, the  $T_g$  of organic–inorganic mixtures is calculated using the Gordon–Taylor equation (Gordon and Taylor, 1952):

$$T_g = \frac{\omega_1 T_{g1} + \frac{1}{k} \omega_2 T_{g2}}{\omega_1 + \frac{1}{k} \omega_2} \quad (5)$$

215 where  $T_g$  represents the glass transition temperature of the dry binary mixture,  $T_{g1}$  and  $T_{g2}$  are the  $T_g$  values of the respective pure compounds,  $\omega_1$  and  $\omega_2$  denote the mass fractions of the two components, and  $k$  is the Gordon–Taylor constant. However, the glass transition temperature of inorganic compounds is limited, we thus employ the glass transition temperature of the organic–water system ( $T_{g,org-water}$ ) as a proxy for the overall mixture  $T_g$  when analyzing the correlation between efflorescence relative humidity  
 220 (ERH) and  $T_g$ . In Eq. (5), the parameters  $\omega_2$  and  $T_{g2}$  are substituted with the mass fraction  $\omega_{org}$  and the glass transition temperature  $T_{g(org)}$  of the organic component, respectively, where  $\omega_{org}$  is the organic mass fraction calculated via the AIOMFAC model at aerosol ERH.  $T_{g,w}$  is the glass transition temperature of pure water (136 K) (Kohl et al., 2005), while  $k_{GT}$  (the Gordon–Taylor constant) is assumed to be 2.5 for organic–water mixtures. (Koop et al., 2011; Zobrist et al., 2008) These two values are substituted for  
 225  $T_{g1}$  and  $k$  in the equation, with  $\omega_1$  calculated as  $(1-\omega_{org})$ . As reported by Koop et al. (Dette and Koop, 2015), the  $T_g$  of  $NaNO_3$  is specified as 290 K. Using this reported  $T_g$  value for  $NaNO_3$  together with  $k_{GT} = 0.5$ , we calculated the  $T_g$  for dry glycerol/ $NaNO_3$  and dry sucrose/ $NaNO_3$  mixtures across a range of mixing ratios. (Figure S1) It should be noted that assuming a constant Gordon–Taylor parameter ( $k_{GT} = 2.5$ ) for all organic–water mixtures represents a generalized simplification. The  $k_{GT}$  value essentially



230 parameterizes the interaction strength between water and the specific organic solute. Consequently, for organic mixtures with highly complex or varying oxygenated functional groups, the actual  $k_{GT}$  value may deviate from this assigned average (typically estimated within a range of  $2.5 \pm 1.0$ ). This variability introduces a potential, albeit minor, source of uncertainty into the  $T_g$ -based model predictions presented in this study.

235 In addition to the approach used here for estimating  $T_g(\omega_{org})$ , the glass transition temperature of the entire mixture, including contributions from inorganic ions, could also be derived from the predicted mixture viscosity. This can be achieved by combining AIOMFAC-VISC viscosity predictions with an assumed fragility parameter within the Vogel–Tammann–Fulcher framework. (DeRieux et al., 2018) However, the pure-component viscosity parameterization for organics implemented in AIOMFAC-web may be less  
240 accurate than more recent approaches. (Armeli et al., 2023) Moreover, for the purposes of the present regression model, estimating  $T_g$  from mixture viscosity is likely redundant, as it does not provide additional predictive power beyond viscosity itself.

