

1 **On Calculating Deposition Coefficients and**
2 **Aspect Ratio Evolution in Approximate**
3 **Models of Ice Crystal Vapor Growth**

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ABSTRACT

Models of ice crystal vapor growth require estimates of the deposition coefficient (α) when surface attachment kinetics limit growth and when ice crystal shape is predicted. Parametric models can be used to calculate α for faceted growth as long as characteristic supersaturation (s_{char}) values are known. However, previously published measurements of s_{char} are limited to temperatures higher than -40°C . Estimates of s_{char} at temperatures between -40 and -70°C are provided here through reanalysis of vapor growth data. The estimated s_{char} follow the same functional temperature dependence as data taken at higher temperatures. Polynomial fits to s_{char} are used as inputs to a parameterization of α suitable for use in cloud models. Comparisons of the parameterization with wind tunnel data show that growth at liquid saturation and constant temperatures between -3 and -20°C can be modeled by ledge nucleation for larger (100s of μm) crystals, however comparisons with free-fall chamber data at -7°C suggest that dislocation growth may be required to model the vapor growth of small crystals ($\sim 20 \mu\text{m}$) at liquid saturation. The comparisons with free-fall chamber data also show that the parameterization can reproduce the measured pressure-dependence of aspect ratio evolution. Comparisons with a hexagonal growth model indicate that aspect ratio evolution based on the theory of Chen and Lamb (1994) produces unrealistically fast column growth near -7°C that is mitigated if a theory based on faceted growth is used. This result indicates that the growth hypothesis used in habit-evolving microphysical models needs to be revised when deposition coefficients are predicted.

28 1. Introduction

29 Cold cloud systems are sensitive to the manner in which ice vapor growth is parameterized
30 (Gierens et al. 2003; Avramov and Harrington 2010), and while our knowledge is sufficient to
31 formulate approximate models the mechanisms controlling ice crystal growth remain poorly
32 understood. Laboratory data for vapor grown ice crystals exist at temperatures above -40°C,
33 but the quantities measured in many laboratory studies (Nelson and Knight 1998; Libbrecht
34 2003b) are often not amenable to direct inclusion in the capacitance analogy that is almost
35 universally used in atmospheric applications. This has led to an unfortunate situation in
36 which the methods used to represent ice growth in atmospheric models are almost entirely
37 divorced from process-oriented measurements. While popular parameterization methods
38 have difficulties reproducing laboratory measurements (Westbrook and Heymsfield 2011;
39 Harrington et al. 2013b), a more fundamental issue is that these methods do not account
40 for the growth of faceted ice. Popular parameterizations are rooted in capacitance theory,
41 which assumes that the vapor density is constant over the crystal surface. The aspect
42 ratio cannot evolve in this model (Nelson 1994, pgs. 83-85) unless it is supplemented with
43 an auxiliary hypothesis (Chen and Lamb 1994). In contrast, faceting requires a uniform
44 flux boundary condition. Moreover, faceting indicates that crystal evolution is controlled
45 by surface attachment kinetics (hereafter “attachment kinetics”) that are supersaturation
46 dependent, leading to growth rates that can be substantially lower than those predicted
47 by the capacitance model (Nelson and Baker 1996). Only a handful of cloud modeling studies
48 include supersaturation-dependent attachment kinetics that are consistent with the theory
49 of faceted growth (MacKenzie and Haynes 1992; Wood et al. 2001; Zhang and Harrington

50 2015). All other studies assume either perfectly efficient attachment kinetics (capacitance
51 growth) or constant attachment efficiencies (deposition coefficients, α), approximations that
52 are only valid for a narrow range of conditions (Nelson 2005). These simplifications are
53 not limited to the world of model parameterizations, but also appear in interpretations of
54 measurements (Fukuta and Takahashi 1999; Magee et al. 2006).

55 The ubiquitous use of diffusion-only growth models is driven by the undeniable complexity
56 of crystal growth. However, there has been a trend to develop approximate models that are
57 consistent with the growth of faceted ice. These methods use laboratory-derived parameters
58 to drive changes in particle shape (Chen and Lamb 1994) and to estimate the attachment
59 efficiencies that control mass growth and shape evolution (Wood et al. 2001; Zhang and
60 Harrington 2014). The models are simple enough that they are amenable to application
61 within cloud models, providing a simplified theoretical approach for treating the influences
62 of attachment kinetics on the overall mass growth rate and the evolution of the habits
63 of single crystalline ice (cf. Zhang and Harrington 2015). Moreover, these methods can
64 also be used to extract approximate estimates of attachment kinetic influences on vapor
65 growth from laboratory measurements, thus directly linking laboratory measurements with
66 model parameterizations. In this paper, we provide a composite data set of characteristic
67 supersaturations (s_{char}) that are needed for supersaturation-dependent models of α . Ledge
68 nucleation has been proposed as the mechanism by which snow grows in atmospheric clouds
69 (Nelson and Knight 1998; Libbrecht 2003b), but this hypothesis has never been explored with
70 a crystal growth model. We show that the model of Zhang and Harrington (2014), referred
71 to herein as the Diffusion Surface Kinetics Ice Crystal Evolution (DiSKICE) model, can
72 reproduce the growth of single crystals at low pressures, and at high (liquid) supersaturations.

73 We also critique existing growth hypotheses that are used to evolve crystal shape (Chen and
74 Lamb 1994; Nelson and Baker 1996). We first review attachment kinetics and vapor growth
75 since these are integral to our analysis.

76 2. Ice Crystal Vapor Growth and Simplified Models

77 The rate of vapor uptake by growing crystals depends on the link between surface attach-
78 ment processes and vapor diffusion. Vapor molecules that impinge upon the surface must
79 find suitable attachment sites before they can incorporate into the bulk crystalline lattice.
80 If suitable attachment sites are uncommon a surface supersaturation, s_{surf} , will develop im-
81 mediately above the growing surface. Diffusion through the background gas supplies vapor
82 to the growing particle and removes the thermal energy generated by bond formation. The
83 rates of diffusion are driven by vapor and thermal gradients between the surface and the
84 ambient environment, and are therefore inextricably linked to the surface attachment rates.
85 The mass growth of crystals is therefore limited by both diffusion and attachment kinetic
86 processes, and is referred to as *diffusion-kinetics limited growth*.

87 a. Surface Processes and the Deposition Coefficients

88 During growth, a number of physical processes occur on the crystal surface that ultimately
89 determine the axis and mass growth rates. Vapor molecules must first adsorb to the crystal
90 surface, though not all molecules will necessarily do so. The efficiency of adsorption is often
91 referred to as a “sticking” probability, α_s , and though it is thought that this quantity is near

92 unity (Lamb and Scott 1974; Lamb and Chen 1995; Nelson 2001) at least one experiment
93 suggests it may be quite low (Asakawa et al. 2014). Adsorbed water molecules will migrate
94 across the surface and will desorb unless they find a suitable attachment site such as a ledge
95 or a surface vacancy. At relatively high temperatures ($> -5^{\circ}\text{C}$), the surface becomes rough
96 on the growth (nanometer) scale and many attachment sites are available for impinging
97 water molecules (Elbaum 1991). A quasi-liquid layer exists on ice surfaces (Bartels-Rauch
98 et al. 2014) and measurements show occurrences of this layer to temperatures as low as -
99 30°C (Constantin et al. 2018). The existence of this layer has been used in theories of habit
100 development (Kuroda and Lacmann 1982). Attachment kinetics can also change in time
101 because of surface transitions: Frozen drops can undergo a faceting transition where small,
102 pyramidal facets quickly grow themselves out of existence leaving only slower growing, larger
103 facets (Gonda and Yamazaki 1984). The formation of grain boundaries in polycrystalline ice
104 at low temperatures ($< -20^{\circ}\text{C}$) can be a source of dislocations that substantially alter crystal
105 growth (Pedersen et al. 2011). Low temperature ice is often complex in shape, is affected
106 by both cubic and hexagonal forms with stacking faults (Carignano 2007; Kuhs et al. 2012),
107 and with varied surface processes that control the growth. For example, crossed plates grow
108 with dislocations that propagate parallel to the grain boundary while the remaining facets
109 have slow growth rates (Furukawa and Kobayashi 1978). Scrolls appear to grow by either the
110 propagation of dislocations (Kobayashi et al. 1976) or by a protrusion mechanism (Nelson
111 and Swanson 2018).

112 Though the above surface processes control crystal growth rates, we lack the requisite
113 measurements to formulate general quantitative models. Consequently, surface processes
114 are typically treated in an aggregate sense, and with a single parameter for each facet called

115 a deposition coefficient, α . The deposition coefficient is the probability that a molecule
 116 impinging on the surface will contribute to bulk mass and axis growth, and it acts as a
 117 growth efficiency. The deposition coefficient has been measured in numerous studies, often
 118 with the approximation that α is constant. The measurements have been scattered from
 119 low (~ 0.001 , Choularton and Latham 1977; Magee et al. 2006) to high (> 0.2 , Skrotzki et al.
 120 2013; Kong et al. 2014) values. However, treating α as a constant is only valid for a small
 121 range of environmental conditions, crystal sizes, and specific, constant surface types.

