



1 **A novel method to derive the aerosol hygroscopicity parameter based only on**
2 **measurements from a humidified nephelometer system**

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11

12 **Abstract**

13 Aerosol hygroscopicity is crucial for understanding roles of aerosol particles in atmospheric
14 chemistry and aerosol climate effects. Light scattering enhancement factor $f(\text{RH}, \lambda)$ is one of the
15 parameters describing aerosol hygroscopicity which is defined as $f(\text{RH}, \lambda) =$
16 $\sigma_{sp}(\text{RH}, \lambda) / \sigma_{sp}(\text{dry}, \lambda)$ where $\sigma_{sp}(\text{RH}, \lambda)$ or $\sigma_{sp}(\text{dry}, \lambda)$ represents σ_{sp} at wavelength λ under
17 certain RH or dry conditions. Traditionally, an overall hygroscopicity parameter κ can be retrieved
18 from measured $f(\text{RH}, \lambda)$, hereinafter referred to as $\kappa_{f(\text{RH})}$, by combining concurrently measured
19 particle number size distribution (PNSD) and mass concentration of black carbon. In this paper, a new
20 method is proposed to directly derive $\kappa_{f(\text{RH})}$ based only on measurements from a three-wavelength
21 humidified nephelometer system. The advantage of this newly proposed novel approach is that it
22 allows researchers to estimate $\kappa_{f(\text{RH})}$ without any additional information about PNSD and black
23 carbon. Values of $\kappa_{f(\text{RH})}$ estimated from this new method agree very well with those retrieved by
24 using the traditional method, the average difference between $\kappa_{f(\text{RH})}$ derived from newly proposed
25 method and traditional method is 0.005 and the square of correlation coefficient between them is 0.99.

26

27 **1. Introduction**

28 Atmospheric aerosol particles play vital roles in visibility, energy balance and the hydrological
29 cycle of the Earth-atmosphere system and have attracted a lot of attention in recent decades. Aerosol



30 particles suspended in the atmosphere directly influence radiative transfer of solar radiation and
31 indirectly affect cloud properties, therefore, have large impacts on climate change. Especially,
32 uncertainties in direct aerosol radiative forcing due to anthropogenic aerosols and in aerosol indirect
33 forcing caused by aerosol interaction with clouds contribute most to the total uncertainty in climate
34 forcing (Boucher et al., 2013). One of the most important factors affecting these uncertainties is the
35 interaction between aerosol particles and ambient atmospheric water vapour (Zhao et al., 2006; Kuang
36 et al., 2016b). Under supersaturated conditions, aerosol particles serve as cloud condensation nuclei
37 (CCN) and hence influence cloud properties. Under subsaturated conditions, with respect to typical
38 aerosol compositions, liquid water content condensed on aerosol particles usually constitute about half
39 of the total aerosol mass at relative humidity (RH) of 80% (Bian et al., 2014). Liquid water usually
40 dominates the total aerosol mass in most aerosol types when RH is above 90% (Bian et al., 2014). The
41 amounts of condensed water content in ambient aerosols and cloud droplets depend both on water
42 uptake abilities of the aerosol components and ambient RH.

43 Traditionally, the Köhler theory (Petters and Kreidenweis, 2007) is widely used to describe the
44 hygroscopic growth of aerosol particles and successfully used in laboratory studies for single
45 component and some multicomponent particles. However, it is found that most atmospheric aerosol
46 particles usually consist of both organic and inorganic constituents (Murphy et al., 1998) rather than
47 consist of a single component. Given this, a modified version of Köhler theory called κ -Köhler theory
48 is proposed by Petters and Kreidenweis (2007) and widely used in recent ten years to study the
49 hygroscopic growth of aerosol particles. The formula of this theory is expressed as the following:

$$50 \quad S = \frac{D^3 - D_d^3}{D^3 - D_d^3(1 - \kappa)} \cdot \exp\left(\frac{4\sigma_{s/a} \cdot M_{water}}{R \cdot T \cdot D_p \cdot g \cdot \rho_w}\right) \quad (1)$$

51 where S is the saturation ratio, D is the diameter of the droplet, D_d is the dry diameter, $\sigma_{s/a}$ is the
52 surface tension of solution/air interface, T is the temperature, M_{water} is the molecular weight of water,
53 R is the universal gas constant, ρ_w is the density of water, and κ is the hygroscopicity parameter.
54 This theory is not only applicable to single-component aerosol particles, but also to multicomponent
55 aerosol particles. With regard to a multicomponent aerosol particle, the Zdanovskii, Stokes, and
56 Robinson assumption can be applied. The hygroscopicity parameter κ of multicomponent aerosol
57 particle can be derived by using the following formula: $\kappa = \sum_i \varepsilon_i \cdot \kappa_i$, where κ_i and ε_i represent the
58 hygroscopic parameter and volume fraction of each component. In recent ten years, this hygroscopicity



59 parameter κ has received much attentions and turns out to be a very effective parameter to study
60 aerosol hygroscopicity. This hygroscopicity parameter κ makes the comparison of the aerosol
61 hygroscopicity at different sites around the world and different time periods more convenient. In
62 addition, hygroscopicity parameter κ also facilitates the intercomparison of aerosol hygroscopicity
63 derived from different techniques and measurements made at different RHs. This hygroscopicity
64 parameter κ is widely used to account the influence of aerosol hygroscopic growth on aerosol
65 optical properties as well as aerosol liquid water contents (Tao et al., 2014;Kuang et al., 2015;Brock
66 et al., 2016;Bian et al., 2014;Zieger et al., 2013) and to examine the role of aerosol hygroscopicity in
67 CCN (Chen et al., 2014;Gunthe et al., 2009;Ervens et al., 2010). Therefore, the derived κ values from
68 field campaigns and laboratory studies will further our understanding in aerosol hygroscopicity and
69 help estimate the influences of aerosol hygroscopic growth on different aspects of atmospheric
70 processes.

71 Currently, several types of instruments are widely used in field campaigns to study the aerosol
72 hygroscopicity through different aspects of aerosol properties. The Humidity Tandem Differential
73 mobility Analyzer (HTDMA) operates below water saturation and directly measures the aerosol
74 hygroscopic growth factor of selected particles which have certain diameters at specified RH points.
75 The aerosol hygroscopicity parameter κ can be directly derived from measurements of HTDMA by
76 applying equation (1) (Liu et al., 2011;Wu et al., 2016). Through relating the aerosol hygroscopicity to
77 CCN properties, measurements of size resolved CCN efficiency spectra can also be used to infer the
78 hygroscopicity parameter κ at different diameters (Gunthe et al., 2009;Petters et al., 2009;Rose et al.,
79 2010;Su et al., 2010). These two methods can both provide insights into the aerosol hygroscopicity at
80 different aerosol diameters, however, they can only be used to derive aerosol hygroscopicity parameter
81 κ within certain size range (usually less than 300 nm). Thus, these two methods are not capable of
82 providing more details about aerosol hygroscopicity of aerosol particles which contribute most to
83 aerosol optical properties and aerosol liquid water contents (their diameters usually ranging from 200
84 nm to 1 μ m) (Ma et al., 2012;Bian et al., 2014). The effect of aerosol water uptake on the aerosol
85 particle light scattering (σ_{sp}) (sometimes aerosol extinction coefficient (Brock et al., 2016)) is usually
86 measured with a humidified nephelometer system. Measurements from a humidified nephelometer
87 system can also be used to calculate the aerosol hygroscopicity parameter κ if the dry aerosol particle
88 number size distribution (PNSD) is measured simultaneously (Chen et al., 2014). The enhancement



89 factor $f(\text{RH}, \lambda)$ which is defined as $f(\text{RH}, \lambda) = \sigma_{sp}(\text{RH}, \lambda) / \sigma_{sp}(\text{dry}, \lambda)$, is usually used as an
90 indicator of how much the RH impacts on σ_{sp} . $\sigma_{sp}(\text{RH}, \lambda)$ or $\sigma_{sp}(\text{dry}, \lambda)$ represents σ_{sp} at
91 wavelength λ at a certain RH or under dry conditions. In this research, $f(\text{RH})$ is referred to as
92 $f(\text{RH}, 550 \text{ nm})$ and $f(80 \%)$ represents the $f(\text{RH})$ at 80 % RH. The nephelometer measures
93 aerosol optical properties of the entire aerosol size distribution, thus, the deduced κ value from
94 measurements of $f(\text{RH}, \lambda)$ represents an overall, optically weighted κ . This κ is more suitable to be
95 used to account the influences of aerosol hygroscopic growth on aerosol optical properties compared
96 to aerosol hygroscopicity derived from HTDMA and CCN measurements.