### 3 Results and Discussion

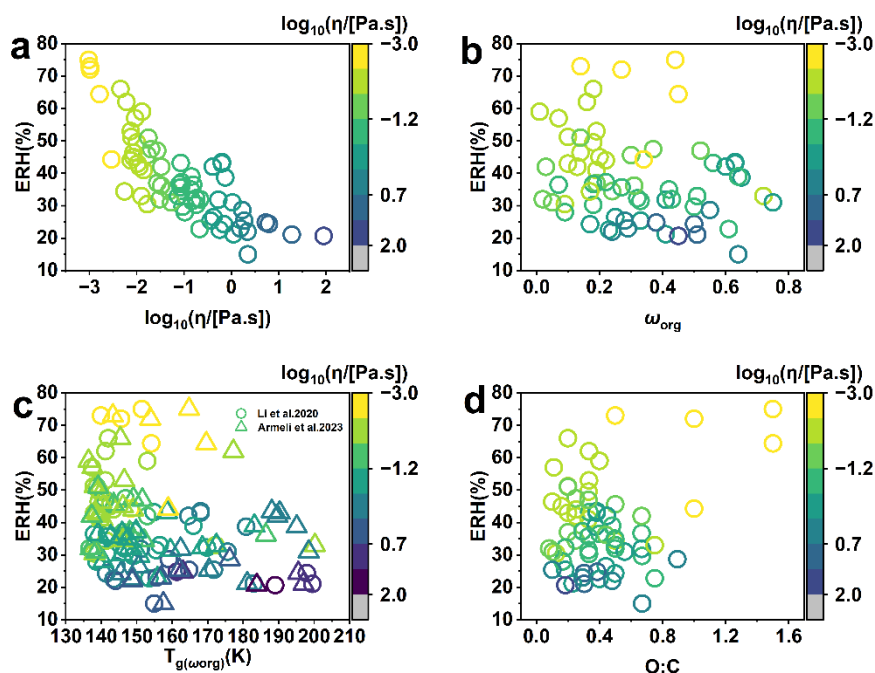
#### 3.1 Correlation between ERH and physicochemical parameters

245 In the study of ERH presented here, we focus on the efflorescence of the inorganic component in internally mixed organic-inorganic aerosols. Therefore, we examined the correlation between ERH and key physicochemical parameters describing the organic components in organic-inorganic mixture aerosols: O:C ratio,  $\omega_{org}$ , and  $T_g(\omega_{org})$  (Figure 1).

Linear regression analysis using the training set (66 data points) showed no significant correlations  
250 between aerosol ERH and any of O:C,  $\omega_{org}$ , or  $T_g(\omega_{org})$  individually ( $R^2 < 0.15$ ; Table S8). Multivariate regression considering the combined influence of three parameters provided only a limited improvement ( $R^2 = 0.25$ , Table S9). According to the nested cross-validated model evaluation reported by Armeli et al. (Armeli et al., 2023) for the full dataset, the SMILES mode with  $T_m$  showed the highest predictive accuracy among the four modes. Therefore, the  $T_g$  values calculated using the SMILES mode with  $T_m$   
255 were selected for comparison with the parameterization developed by Li et al. (Li et al., 2020). The results, presented in Figure 1c, demonstrate that neither methodology yields a statistically significant correlation. Furthermore, analyses employing alternative FGM and SMILES models for  $T_g$  estimation, and  $T_g$  values for organic-inorganic mixtures were also investigated (Figure S2). Similarly, no discernible

correlation with ERH was observed.

When we color-coded the training dataset in Figure 1b-1d based on aerosol viscosity (the logarithm of viscosity estimated using the AIOMFAC,  $\log_{10} \eta$ ), a clear trend emerged: aerosols with lower viscosity generally exhibited higher ERH, while those with higher viscosity tended to have lower ERH. This trend is further supported by the scatter plot of aerosol ERH versus viscosity (Figure 1a), which shows an apparent negative correlation. These results indicate that viscosity may play a key regulatory role in aerosol efflorescence and could be a critical predictor for ERH.



**Figure 1.** Relationships between the ERH of organic-inorganic mixed aerosols. (a) the logarithm of viscosity ( $\log_{10}(\eta/[\text{Pa} \cdot \text{s}])$ ), (b) the organic mass fraction ( $\omega_{\text{org}}$ ), (c) the glass transition temperature ( $T_g(\omega_{\text{org}})$ , K) calculated using the methods of Li et al. (Li et al., 2020) and Armeli et al. (Armeli et al., 2023), respectively, and (d) O:C ratio. The color bar denotes the aerosol viscosity in  $\log_{10}(\eta/[\text{Pa} \cdot \text{s}])$ .