122 The only available models of α are valid for faceted growth and they are supersaturation
 123 dependent. A parametric model of α was proposed by Nelson and Baker (1996),

$$124 \quad \alpha(s_{surf}, T) = \alpha_s \left(\frac{s_{surf}}{s_{char}} \right)^m \tanh \left(\frac{s_{char}}{s_{surf}} \right)^m \quad (1)$$

125 where α_s is the sticking probability which is assumed to be unity herein (see above), m is an
 126 adjustable parameter, s_{surf} is the surface supersaturation, and s_{char} is a laboratory-measured
 127 “characteristic” supersaturation. This latter quantity is, effectively, a scaling parameter that
 128 controls the supersaturation dependence of α ¹. The model treats growth by surface ledges
 129 in the sense that α rises commensurately with s_{surf} : As s_{surf} becomes larger the density of
 130 ledges rises leading to more efficient growth. The transition from inefficient (α near zero)
 131 to efficient growth (α near unity) is controlled by s_{char} . The parameter, m , determines the
 132 growth mechanism with a value of $m=1$ corresponding to growth by permanent dislocations

¹It is worth noting that this scaling supersaturation is referred to as a “critical” supersaturation in the theory of ledge nucleation (e.g. Nelson 2001, Eq. 2) and as a “transition” supersaturation when growth is controlled by spiral dislocations (Lamb 2000; Magee et al. 2006). However, we avoid using these terms since each of these two quantities has a specific theoretical definition, whereas the scaling supersaturation derived from measurements is often a parametric value.

as originally derived in the pioneering work of Burton et al. (1951), but given in the above form by Lamb and Scott (1974) (their Eq. 43). Values of $m \geq 10$ were shown by Nelson and Baker (1996) to be representative of growth by ledge nucleation. Dislocations are a permanent source of ledges and produce relatively efficient growth even at low s_{surf} whereas ledge nucleation causes a rapid onset of growth when s_{surf} is near s_{char} . Growth is not strongly dependent on m once the value is larger than 10 (see Zhang and Harrington 2015, their Fig. 1). In the studies below, we use $m = 1$ for dislocation growth and $m = 10$ for ledge nucleation as these are thought to be the primary growth mechanisms for faceted ice.

141 b. *Diffusion-Kinetics Limited Growth Model*

142 Including α (Eq. 1) in calculations of the mass and axis growth rates requires a model
143 for the gas-phase diffusion of vapor and thermal energy. While methods exist for explicitly
144 solving the diffusion-kinetic growth problem for faceted single crystals (Nelson and Baker
145 1996; Wood et al. 2001), these methods are complex and simplified methods rooted in the
146 capacitance model are an attractive alternative. Zhang and Harrington (2014) developed a
147 modified version of the capacitance model that calculates α for the major and minor axes
148 of spheroidal ice crystals, which are used to represent the general shape of atmospheric ice.
149 The semi-dimensions are defined in relation to the hexagonal structure of single crystalline
150 ice where a is half the basal plane maximum width, and c is half the prism plane height.
151 The model deviates from the capacitance model in that axis-dependent vapor fluxes are used
152 to determine s_{surf} and, therefore, the deposition coefficients for each axis (α_a and α_c) using
153 Eq. 1. This mimics the boundary condition for faceted growth, allowing different vapor

154 densities over the a and c axes and leads to (see Zhang and Harrington 2014, for details),

155

$$\frac{dm_i}{dt} = 4\pi C(c, a)\rho_{eq}s_i \left[\frac{1}{D_{eff}} + \frac{\rho_{eq}l_s}{K'_T T} \left(\frac{l_s}{R_v T} - 1 \right) \right]^{-1}. \quad (2)$$

156 In the above equation m_i is the crystal mass, $C(c, a)$ is the capacitance (see Westbrook et al.

157 2008, for comprehensive formulations), ρ_{eq} is the ice equilibrium vapor density, s_i is the

158 ambient ice supersaturation (hereafter supersaturation), R_v is the water vapor gas constant,

159 l_s is the sublimation enthalpy, T is the temperature, and K'_T is the thermal conductivity.

160 Equation 2 is identical to the capacitance model except that the diffusivity of vapor in air

161 (D_v) is replaced with a modified form, D_{eff} , that combines the influences of gas-phase vapor

162 diffusion and attachment kinetics (Zhang and Harrington 2014, their Eq. 11),

163

$$D_{eff} = \frac{2}{3} \frac{D_v}{\frac{4D_v C}{\alpha_a \bar{v}_v a c} + \frac{C}{C_\Delta}} + \frac{1}{3} \frac{D_v}{\frac{4D_v C}{\alpha_c \bar{v}_v a^2} + \frac{C}{C_\Delta}} \quad (3)$$

164 where \bar{v}_v is the vapor mean molecular speed and $C_\Delta = C(a + \Delta, c + \Delta)$ is capacitance

165 evaluated at a distance Δ (approximately the mean free path) away from the crystal surface.

166 The axis-dependent α values, therefore, enter directly into the mass growth rate through

167 D_{eff} and are calculated as in Zhang and Harrington (2014). Since the deposition coefficients

168 are usually less than unity, D_{eff} is less than D_v leading to mass growth rates that can be

169 substantially lower than the capacitance model. At high supersaturation α can approach

170 unity, and D_{eff} then becomes very nearly D_v . Therefore, the capacitance model is recovered

171 when α is near unity and growth becomes diffusion-limited. Since Eqs. 2 and 3 are applicable

172 to the diffusion and attachment kinetics growth limits we refer to it as the Diffusion Surface

173 Kinetics Ice Crystal Evolution (DiSKICE) model.

174 *c. Aspect Ratio Evolution*

175 Aspect ratio evolution requires a theory for the distribution of mass along each axis,
176 and the theories of Chen and Lamb (1994) and of Nelson and Baker (1996) are generally
177 used. Chen and Lamb (1994) hypothesized that the ratio of the axis growth rates is directly
178 proportional to the aspect ratio ($\phi = c/a$) and the ratio of the deposition coefficients (Γ),

179

$$\frac{dc}{da} = \frac{\alpha_c}{\alpha_a} \frac{c}{a} = \Gamma \phi \quad (\text{aspect-ratio based hypothesis}). \quad (4)$$

180 This hypothesis has been used in the development of habit-evolving microphysical models
181 (Chen and Lamb 1999; Hashino and Tripoli 2007; Chen and Tsai 2016; Jensen et al. 2017),
182 and in the interpretation of laboratory measurements (Sulia and Harrington 2011; Connolly
183 et al. 2012).

184 Crystals growing by ledge nucleation likely have ledges that form near crystal edges where
185 s_{surf} is greatest (Frank 1982, see Fig. 18). In this case, the ratio of the axis growth rates
186 depends only on the deposition coefficients (Nelson and Baker 1996),

187

$$\frac{dc}{da} = \frac{\alpha_c}{\alpha_a} \quad (\text{facet-based hypothesis}). \quad (5)$$

188 This hypothesis is less commonly used in cloud modeling (Wood et al. 2001), but it has
189 been used to interpret laboratory growth data (Nelson and Knight 1998; Nelson 2001). It is
190 worth noting that this hypothesis also breaks down in some situations. Ledge sources can be
191 at locations besides crystal edges (Nelson 2001), crystal hollowing may require a transition
192 from dislocation to ledge nucleation growth (Nelson and Knight 1998), crystals often have
193 non-faceted “rough” regions that may affect habit development (Pfalzgraff et al. 2010), and
194 molecules may migrate across crystal edges thus influencing ledge nucleation rates (Frank

195 1982). At this stage, however, there is no theoretical approach for including these more
196 detailed processes in a simplified model of single crystal habit development.

197 As the above discussion implies, our theoretical knowledge is insufficient to explicitly
198 model the development of secondary habit features that appear at high supersaturations,
199 such as dendritic branching and hollowing. These features are normally treated through
200 an “effective” density (ρ_{eff}) that is lower than the bulk density of ice and accounts for
201 the branches and hollowed regions that are not represented in the modeled particle. This
202 approach has a long history in modeling (Miller and Young 1979; Chen and Lamb 1994;
203 Thompson et al. 2008) and in data analysis (Fukuta 1969; Fukuta and Takahashi 1999).
204 Chen and Lamb (1994) approximate the density added during growth (the *deposition density*)
205 with an empirical equation that depends on temperature and supersaturation (their Eq. 42).
206 This form allows the effective density to decline during growth and is used in the simulations
207 presented below.

208 For the sake of completeness, we note that laboratory evidence suggests the aspect ratio
209 should be treated as a constant during sublimation with a sublimation coefficient of unity
210 (Nelson 1998), an approach we advocate here. This result has a physical basis: Measurements
211 suggest that crystal roughening during sublimation (Nelson 1998; Magee et al. 2014) causes
212 the sublimation coefficient to approach unity. As a consequence, the vapor density becomes
213 constant along the surface leading to a constant aspect ratio (shape is preserved) during
214 sublimation (Ham 1959).

215 **3. Characteristic Supersaturations**

216 *a. Synopsis of previously published data*

217 Characteristic supersaturations are required as input to the α -dependent growth model,
218 however available measurements of s_{char} (Fig. 1) are sparse and restricted to temperatures
219 above -40°C . Various laboratory devices along with different models of α have been used in
220 prior measurements of s_{char} . We therefore expect variations in s_{char} that are due solely to
221 differences in the measurement techniques and analytical models employed in a given study.
222 It is beyond the scope of this work to assess and interpret these differences; instead we have
223 collected prior published data and use them to provide a proxy data-set.