97 Traditionally, as mentioned before, the way of deriving κ values from $f(\text{RH})$ measurements
98 require measurements of PNSD at dry state and may also need the mass concentrations of black carbon
99 (BC) to account the influence of BC on aerosol refractive index. However, the instruments of
100 measuring the PNSD and BC at dry state are expensive, and during field campaigns their information
101 is sometimes not available. In this paper, with measurements from a field campaign on the North China
102 Plain (NCP), we first derived κ values from measurements of $f(\text{RH})$ with the traditional method
103 and then compared them with the κ values derived from High Humidity Tandem Differential
104 Mobility Analyzer (HH-TDMA). HH-TDMA is a system very similar with HTDMA but is capable of
105 operating at higher RH points (Liu et al., 2011). The relationships between κ values derived from
106 $f(\text{RH})$ measurements and parameters used to fit measured $f(\text{RH})$ curves are further examined and
107 analyzed. Finally, basing on finished analysis about the relationship between κ and $f(\text{RH})$ fitting
108 parameters, a novel method to directly derive the aerosol hygroscopicity parameter κ based only on
109 measurements from a humidified nephelometer system is proposed. This newly proposed approach
110 makes it more convenient and cheaper for researchers to conduct aerosol hygroscopicity research with
111 measurements of $f(\text{RH})$.

112 2. Site description and instruments

113 In this study, the main part of used datasets is from the field campaign conducted at Wangdu
114 ($38^{\circ}40'N$, $115^{\circ}08'E$) during summer on the North China Plain (NCP). This field campaign was jointly
115 conducted by Peking University, China and Leibniz-Institute for Tropospheric Research, Germany.
116 Wangdu site is located in the suburban district of Wangdu County, Hebei Province, China and situated
117 adjacent to farmland and residential areas, it belongs to the typical region of the NCP. This observation
118 campaign lasted for about one month from 4 June, 2014 to 14 July, 2014. The measured $f(\text{RH}, \lambda)$



119 dataset was available from June 21st, 2014, to July 1st, 2014.

120 For datasets from Wangdu campaign. The chemical compositions of the aerosol particles with an
121 aerodynamic diameter of less than 2.5 μm (PM_{2.5}) were analyzed based on the samples collected on
122 quartz and Teflon filters. Other instruments share one inlet which is placed on the roof of the container.
123 Regarding this inlet system, aerosol particles first entered an impactor which selected the aerosol
124 particles with an aerodynamic diameter of less than 10 μm , and then passed through a dryer which is
125 capable of reducing the RH of the sample air to lower than 30 %. In succession, the sample air passed
126 through a splitter and was allotted to different instruments according to their required flow rates. The
127 PNSD at dry state ranging from 3nm to 10 μm was observed jointly by a Twin Differential Mobility
128 Particle Sizer (TDMPS, Leibniz-Institute for Tropospheric Research (IfT), Germany; Birmili et al.
129 (1999)) and an Aerodynamic Particle Sizer (APS, TSI Inc., Model 3321) with a temporal resolution of
130 10 minutes. The absorption coefficient at 637 nm was measured using a Multi-angle Absorption
131 Photometer (MAAP Model 5012, Thermo, Inc., Waltham, MA USA) with a temporal resolution of 1
132 minute, and further used to calculate the mass concentrations of black carbon (BC) with a constant
133 mass absorption efficiency (MAE) of 6.6 m^2g^{-1} . The growth factors of aerosol particles at six
134 selected particle diameters (30 nm, 50 nm, 100 nm, 150nm, 200 nm and 250 nm) at 98% RH condition
135 were obtained from the measurements of the HH-TDMA (Leibniz-Institute for Tropospheric Research
136 (IfT), Germany; Hennig et al. (2005)). The $f(\text{RH}, \lambda)$ curves of aerosol particles with RH ranging from
137 about 50% to 90% were measured by a humidified nephelometer system which consists of two three-
138 wavelength integrating nephelometers (TSI Inc., Model 3563) and a humidifier. The humidifier was
139 used to moisten the air which will be sampled into the second nephelometer. Details of this humidified
140 nephelometer system please refer to (Kuang et al., 2016a).

141 PNSDs at dry state and mass concentrations of BC derived from MAAP measurements measured
142 at both Wuqing from 12 July to 14 August in 2009 and Xianghe from 9 July to 8 August in 2013 are
143 also used in this study to examine the influence of PNSD and BC on derivation of κ values from
144 $f(\text{RH})$ measurements and other relationships. Additionally, σ_{sp} values which were observed during
145 these three field campaigns introduced before with a three-wavelength integrating nephelometer (TSI
146 Inc., Model 3563) are also used in Sect.4.3. Both Wuqing and Xianghe are representative regional
147 background sites of the NCP and locates in the northern part of the NCP. Details about these two
148 campaigns can be found in papers published by Kuang et al. (2015) and Ma et al. (2016).



149 3. Methodology

150 3.1 Calculations of hygroscopicity parameter κ from measurements of $f(\text{RH})$ and HH-TDMA

151 Research of Chen et al. (2014) demonstrated that if the PNSD at dry state is measured, then
152 measurements of $f(\text{RH})$ can be used to derive the aerosol hygroscopicity parameter κ by
153 conducting an iterative calculation with the Mie theory and the κ -Köhler theory. To reduce the
154 influence of random errors of observed $f(\text{RH})$ at a certain RH, all valid $f(\text{RH})$ measurements in a
155 complete humidifying cycle is used in the derivation algorithm. The retrieved κ is the κ value which
156 can be used to best fit the observed $f(\text{RH})$ curve. Details about this retrieval algorithm is described
157 in Chen et al. (2014). Of particular note is that in this research the mass concentration of BC is also
158 considered in the retrieval algorithm to account for the influence of BC on refractive indices of aerosol
159 particles at different sizes. The BC is considered to be homogeneously mixed with other aerosol
160 components, and the mass size distribution of BC used in Ma et al. (2012) which is observed on the
161 NCP is used in this research to account the mass distributions of BC at different particle sizes. The
162 used refractive index and density of BC are $1.80 - 0.54i$ and 1.5g cm^{-3} (Kuang et al., 2015). Used
163 refractive indices of non light-absorbing aerosol components (other than BC) and liquid water are
164 $1.53 - 10^{-7}i$ (Wex et al., 2002) and $1.33 - 10^{-7}i$ (Seinfeld and Pandis, 2006), respectively.

165 The HH-TDMA measures hygroscopic growth factors of particles at different sizes at 98% RH
166 condition. The measured hygroscopic factors can be directly related to κ with equation (1). For a
167 specified size of selected aerosol particles, a distribution of growth factors can be measured, and thus
168 can be used to derive a probability distribution of κ and finally come to the calculation of average κ
169 value corresponding to this size of aerosol particles. The method on how to derive average κ value
170 of certain size of aerosol particles from HH-TDMA measurements is elaborately described in Liu et al.
171 (2011).