### 3.2 ERH prediction using viscosity

Building on the correlation between viscosity and ERH observed in Figure 1, we developed a linear regression model using the training set (Figure 2). The fitted equation is

$$ERH(\%) = (-10.23 \cdot \log_{10}(\eta/(\text{Pa} \cdot \text{s})) + 27.40) \cdot 100\% \quad (6)$$

The  $R^2$  of the linear fit is 0.65, which suggests a significant correlation between aerosol viscosity and aerosol ERH. This indicates that the influence of organic components on aerosol efflorescence is primarily mediated through their effect on the aerosol viscosity. Interestingly, we found that although

aerosol efflorescence occurs due to the crystallization of aqueous inorganic salt, the correlation between aerosol viscosity and the normalized aerosol ERH defined as  $ERH/ERH_{\text{Inorg}}$  (where  $ERH_{\text{Inorg}}$  is the efflorescence relative humidity of the corresponding inorganic salt system) was even weaker ( $R^2 = 0.59$ , Figure S3). Moreover, the linear fit for organic–ammonium sulfate systems exhibits a similar correlation between ERH and viscosity, with comparable  $R^2$ , slope, and intercept to those obtained for the full dataset (Figure S4). Since the AIOMFAC calculation accounts for inorganic salts, this result emphasizes that the overall aerosol viscosity plays a dominant role in governing efflorescence, overwhelming differences between individual inorganic salts.

Our linear model can predict whether nucleation and crystal growth will occur at a given RH, and whether hygroscopic hysteresis is expected between humidification and dehumidification cycles. Notably, the model predicts that aerosol ERH approaches zero when viscosity reaches  $4.76 \times 10^2 \text{ Pa} \cdot \text{s}$ , and the corresponding 95% confidence interval is  $[1.05 \times 10^2, 3.24 \times 10^3] \text{ Pa} \cdot \text{s}$ . We therefore infer that organic–inorganic aerosols with viscosities exceeding  $4.76 \times 10^2 \text{ Pa} \cdot \text{s}$  are unlikely to undergo efflorescence. The complete viscosity-ERH model is expressed as:

$$ERH(\eta) = \begin{cases} (-10.23 \cdot \log_{10} \eta + 27.4) \cdot 100\%, & \eta < 4.76 \times 10^2 \text{ Pa} \cdot \text{s} \\ 0, & \eta \geq 4.76 \times 10^2 \text{ Pa} \cdot \text{s} \end{cases} \quad (7)$$

It should be emphasized that the empirical data utilized to derive this regression model were obtained at a constant temperature of 298 K. Given the well-established temperature dependence of both ERH and viscosity, the applicability of the current model is limited to standard room temperature conditions (298 K). Extrapolating this model to broader temperature regimes would necessitate further temperature-specific calibrations. In addition, the ERH model presented here is an empirical parameterization based on experimentally reported mean ERH values; therefore, it is intended to capture the occurrence of efflorescence, rather than the full statistical distribution of crystallization events. The model uncertainty reflects the scatter of the experimental ERH data, the stochastic nature of nucleation, and the uncertainty in AIOMFAC-predicted viscosity. The current model is mainly applicable to internally mixed organic–inorganic aerosols containing ammonium sulfate, ammonium nitrate, sodium chloride, or potassium chloride, with organic components primarily consisting of water-soluble C, H, and O, containing compounds, such as dicarboxylic acids and sugars.

We further performed multivariate regression incorporating viscosity ( $\log_{10} \eta$ ), equivalent O:C ratio,  $\omega_{\text{org}}$ , and  $T_g(\omega_{\text{org}})$  as independent variables. The four-variable model demonstrated an improved  $R^2$  of 0.69,

which is significantly different from the ERH-viscosity model (Sig. F change < 0.05, Table S10) due to the major contribution of viscosity and  $T_g(\omega_{org})$  (both showed  $p < 0.05$ , Table S10). We therefore established a bivariate viscosity- $T_g$ -ERH model, and the analysis of variance (ANOVA) showed a Sig. F change (0.685) much greater than 0.05 with the same  $R^2$  of 0.69. This indicated that there was no significant difference between the viscosity- $T_g$ -ERH model and the four-variable model (Table S11), and O:C and  $\omega_{org}$  did not significantly improve predictive performance. The viscosity- $T_g$ -ERH model is expressed as:

$$ERH(\eta, T_g) = (0.25 \cdot (T_g(\omega_{org})/K) - 12.58 \cdot \log_{10}(\eta/(Pa \cdot s)) - 13.28) \cdot 100\% \quad (8)$$

Although adding predictors typically increases the coefficient of determination ( $R^2$ ) by capturing additional variance, the inclusion of  $T_g(\omega_{org})$  improved  $R^2$  by only 0.04. In contrast, the viscosity-ERH model retains strong predictive power, while requiring fewer assumptions, demonstrating broader applicability and greater robustness for predicting ERH across diverse internally mixed organic-inorganic aerosols.