224 The available data clearly indicate that s_{char} increases commensurately with the super-
225 cooling ($T_{\text{o}}-T$, $T_{\text{o}} = 273.15\text{K}$, Fig. 1). Values of s_{char} are relatively low ($< 2\%$) at tempera-
226 tures above -20°C and their variation with temperature is consistent with the primary habits
227 of ice for most of the data sets: At a given temperature, the major growth axis has lower
228 values of s_{char} than the minor axis, and this difference will produce a larger α in Eq. 1 and
229 faster major axis growth. At temperatures above -20°C the data of Libbrecht and Rickerby
230 (2013) generally have the highest s_{char} and these data show basal and prism values for s_{char}
231 that are the opposite of the other published measurements. Libbrecht and Rickerby (2013)
232 point out that this result may be due to processes occurring on the crystal surface that are
233 not captured by the standard models of α . Nevertheless, using these data directly in Eq. 1
234 produces the wrong primary habits and they are not included in our studies above -20°C .

235 Values of s_{char} become progressively larger as the temperature falls below -20°C indicat-
236 ing that growth becomes more strongly limited by attachment kinetics. The data sets are

237 relatively consistent with one another though substantial outliers occur near -30°C. Only
238 two published measurements of the basal and prism facet s_{char} exist, and they are consis-
239 tent with one another below -20°C (Libbrecht 2003b; Libbrecht and Rickerby 2013). The
240 measured s_{char} are nearly the same on the basal and prism facets, with habit development
241 controlled by the leading coefficient in the ledge nucleation rate equation (Libbrecht 2003b).
242 These results contrast with those of (Nelson and Knight 1998) in which the primary habits
243 are controlled by the difference in s_{char} between the basal and prism facets at higher tem-
244 peratures. Zhang and Harrington (2014) used the data of (Libbrecht 2003b) to derive values
245 of s_{char} consistent with the formulation of Nelson and Knight (1998), however these derived
246 values produce thick columns and plates at high supersaturation, whereas thinner particles
247 are often observed. Libbrecht (2003a) has hypothesized that the production of thinner plates
248 may be due to structure-dependent α that is currently not accounted for in theories. Un-
249 fortunately, there is no consensus on the modeling of the primary habits at temperatures
250 below -20°C, though the values of s_{char} are relatively consistent among the data sets. At
251 lower temperatures we therefore estimate the reduction in the mass growth rate by using a
252 single, particle-averaged value of s_{char} following Zhang and Harrington (2014).

253 The use of a single s_{char} to characterize attachment kinetic influences on growth has
254 precedence. Our prior work (Zhang and Harrington 2014, their Fig. 10) showed that the
255 mass uptake of non-spherical single crystals is very nearly reproduced if a particle-average
256 value of s_{char} and an equivalent volume sphere are used in the mass growth calculations.
257 Figure 2 shows a similar result for the instantaneous mass growth rate (normalized to the
258 capacitance rate) for a large range of aspect ratios, supersaturations, and for both ledge
259 nucleation and dislocation growth. The relative errors between the solutions using separate

260 s_{char} for each axis, and the solutions using an average s_{char} , are generally less than 5%,
261 however the relative error becomes substantial (20% or larger) when crystals are small ($r <$
262 10 μm) and the supersaturation is near s_{char} (not shown). Given that an average value
263 of s_{char} and an equivalent volume sphere provide an accurate estimate of the overall mass
264 growth rate for highly anisotropic single-crystal growth, we hypothesize that it is reasonable
265 to employ this method for the growth of crystals with more complex facet morphologies, as
266 is observed near liquid saturation and in polycrystalline ice. This hypothesis underlies the
267 estimates of s_{char} provided in the following subsection.

268 *b. Estimating Characteristic Supersaturations at $T < -40^\circ\text{C}$*

269 To our knowledge, no data exist for s_{char} and the growth rates of individual crystal facets
270 at temperatures below -40°C . However, mass growth rate data do exist at these temperatures
271 for individually grown crystals (Magee et al. 2006). These growth data are particularly useful
272 as they record the vapor growth and sublimation of individual crystals, formed from frozen
273 small droplets (radius of 5 to 7 μm), at nearly constant temperatures of -59.8 , -50 , -44 , and
274 -42°C . Estimates of α were low ($\simeq 0.006 \pm 0.002$) and different values of α were required
275 to fit each growth and sublimation portion of the time-series (Magee et al. 2006). Crystal
276 growth was likely inhibited by attachment kinetics indicating that it would be useful to
277 estimate an particle-average value of s_{char} from the data.

278 An example of the growth data is shown in Fig. 3a for a crystal undergoing cycles of
279 sublimation and growth at a temperature of -59.8°C and a pressure of 972 hPa. The
280 mass evolution of the crystal depends on its initial size, which can be determined to about

281 1 μm . This size uncertainty dominates the errors in determining α and therefore s_{char} ,
282 whereas supersaturation errors have a smaller influence on α (Magee et al. 2006). We use
283 the DiSKICE model to fit the measured mass ratio time-series by minimizing on a value
284 of s_{char} . We assume spherical particles because the grown crystals were small (5 to 20 μm
285 in radius), and therefore had likely not developed pronounced habits. In the model we
286 also assume that the crystals grow by ledge nucleation. Equally accurate fits are possible
287 if dislocation growth is assumed, however a low sticking efficiency ($\alpha_s \sim 0.004$) is then
288 required. Though this result is conceivable, it seems unlikely as high values of α (above 0.1)
289 that have been estimated at $T < -40^\circ\text{C}$ (Skrotzki et al. 2013; Pokrifka 2018) would then
290 not be possible even for dislocation growth at liquid saturation.

291 The model fits to the measured mass ratios are shown in Fig. 3a for the most probable
292 initial radius ($r_0 = 7 \mu\text{m}$), and the upper ($r_0 = 8.4 \mu\text{m}$) and lower ($r_0 = 5.4 \mu\text{m}$) uncertainty
293 bounds. The fit has the same accuracy as that of Magee et al. (2006) except that α varies
294 with time (Fig. 3b), rising and decreasing commensurately with the supersaturation. The
295 rapid decline in α with decreasing supersaturation is the reason the model captures the
296 relatively flat region in the mass growth time-series (such as 500 to 1000 seconds) that is
297 not reproducible with a diffusion-limited growth model. More critically, the values of s_{char}
298 (on Fig. 3b) required to fit each growth and sublimation period are relatively similar to one
299 another. These results suggest that a similar surface process is occurring on these crystals
300 throughout the growth cycle. Finally, note that the predicted time-series of α falls within
301 the range (grey shades) determined by Magee (2006) with α that are similar to his fitted
302 values (dashed lines).

303 The values of s_{char} determined from the above growth time-series, and the other exper-

304 iments at -50, -44, and -42°C (not shown), are given on Fig. 1. Note that values for the
305 sublimation portions of the time-series are not shown since the present study focuses on
306 depositional growth. The values of s_{char} determined from the fitting procedure are some-
307 what lower than prior measurements near -40°C, but show a similar temperature trend for
308 s_{char} as the higher temperature data. The approximate consistency of our results with prior
309 data is encouraging, but comparisons with independent data would be useful. In particular,
310 experiments run over a range of supersaturation for fixed temperatures and single crystalline
311 ice would be ideal.

312 It is important to point out that one should exercise caution in the use of the estimates
313 of s_{char} provided herein since it is always possible that experimental artifacts affected the
314 resulting growth curves. It is also not known whether the measured crystals developed facets
315 quickly after the periods of sublimation, though the observations of Gonda and Yamazaki
316 (1978) suggest that facets appear rapidly after droplet freezing and the electron microscope
317 studies of Pfalzgraff et al. (2010) show that facets rapidly re-appear (within 90 seconds to a
318 few minutes) during re-growth following sublimation. In addition, (Magee et al. 2006) found
319 that low values of α were required for sublimation, a result which is inconsistent with other
320 measurements (Nelson 1998; Magee et al. 2011).

321 *c. Comparisons with effective capacitance measurements*

322 Bailey and Hallett (2004) reported on thermal gradient diffusion chamber measurements
323 of crystals grown on a substrate. From these growth measurements capacitance values nor-
324 malized to the maximum dimension (L_i) were extracted for hexagonal plates and columns

325 (Bailey and Hallett 2010). These values were estimated by using the measured mass growth
326 rate and then solving for the capacitance in the capacitance mass growth model. The nor-
327 malized capacitance is a useful metric for vapor growth since it is independent of size, and
328 depends only on the aspect ratio for diffusion-limited growth. Normalized capacitance values
329 extracted from growth data would therefore fall approximately along the solid black lines
330 in Fig. 4 if growth is diffusion-limited. However, Bailey and Hallett (2010) found that the
331 measured normalized capacitance was substantially lower than capacitance theory (Bailey
332 and Hallett 2010, their Fig. 12). Their data span a relatively large range of L_i (50 to 400
333 μm), pressure p (500 to 150 hPa), and supersaturation s_i (1 to 13% at -40°C , and 5 to 25%
334 at lower temperatures), and an apparent aspect ratio dependence reminiscent of capacitance
335 theory appears in the data. The approximate ranges of their data are shown on Fig. 4.