172 3.2 Parameterization schemes for $f(\text{RH})$

173 Due to the complex chemical compositions of ambient aerosol particles and the challenge in
174 precisely measuring their molecular compositions, it is difficult to directly describe the influence of
175 RH on σ_{sp} . Some simplified parameterization schemes are usually used to describe $f(\text{RH})$ as a
176 function of RH. The most frequently used parameterization scheme is a power-law function which is
177 known as “gamma” parameterization (Quinn et al., 2005) and the formula of this single-parameter
178 representation is written as the following:



179
$$f(\text{RH}) = \left[\frac{(100 - \text{RH}_0)}{100 - \text{RH}} \right]^\gamma \quad (2)$$

180 where RH_0 is the RH of dry condition, and γ is a parameter fitted to the observed $f(\text{RH})$. In this
181 study, we estimated γ values with observed $f(\text{RH})$ curves and for the first time to our knowledge,
182 we further examined the relationship between γ and κ retrieved from $f(\text{RH})$ measurements.

183 Recently, a new physically based single-parameter representation was proposed by Brock et al.
184 (2016) to describe $f(\text{RH})$. Their results demonstrated that this proposed parameterization scheme can
185 better describe $f(\text{RH})$ than the widely used gamma power-law approximation (Brock et al., 2016).
186 The formula of this new scheme is written as:

187
$$f(\text{RH}) = 1 + \kappa_{sca} \frac{\text{RH}}{100 - \text{RH}} \quad (3)$$

188 where κ_{sca} is a parameter fitted to observed $f(\text{RH})$. Here, we give a brief introduction about the
189 physical understanding of this alternative parameterization scheme. Regardless of the curvature effects
190 for particle diameters larger than 100 nm, the hygroscopic growth factor for aerosol particles can be
191 approximately expressed as the following (Brock et al., 2016): $gf_{diam} \cong \left(1 + \kappa \frac{\text{RH}}{100 - \text{RH}} \right)^{1/3}$.
192 Moreover, σ_{sp} is usually proportional to total aerosol volume (Pinnick et al., 1980) which means that
193 the relative change in σ_{sp} due to aerosol water uptake is roughly proportional to relative change in
194 aerosol volume. The enhancement factor in volume can be expressed as the cube of gf_{diam} , thus lead
195 to the formula form of $f(\text{RH})$ expressed in equation (3). More details about the discussion of this
196 new expression form of $f(\text{RH})$ can be found in the paper published by Brock et al. (2016). In this
197 paper, the performance of this newly proposed scheme is investigated, values of κ_{sca} are estimated
198 from observed $f(\text{RH})$ curves and their relationship with κ values retrieved from $f(\text{RH})$
199 measurements is also examined.

200 These two parameterization schemes which are introduced here are referred to as γ -Method and
201 κ_{sca} -Method respectively in the following paragraphs.

202 4. Results and discussions

203 4.1 Derived κ values from $f(\text{RH})$ and HH-TDMA measurements

204 During this field campaign, the aerosol physical, chemical and optical properties are
205 synergistically observed with different types of instruments. They provide valuable datasets to perform
206 an insightful analysis about aerosol hygroscopicity and its relationship with other aerosol properties.



207 The time series of σ_{sp} at 550 nm at dry state are shown in Fig.1a, and values of σ_{sp} at 550 nm shown
208 in Fig.1a are corrected from measurements of TSI 3563 nephelometer (the truncation errors are
209 corrected using Mie theory with measured PNSD and mass concentrations of BC). The results show
210 that this observation period has experienced varying degrees of pollution levels, with σ_{sp} at 550 nm
211 ranging from 15 to 1150 Mm^{-1} . The aerosol chemical compositions also change a lot during the
212 observation period. The relative contributions of mass concentrations of organic matter to total PM2.5
213 mass concentrations range from 2% to 42%. Moreover, the relative contributions of mass
214 concentrations of sulfate, nitrate and ammonium to total PM2.5 mass concentrations range from 5 to
215 50 %, 2 to 27 % and 1 to 21 %, respectively (Kuang et al., 2016a). These results imply that during this
216 observation period, the aerosol hygroscopicity changes a lot whereafter corroborated by $f(RH)$
217 measurements. Overall, $f(80\%)$ values range between 1.1 and 2.3 with an average of 1.8. Periods
218 when deliquescent phenomena occur, $f(80\%)$ values are relatively higher with a variation range of
219 1.7 to 2.3 and their average is 2.0. This is because of the dominance of ammonium sulfate during
220 periods when deliquescent phenomena occur. More detailed analysis about the frequently observed
221 deliquescent phenomena during this field campaign please refer to (Kuang et al., 2016a).

222 Furthermore, κ values derived from $f(RH)$ measurements by combining measurements of
223 PNSD at dry state and mass concentrations of BC are shown in Fig.1b. During deliquescent phenomena
224 periods, $f(RH)$ jumps when sample RH in the cavity of the nephelometer mainly ranges from 60 %
225 to 65 %. Therefore, for a humidifying cycle which shows jump phenomenon in $f(RH)$, only $f(RH)$
226 points when RHs are greater than 70% (after the jump point) are used to retrieve κ . The results
227 demonstrate that κ derived from $f(RH)$ measurements (hereinafter referred to as $\kappa_{f(RH)}$) lie
228 between 0.06 and 0.43. The lowest $\kappa_{f(RH)}$ values are found when the air quality is relatively clean
229 (σ_{sp} at 550 nm is lower than 100 Mm^{-1}) on 27 and 28 June. During these two days, organic matter
230 dominates the mass concentration of PM2.5 which results in the low hygroscopicity of aerosol particles.
231 On the contrary, the largest $\kappa_{f(RH)}$ values are found during periods when deliquescent phenomena
232 occur and inorganic chemical compositions dominate the mass concentrations of PM2.5, especially,
233 sulfate is highly abundant during these periods. Of particular note is that during relatively polluted
234 periods (σ_{sp} at 550 nm larger than 100 Mm^{-1}) aerosol particles are generally very hygroscopic which
235 imply that aerosol water uptake can exert significant impacts on regional direct aerosol radiative effect



236 and ambient visibility during this observation period. On the whole, the average $\kappa_{f(\text{RH})}$ during this
237 observation period is 0.28.

238 On the basis of the average size-resolved κ distribution from Haze in China (HaChi) campaign
239 (Liu et al., 2014), κ values change a lot for aerosol particles whose diameters are less than 250 nm,
240 however, κ values vary relatively smaller for aerosol particles whose diameter range from 250 nm to
241 1 μm . In addition, the results from HaChi campaign also demonstrate that aerosol particles whose
242 diameter range from 200 nm to 1 μm usually contribute more than 80% to σ_{sp} at 550 nm during
243 summer on the NCP (Ma et al., 2012). To compare κ values derived from measurements of
244 humidified nephelometer system and HH-TDMA, average κ values corresponding to aerosol
245 particles at only 250 nm which are derived from HH-TDMA measurements are also shown in Fig.1b.
246 In the following, average κ which is derived from HH-TDMA measurements for aerosol particles
247 having particle size of 250 nm is referred to as κ_{250} . During this observation period, values of κ_{250}
248 range from 0.11 to 0.56, with an average of 0.34 which is very close to average κ_{250} observed during
249 HaChi campaign (Liu et al., 2011). The results shown in Fig.1b suggest that, in general, $\kappa_{f(\text{RH})}$ values
250 agree well with κ_{250} values, however, are usually lower than κ_{250} values. To quantitatively compare
251 these two types of κ values, they are plotted against each other and shown in Fig.2. It can be seen that
252 they are highly correlated but the κ_{250} values are systematically higher than $\kappa_{f(\text{RH})}$ values, and the
253 average difference between κ_{250} and $\kappa_{f(\text{RH})}$ is 0.06. The statistical relationship between κ_{250} and
254 $\kappa_{f(\text{RH})}$ is also shown in Fig.2. This relationship may be useful for researchers if they want to estimate
255 the influences of aerosol water uptake on aerosol optical properties and aerosol liquid water contents
256 when only HH-TDMA or HTDMA measurements are available.