Moreover, the viscosity- $T_g$ -ERH model exhibits practical limitations. The  $T_g(\omega_{org})$  does not account for the effects of inorganic salts due to the limited availability of apparent  $T_g$  values for inorganic components. As shown in Figure S1, we estimated the  $T_g$  of organic/ $NaNO_3$  mixtures by incorporating the  $T_g$  of pure  $NaNO_3$  into the Gordon-Taylor framework, using a  $k_{GT}$  value of 0.5 extrapolated from Dette and Koop. (Dette and Koop, 2015) The result indicates that, regardless of whether the  $T_g$  of pure inorganic salt or water is used, the calculated  $T_g(\omega_{org})$  exhibits no apparent correlation with ERH.

We also note that other parameters, such as aerosol pH, can affect the ERH of organic/inorganic aerosol (Sun et al., 2023); however, to the best of our knowledge, no consistent trend has been established regarding the direction or magnitude of this effect, since the influence of pH is highly dependent on the specific organic composition. Given that the modeled viscosity accounts for organic functional groups, the influence of pH may be implicitly reflected within the overall model uncertainty. Future studies should incorporate more inorganic  $T_g$  data and systematically investigate pH effects to provide more accurate  $T_g(\omega_{org})$  and pH values and better evaluate the role of both parameters in the viscosity-ERH relationship.

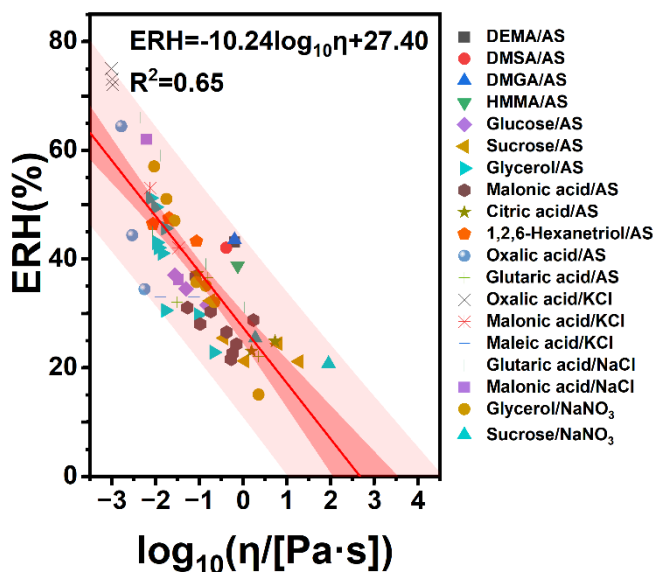


Figure 2. Linear fitting of aerosol ERH for different organic-inorganic aerosols by using viscosity ( $\log_{10} \eta$ ) as the predictor. Data points are color- and shape-coded to distinguish between different organic and inorganic aerosol species. The red solid line represents the linear regression fit between viscosity and ERH. The dark pink band indicates the 95% confidence interval, while the light pink band shows the prediction interval. AS refers to ammonium sulfate, DEMA represents Diethyl malonic acid, DMSA represents 2,2-Dimethyl succinic acid, DMGA represents 3,3-Dimethyl glutaric acid, HMMA represents DL-4-Hydroxy-3-methoxymandelic acid and DHBA represents 2,5-Dihydroxybenzoic acid.

We then validated the established viscosity-ERH model using an independent validation set (32 data points) compiled from the literature. As depicted in Eq. (7), a threshold viscosity of  $4.76 \times 10^2$  Pa·s distinguishes between crystallized and amorphous solid phases in atmospheric aerosols. Figure 3a compares the experimentally measured ERH in the validation set with model predictions. For viscosities  $< 4.76 \times 10^2$  Pa·s (data points to the left of the red dashed line), all validation data fall within the predicted confidence intervals, confirming the reliability of using viscosity to predict the ERH of organic-inorganic mixture aerosols. This result suggests that when aerosol viscosity and ambient RH fall below the regression line (red), aerosols tend to undergo nucleation and crystal growth. For viscosities  $> 4.76 \times 10^2$  Pa·s (data points to the right of the red dashed line), consistent with model predictions, aerosols in the validation set remain in a viscous liquid state and ultimately form amorphous solids without crystallization. As shown in Figure 3b, the data points tightly cluster around the  $y = x$  reference line (black dashed line), with the fitted line (red solid line) closely aligned and a slope near unity ( $R^2 = 0.95$ ), indicating negligible systematic prediction bias and strong agreement between model predictions and experimental measurements. By performing the fitting analysis excluding the (0,0) data points (Figure S5), the slope remains close to unity, while the coefficient of determination decreases but remains

significant ( $R^2 = 0.68$ ). We also validated the viscosity- $T_g$ -ERH model using the same data set, excluding systems for which  $T_g$  could not be estimated. The viscosity- $T_g$ -ERH model also demonstrated good agreement between predicted and measured ERH values (Figure S6).