336 There are many possible reasons why the extracted values of the normalized capacitance
337 are lower than capacitance theory (see Bailey and Hallett 2010), but one main reason is
338 that attachment kinetics are not included in the capacitance model. Therefore the extracted
339 values of $\frac{C}{L_i}$ are convolved with the attachment kinetics, producing an effective normalized
340 capacitance (c_{eff}) that can be computed with DiSKICE. Values of c_{eff} are computed using
341 DiSKICE assuming ledge nucleation growth and using values of s_i , L_i , and p in the middle
342 of the measured range. The spread of model solutions is calculated using the range of
343 measured s_i (light shades), and L_i and p (dark shades). The calculated c_{eff} using the
344 s_{char} values derived in §2b (green lines, Fig. 4) are often consistent with the range of the
345 measurements, though at -40 and -50°C the calculated values are too high for planar crystals
346 and at -60°C the calculated values are at the lower end of the measured range. Including
347 the variability in s_i , L_i , and p produces a range of solutions that has a spread similar to the

348 red and blue shaded regions (discussed below) and often encompasses the measured range
349 (omitted for clarity).

350 Because some of the c_{eff} calculations fall outside of the observed range, it is useful to
351 calculate adjusted values of s_{char} so that spread in the solutions is confined primarily within
352 the measured range of c_{eff} . This was done by adjusting s_{char} until the accumulated error in
353 the spread of the model solutions as compared to the observed range of c_{eff} is a minimum.
354 These are shown as the red and blue shaded regions on Fig. 4 along with model solutions
355 using the mid-range values of s_i , L_i , and p (solid red and blue lines). The spread in the
356 solutions for c_{eff} is weakly dependent on p and L_i , but is dominated by s_i with a spread
357 that is similar to that of the measured range; these are expected results since attachment
358 kinetics are dominated by the supersaturation dependence through α . The adjusted values
359 of s_{char} , along with a large uncertainty based on the spread of s_i values, are given in Fig. 4
360 and Fig. 1. These adjusted values are generally consistent with those from the data of Magee
361 et al. (2006), and also provide a data point at -70°C.

362 Naturally, one should bear in mind that these adjusted values of s_{char} are very rough
363 estimates calculated from measured growth rates of crystals of various sizes over a range
364 of pressures and supersaturations. Furthermore, since numerous crystals were grown from
365 the substrate in these experiments, the vapor diffusion field in the chamber deviated from
366 that of a classic flat-plate diffusion chamber. The simulations of Westbrook et al. (2008)
367 show that crystal crowding could have reduced the measured growth rates by as much as
368 a factor of three, though Bailey and Hallett (2010) point out that they only analyzed data
369 from conditions that were not crowded (their Fig. 11). Though Bailey and Hallett (2010)
370 evidently took great care to exclude crowded conditions, the influence of the population of

371 crystals on the three-dimensional vapor field, and the horizontal diffusion that must occur,
372 is not known. Moreover, as a reviewer of the current paper pointed out, the crystals grown
373 from the central strand are asymmetric since one end is attached to the substrate. The
374 attached end of the crystal therefore cannot grow naturally and deplete the vapor supply,
375 and it is unknown how important this effect would be to the measured growth rates.

376 *d. Polynomial Fits to Characteristic Supersaturation Data*

377 A subset of the data shown in Fig. 1 is used to produce polynomial fits (Table 1) to
378 s_{char} as a function of temperature (black curves). The data subsets used in the fits were
379 selected as follows: At temperatures above -20°C, the data from Nelson and Knight (1998)
380 and Libbrecht (2003b) were used. The data from Libbrecht and Rickerby (2013) were not
381 used for reasons discussed earlier, and Sei and Gonda (1989) was excluded because the
382 growth was possibly influenced by thermal substrate effects (Nelson 1993). At temperatures
383 below -20°C the data of Libbrecht (2003b) are used along with s_{char} estimated in §2b and
384 c above. The basal facet s_{char} for temperatures between -20 and -30°C (black solid curve)
385 is taken from the estimates made by Wood et al. (2001). We use a particle-average s_{char}
386 and an equivalent volume sphere at temperatures below -30°C, since these values are rough
387 approximations of attachment kinetic influences on crystal growth. The only exceptions are
388 model tests of columnar growth for which an ad hoc reduction in the basal s_{char} is applied
389 (black dotted curve) based on the analysis of Zhang and Harrington (2014).

390 4. Single Crystal Evolution at Low and High Supersat- 391 uration

392 The evolution of the primary habits of single crystalline ice depends on the growth
393 hypothesis (Eq. 4 or 5) that is employed. A number of studies have shown that the aspect-
394 ratio based hypothesis of Chen and Lamb (1994) can reproduce the evolution of the primary
395 habits of ice at liquid saturation, however those works were predicated on the assumption
396 that the ratio of the deposition coefficients (Γ) is a constant at a given temperature. It is not
397 immediately evident which hypothesis (Eq. 4 or 5) provides a more accurate representation
398 of aspect ratio evolution when deposition coefficients are predicted. Moreover, it is not
399 clear whether ledge nucleation or dislocation growth provides a better representation of axis
400 evolution at high supersaturations. Below, we analyze the two growth hypotheses and extend
401 the studies to secondary habits (branching and hollowing) through a reduced density.

402 a. *Assessment of Axis Growth Hypotheses*

403 The hexagonal ice growth model developed by Wood et al. (2001) was used in prior work
404 to assess the axis-dependent growth of crystals using DiSKICE (Zhang and Harrington 2014).
405 The hexagonal model solves the Laplace equation on a triangular grid covering the basal and
406 prism facets of hexagonal ice using the constant-flux boundary condition for faceted growth.
407 The model is limited in that simulations of branched and hollowed crystals are not possible.
408 Nevertheless, the hexagonal model reproduces the general features of faceted growth and
409 provides a convenient comparison basis for simplified theories. For the simulations below,

410 the hexagonal model is set-up as in Zhang and Harrington (2014) with ledge nucleation
411 growth occurring where s_{surf} is a maximum and dislocation growth at the facet centers.
412 Characteristic supersaturations used in DiSKICE and the hexagonal model were calculated
413 from the polynomial fits in Fig. 1, though the short dashed curve is used for the basal facets
414 at temperatures below -20°C so that a columnar aspect ratio develops (discussed below).

415 Results of the comparison between the hexagonal and DiSKICE model in the work of
416 Zhang and Harrington (2014) were encouraging in the sense that the general dependence of
417 growth on aspect ratio and α were captured by the DiSKICE model. However, those studies
418 were limited in a number of ways: They only examined growth by dislocations at high s_i and
419 ledge nucleation growth at low s_i , but did not examine the facet-based growth hypothesis
420 (Eq. 5), nor growth where branching and hollowing are treated through a reduction in the
421 particle density. While the aspect-ratio based hypothesis was found to compare well to
422 the hexagonal model, there were indications that dislocations produced excessive columnar
423 growth near -7°C. This excessive growth is inherent in the aspect-ratio based hypothesis,
424 which becomes markedly clear for ledge nucleation growth as is shown in Fig. 5a. After
425 10 minutes of growth at liquid saturation with the aspect-ratio based hypothesis, columnar
426 crystals at temperatures between -5 and -9°C are nearly an order of magnitude longer,
427 and significantly thinner, than the hexagonal model solutions. This result also occurs at
428 temperatures below -20°C, however the excessive growth is weaker here because of the larger
429 values of s_{char} . Aspect ratio evolution is also excessive for planar crystals, and so it appears
430 that at high supersaturation the aspect-ratio based hypothesis produces crystals that are
431 too large and with extreme aspect ratios. This excessive growth is substantially muted
432 at low supersaturations, and only an indication is seen near -7°C (Fig. 5b). In contrast,

433 the facet-based hypothesis produces a much better match to the hexagonal model at high
434 supersaturations, though at low supersaturations there is less of a distinction between the
435 two hypotheses. Growth by dislocations produces a similar result at high (liquid) saturation
436 (Fig. 6) and low saturation (not shown), though the enhancement of axis growth is far lower
437 than it is for ledge nucleation.

438 The reasons for enhanced growth in the aspect-ratio based hypothesis can be understood
439 by examining the time evolution of the semi-axis lengths, α , and the axis-dependent vapor
440 fluxes at -7°C , where enhanced growth is the most excessive. The evolution of the c and
441 a axes is clearly better represented by the facet-based hypothesis at all supersaturations
442 (Fig. 7a and b). In contrast, the aspect-ratio based hypothesis becomes progressively worse
443 at higher supersaturations, with time-dependent values of a and c diverging substantially
444 from the hexagonal model solution. At high supersaturation ($s_i = 7\%$), runaway growth is
445 produced for the c -axis while the a -axis essentially ceases growth after about 200 seconds,
446 producing columns with extreme aspect ratios ($\phi = 190$).

