



Theory for Optically Observing Cloud Aerosol Critical Activation

Diameter

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Abstract:

Aerosol activation is the fundamental microphysical process governing cloud droplet formation. The critical activation diameter, D_a , is a key parameter that characterizes cloud supersaturation and aerosol activation and serves as a fundamental variable in simulations of aerosol–cloud interactions. Kuang et al. (2025) proposed a machine-learning framework for the instantaneous retrieval of D_a from spectral scattering measurements of interstitial and activated aerosols. However, the lack of a rigorous theoretical foundation has limited the understanding of the method's physical basis, applicability, and limitations in field observations. Here, we establish the missing theoretical foundation by integrating aerosol activation theory with Mie scattering theory to derive a unified closed-form solution linking optical scattering to the D_a . The resulting theory provides the first analytical explanation for the approximately linear relationship between D_a and the scattering fraction of interstitial aerosols reported by Kuang et al. (2025), while also revealing the strongly nonlinear behavior near the limiting scattering regimes. Furthermore, the theoretical detection limit of the optical approach is quantified, and its potential applicability to different cloud types is evaluated. Overall, this work provides a rigorous theoretical foundation for optical observations of cloud aerosol activation. The unified analytical solution bridges all limiting regimes, clarifies the physical basis of optical retrievals of the critical activation diameter, and offers theoretical guidance for the design and optimization of future optical instruments for cloud activation measurements.



31 1. Introduction

32 Aerosol-cloud interactions are fundamental for cloud physics and play significant roles in weather
 33 modifications (Kreidenweis et al., 2018) and climate change projections (Fan et al., 2016). Aerosol
 34 activation plays a critical role in aerosol-cloud interactions which determines the number of cloud
 35 droplets and thus constitutes the central concept of the classical Twomey theory (Twomey, 1959). The
 36 critical activation diameter (D_a) is the most important parameter in quantifying cloud droplet number
 37 concentrations with given aerosol particle number size distributions in models (Abdul-Razzak et al.,
 38 1998; Abdul-Razzak and Ghan, 2000, 2002) and also the key parameter in constraining aerosol
 39 activation properties and estimating effective supersaturation in clouds (Ditas et al., 2012; Hammer et
 40 al., 2014; Zíková et al., 2020; Wainwright et al., 2021; Dedrick et al., 2025).

41 Theoretically, on the basis of Köhler activation theory, D_a is mostly determined by aerosol
 42 hygroscopicity and the supersaturation experienced by aerosols. In clouds, supersaturation might
 43 fluctuate quickly, which result in quick changes of aerosol activation processes, therefore high-time-
 44 resolution observations are necessary for D_a observations. Several methods have been proposed to
 45 observe D_a in clouds/fogs. The direct way is to observe size-resolved activation ratio using aerosol
 46 size distribution techniques in combination with different sampling inlets and adryer system. For
 47 example, the combination of total suspended particle (TSP) inlet with the cloud droplet sampling, such
 48 as Ground-based Counterflow Virtual Impactor (GCVI) (Noone et al., 1988) and Pumped Counterflow
 49 Virtual Impactor (PCVI) (Boulter et al., 2006; Motos et al., 2019a). Another way could also achieve
 50 this purpose by integrating TSP and normal impactors which allow interstitial aerosols to be sampled
 51 (Henning et al., 2004; Motos et al., 2019b; Zíková et al., 2020), for example, the impactors with
 52 aerodynamic cutoff diameter of 1 μm and 2.5 μm (PM_{10} and $\text{PM}_{2.5}$). Despite the differences of sampling
 53 inlets, it is not the fundamental reason that affects the high time resolution and even vertical
 54 observations of D_a because multiple inlets could be used simultaneously to overcome the inlet
 55 limitations. The key issue is that the accurate aerosol size distribution usually needs minutes level to
 56 finish the scanning procedure, for example, the most popular scanning mobility particle sizer (SMPS).
 57 Other optical techniques could achieve the seconds level observations of aerosol size distributions;
 58 however, their size ranges are generally limited. For example, the aerodynamic particle sizer (APS)
 59 could only cover size range of larger than about 500 nm, and some other low-cost optical particle
 60 counter which are rarely capable of extending the size range down to 100 nm (Gao et al., 2013). In
 61 addition, these optical techniques usually bear uncertainties of diameter conversion which could be
 62 significantly impacted by aerosol properties such as refractive index and density.



Other methods were also used to indirectly infer D_a and could not achieve online estimations. For example, using the observed cloud number concentrations and aerosol size distributions to estimate D_a . This approach integrates the out-of-cloud or below-cloud aerosol PNSD from the largest diameter downward until the integral matches the measured in-cloud droplet number concentration (Mazoyer et al., 2019; Dedrick et al., 2025). However, this method suffers from biases caused by spatial and temporal variabilities of aerosols in clouds and neglecting the non-linear activated-curve transition caused by chemical heterogeneity of aerosol population. In addition, the Hoppel Minimum Method was also used to estimate D_a . It estimates D_a experimentally from aerosol size distribution measurements by identifying the Hoppel minimum, a morphological valley separating the Aitken and accumulation modes (Gong et al., 2023). While the applicability of this method in different regions and seasons needs further verification. Especially, this distinctive bimodal structure might be frequently obscured by primary emissions and secondary particle formation in complex continental environments (Cai et al., 2020; Hou et al., 2026).