To further elucidate the mechanistic basis of the model, we examined (i) viscosity changes for fixed chemical systems across RH (20 - 60%) and (ii) the correlation between viscosity at fixed RH and ERH. The slopes of the viscosity–RH relationships for representative systems range from -10 to -60 (Table S12), while the viscosity–ERH correlation weakens substantially when RH is held constant. These results indicate that the observed viscosity–ERH relationship primarily arises from the RH dependence of viscosity and confirm that efflorescence is governed by nucleation and crystal growth kinetics controlled by viscosity at the ERH. Accordingly, the viscosity–ERH model can be interpreted as a representative linear relationship that captures variability due to differences in organic composition and organic–inorganic interactions. It further defines a boundary condition for aerosol phase transitions: below the fitted viscosity–ERH line, aerosols tend to undergo efflorescence, whereas above this line (or beyond the threshold viscosity), aerosols tend to remain viscous without crystallization.

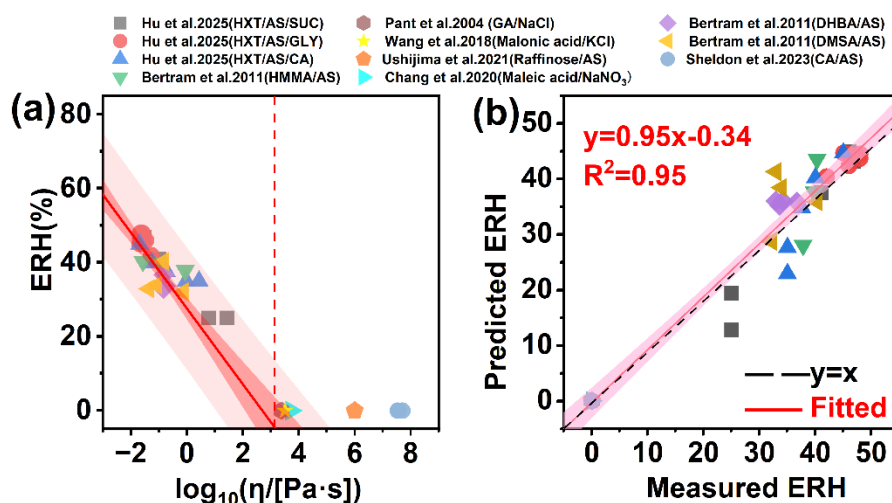


Figure 3. (a) Comparison between the validation set data and viscosity-ERH model prediction (Eq. 6). (b) Comparison of measured ERH values versus model predicted ERH values based on the viscosity-ERH model. The black dashed line denotes the  $y = x$  reference line, and the red line represents the linear regression fit to the data, along with the corresponding coefficient of determination,  $R^2=0.95$ . The pale pink shading represents the 95% confidence interval. DEMA represents Diethyl malonic acid, DMSA represents 2,2-Dimethyl succinic acid, DMGA represents 3,3-Dimethyl glutaric acid, HMMA represents DL-4-Hydroxy-3-methoxymandelic acid and DHBA represents 2,5-Dihydroxybenzoic acid, CA represents citric acid, SUC represents Sucrose, GA represents glutaric acid, HXT represents 1,2,6-hexanetriol.