257 A model experiment is conducted to better understand the relationship between κ_{250} and $\kappa_{f(\text{RH})}$.
258 During HaChi campaign, size-resolved κ distributions are derived from measured size-segregated
259 chemical compositions (Liu et al., 2014) and their average is used in this experiment to account the
260 size dependence of aerosol hygroscopicity which is shown in Fig. 3a. With this fixed average size-
261 resolved κ distribution, all observed PNSDs at dry state along with mass concentrations of BC which
262 are observed at three different representative background sites of the NCP during summer are used to
263 simulate the retrieval of $\kappa_{f(\text{RH})}$ under different PNSD and BC conditions. Field campaigns conducted
264 at these three site is introduced in Sect.2. The treatment of BC is the same with the way in the process



265 of deriving $\kappa_{f(\text{RH})}$ which is introduced in Sect.4.1. The used PNSDs shown in Fig.3b show that large
266 varying types of PNSDs are considered in the simulative experiment. The probability distribution of
267 simulated $\kappa_{f(\text{RH})}$ is also shown in Fig.3a. The standard deviation of retrieved $\kappa_{f(\text{RH})}$ is about 0.01
268 which suggests that if the size-resolved κ distribution is fixed, then $\kappa_{f(\text{RH})}$ varies little. Due to
269 $\kappa_{f(\text{RH})}$ represents an overall, optically weighted κ , it is clearly shown in Fig.3a that in most cases
270 $\kappa_{f(\text{RH})}$ values are located between κ values of aerosol particles ranging from 200 nm to 1 μm .
271 Moreover, about 70% of simulated $\kappa_{f(\text{RH})}$ values are less than κ_{250} which to some extent explains
272 the observed difference between κ_{250} and $\kappa_{f(\text{RH})}$ mentioned before. However, the simulated
273 average difference between κ_{250} and average $\kappa_{f(\text{RH})}$ is about 0.01 which is far less than the observed
274 averaged difference between κ_{250} and $\kappa_{f(\text{RH})}$ which is 0.06. Except that uncertainties from
275 measurements of instruments, for example, the uncertainty of RH in measurements of HH-TDMA and
276 uncertainties of measuring $f(\text{RH})$ (details about the uncertainty sources of $f(\text{RH})$ measurements
277 can be found in the paper published by Titos et al. (2016)), there are other two reasons may be
278 associated with this discrepancy. The first one is that configurations of size-resolved κ distributions
279 and PNSDs during this field campaign are far different from the model experiment. The second one is
280 that in the real atmosphere, κ values at different RH conditions may be different (You et al., 2014)
281 and most of $f(\text{RH})$ measurements are conducted when RH is lower than 90%, however, the
282 measurements of HH-TDMA are conducted when RH is equal to 98% . Overall, the observed general
283 consistency between κ values derived from measurements of $f(\text{RH})$ and HH-TDMA confirms the
284 reliability of κ values derived from $f(\text{RH})$ measurements.

285 **4.2 Relationships between κ derived from $f(\text{RH})$ measurements and $f(\text{RH})$ fitting** 286 **parameters**

287 In the previous section, the overall properties of ambient aerosol particles are introduced, derived
288 $\kappa_{f(\text{RH})}$ values are characterized and compared with κ_{250} values. These results demonstrated that
289 derived $\kappa_{f(\text{RH})}$ values can commendably represent variations of aerosol hygroscopicity of ambient
290 aerosol populations. In this section, the relationship between derived $\kappa_{f(\text{RH})}$ values and $f(\text{RH})$
291 fitting parameters are further examined to investigate their relationships.

292 Two parameterization schemes of $f(\text{RH})$ are discussed in this paper, including the currently
293 widely used γ -Method and newly proposed κ_{sca} -Method by Brock et al. (2016) and both methods
294 are introduced in Sect.3.2. A fitting example of these two methods is shown in Fig.4A. For $f(\text{RH})$



295 cycles observed during this field campaign, they are fitted by using both γ -Method and κ_{sca} -Method,
296 corresponding values of γ and κ_{sca} are also deduced. For cycles during deliquescent periods, only
297 $f(\text{RH})$ points with RH higher than 70% are used to perform fitting processes. The fitting performance
298 of these two methods are further investigated by conducting the comparison between measured and
299 fitted $f(85)\%$ values. Probability distributions of the ratio between fitted and measured $f(85)\%$ by
300 using these two methods are shown in Fig.4B. The results indicate that in most cases both γ -Method
301 and κ_{sca} -Method fit observed $f(\text{RH})$ cycles well with γ -Method performs slightly better which is
302 contrary to the results introduced by Brock et al. (2016), their results demonstrate that κ_{sca} -Method
303 can better describe observed $f(\text{RH})$ than γ -Method. That is to say, $f(\text{RH})$ curves observed at
304 different places or time periods may require different parameterization schemes to fit them best,
305 however, in general both γ -Method and κ_{sca} -Method are good approaches to fit observed $f(\text{RH})$
306 curves.

307 Concerning γ -Method, previous studies usually examine the relationship between γ and aerosol
308 chemical compositions and established several parameterization schemes to fit γ with mass fractions
309 of different aerosol chemical compositions, including organic materials, sulfate and nitrate (Quinn et
310 al., 2005; Titos et al., 2014; Zhang et al., 2015). However, to obtain a reliable estimation of γ , complete
311 information of aerosol chemical compositions may be required which is difficult to get, and it is also
312 hard to find a comprehensive description of γ based on those complicated chemical compositions.
313 Single aerosol hygroscopicity parameter κ can represent overall hygroscopicity of aerosol particles
314 which contains influences of different chemical compositions on aerosol hygroscopicity, therefore may
315 be used to better fit γ . In view of this, the relationship between $\kappa_{f(\text{RH})}$ and γ is investigated and
316 shown in Fig.4C. It is found that a pretty good linear relationship exists (square of correlation
317 coefficient is 0.95) between $\kappa_{f(\text{RH})}$ and γ , especially when $\kappa_{f(\text{RH})}$ is larger than 0.15. This
318 correlation is far better than previously found relationships between γ and aerosol chemical
319 compositions (Quinn et al., 2005; Titos et al., 2014; Zhang et al., 2015) and statistical parameters which
320 can be used to parameterize γ with $\kappa_{f(\text{RH})}$ is also shown in Fig.4C. During this field campaign, fitted
321 γ ranges from 0.15 to 0.63 with an average of 0.46.