In summary, a new theory and method is urgently needed to achieve the high-time-resolution D_a observations in clouds. The optical machine learning method proposed in Kuang et al. (2025) matches this requirement. The physical thinking of this method has been made clear in Kuang et al. (2025), however, the detailed theoretical foundations of this method need further clarifications and its potential applications and limitations in field cloud measurements need to be further explored in a theoretical way to guide the future optical instrument design. This study addresses this aspect by clearly deducing the relationships between scattering properties of inactivated and activated aerosols and D_a directly from the combination of the Mie theory and activation theory. The theoretical analysis explains clearly the physical connections between D_a and the dry-state scattering fractions of interstitial aerosols in total aerosol populations and could pave the way for sophisticated optical instrument development that aims for aerosol activation observations in clouds.

2. Theoretical formulations

2.1 Analytical Formulation of the Interstitial aerosol Scattering Fraction

The results of Kuang et al. (2025) demonstrated clear observational evidence that the D_a is almost linearly correlated with the submicron scattering fraction of interstitial aerosols (f_{sp}) in the total aerosol populations which depends on the scattering Ångström exponent (SAE) which are determined mainly by the shape of aerosol size distribution. The qualitative physical intuition behind this phenomenon is that the larger the D_a , the higher the f_{sp} and their relationships are likely shaped by aerosol size distributions (Kuang et al., 2018) which could be constrained by spectral scattering measurements of



95 aerosols. However, the theoretical understanding behind these linear relationships as well as the
 96 application boundaries of this optical method need to be further clarified. Here, we address this by
 97 establishing a theoretical framework that starts from connecting f_{sp} and D_a on the basis of Mie theory
 98 and aerosol activation theory.

99 The f_{sp} is formally defined as the scattering ratio of interstitial aerosols to the total scattering of
 100 submicron aerosols:

$$101 \quad f_{sp} = \frac{\int_{-\infty}^{\infty} [1 - F_{act}(D_p)] \frac{d\sigma_{sp}(\lambda)}{d\ln D_p} d\ln D_p}{\int_{-\infty}^{\infty} \frac{d\sigma_{sp}(\lambda)}{d\ln D_p} d\ln D_p} \quad (1)$$

102 where $F_{act}(D_p)$ represents size-resolved aerosol activation ratios and $\sigma_{sp}(\lambda)$ represents the aerosol
 103 scattering coefficient at wavelength λ , $F_{act}(D_p)$ could be represented using the error function:

$$104 \quad F_{act}(D_p) = \frac{MAF}{2} \left[1 + \operatorname{erf} \left(\frac{\ln D_p - \ln D_a}{\sqrt{2} \ln \sigma_a} \right) \right] \quad (2)$$

105 where MAF is the maximum activation fraction and $\ln \sigma_a$ represents activation broadening associated
 106 with chemical composition heterogeneity (Tao et al., 2018), turbulence (Shawon et al., 2021; Anderson
 107 et al., 2023) and even supersaturation history that droplets have been experienced (Kuang et al., 2025).
 108 Note that this form is slightly different with the typical error function that were usually used in fitting
 109 size-resolved activation ratio measurements (Rose et al., 2008) which is discussed in Sect.S1 and the
 110 slight change is for the purpose of explicitly integrating.

111 While the explicit formula of $F_{act}(D_p)$ is convenient, the challenge remains for size-resolved
 112 aerosol scattering due to the nonlinear responses of aerosol scattering to size changes. It was well
 113 known that aerosol scattering is generally proportional to volume for submicron aerosols (Kuang et al., 2018)
 114 and size-resolved aerosol scattering is mainly shaped by the aerosol volume size distribution
 115 (Kuang et al., 2020) which is in general dominantly contributed by accumulation mode aerosols (Bian
 116 et al., 2014) for submicron aerosols. These characteristics resulted in that submicron aerosol volume
 117 size distributions could be generally represented using a lognormal distribution. Therefore, we first
 118 express the volume size distribution $v(D_p)$ as:

$$119 \quad v(D_p) = \frac{V_{tot}}{\sqrt{2\pi} \ln \sigma_g} \exp \left(-\frac{(\ln D_p - \ln D_{g,v})^2}{2 \ln^2 \sigma_g} \right) \quad (3)$$



120 where V_{tot} is the total aerosol volume concentration, $D_{g,v}$ is the volume geometric mean diameter, and
 121 σ_g is the geometric standard deviation. According to the Mie theory, the scattering efficiency
 122 parameter Q_{sca} for a dielectric sphere is an infinite multipole expansion:

$$123 \quad Q_{sca}(x, m) = \frac{2}{x^2} \sum_{n=1}^{\infty} (2n+1) (|a_n(x, m)|^2 + |b_n(x, m)|^2) \quad (4)$$