### 380 3.3 Nucleation kinetics mechanism for viscosity-dependent ERH

Aerosol efflorescence is a kinetically controlled process involving: (1) formation of critical crystal nuclei in supersaturated droplets; (2) subsequent crystal growth, and (3) water evaporation. (Zhang et al., 2012)

In highly viscous aerosols, mass transfer limitations significantly alter nucleation dynamics, making nucleus formation the rate-limiting step. (Ji et al., 2017; Wang et al., 2017) Turnbull and others showed  
385 that the nucleation rate can be expressed as the product of a pre-exponential factor and two thermally activated rate terms (Hollomon and Turnbull, 1953; Kelton et al., 1983; Turnbull, 1969):

$$\tau_x^{-1} = v \exp \left[ \frac{-W(T)}{kT} \right] \exp \left[ \frac{-\Delta G(T)}{kT} \right] \quad (9)$$

where  $W(T)$  represents the kinetic barrier associated with atomic rearrangement, and  $\Delta G(T)$  is the thermodynamic free energy barrier for nucleus formation. The term  $\exp(W(T)/kT)$  represents the  
390 thermally activated atomic rearrangement rate (i.e., atomic mobility) in liquids and is inversely related to viscosity ( $\eta^{-1}$ ). Here, we introduce an approximate proportional relationship between the kinetic barrier term and the logarithm of viscosity:

$$\frac{-W(T)}{kT} = -A \log_{10} \eta \quad (10)$$

where A is a constant linking  $-W(T)/kT$  with  $-\log_{10} \eta$ . Substituting this relation into Eq. (9) and  
395 taking  $\log_{10}$  gives:

$$\log_{10}(\tau_x^{-1}) = -A \log_{10} \eta + \log_{10} v - \frac{\Delta G(T)}{2.303 kT} \quad (11)$$

The empirical relationship between ERH and aerosol viscosity (Eq.(6)) is consistent with Eq. 11, indicating that  $\log_{10} \eta$  serves as a proxy for the kinetic energy barrier  $W(T)$ , while ERH reflects the effective nucleation rate. The intercept achieved in Eq.(6) can therefore be associated with  $\log_{10} v -$

400  $\frac{\Delta G(T)}{2.303 kT}$ . Overall, the viscosity-ERH model parameterization aligns with this framework, showing that

aerosol ERH decreases with increasing  $\log_{10} \eta$ , corresponding to reduced atomic rearrangement rates.

Since ambient RH governs aerosol water activity ( $a_w$ ), the strong correlation between ERH and aerosol viscosity indicates that water activity and molecular mobility jointly control inorganic salt crystallization.

Under low water activity, elevated viscosity suppresses molecular rearrangement, hinders nucleus  
405 formation, reduces nucleation rates, and consequently suppress or inhibits efflorescence. At this critical

viscosity threshold, the aerosol transitions from a liquid to a semi-solid phase, creating an immense kinetic barrier to molecular diffusion. Consequently, the nucleation induction time is prolonged so significantly that efflorescence is effectively inhibited, even over the extended residence times typical of aerosols in the atmosphere (days to weeks).

410 Thermodynamics-based approaches, such as that described by Hodas et al. (Hodas et al., 2016), predict salt crystallization and ERH based on the degree of supersaturation relative to the ion activity product (IAP). These frameworks capture the fundamental requirement that crystallization can only occur when the aqueous phase becomes thermodynamically supersaturated with respect to the relevant solid phase. In this study, we consider aerosol conditions near ERH, where particles are expected to exhibit high ionic  
415 strength due to reduced water content. Under these conditions, we evaluate aerosol viscosity and its influence on phase transitions. Efflorescence is therefore interpreted as a coupled thermodynamic–kinetic process: while supersaturation provides the necessary driving force for crystallization, the actual onset of efflorescence is modulated by kinetic barriers associated with nucleation and crystal growth. In this context, aerosol viscosity plays a key role by limiting molecular diffusion and ion mobility.

420 Thus, thermodynamic and kinetic perspectives are complementary: thermodynamic conditions determine whether crystallization is possible, whereas viscosity provides a practical proxy for the kinetic limitations that govern when crystallization occurs.

While our results demonstrate a clear empirical correlation between viscosity and ERH, it is crucial to acknowledge that predicting ERH solely based on bulk viscosity has inherent limitations. As highlighted  
425 by previous studies (Ciobanu et al., 2010; Kuwata and Martin, 2012), in complex organic-inorganic systems, liquid-liquid phase separation (LLPS) frequently occurs, forming core-shell morphologies. In such core-shell aerosols, the inner aqueous inorganic-rich phase may maintain a low, liquid-like viscosity comparable to that of pure inorganic droplets, yet still exhibit a suppressed ERH. This suppression can arise because the organic coating alters the interfacial energy or acts as a physical barrier, independent of  
430 the inner core’s viscosity. (Ciobanu et al., 2010) Therefore, while bulk viscosity acts as a critical kinetic constraint, the observed ERH is ultimately governed by a complex interplay of particle size, phase morphology, and viscosity.