322 Furthermore, during this field campaign, fitted κ_{sca} ranges from 0.05 to 0.3 with an average of
323 0.2. The relationship between $\kappa_{f(\text{RH})}$ and κ_{sca} is also investigated and it is found a strong linear



324 relationship also exists (square of correlation coefficient is 0.98) between $\kappa_{f(\text{RH})}$ and κ_{sca} . This
325 linear relationship is even better than the linear relationship between $\kappa_{f(\text{RH})}$ and γ . Not only that, the
326 statistically fitted line almost passes through zero point which implies that a proportional relationship
327 may exist between $\kappa_{f(\text{RH})}$ and κ_{sca} . This strong correlation should be intrinsic due to the idea of
328 κ_{sca} -Method is from the linkage between total aerosol volume and σ_{sp} as introduced in Sect.3.2 and
329 the increase of total aerosol volume due to aerosol water uptake is directly linked to the overall aerosol
330 hygroscopicity parameter κ . It seems that this promising linear relationship can help bridge the gap
331 between $f(\text{RH})$ and κ . However, results from Brock et al. (2016) demonstrated that the relationship
332 between $\kappa_{f(\text{RH})}$ and κ_{sca} is much more sophisticated and it is affected by both aerosol
333 hygroscopicity and PNSD at dry state. In the paper published by Brock et al. (2016), κ_{ext} and κ_{chem}
334 are used and correspond to κ_{sca} and $\kappa_{f(\text{RH})}$ in this research, the difference between κ_{ext} and κ_{sca}
335 is that κ_{ext} is used to fit the light enhancement factor of aerosol extinction coefficient, κ_{chem} and
336 $\kappa_{f(\text{RH})}$ actually means the same because both them are overall and size independent hygroscopicity
337 parameters. Results from Brock et al. (2016) concluded that the ratio $\kappa_{ext}/\kappa_{chem}$ generally lies
338 between 0.6 to 1 which implies that the ratio $\kappa_{sca}/\kappa_{f(\text{RH})}$ (in the following, this ratio is referred to as
339 R_κ) also should have large variations and shares the similar variation range. By revisiting the
340 relationship between $\kappa_{f(\text{RH})}$ and κ_{sca} found in this research, it can be found that R_κ during this
341 field campaign ranges from 0.6 to 0.77 with an average of 0.7. This result suggests that if we directly
342 establish a linkage between $\kappa_{f(\text{RH})}$ and κ_{sca} with an average R_κ can result in a non-negligible bias
343 (relative difference can reach about 15%). Besides, this range of R_κ only represents the relationship
344 between $\kappa_{f(\text{RH})}$ and κ_{sca} during a short time period and at only one site.

345 To better understand the relationship between $\kappa_{f(\text{RH})}$ and κ_{sca} , all PNSDs at dry state (shown
346 in Fig.3a) along with mass concentrations of BC observed from three different representative
347 background sites of the NCP during summer which is introduced in Sect.2 are used to simulate the
348 relationship between $\kappa_{f(\text{RH})}$ and κ_{sca} with Mie and κ -Köhler theories. During simulating processes,
349 for each PNSD, we change $\kappa_{f(\text{RH})}$ from 0.01 to 0.6 with an interval of 0.01 to examine the influence
350 of aerosol hygroscopicity on R_κ . The way of treating BC is same with the simulation experiment
351 introduced in Sect.4.1. Simulated results are shown in Fig.5a and the probability distribution of
352 simulated R_κ values is shown in Fig.5b. The results show that R_κ primarily ranges from 0.55 to 0.82
353 with an average of 0.69 which is very close to the average R_κ measured during the field campaign of



354 this research. These results also indicate that the relationship between $\kappa_{f(\text{RH})}$ and κ_{sca} is much more
355 complex than a simple linear relationship and more information about aerosol properties are necessary
356 to gain insights into the variation of R_{κ} .

357 **4.3 A novel method to directly derive κ from measurements of a humidified nephelometer** 358 **system**

359 A robust linear relationship is first found between $\kappa_{f(\text{RH})}$ and κ_{sca} in Sect.4.2 and then it turns
360 out that the relationship between $\kappa_{f(\text{RH})}$ and κ_{sca} is much more complex than that shown in Fig.4D.
361 The complexity comes from large variations of R_{κ} and both PNSD at dry state and aerosol
362 hygroscopicity have impacts on R_{κ} . Generally, used nephelometer of a humidified nephelometer
363 system have three wavelengths (Titos et al., 2016) and the spectral dependence of σ_{sp} is usually
364 described by the following Ångström formula: $\sigma_{sp}(\lambda) = \beta\lambda^{-\alpha_{sp}}$, where β is the particle number
365 concentration dependent coefficient, λ is the wavelength of light and α_{sp} represents the Ångström
366 exponent of σ_{sp} (Zieger et al., 2014). Thus, Ångström exponent can be directly inferred from the
367 measurements of σ_{sp} at different wavelengths. Of particular note is that Ångström exponent can
368 not only be used to account the spectral course of σ_{sp} , it also reveals information about PNSD. In
369 general, larger value of Ångström exponent corresponds to smaller aerosol particles. That is,
370 Ångström exponent can be a proxy of PNSD at dry state and be used in the processes of estimating
371 the impacts of PNSD on R_{κ} . On the other hand, with regard to aerosol hygroscopicity, although R_{κ}
372 varies within certain range, value of κ_{sca} can still be used to represent the overall hygroscopicity of
373 aerosol particles. Given this, simulated R_{κ} values introduced in the last paragraph of Sect.4.2 are
374 spread into a two dimensional gridded plot. The first dimension is Ångström exponent with an
375 interval of 0.02 and the second dimension is κ_{sca} with an interval of 0.01, average R_{κ} value within
376 each grid is represented by color and shown in Fig.6a. Values of Ångström exponent corresponding
377 different PNSDs are calculated from concurrently measured σ_{sp} values at 450 nm and 550 nm from
378 TSI 3563 nephelometer. Basing on results shown on Fig.6a, the different impacts of aerosol
379 hygroscopicity and PNSD at dry state on R_{κ} can be clearly distinguished. The results demonstrate
380 that PNSD at dry state dominates the variation of R_{κ} , nevertheless, aerosol hygroscopicity has non-
381 negligible impacts. Overall, larger value of Ångström exponent corresponds to higher R_{κ} . However,



382 aerosol hygroscopicity exhibits different influences on R_{κ} when Ångström exponent values are
383 different. Generally speaking, higher κ_{sca} corresponds to lower R_{κ} if Ångström exponent is
384 smaller than about 0.8 and higher κ_{sca} corresponds to higher R_{κ} if Ångström exponent is larger
385 than about 1.6.

386 In addition, the percentile value of standard deviation of R_{κ} values within each grid divided by
387 their average is shown in Fig.6b. In most cases, these percentile values are less than 6% (about 90%)
388 which demonstrates that R_{κ} varies little within each grid shown in Fig.6a. This implies that results of
389 Fig.6a can be used as a look up table to estimate R_{κ} . As what's introduced before, currently widely
390 used nephelometer of a humidified nephelometer system usually have three wavelengths (Titos et al.,
391 2016), thus can provide information about Ångström exponent, and κ_{sca} can be directly fitted from
392 observed $f(\text{RH})$ curve. Even only one $f(\text{RH})$ point is measured κ_{sca} can still be calculated from
393 equation (3). Therefore, using results shown in Fig.6a as a look up table, R_{κ} values can be directly
394 predicted from measurements of a humidified nephelometer system. With this method, R_{κ} values
395 during this Wangdu field campaign are predicted (values of Ångström exponent are calculated from
396 measured σ_{sp} values at 450 nm and 550 nm under dry conditions) and compared with measured R_{κ}
397 values, the results are shown in Fig.7a. The Ångström exponent during this field campaign ranges
398 from 0.63 to 1.96 with an average of 1.4. It can be seen from Fig.7a that majority of points lie nearby
399 1:1 line and 92% points have relative differences less than 6% which is consistent with results shown
400 in Fig.6b. This result is quite promising and can be further used to derive $\kappa_{f(\text{RH})}$ values by combining
401 fitted κ_{sca} and predicted R_{κ} . The results of predicted $\kappa_{f(\text{RH})}$ values are shown in Fig.7b and a robust
402 correlation between predicted $\kappa_{f(\text{RH})}$ values and $\kappa_{f(\text{RH})}$ values retrieved by using the traditional
403 method introduced in Sect.3.1 is achieved (square of correlation coefficient is 0.99). All points shown
404 in Fig.7b lie nearby 1:1 line, average difference between $\kappa_{f(\text{RH})}$ derived from newly proposed method
405 and traditional method is -0.005. This result demonstrates a quite good estimation of $\kappa_{f(\text{RH})}$ can be
406 achieved by using only measurements from a humidified nephelometer system.