124 where $x = \pi D_p / \lambda$ is the dimensionless size parameter, m is the complex refractive index, and a_n and
 125 b_n are the Mie scattering coefficients. At visible wavelengths, accumulation mode aerosols that
 126 contribute most (Kuang et al., 2018) to submicron aerosol scattering (100-800 nm) fall within the
 127 primary Mie resonance transition zone ($x \sim 1$). Applying a first-order logarithmic Taylor expansion
 128 near the accumulation mode centroid simplifies results of Equation (4) to a generalized power-law
 129 approximation:

$$130 \quad Q_{sca}(D_p) \approx C(m, \lambda) \cdot D_p^{k_{mie}} \quad (5)$$

131 where $k_{mie} = \partial \ln Q_{sca} / \partial \ln D_p$ is the local scaling exponent and sometimes approximated as 1 for
 132 simplification (Kuang et al., 2020). The parameter C stems analytically from the multipole expansion
 133 and accounts for roles of wavelength and refractive index. Given these analytical formulations, the
 134 scattering cross-section of a single particle $C_{sca}(D_p)$ relates to its volume ($V = \frac{\pi}{6} D_p^3$) via $C_{sca}(D_p) =$
 135 $Q_{sca}(D_p) \cdot \pi \cdot (D_p/2)^2 = Q_{sca}(D_p) \cdot \frac{3}{2D_p} \cdot V$. Therefore, size-resolved aerosol scattering could be
 136 expressed as:

$$137 \quad \frac{d\sigma_{sp}}{d \ln D_p} = \frac{3}{2D_p} Q_{sca}(D_p) v(D_p) = \frac{3C}{2} D_p^{k_{mie}-1} v(D_p) \quad (6)$$

138 Substitute Eq.4 and 6 into Eq.1, an integral equation that relates f_{sp} to D_a yields. Multiplying a log-
 139 normal distribution by D_p^n ($n = k_{mie} - 1$) yields an identically shaped, shifted distribution with its
 140 geometric mean scaled by $\exp(n \ln^2 \sigma_g)$. Setting the new optical peak diameter as the scattering
 141 equivalent mean diameter ($D_{g,sca}$), which could be explicitly formulated as:

$$142 \quad D_{g,sca} = D_{g,v} \exp((k_{mie} - 1) \ln^2 \sigma_g) \quad (8)$$

143 when $k_{mie} = 1$, the scattering peak diameter aligns perfectly with the volume peak diameter ($D_{g,sca} =$
 144 $D_{g,v}$).



145 To solve this integral, we normalize the scattering kernel by defining the Optical Scattering Size
 146 Distribution (OSSD), $p_{sca}(\ln D_p)$, as a probability density function:

$$147 \quad p_{sca}(\ln D_p) = \frac{\frac{d\sigma_{sp}(\lambda)}{d\ln D_p}}{\int_{-\infty}^{\infty} \frac{d\sigma_{sp}(\lambda)}{d\ln D_p} d\ln D_p} \quad (9)$$

148 Since $p_{sca}(\ln D_p)$ would be integrated as 1 over the whole space and forms a log-normal distribution
 149 centered at $\ln D_{g,sca}$ with a standard deviation of $\ln \sigma_g$, the f_{sp} formula simplifies to:

$$150 \quad f_{sp} = 1 - \int_{-\infty}^{\infty} F_{act}(D_p) p_{sca}(\ln D_p) d\ln D_p \quad (10)$$

151 Substituting the activation function $F_{act}(D_p)$ into Equation (10) yields:

$$152 \quad f_{sp} = 1 - \frac{\text{MAF}}{2} - \frac{\text{MAF}}{2} \int_{-\infty}^{\infty} \text{erf}\left(\frac{\ln D_p - \ln D_a}{\sqrt{2} \ln \sigma_a}\right) p_{sca}(\ln D_p) d\ln D_p \quad (11)$$

153 To solve this integral analytically, we utilize the mathematical convolution identity for Gaussian and
 154 error functions:

$$155 \quad \int_{-\infty}^{\infty} \text{erf}(Ax + B) \frac{1}{\sqrt{2\pi}\sigma_s} e^{-\frac{(x-\mu_s)^2}{2\sigma_s^2}} dx = \text{erf}\left(\frac{A\mu_s + B}{\sqrt{1 + 2A^2\sigma_s^2}}\right) \quad (12)$$

156 By mapping the physical variables to the mathematical identity using $x = \ln D_p$, the log-normal OSSD
 157 p_{sca} translates to a normal distribution with mean $\mu_s = \ln D_{g,sca}$ and variance $\sigma_s^2 = \ln^2 \sigma_g$. The
 158 arguments of the error function map to $A = 1/(\sqrt{2} \ln \sigma_a)$ and $B = -\ln D_a/(\sqrt{2} \ln \sigma_a)$. We define the
 159 orthogonal sum of the size distribution width and the activation width as the total broadening factor
 160 σ_d :

$$161 \quad \sigma_d = \sqrt{\ln^2 \sigma_g + \ln^2 \sigma_a} \quad (13)$$

162 Utilizing the property $\text{erf}(-x) = -\text{erf}(x)$ to handle the logarithmic terms, the exact closed-form
 163 expression for f_{sp} could be finally derived as:

$$164 \quad f_{sp} = 1 - \frac{\text{MAF}}{2} + \frac{\text{MAF}}{2} \text{erf}\left(\frac{\ln D_a - \ln D_{g,sca}}{\sqrt{2} \sigma_d}\right) \quad (14)$$

165 Which explicitly relates f_{sp} to D_a . As discussed before both $D_{g,sca}$ and σ_g are essential volume size
 166 distribution parameters.