447 Strong growth along the c -axis in the aspect-ratio based hypothesis indicates that α
448 is large (Fig. 8). While the facet-based hypothesis produces α values that follow a similar
449 functional form to those predicted by the hexagonal model, with α for each axis decreasing in
450 time, the aspect-ratio based hypothesis produces a rise in α for the c -axis that is accentuated
451 at higher supersaturations. Interestingly, α for the a -axis follows a similar functional form
452 to the hexagonal model solution. The increasing values of α with time indicate that s_{surf} ,
453 and hence the vapor flux onto the c -axis, must be rising unrealistically in time, which is
454 indeed the case (Fig. 9). This increase in the vapor flux onto the c -axis is driven by the
455 aspect ratio dependence of the growth hypothesis (Eq. 4). The equation originates from the

ratio of the vapor fluxes along the c and a -axes in the capacitance model (Chen and Lamb 1994, Eq. 25), and therefore the inclusion of ϕ essentially multiplies the vapor flux from the facet-based hypothesis by the aspect ratio, causing an increase in the vapor flux by a factor of ϕ . When α values are predicted, including this scaling by ϕ causes a positive feedback where increasing aspect ratio amplifies the fluxes onto the major axis. This result does not occur in the original model of Chen and Lamb (1994) because the ratio of the deposition coefficients is constant at a given temperature. The artificial feedback with aspect ratio evolution that occurs when α is predicted is a general feature of the aspect-ratio based hypothesis, and it indicates a flaw in the hypothesis that becomes progressively worse at higher supersaturations. Consequently, models that evolve crystal shapes based on this hypothesis (Hashino and Tripoli 2007; Harrington et al. 2013a; Chen and Tsai 2016) would have to be modified to use the facet-based hypothesis if α is predicted.

b. Comparison with Laboratory Measurements at Liquid Saturation

While the above comparisons, and the studies of Zhang and Harrington (2014), indicate that DiSKICE provides a suitable approximation for single crystal growth as compared to the hexagonal growth model of Wood et al. (2001), no comparisons to growth data have been done. Few comprehensive data sets exist to which analytical growth models can be compared, and this is especially true at low ice supersaturations. However, a few data sets exist from wind tunnel measurements of crystals grown at liquid saturation. The comprehensive data set of Fukuta and Takahashi (1999) is particularly useful because freely suspended crystals were grown for long periods of time (up to 30 min) and data are reported for the axis

477 lengths, crystal mass, and fall speed (reproduced in Figs. 10 and 11). Crystals grown in
478 these experiments had a range of initial sizes, but the model simulations below use spheres
479 with an initial radius of 10 μm based on the studies of Sulia and Harrington (2011). All of
480 the simulations below are integrated for up to 15 minutes at liquid saturation, and a constant
481 temperature and pressure (1000 hPa). Effective density, fall speed and ventilation effects
482 are computed following the axis-dependent approach described in Chen and Lamb (1994).
483 Ventilation effects are particularly important here, as they strongly impact the growth rates
484 for larger crystals. As discussed by Chen and Lamb (1994), ventilation effects tend to not
485 only increase the overall mass growth rate, but the major axis growth rate is also amplified
486 leading to thinner crystals.

487 Prior comparisons with the above data using the Chen and Lamb (1994) model showed
488 that the mass, axis lengths, and fall speed could be captured with relatively high accuracy
489 (Sulia and Harrington 2011; Harrington et al. 2013b). Typical results from those comparisons
490 are reproduced in Figs. 10 and 11. These results provide a benchmark for the DiSKICE model
491 at liquid saturation, and they also underscore an important point: Diffusion-limited growth
492 captures the mass evolution of the observed crystals, whereas the deposition coefficient ratio
493 (Γ) primarily controls aspect ratio evolution.

494 Simulations of crystal growth with DiSKICE used the facet-based hypothesis and ledge
495 nucleation growth with s_{char} from the polynomial fits given in Table 1. The DiSKICE
496 simulations produce results that are similar to those of the Chen and Lamb (1994) model
497 and also compare well with the observed evolution of axis length (Fig. 10), mass and fall
498 speed (Fig. 11). Like the Chen and Lamb (1994) model, relative errors in the simulated
499 crystal properties can often be large (20 to 50% for individual data points), but the general

500 qualitative features of habit evolution and crystal fall speed are reproduced by DiSKICE:
501 The model captures the strong increase in the a -axis length and the small c -axis length
502 where thin dendritic crystals are observed near -15°C . These crystals have a low effective
503 density (around 100 kg m^{-3}) and a large area leading to low fall speeds. The model predicts
504 the greatest fall speeds, and lowest masses, near the habit transition temperatures of -10
505 and -22°C , where crystals are relatively isometric with low drag and high effective density.
506 Similar to the Chen and Lamb (1994) model, columns are predicted to be too thin with
507 a -axis lengths that are smaller than observed.

508 Simulations assuming dislocation growth on the basal and prism facets produce crystals
509 that are too thick in comparison to the measurements (Fig. 10). This result occurs because
510 dislocations, unlike ledge nucleation, produce relatively high α along both axes (Fig. 12a).
511 Consequently, both the a and the c axes grow with high efficiency.

512 There are two other possible mechanisms that could produce thin crystals. It is certainly
513 possible that dislocation growth could occur on the primary growing axis whereas ledge
514 nucleation could occur on the weakly growing facet. However, simulations of this process
515 produce crystals that are far too thin in comparison to the measurements (not shown).
516 It is also possible that both dislocations and ledge nucleation occur on each facet, and
517 that the growth mechanism with the largest α controls the growth (Nelson and Knight
518 1998). DiSKICE simulations with α chosen based on the most efficient growth mechanism
519 produce thicker crystals reminiscent of dislocation growth. A key result of these simulations
520 is that only ledge nucleation for each axis can reproduce crystal growth at liquid saturation.
521 However it should be borne in mind that real crystals may indeed grow by the aforementioned
522 mechanisms, and that the inability of DiSKICE to reproduce those growth mechanisms

523 may indicate a limitation of the model. Nevertheless, from a practical parameterization
524 perspective, ledge nucleation can be used to reproduce the growth of thin crystals at liquid
525 saturation.

526 It is curious that the model of Chen and Lamb (1994) and DiSKICE produce results
527 that are similar to one another at liquid saturation even though the models are driven by
528 different data sets. The ratio, Γ , used in Chen and Lamb (1994) is derived from the α
529 measurements of Lamb and Scott (1974) whereas DiSKICE uses polynomial fits to s_{char}
530 that are primarily due to the measurements of Nelson and Knight (1998) at these tempera-
531 tures. Since the models use different growth hypotheses, namely the facet-based hypothesis
532 for DiSKICE and the aspect-ratio based hypothesis for Chen and Lamb (1994), the results
533 presented above suggest that α_c/α_a for ledge nucleation growth should approach $\Gamma(T)c/a$
534 in the limit of liquid saturation. While we have not discovered an analytical proof of this
535 assertion, it is consistent with the model results. For instance, Fig. 12b shows the ratio
536 α_c/α_a and $\Gamma(T)c/a$ from the DiSKICE and Chen and Lamb (1994) simulations, respectively.
537 The ratios are nearly identical after 15 minutes of growth. The near equality of these ratios
538 provides circumstantial evidence for why the Chen and Lamb (1994) model is accurate at
539 liquid saturation: The ratio $\Gamma(T)c/a$ provides a parameterization of the change in the depo-
540 sition coefficient ratio. Moreover, these results provide a tantalizing hint of an underlying
541 commonality in the measurements of Lamb and Scott (1974) and Nelson and Knight (1998).

542 *c. Aspect ratio dependence on pressure*

543 In a series of experiments using a free-fall chamber Gonda (1980) measured the de-
544 pendence of aspect ratio on the vapor diffusion coefficient (D_v) at liquid saturation and
545 temperatures of -7°C and -15°C . Crystals formed from frozen liquid droplets fell about 5 to
546 perhaps 15 cm onto a window where the crystals could be imaged. Measurements showed
547 that the aspect ratios of the hexagonal crystals deviated further from unity for higher pres-
548 sure or lower values of D_v (Fig. 13), a result that is consistent with theory: The rate of vapor
549 diffusion to a growing crystal depends on the background gas pressure. At relatively high
550 pressure (low D_v), the diffusive resistance to vapor transport is large, keeping s_{surf} below
551 the value needed for minor axis growth. The aspect ratio of the crystal therefore evolves
552 away from unity in time. Conversely, at very low pressures (high D_v) the background gas
553 provides little resistance to the flow of vapor. In this case s_{surf} is closer to the ambient value
554 (liquid saturation), which is above s_{char} for each axis and, therefore, isometric crystals are
555 produced.