407 It should be noted that the look up table shown in Fig.6a already covers large variation ranges of
408 Ångström exponent and κ_{sca} . Which means that this look up table can be used under different
409 conditions. However, it should be pointed out that the look up table shown in Fig.6a is from simulations
410 of measured continental aerosols without influences of desert dust, and it might not be suitable for



411 being used to estimate $\kappa_{f(\text{RH})}$ when sea salt or dust particles prevail. In summary, this approach allows
412 researchers to directly derive aerosol hygroscopicity from measurements of $f(\text{RH})$ without any
413 additional information about PNSD and BC which is quite convenient for researchers to conduct
414 aerosol hygroscopicity researches with measurements from a humidified nephelometer system.

415 5. Conclusions

416 During the field campaign introduced in this paper, which is conducted in summer at a
417 background site of the NCP, integrative aerosol information including aerosol chemical, optical and
418 physical properties are observed. Among them, aerosol hygroscopicity is crucial for understanding
419 roles of aerosol particles in air pollution and aerosol climate effects. In this paper, values of aerosol
420 hygroscopicity parameter κ are first derived from measurements of $f(\text{RH})$ by combining
421 measurements of PNSD at dry state and BC. The results show that during this field campaign, aerosol
422 hygroscopicity varies a lot, and $\kappa_{f(\text{RH})}$ ranges from 0.06 to 0.43 with an average of 0.28. Retrieved
423 $\kappa_{f(\text{RH})}$ values are further compared with κ_{250} which is derived from measurements of HH-TDMA
424 and good consistency is achieved. Results show that κ_{250} is systematically higher than $\kappa_{f(\text{RH})}$ and
425 the average of their difference is 0.06. A simulative experiment is conducted to better understand their
426 difference and partially explained the observed discrepancy, however, still not enough and possible
427 reasons are discussed in Sect.4.1.

428 Relationships between $\kappa_{f(\text{RH})}$ and $f(\text{RH})$ fitting parameters γ and κ_{sca} are further
429 investigated in Sect.4.2 which is for the first time to our knowledge. Good linear relationship exists
430 between $\kappa_{f(\text{RH})}$ and γ , and the correlation between $\kappa_{f(\text{RH})}$ and γ is far better than previously found
431 relationships between γ and aerosol chemical compositions. This results demonstrate that κ should
432 be a better choice to parameterize $f(\text{RH})$ fitting parameters than mass fractions of aerosol chemical
433 compositions which is so far widely used. The relationship between $\kappa_{f(\text{RH})}$ and κ_{sca} is then also
434 examined, and it is found that a better linear relationship than the relationship between $\kappa_{f(\text{RH})}$ and γ
435 exists between $\kappa_{f(\text{RH})}$ and κ_{sca} , and $\kappa_{f(\text{RH})}$ may be proportional to κ_{sca} . However, through
436 detailed analysis about the relationship between $\kappa_{f(\text{RH})}$ and κ_{sca} , it turns out that their relationship
437 is more complicated than what is found at the very beginning and the ratio $\kappa_{sca}/\kappa_{f(\text{RH})}$ (R_κ) varies a
438 lot. Results show that both PNSD at dry state and aerosol hygroscopicity have impacts on value of R_κ .

439 In Sect.4.3, by introducing Ångström exponent as a proxy for PNSD, impacts of PNSD and
440 aerosol hygroscopicity on R_κ are distinguished and then discussed. In succession, a look up table



441 based on Ångström exponent and κ_{sca} is developed to estimate R_{κ} . With this look up table, R_{κ}
442 can be directly estimated from measurements of a humidified nephelometer system. This method is
443 further verified with measurements of this field campaign. Results show that great consistency is
444 achieved between predicted and measured R_{κ} values (92% points have relative difference less than
445 6%). Given this, the linkage between $\kappa_{f(RH)}$ and κ_{sca} is directly established and further used to
446 estimate $\kappa_{f(RH)}$. The comparison results demonstrate a pretty good agreement is achieved, all points
447 lie nearby 1:1 line. The average absolute difference between $\kappa_{f(RH)}$ derived from newly proposed
448 method and traditional method is -0.005 and the square of correlation coefficient between them is 0.99.
449 This newly proposed novel approach allow researchers to estimate $\kappa_{f(RH)}$ without any additional
450 information about PNSD and BC. This new finding directly links κ and $f(RH)$ and will make the
451 humidified nephelometer system more convenient when it comes to aerosol hygroscopicity research.
452 Finally, findings in this research may facilitate the intercomparison of aerosol hygroscopicity derived
453 from different techniques, help for parameterizing $f(RH)$ and predicting CCN properties with optical
454 measurements.

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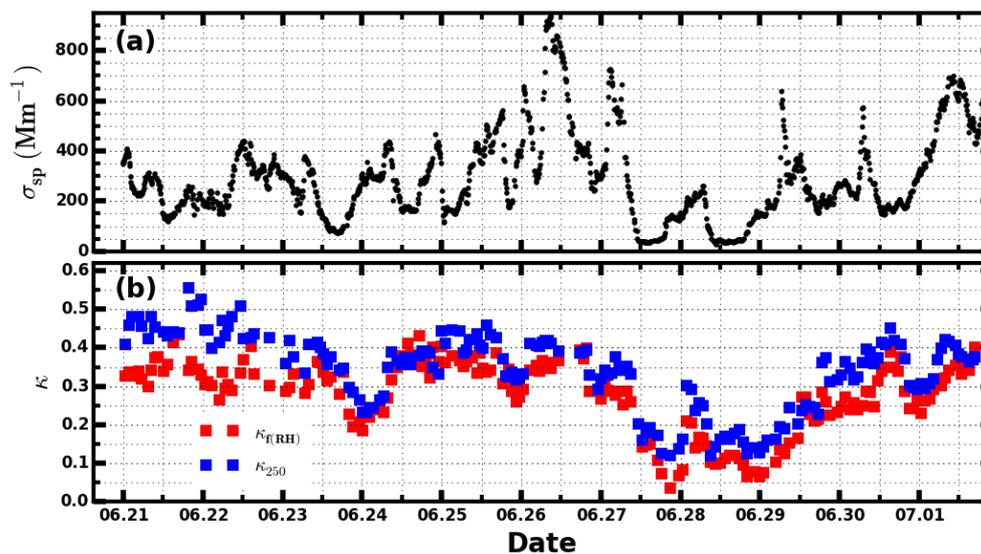
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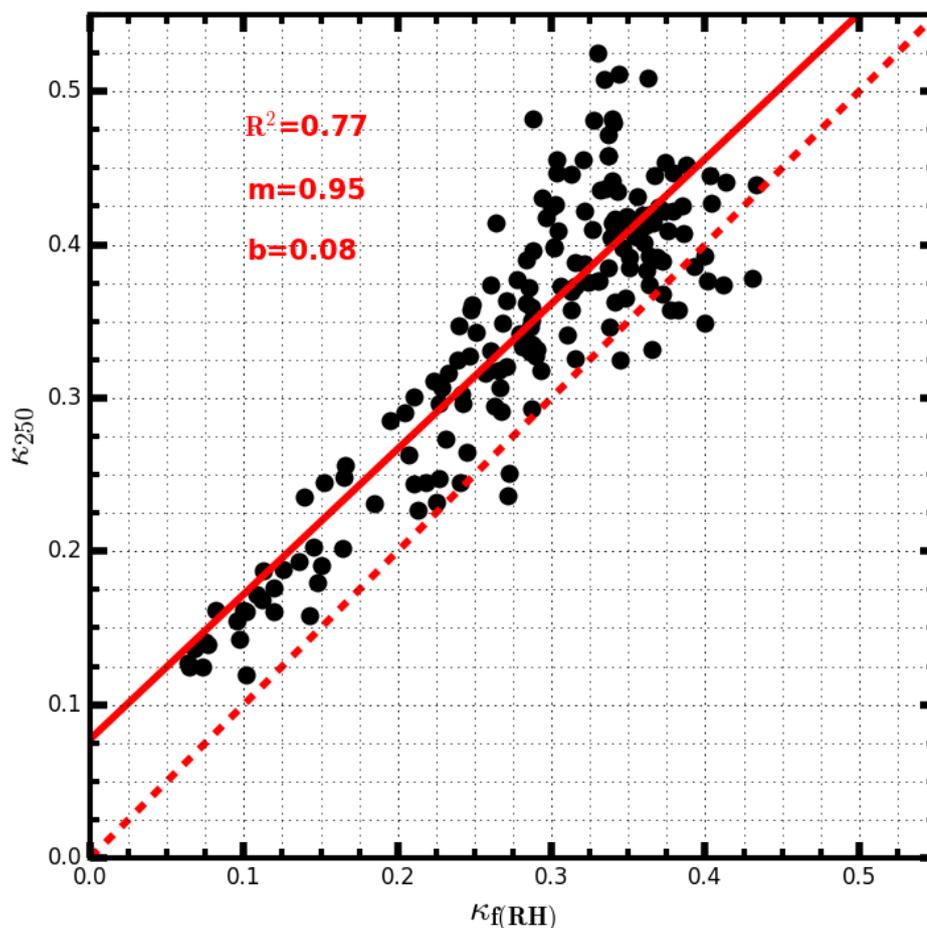
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600 **Figure 1.** (a) The time series of σ_{sp} at 550 nm; (b) The time series of κ values derived from $f(\text{RH})$
601 measurements ($\kappa_{f(\text{RH})}$) by combining information of PNSD and BC, and time series of average κ values of aerosol
602 particles at 250 nm (κ_{250}) which is calculated from measurements of HH-TDMA.