167 2.2 Linking Scattering Spectral dependence to Aerosol volume Size Distributions

168 While Equation (14) establishes a physical linkage, an optical retrieval requires parameterizing
 169 $D_{g,v}$ with optical measurables. It was well known that a strong relationship exists between the aerosol
 170 mean size and the scattering spectral dependence of aerosols, and this dependence was usually
 171 characterized by SAE as aforementioned. If scattering spectra cover wide ranges of wavelengths or
 172 angles could be measured, aerosol size distribution could even be directly retrieved through Mie
 173 scattering closure (Espinosa et al., 2017). However, aerosol scattering of limited wavelengths was
 174 usually measured and made only constant SAE available. For example, multiple wavelength
 175 nephelometer was usually applied for this purpose. Given this, we define a proxy function χ that
 176 expresses $D_{g,v}$ using SAE and the σ_g :

$$177 \log_{10}(D_{g,v}) = \chi(\text{SAE}, \sigma_g) \quad (15)$$

178 This definition helps transform the size parameter into an optically derivable quantity. Substituting
 179 this proxy for $D_{g,v}$ and inverting Equation (14) yields the closed-form expression for the critical
 180 activation diameter D_a :

$$181 D_a = 10^{\chi(\text{SAE}, \sigma_g)} \cdot e^{(k_{mie}-1)\ln^2 \sigma_g} \cdot \exp \left[\sqrt{2} \sigma_d \cdot \text{erf}^{-1} \left(\frac{2f_{sp} - 2 + \text{MAF}}{\text{MAF}} \right) \right] \quad (16)$$

182 This derivation applies across the entire f_{sp} domain of $[1 - \text{MAF}, 1]$ and requires only optical inputs
 183 (f_{sp} , SAE) alongside aerosol parameters (σ_g , σ_d , MAF). The formula that characterizes the function
 184 $\chi(\text{SAE}, \sigma_g)$ would be discussed in Sect 3.1.

185 By the way, in measurements of size-resolved aerosol activation ratios using a cloud condensation
 186 nuclei counter at the ground site, the MAF can sometimes be substantially lower than unity because
 187 of the presence of externally mixed hydrophobic aerosol particles (Tao et al., 2018). However, these
 188 measurements are typically limited to particle diameters below 400 nm, which may bias the estimated
 189 MAF. Comprehensive field measurements of size-resolved aerosol activation over a broader particle
 190 size range, particularly under in-cloud conditions, are therefore needed to better constrain the
 191 MAF in future studies.

192

193 2.3 Regime-Specific Approximate Formulas

194 Although the exact unified solution of Eq. (16) fully characterizes the entire parameter space, its
 195 mathematical complexity hinders direct physical interpretation. To elucidate the underlying linkage



regimes, we derive three asymptotic approximations by examining the limiting behavior of the inverse error function. It should be emphasized that these asymptotic expressions are intended to provide physical insight rather than numerical reconstruction. Constructing a piecewise solution from these local approximations would introduce truncation errors at the regime boundaries.

(1). Linearization (transition regime): Near $f_{sp} \approx 1 - MAF/2$ or $0.3 < f_{sp} < 0.7$, a Taylor expansion simplifies Eq. (16) to a linear form $D_a \approx k \cdot f_{sp} + b$, where:

$$k = 10^{\chi(\text{SAE}, \sigma_g)} \cdot e^{(k_{mie}-1)\ln^2 \sigma_g} \cdot \frac{\sqrt{2\pi(\ln^2 \sigma_g + \ln^2 \sigma_a)}}{MAF} \quad (17)$$

and

$$b = 10^{\chi(\text{SAE}, \sigma_g)} \cdot e^{(k_{mie}-1)\ln^2 \sigma_g} \cdot \left[1 - \frac{\sqrt{2\pi}\sigma_d}{MAF} \left(1 - \frac{MAF}{2} \right) \right] \quad (18)$$

Asymptotic regimes (Boundary Zones): Near the physical boundaries ($y \rightarrow \pm 1$), the linear scheme breaks down. Using the asymptotic approximation $\text{erf}^{-1}(y) \approx \text{sgn}(y)\sqrt{-\ln(1 - |y|)}$, we define the boundary behaviors.