556 We simulated a scenario similar to the experiments of Gonda (1980) by allowing initially
557 spherical crystals (radius of 2 μm following Nelson 2001) to grow while falling 10 cm. The
558 model of Chen and Lamb (1994) cannot reproduce the dependence of crystal aspect ratio on
559 D_v since Γ is constant (not shown). In contrast, the DiSKICE model produces an aspect ratio
560 dependence on D_v that is similar to the measurements (Fig. 13). Moreover, either dislocation
561 growth or ledge nucleation can fit the data though different s_{char} values are required for each
562 growth mechanism. At -7°C dislocation growth requires s_{char} for the basal and prism facets
563 of 0.18 and 0.48%, respectively, values that are similar to those from prior measurements

564 (0.2 and 0.44%, Fig. 1). Ledge nucleation requires nearly the same s_{char} for each axis, which
565 is inconsistent with prior measurements. This result is consistent with Nelson (2001), who
566 concluded that dislocation growth likely controlled the growth at -7°C. At -15°C, however,
567 the s_{char} required for ledge nucleation to fit the growth data (0.59 and 1.5% for the prism and
568 basal facets, respectively) are consistent with prior measurements (0.54 and 2.1%, Fig. 1).
569 Dislocation growth requires s_{char} for the basal facet to be an order of magnitude smaller
570 than observed. This result is consistent with the findings of Nelson (2001) at low D_v , which
571 indicated that ledge nucleation was occurring, but not at high D_v , where it appeared that
572 dislocation growth occurred. The inconsistency may indicate a limitation of our model at
573 higher D_v , but further laboratory studies are needed.

574 Because the experiments of Gonda (1980) produced only small crystals (less than 20 μm)
575 formed from frozen droplets, it is likely that dislocation growth dominated much of the early
576 growth of these crystals. However, explaining the thin crystals from the wind tunnel data of
577 Fukuta and Takahashi (1999) requires ledge nucleation. Taken together, these results suggest
578 that the early growth of small crystals may be dominated by dislocation growth while ledge
579 nucleation dominates the growth at latter stages when crystals are large. This conclusion is
580 broadly consistent with the discussions of Nelson (2001) and with the results of Gonda and
581 Yamazaki (1984) who showed that crystals formed from frozen drops initially grow efficiently
582 until facets become large enough that ledge nucleation dominates the growth.

583 5. Summary and Concluding Remarks

584 In this paper we have provided a composite data set for the characteristic supersatu-
585 rations, s_{char} , that are needed as input for supersaturation-dependent α . We have shown
586 these s_{char} values, when used in conjunction with the model of Zhang and Harrington (2014),
587 can reproduce the growth of ice crystals at liquid saturation as observed in a wind tunnel.
588 Only ledge nucleation for both axes is capable of reproducing the mass, lengths, and the fall
589 speed of the measured crystals. Axis evolution based on the aspect-ratio based hypothesis
590 of Chen and Lamb (1994) produces columnar growth with unrealistic aspect ratios when α
591 is predicted. This occurs because of a positive feedback with the aspect ratio in that param-
592 eterization method. Only the facet-based hypothesis of Nelson and Baker (1996) is capable
593 of reproducing aspect ratio evolution from a hexagonal model at both low and high super-
594 saturations. While the diffusion-limited model of Chen and Lamb (1994) will not produce
595 variations in aspect ratio with pressure, the model of Zhang and Harrington (2014) is capable
596 of reproducing pressure-dependent growth of small crystals in comparison to measurements.

597 The growth of ice at low temperatures ($T < -30^{\circ}\text{C}$) has been infrequently measured,
598 though the experiments of Libbrecht (2003b) indicate that the growth of basal and prism
599 facets is driven primarily by ledge nucleation. However, no measurements of s_{char} exist at
600 temperatures below -40°C . To help fill this gap, we reanalyzed prior measurements (Magee
601 et al. 2006) to extract average values of s_{char} between -40 and -60°C . Comparisons to the
602 normalized capacitance values of Bailey and Hallett (2010) for hexagonal plates and columns
603 indicate that our values of s_{char} are generally consistent with their data.

604 It is critical to bear in mind the approximate nature of the analyses at $T < -40^{\circ}\text{C}$. At

605 present, only two data sets have been published with precise measurements of facet growth
606 down to -40°C, and both data sets indicate that s_{char} increases with decreasing temperature
607 (Libbrecht 2003b; Libbrecht and Rickerby 2013). While the s_{char} values from our analysis
608 are consistent with these measurements, it is not known whether our values are truly repre-
609 sentative of the average growth of basal and prism facets. The morphology of crystals grown
610 in the studies of Magee et al. (2006) is unknown, and since those crystals were formed from
611 frozen droplets they likely were polycrystalline (Bacon et al. 2003), may have had mesoscopic
612 surface features (Magee et al. 2014), and likely underwent a transition as facets emerged.
613 These processes, and others, would be convolved together in the values of s_{char} that we have
614 derived.

615 It is also difficult to relate these known growth mechanisms, and measured growth rates,
616 to the measures of crystal roughness reported in the literature (Neshyba et al. 2013; Magee
617 et al. 2014; Schnaiter et al. 2016). Magee et al. (2014) showed mesoscopic features on
618 crystal facets, yet growth was at times limited by attachment kinetics. Moreover, Pedersen
619 et al. (2011) found weak growth of crystal facets until a grain boundary is formed through
620 the contact of two dissimilar facets. More recently, Voigtländer et al. (2018) indicated
621 that crystals cycled between growth and sublimation show reduced growth rates in later
622 cycles, and that surface roughening can increase during cycled growth. This latter result
623 is consistent with prior measurements that show faceting disappears and crystals roughen
624 during sublimation (Nelson 1998; Magee et al. 2014). Taken together, these results indicate
625 that our understanding of ice vapor growth is still in its infancy. Approximate models, such
626 as the one posed in this paper, must be used with caution and should be interpreted as a
627 place-holder for a more precise theory of ice growth.

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3 Evolution of the measured mass ratio and ice saturation ratio (a) and the deposition coefficient (b) at -59.8°C and 972 hPa. The measured mass-ratio (m/m_{∞} where m_{∞} = initial mass) is given by the black diamonds and ice saturation ratio (S_i) is given by the blue circles. Best fits to the mass-ratio time-series using the DiSKICE model are shown by the solid lines. The fits used the most likely initial radius (black line, $r_{\infty} = 7 \mu\text{m}$) along with the upper (blue line, $r_{\infty} = 8.4 \mu\text{m}$) and lower (red line, $r_{\infty} = 5.4 \mu\text{m}$) limits, since uncertainty in the initial radius dominates the error. Evolution of the predicted deposition coefficient uses the three initial radii given in (a). Also shown are the best estimate α (black dashed lines and number) and the range (shaded region) from Magee (2006). Best-fit values of s_{char} for sublimation (sub) and deposition (dep) from each DiSKICE time-series is given on the figure and color-coded to match the simulations.

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879 13 Aspect ratio of crystals as a function of the vapor diffusivity, D_v , at -7 and -15°C (red
880 and blue colors, respectively). Solid circles indicate the free-fall chamber measurements
881 of Gonda (1980) made after crystal vapor growth at liquid saturation. Simulated crystals
882 began as spheres with an initial radius of 2 μm and grew during free-fall over a distance of
883 10 cm. Dislocation growth (dashed line) used a and c -axis s_{char} , respectively, of 0.48% and
884 0.18% at -7°C, and 0.04% and 1.9% at -15°C. Ledge nucleation growth (solid line) used a
885 and c -axis s_{char} , respectively, of 0.48% and 0.35% at -7°C, and 0.59% and 1.5% at -15°C. 58

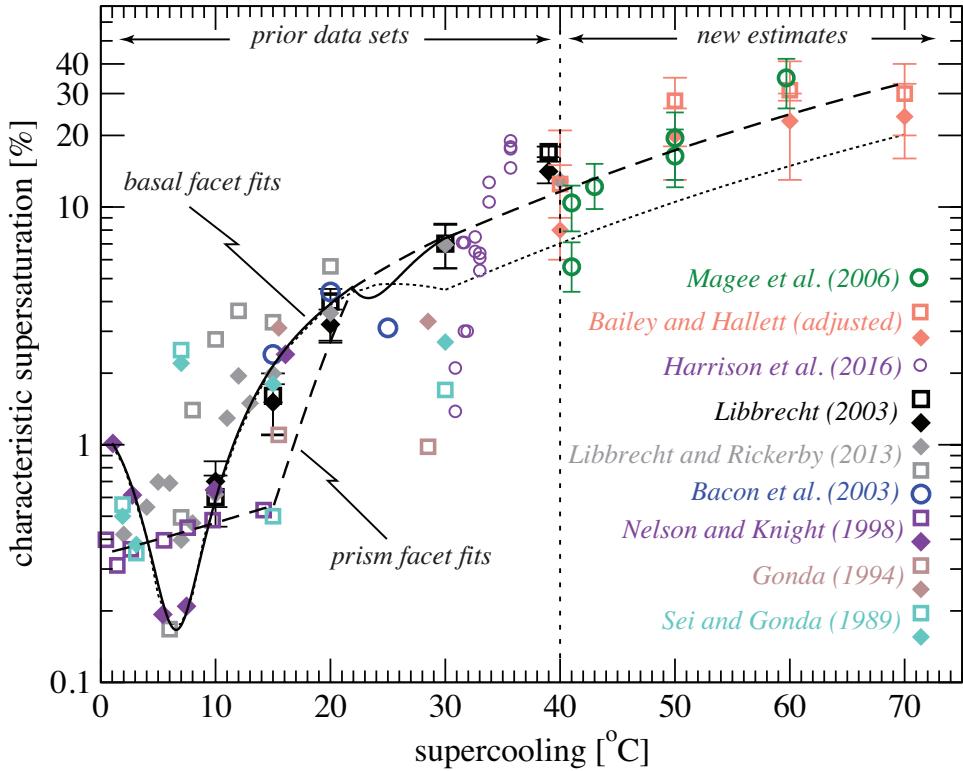


FIG. 1. Characteristic supersaturations as a function of supercooling ($\Delta T = T_{\circ} - T$) from laboratory-measured vapor growth rates (Sei and Gonda 1989; Gonda et al. 1994; Nelson and Knight 1998; Bacon et al. 2003; Libbrecht 2003b; Libbrecht and Rickerby 2013; Harrison et al. 2016). Prior values of s_{char} exist for temperatures above -40°C only. The data set is extended to temperatures below -40°C using estimates from the growth data of Magee et al. (2006) and Bailey and Hallett (2010). Values of s_{char} for the basal and prism facets are denoted by diamond and square symbols, respectively. A circle is used to denote s_{char} that is representative of the overall particle growth. Symbol color indicates the data source. Best fits to the data using polynomials are indicated by the black line (solid, basal facet; dashed, prism facet). The black dotted curve is an ad hoc modification of the basal s_{char} to produce columnar growth at $T < -20^{\circ}\text{C}$.