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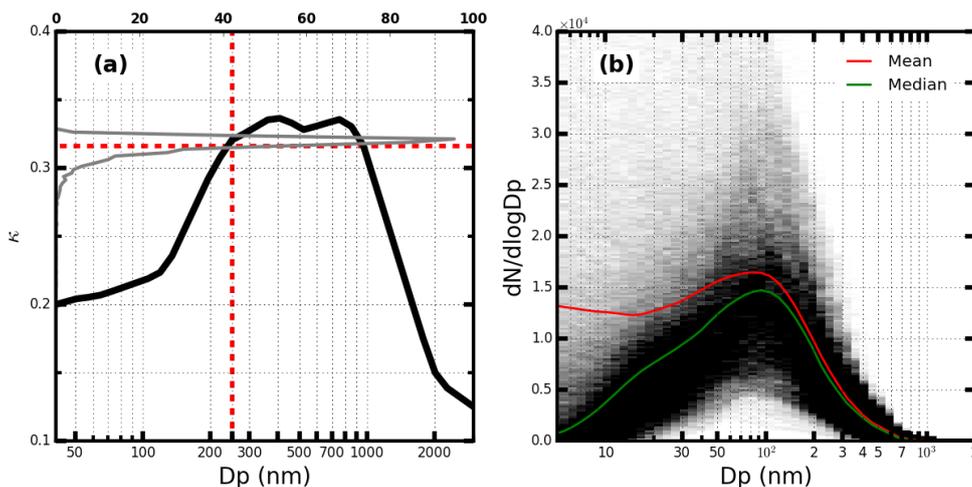
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607 **Figure 2.** The comparison between κ values derived from $f(\text{RH})$ measurements ($\kappa_{f(\text{RH})}$) and average κ values
608 for aerosol particles with a diameter of 250 nm (κ_{250}) which are derived from measurements of HH-TDMA. R^2 is
609 the square of correlation coefficient, m is the slope and b is the intercept.

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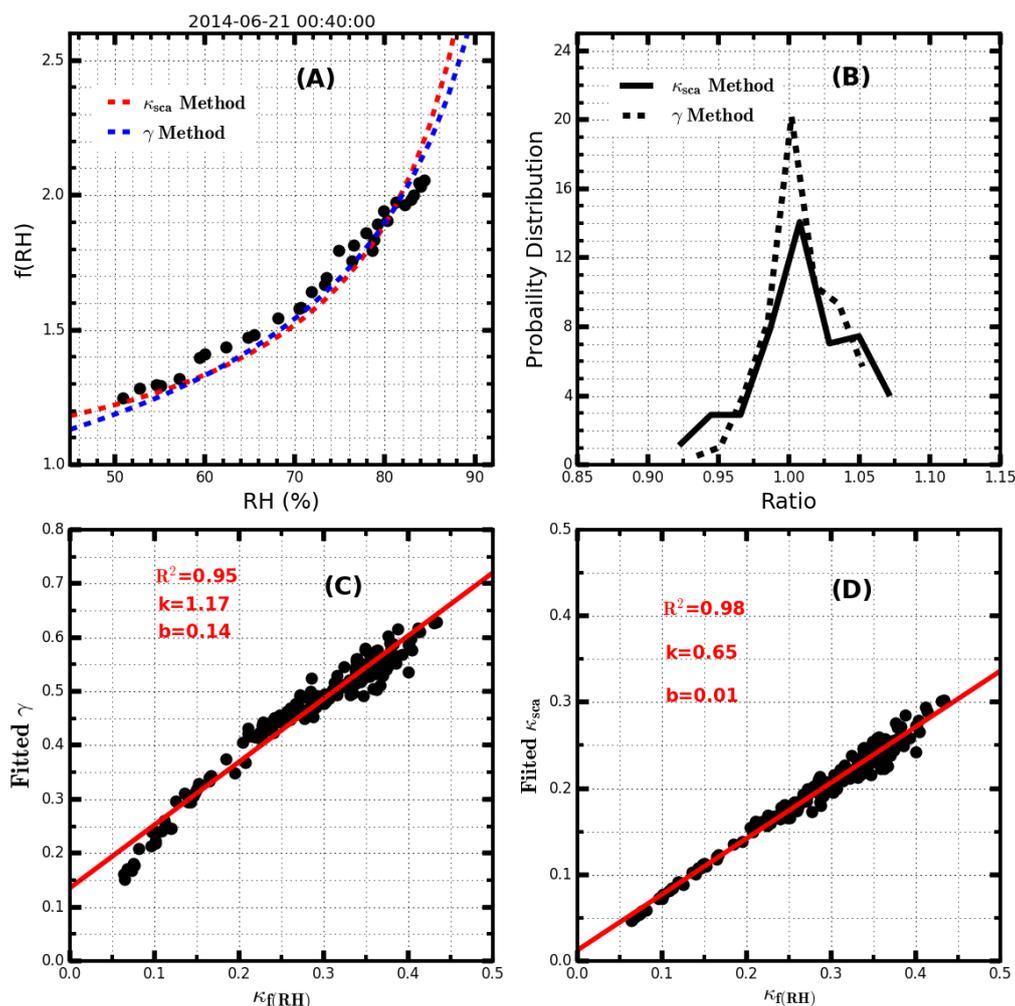
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614 **Figure 3.** (a) The thick black line represents the average size-resolved κ distribution from HaChi campaign. The
615 solid gray line represents the probability distribution of retrieved κ values with this size-resolve κ distribution by
616 using all PNSDs shown in figure (b), and the horizontal dashed line represents their average. The vertical dashed red
617 line represents the position of 250 nm. (b) All PNSDs which are observed from three different representative
618 background sites of the NCP during summer, they are used to model relationship between size-resolved κ and
619 retrieved κ values from $f(\text{RH})$ measurements, and the gray color represents the frequency of PNSD, darker point
620 corresponds to higher frequency, red and green line represent mean of median values of all observed PNSDs.
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623 **Figure 4.** (A) Fitting example of two discussed parameterization schemes, and title shows the observation time of
 624 this $f(RH)$ curve; (B) The fitting performance of two discussed parameterization schemes, x-axis represents the ratio
 625 between fitted $f(85\%)$ and measured $f(85\%)$ and y-axis represents the probability distribution. (C) The linear
 626 relationship between values of $\kappa_f(RH)$ and fitted γ , R^2 is the square of correlation coefficient, k is the slope and
 627 b is the intercept; (D) The linear relationship between values of $\kappa_f(RH)$ and fitted κ_{sca} .