(2). For low interstitial scattering fractions ($f_{sp} \rightarrow 1 - MAF$):

$$D_a \approx 10^{\chi(\text{SAE}, \sigma_g)} \cdot e^{(k_{mie}-1)\ln^2 \sigma_g} \cdot \exp \left[-\sqrt{-2\sigma_d^2 \cdot \ln \left(\frac{2(f_{sp}-1+MAF)}{MAF} \right)} \right] \quad (19)$$

(3). For high interstitial scattering fractions ($f_{sp} \rightarrow 1$):

$$D_a \approx 10^{\chi(\text{SAE}, \sigma_g)} \cdot e^{(k_{mie}-1)\ln^2 \sigma_g} \cdot \exp \left[\sqrt{-2\sigma_d^2 \cdot \ln \left(\frac{2(1-f_{sp})}{MAF} \right)} \right] \quad (20)$$

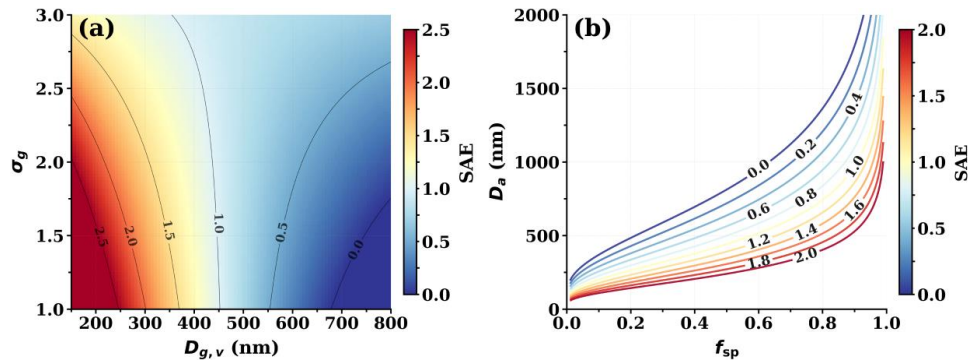
For the observations shown in (Kuang et al., 2025), most of f_{sp} points located in the range of 0.3 to 0.8 thus corresponds to the linearization regime which explains the general linear relationships between D_a and f_{sp} . In the following, the theoretical formulation was further verified directly using the field measurements and the applications of this optical method in observing cloud critical activation diameter was discussed.

3 Results and discussions

3.1 Theory Verifications using Field measurements



220 The theoretical relationship between D_a and f_{sp} is further extended and evaluated using
 221 simultaneous measurements of D_a and f_{sp} , and aerosol size distributions obtained during the AQ-
 222 SOFAR campaign conducted at the Gucheng site on the North China Plain. Details of the
 223 measurements of D_a and f_{sp} are provided in Kuang et al. (2025). Briefly, aerosol size distributions
 224 and multiwavelength scattering coefficients of both submicron interstitial aerosols and total aerosols
 225 were measured using a combination of the advanced aerosol–fog sampling system (Kuang et al., 2024),
 226 aerosol sizing instruments, and a three-wavelength integrating nephelometer operating at 450, 525,



227 **Figure 1. (a)** Variations of SAE under different $D_{g,v}$ and σ_g , with the square region represents general
 228 areas of fitted $D_{g,v}$ and σ_g during fog events of the AQ-SOFAR campaign as shown in Fig.S2; **(b)** The
 229 relationships between D_a and f_{sp} under different SAE conditions.
 230 and 635 nm.

231 There is no explicit analytical expression for $\chi(\text{SAE}, \sigma_g)$. Therefore, we investigated the
 232 dependence of SAE on $D_{g,v}$ and σ_g through Mie scattering calculations of $\sigma_{sp}(450)$ and $\sigma_{sp}(525)$
 233 under different combinations of $D_{g,v}$ and σ_g . The simulated results are presented in Fig. 1a, and details
 234 of the calculations are provided in Sect. S2. As shown in Fig. 1a, SAE decreases monotonically with
 235 increasing $D_{g,v}$ and decreasing σ_g . The fitted $D_{g,v}$ and σ_g values derived from the AQ-SOFAR
 236 campaign are also shown in Fig. S2a. These observations exhibit a positive correlation ($R^2=0.23$),
 237 indicating that variations in $D_{g,v}$ and σ_g are not independent. To account for this covariance, we
 238 calculated SAE using the observed pairs of $D_{g,v}$ and σ_g and subsequently parameterized its
 239 dependence on both variables with the following empirical relationship:

$$237 \quad \text{SAE} = a \cdot \log_{10} D_{g,v} + b \cdot \sigma_g + c \cdot \sigma_g \cdot \log_{10} D_{g,v} + d \quad (21)$$

238 The fitted coefficients are $a=-7.11$, $b=-4.846$, $c=1.836$ and $d=19.84$. The parameterizations reproduced
 239 the simulated SAE well with a coefficient of determination (R^2) of 0.96 (Fig.S2b). Using Eq. (16) we



240 further examined the relationship between D_a and f_{sp} by fixing $MAF=1$, $\sigma_a=1.2$ and $\sigma_g=1.8$,
 241 corresponding to the mean values observed during the three consecutive fog events in the AQ-SOFAR
 242 campaign. The calculated results are shown in Fig. 1b. As shown, the relationship between D_a and f_{sp}
 243 is approximately linear over the f_{sp} range of 0.2-0.8. Specifically, for the linear expression $D_a = k \cdot$
 244 $f_{sp} + b$ in the f_{sp} range of 0.2-0.8, the fitted coefficients are $k=703.4$ and $b=103.2$ for SAE of 1 and
 245 $k=519.0$ and $b=76.1$ for SAE of 1.5. To evaluate the applicability of this simplified relationship, the
 246 calculated D_a values derived from the mean observed SAE of 1.2 ($D_a = 625.9 \cdot f_{sp} + 84.6$) were
 247 compared with the observed D_a values. The comparison (Fig. 2a) yields an R^2 of 0.57, with most data
 248 points falling within the $\pm 20\%$ deviation lines, demonstrating that the simplified linear relationship
 249 provides a reasonable approximation under typical fog conditions. This agreement also provides a
 250 direct field validation of the proposed theoretical framework using independent measurements of
 251 aerosol size distributions and aerosol activation characteristics. Nevertheless, the relationship becomes
 252 nonlinear, while remaining monotonic, as f_{sp} approaches either 0 or 1.