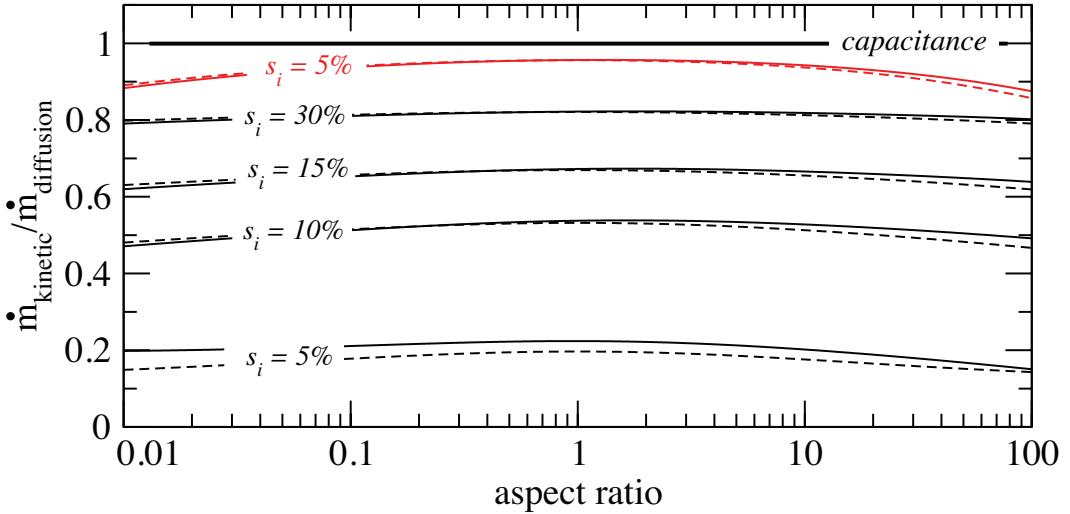


FIG. 2. Ratio of diffusion-kinetics limited growth rate to the maximum (capacitance) growth rate as a function the aspect ratio for a major crystal axis length of 200 μm . A temperature of -35°C and pressure of 500 hPa were used. The solid lines used an s_{char} of 11% and 5.5% for the major and minor crystal axes, respectively. The dashed lines used an average s_{char} and assumed an equivalent volume spherical crystal. Black curves indicate ledge nucleation whereas the red curves indicates dislocation growth at $s_i = 5\%$.

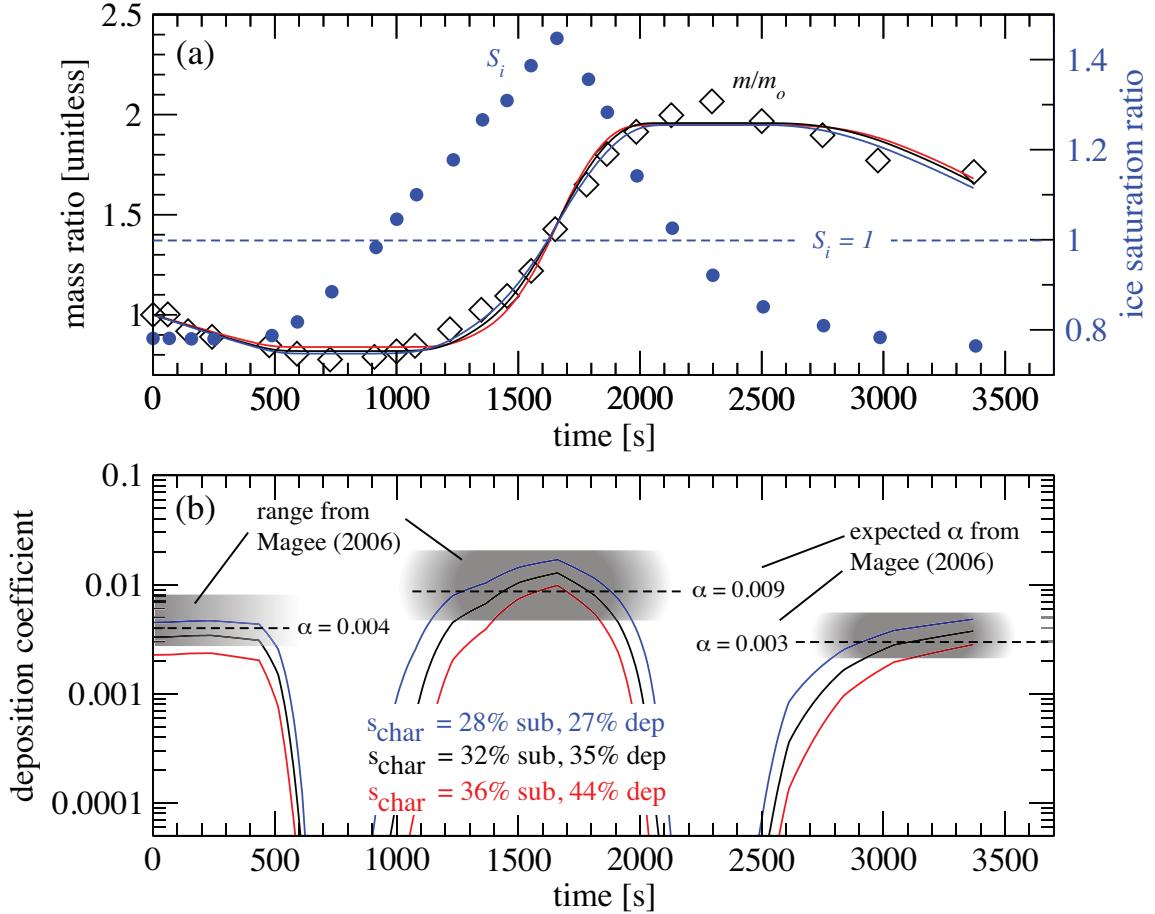


FIG. 3. Evolution of the measured mass ratio and ice saturation ratio (a) and the deposition coefficient (b) at -59.8°C and 972 hPa. The measured mass-ratio (m/m_o where m_o = initial mass) is given by the black diamonds and ice saturation ratio (S_i) is given by the blue circles. Best fits to the mass-ratio time-series using the DiSKICE model are shown by the solid lines. The fits used the most likely initial radius (black line, $r_o = 7 \mu\text{m}$) along with the upper (blue line, $r_o = 8.4 \mu\text{m}$) and lower (red line, $r_o = 5.4 \mu\text{m}$) limits, since uncertainty in the initial radius dominates the error. Evolution of the predicted deposition coefficient uses the three initial radii given in (a). Also shown are the best estimate α (black dashed lines and number) and the range (shaded region) from Magee (2006). Best-fit values of s_{char} for sublimation (sub) and deposition (dep) from each DiSKICE time-series is given on the figure and color-coded to match the simulations.