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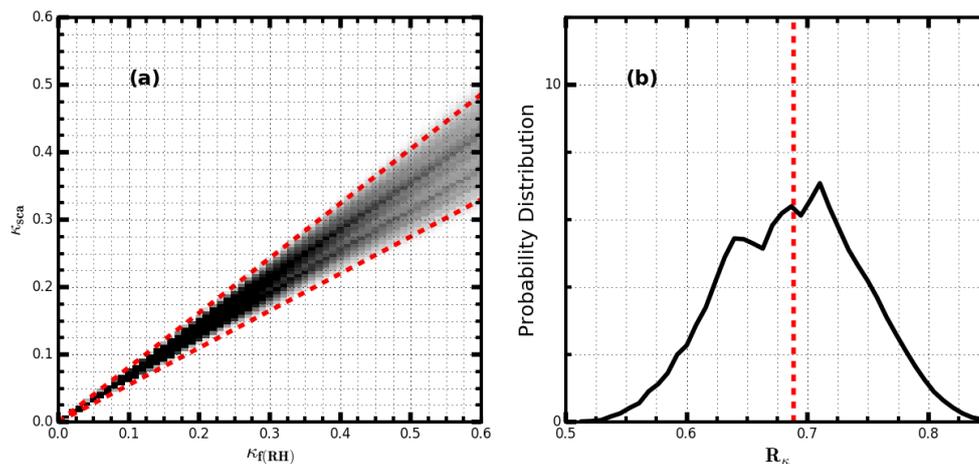
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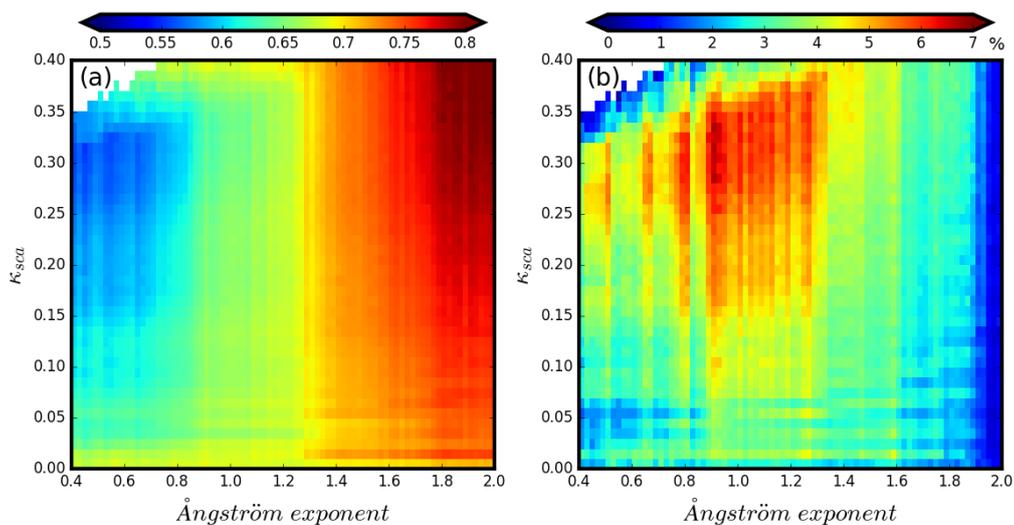
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636 **Figure 5.** (a) Simulated relationships between $\kappa_{f(\text{RH})}$ and κ_{sca} under different PNSD conditions (all PNSDs shown
637 in Fig.3a are used as inputs to conduct the simulation experiment), gray color represents the frequency and darker
638 point corresponds to higher frequency, the slope of two dashed lines are 0.55 and 0.81; (b) The probability distribution
639 of R_{κ} ($\kappa_{sca}/\kappa_{f(\text{RH})}$).

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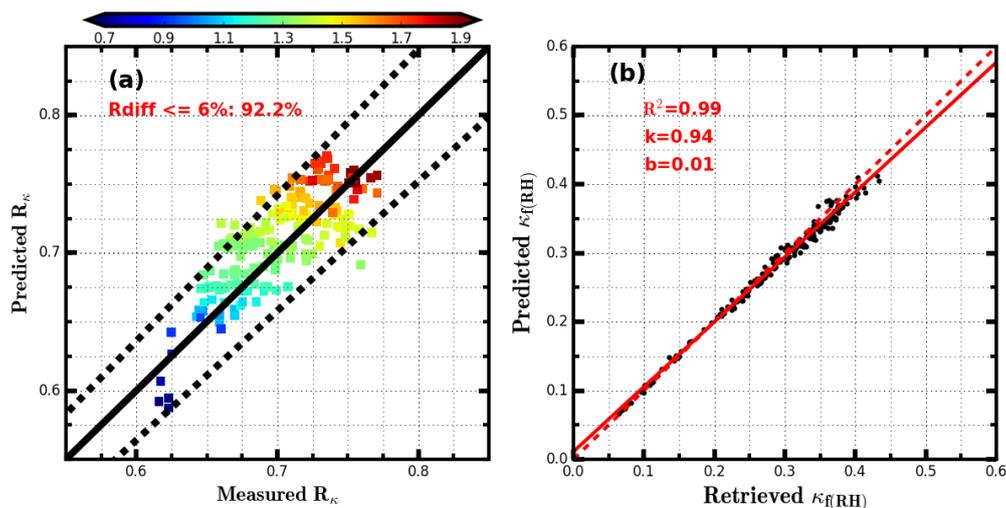


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652 **Figure 6.** (a) Colors represent R_k values and the color bar is shown on the top of this figure, x-axis represents
 653 Ångström exponent and y-axis represents κ_{sca} . (b) Meanings of x-axis and y-axis are same with them in (a),
 654 however, color represents the percentile value of the standard deviation of R_k values within each grid divided by
 655 their average.

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659 **Figure 7.** (a) The comparison between measured and predicted R_k values, colors represent values of Ångström
 660 exponent, texts with red color show the percentile of points with relative difference (Rdiff) less than 6% , two dashed
 661 line are lines with absolute relative difference (Rdiff) equal to 6% ; (b) the comparison between retrieved $\kappa_{f(RH)}$
 662 values by using traditional method introduced in Sect.3.1 and predicted $\kappa_{f(RH)}$ by using the new method introduced
 663 in Sect.4.3, R^2 is the square of correlation coefficient, k is the slope and b is the intercept.