#### 4 Conclusions and Implications

Our results demonstrate that viscosity is a critical parameter governing aerosol phase-transition kinetics.



435 Although the model incorporating both viscosity and  $T_g$  improves explanatory power, the viscosity-ERH model developed herein offers broader applicability for complex atmospheric aerosols, providing a more practical framework for predicting aerosol phase behavior. According to the viscosity-ERH model, efflorescence is completely suppressed when viscosity exceeds  $4.76 \times 10^2 \text{ Pa}\cdot\text{s}$ .

Regional viscosity predictions by Zhang et al. suggest that in southern and northeastern China (Zhang et al., 2024), SOA are predominantly liquid or have low viscosities ( $<10^4 \text{ Pa}\cdot\text{s}$ ); in central and northeastern China, SOA typically exhibit semi-solid viscosities ( $10^5 - 10^8 \text{ Pa}\cdot\text{s}$ ); while highly viscous or glassy SOA ( $\eta > 10^8 \text{ Pa}\cdot\text{s}$ ) are mainly present in the northwest. **Although SOA viscosity may significantly increase by several orders of magnitude as water content (and consequently aerosol water activity) decreases** (Hosny et al., 2016; Renbaum-Wolff et al., 2013; Saukko et al., 2012; Song et al., 2016a; Yli-Juuti et al., 445 2017), SOA tend to become more oxidized and internally mixed with secondary inorganic aerosols during atmospheric aging. This aging process significantly lowers the overall viscosity of SOA due to the greater hygroscopicity of more oxygenated organics and inorganic salts. (Dette and Koop, 2015) Consequently, in more developed and urbanized regions like southern and northeastern China, atmospheric aerosols are more likely to undergo efflorescence and crystallization, leading to hysteresis in aerosol size and water 450 contents during humidification and dehumidification processes. **From a broader atmospheric perspective, the implications of aerosol phase state should be considered on regional to global scales. Shiraiwa et al. (Shiraiwa et al., 2017) showed that SOA phase state varies strongly with relative humidity and temperature, with SOA in the planetary boundary layer being predominantly liquid in humid tropical and polar regions, semi-solid in mid-latitudes, and solid over dry lands, while SOA in the middle and upper** 455 **troposphere are expected to be mostly glassy.** The hysteresis induced by aerosol efflorescence has important implications for modeling aerosol size evolution, optical properties, and chemical reactivity.

It is important to note that the aerosol viscosities in this study were derived from thermodynamic model predictions, as in-situ measurement techniques for atmospheric aerosol viscosity remain largely unavailable. Although substrate-based methods—such as poke-flow cytometry and fluorescence lifetime 460 imaging microscopy (FLIM)—have been successfully employed to determine SOA viscosity under laboratory conditions, substrate effects must be carefully considered. (Grayson et al., 2015; Hosny et al., 2013) This highlights an urgent need to develop techniques that can directly measure aerosol viscosity at a given RH in ambient environments and to validate viscosity-dependent phase transition predictions. Optical tweezers is a promising tool for quantitative viscosity measurements of levitated particles;

465 however, the technique requires further advancement to measure submicron organic aerosols in real atmospheric conditions. (Reid et al., 2014) Further quantitative evaluation is needed to assess the extent to which viscosity-based predictive tools for aerosol phase state influence simulations of climate and air quality.

#### **Code and data availability**

470 The data supporting this article have been included in the main text and as part of the Supplementary Information: Tables S1-S12, Fig. S1-S6.

#### **Supplement**

The supplement related to this article is available online at:

#### **Author contributions**

475 SC and QH analyzed and visualized the data and wrote the original draft. SC conducted the experiments and model calculations. QH created the original research framework. QH and YL reviewed and edited the paper. QH, PL, and YZ supervised the study. All authors discussed the results and contributed to the article editing.

#### **Competing interests**

480 The contact author has declared that neither they nor their co-author has any competing interests.

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485 the preparation of this work the authors used Generative Pretrained Transformer 4 (GPT-4) to detect and correct grammatical errors in the draft. After using this tool/service, the authors reviewed and edited the

content as needed and take full responsibility for the content of the publication.

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