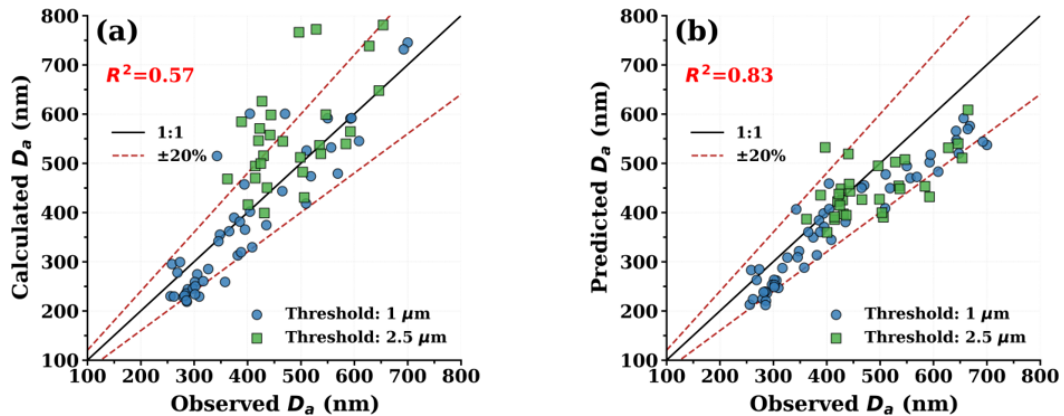


Figure 2. Comparisons between observed D_a and calculated D_a using 1 and 2.5 μm as the threshold of interstitial aerosols with the derived linear formula (a) and with the machine learning framework that accounts for SAE variations (b).

253 As demonstrated by Kuang et al. (2025), a machine-learning framework can account for the
 254 dependence on the scattering Ångström exponent (SAE) as well as the nonlinear behavior near regime
 255 boundaries. Here, the dataset-generation framework developed by Kuang et al. (2025) was adopted to
 256 further extend the analysis. In brief, aerosol particle number size distributions (PNSDs) measured
 257 during six field campaigns over the North China Plain (Kuang et al., 2018) were used to represent a
 258 wide range of atmospheric aerosol conditions. For each measured PNSD, 100 activation ratio curves
 259 were randomly generated using Eq. (2). The scattering coefficients of submicron interstitial aerosols
 260 and total aerosols (interstitial + activated) at 450, 525, and 635 nm were then calculated using the same



261 Mie-theory procedure as described in Kuang et al. (2025), corresponding to the three wavelengths
262 measured by the Aurora 3000 nephelometer during the AQ-SOFAR campaign. A random forest model
263 (Model 1) was subsequently trained using the simulated dataset, with only f_{sp} at 525 nm and the total
264 aerosol scattering coefficients at 450, 525, and 635 nm as input variables and the D_a as the target
265 variable. By incorporating scattering measurements at three wavelengths, the model implicitly
266 accounts for variations in SAE and, therefore, aerosol size distribution. The trained model was then
267 applied to the AQ-SOFAR observations using the measured f_{sp} at 525 nm together with the total
268 aerosol scattering coefficients at the three wavelengths as inputs. As shown in Fig. 2b, the predicted
269 D_a agrees well with the observations compared with the linear formula of $SAE=1.2$, yielding $R^2=0.83$,
270 substantially higher than that obtained using the analytical parameterization alone (Fig. 2a). The mean
271 relative deviation is approximately 11%.

272 However, as indicated by Eq. (16), the relationship between D_a and f_{sp} is influenced not only by
273 SAE but also by MAF, σ_g and σ_a . Therefore, constraining only f_{sp} and SAE is insufficient to uniquely
274 determine D_a , and additional constraints are needed. Physically, MAF, σ_g and σ_a primarily affect the
275 shapes of the dry-state size distributions of the interstitial and activated aerosol populations. As
276 demonstrated by Kuang et al. (2025), these effects can be further constrained through the spectral
277 scattering characteristics of the two aerosol populations. Compared with Model 1, Model 2 (the Model
278 proposed in Kuang et al. (2025)) additionally incorporates dry scattering coefficients of interstitial
279 aerosols, $\sigma_{sp,inter}(dry, \lambda)$, together with those of total aerosols, $\sigma_{sp,all}(dry, \lambda)$ at 450, 525, and 635
280 nm. These six scattering parameters simultaneously contain information on f_{sp} (through
281 $\sigma_{sp,inter}(dry, \lambda)/\sigma_{sp,all}(dry, \lambda)$) and the spectral scattering characteristics of both the interstitial and
282 activated aerosol populations (through spectral dependence subtraction of $\sigma_{sp,all}(dry, \lambda) -$
283 $\sigma_{sp,inter}(dry, \lambda)$). Together, they provide stronger constraints on the aerosol size distributions
284 governing the D_a and f_{sp} relationship. Consequently, the prediction performance is further improved,
285 with R^2 increasing to 0.88. All predicted values fall within the $\pm 10\%$ deviation lines (Fig. 1d in Kuang
286 et al., 2025), and the mean relative deviation decreases to 8.5%. The relatively modest improvement
287 of Model 2 over Model 1 suggests that accounting for the nonlinearity of the D_a and f_{sp} relationship
288 and its dependence on SAE contributes most to the retrieval accuracy, whereas additional constraints
289 on the size distributions of the interstitial and activated aerosol populations provide a secondary
290 improvement.



291 In summary, this study establishes the theoretical basis for optically retrieving D_a , whereas the
 292 machine-learning framework developed by Kuang et al. (2025) provides a practical implementation
 293 by incorporating the nonlinear and multidimensional dependencies revealed by the theory.

294 3.2 Theoretical detection limit and practical applicability

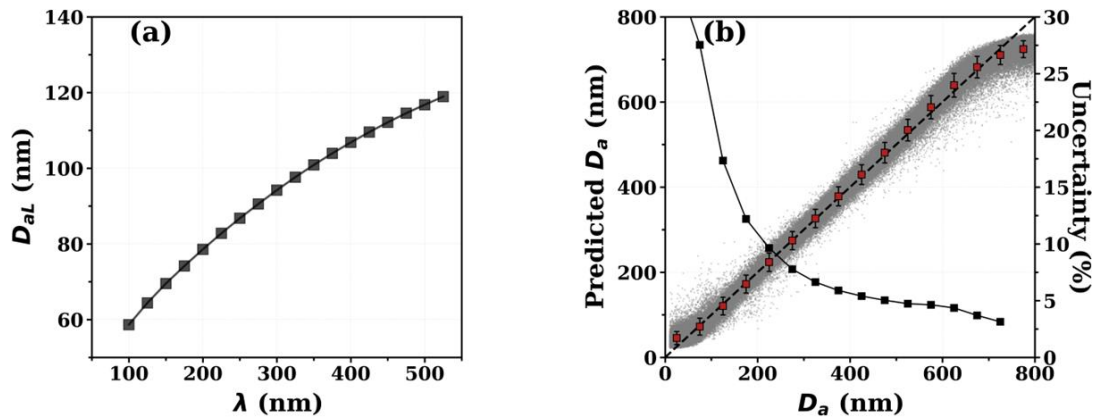


Figure 3. (a) Simulated relationships between D_{al} and λ ; (b) Comparisons of all prescribed D_a and predicted D_a values represented by scatter points with dry state scattering coefficients of interstitial and total aerosols at wavelengths of 100, 200 and 300 nm as inputs; Scatter points are further binned with interval of 50 nm, averages and standard deviations represented by red squares and their error bars, black squares represent relative uncertainty of the right axis at each bin.

295 Although the optical retrieval of D_a from aerosol scattering measurements was first proposed and
 296 demonstrated by Kuang et al. (2025), its physical limitations and applicable scenarios have not been
 297 theoretically discussed. Here, we extend the previous work based on the theoretical framework
 298 developed in this study.

299 The first aspect concerns the lower detection limit of optically retrieving D_a . The physical basis
 300 of the proposed method is the monotonic decrease of f_{sp} with decreasing D_a . However, the size-
 301 resolved cumulative contribution to aerosol scattering reported by Kuang et al. (2018) indicates that,
 302 at a wavelength of 550 nm, particles smaller than approximately 100 nm contribute negligibly to total
 303 aerosol scattering. Consequently, f_{sp} approaches zero as D_a decreases to around 100 nm, causing the
 304 monotonic relationship between D_a and f_{sp} to break down. Therefore, activation diameters below this
 305 threshold cannot be uniquely inferred from scattering measurements. Here, we define the activation
 306 diameter corresponding to $f_{sp}=0.1\%$ (detectable changes) as the lower detection limit, D_{al} . As shown
 307 in Fig. S3, D_{al} depends strongly on the measurement wavelength. Shorter wavelengths increase the
 308 scattering contribution of smaller particles and therefore reduce the detection limit. The quantitative



relationship between D_{al} and wavelength is presented in Fig. 3a. The results indicate that D_{al} decreases to approximately 60 nm when the measurement wavelength reaches 100 nm. Such wavelengths may become feasible for aerosol scattering measurements in the future, given the rapid development of ultraviolet and vacuum-ultraviolet light sources in other technological fields.

As demonstrated in Sect. 3.1, the machine-learning framework incorporates not only f_{sp} but also the spectral scattering characteristics of both the interstitial and total aerosol populations, thereby providing stronger constraints on the underlying aerosol size distributions. Such additional information is expected to further improve the retrieval accuracy near the detection limit. To examine this possibility, Model 2 was retrained using simulated scattering coefficients at 100, 200, and 300 nm instead of 450, 525, and 635 nm. As shown in Fig. 3b, the lower detection limit is further reduced to approximately 50 nm, while the retrieval uncertainty remains below 10% over the D_a range of 200–700 nm.

Assuming $\kappa=0.4$ for 50 nm particles, corresponding to highly hygroscopic marine aerosols (Huang et al., 2022) and substantially larger than the values typically observed for continental aerosols with average κ of ultrafine aerosols generally below 0.2 (Peng et al., 2020), the corresponding critical supersaturation is approximately 0.52%. Cloud supersaturation is jointly controlled by cloud dynamics and microphysical processes, and available observations and estimations indicate that it is typically below 1% (Romps, 2025), exceeding 2% only in vigorous deep convective systems (Fan et al., 2025; Fan et al., 2026). The ranges summarized in Chapter 17 of Seinfeld and Pandis (2016) further show that maximum supersaturations below approximately 0.5% encompass fogs, stratus clouds, and a large fraction of marine and continental cumulus clouds. Existing aircraft and ground-based observations likewise rarely report supersaturations exceeding 0.5% (Hammer et al., 2014; Mazoyer et al., 2019; Wainwright et al., 2021; Zíková et al., 2020; Kuang et al., 2024; Politovich and Cooper, 1988; Hudson et al., 2010; Hudson et al., 2015; Gong et al., 2023; Dedrick et al., 2025). Therefore, the proposed optical approach is expected to be applicable to most atmospheric cloud and fog environments.

Another important advantage of the proposed approach that should be emphasized is its high temporal resolution. Clouds are highly heterogeneous and evolve rapidly because of turbulent mixing, entrainment, and variations in cloud dynamics. Consequently, supersaturation and aerosol activation can vary substantially on timescales of only a few seconds, and rapid transitions between supersaturated and subsaturated conditions might frequently occur near cloud boundaries. Capturing these transient processes requires high-time-resolution observations of D_a . In contrast to retrievals



341 based on aerosol size distribution measurements, which generally require tens of seconds to several
 342 minutes, most aerosol scattering instruments provide reliable measurements with time resolutions of
 343 second level (Moise et al., 2015). As suggested by Kuang et al. (2025), this high-time-resolution
 344 capability makes the proposed method particularly suitable for high-resolution cloud observations and
 345 may ultimately enable vertical profiling of cloud supersaturation and aerosol activation when
 346 combined with airborne or uncrewed aerial platforms.

347 **4. Conclusions**

348 Aerosol activation is the fundamental microphysical process governing cloud droplet formation
 349 and forms the basis of the Twomey effect, thereby playing a central role in aerosol–cloud interactions.
 350 However, direct in-situ observations of aerosol activation in clouds remain extremely limited because
 351 of the lack of efficient observational techniques. In this study, we have established the theoretical
 352 framework for optically retrieving the aerosol critical activation diameter by bridging aerosol
 353 activation theory and Mie scattering theory. The derived formulation explains the nearly linear
 354 relationship between the critical activation diameter and the submicron scattering fraction of interstitial
 355 aerosols reported by Kuang et al. (2025), thereby providing the physical foundation for optical
 356 observations of aerosol activation in clouds.

357 Based on the proposed theory, the applicability and detection limit of the optical retrieval are
 358 further quantified. The results demonstrate that the proposed approach is capable of retrieving the
 359 critical activation diameter over the supersaturation range encountered in most atmospheric clouds.
 360 Although explicit analytical expressions describing the relationship between the critical activation
 361 diameter and the interstitial aerosol scattering fraction were derived, the machine-learning framework
 362 developed by Kuang et al. (2025) remains the preferred approach for practical applications because it
 363 effectively incorporates the nonlinear and multidimensional dependencies identified by the theory.

364 It should be noted that the proposed theory fundamentally determines the diameter above which
 365 particles grow under supersaturated conditions, whereas the diameter used to distinguish cloud
 366 droplets from aerosols depends on the inlet cut size and should not be regarded as an intrinsic activation
 367 threshold. As shown in Fig. S5 of Kuang et al. (2024), the wet diameter corresponding to cloud droplet
 368 activation varies strongly with supersaturation, exceeding $2.5 \mu\text{m}$ for supersaturations below
 369 approximately 0.06% but decreasing to below $1 \mu\text{m}$ when supersaturation exceeds approximately
 370 0.15%. Consequently, careful selection of the impactor cut size is essential for interpreting optical
 371 observations of aerosol activation.



372 This characteristic also suggests a broader application of the proposed theory. When a PM₁ inlet
373 is employed, for example, particles that grow beyond 1 μm under low-supersaturation conditions
374 ($<0.15\%$) should be interpreted as hygroscopic aerosols rather than activated cloud droplets. Therefore,
375 measurements with multiple impactor inlets combined with multiwavelength scattering observations
376 could simultaneously constrain aerosol hygroscopic growth and activation with high temporal
377 resolution. Such observations would provide a new capability for investigating aerosol–water
378 interactions across the transition from subsaturated to supersaturated conditions and offer valuable
379 insights into cloud microphysical processes.

380

381 **Competing interests.** The contact author has declared that none of the authors has any competing
382 interests.

383

384 **Availability Statement.** All data presented are shown in Figures of this manuscript and more specific
385 data will be made available on request.

386

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389

390 **Author contribution.** YK conceived the theory and wrote the manuscript. QL helped plotted the
391 figures.

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