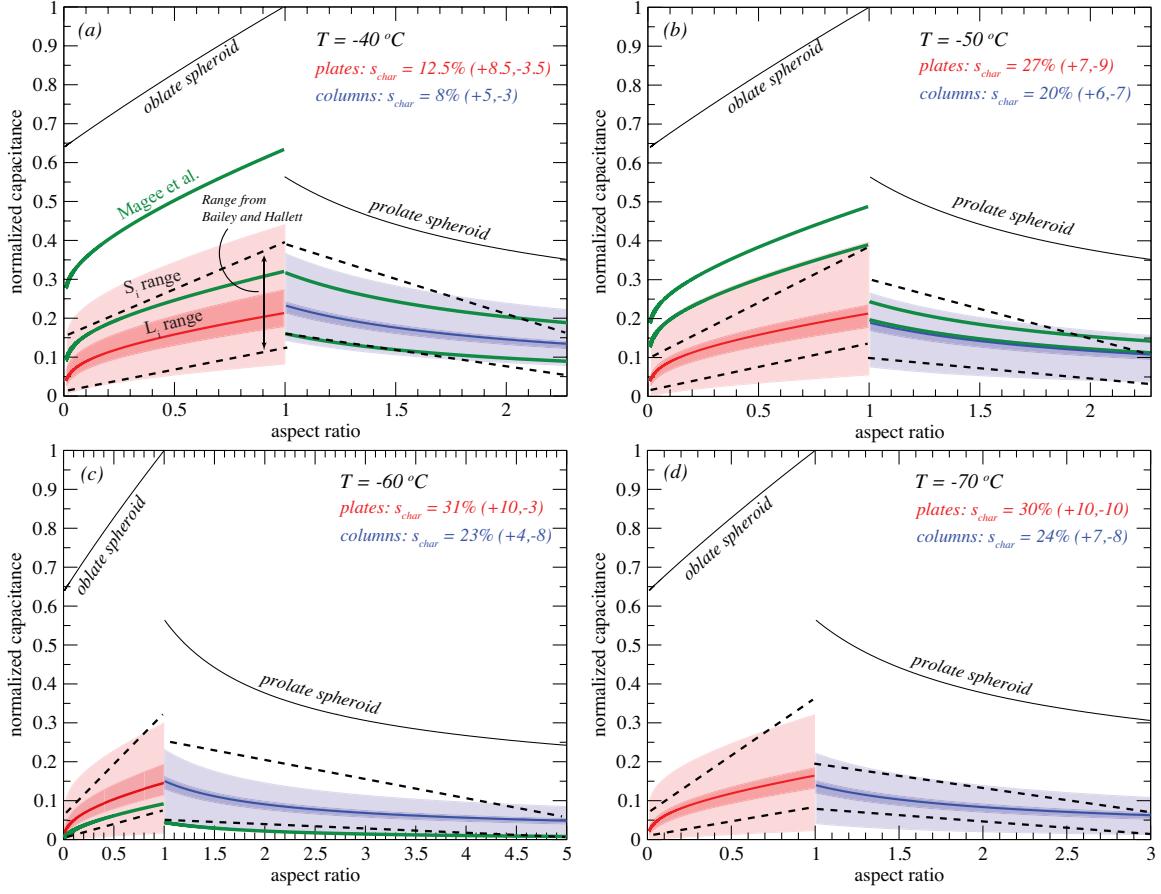


FIG. 4. Normalized effective capacitance (C/a for plates, and $C/(2c)$ for columns) as a function of aspect ratio (ϕ) at temperatures between -40 and -70°C. The solid black lines are capacitance model results for oblate and prolate spheroids. Black dashed lines indicate the approximate range of measured values given in Bailey and Hallett (2010). Green lines used s_{char} values (two at -40 and -50°C) from the data of Magee et al. (2006). Calculations used mid-range values of crystal maximum length (L_i) of 200 μm , pressure (p) of 300 hPa, and ice supersaturations (s_i) of 7%, 10%, 15% and 15% for temperatures of -40, -50, -60, -70°C, respectively. Adjusted solutions using these mid-range values and the DiSKICE model for plates (columns) are indicated by the thick red (blue) lines. Dark shaded regions indicate the variability due to the range of L_i (50 to 400 μm) and p (500 to 150 hPa). Light shaded regions indicate the variability due to the range of s_i (generally 5 to 25%). Variability ranges are similar for the green curves (not shown for clarity). Adjusted values of s_{char} and the variability range (located in the upper right of each panel) are colored to match the lines.

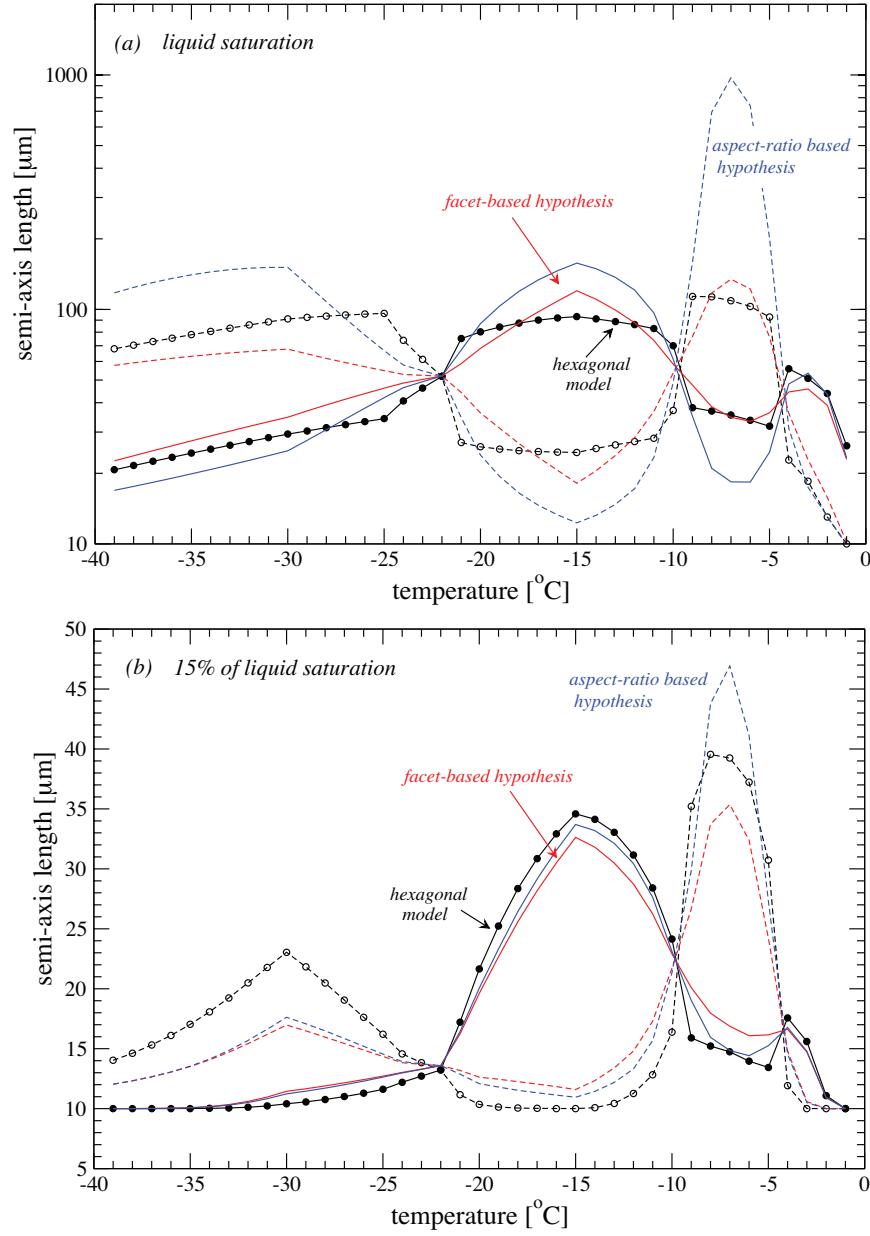


FIG. 5. Comparison of simulated semi-axis lengths from the DiSKICE and hexagonal models, assuming ledge nucleation, after 10 minutes of growth at (a) high (liquid) saturation, and (b) low saturation (15% of the ice saturation ratio at liquid saturation). The a-axis length is given by the solid lines and the c-axis by the dashed lines. Black lines with circles indicate the hexagonal model solutions, red and blue lines indicate DiSKICE solutions with the facet-based and aspect-ratio based hypotheses, respectively.

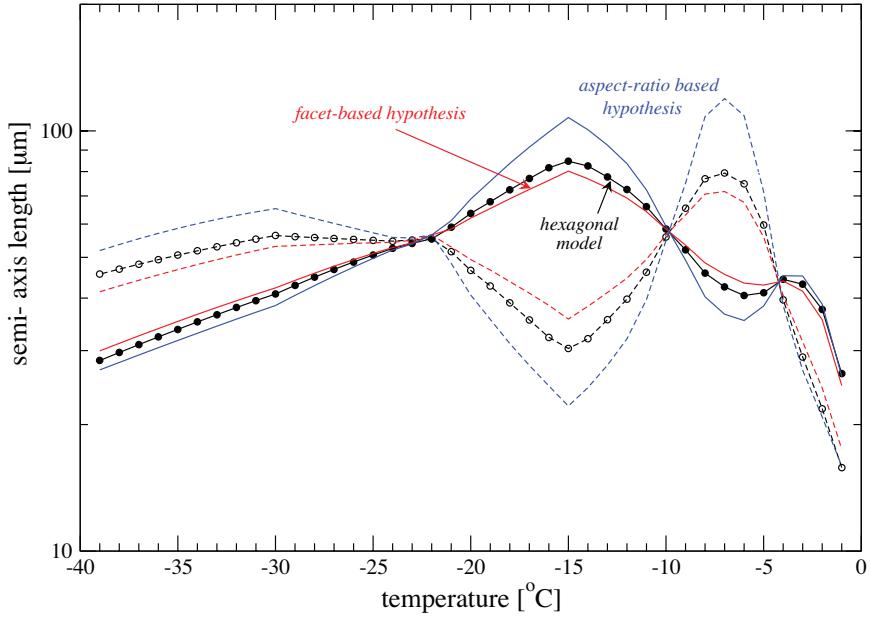


FIG. 6. Comparison of simulated semi-axis lengths from the DiSKICE and hexagonal models, assuming dislocations, after 10 minutes of growth at liquid saturation. The a-axis is given by the solid lines and the c-axis by the dashed lines. Black lines with circles indicate the hexagonal model solutions, red and blue lines indicate DiSKICE solutions with the facet-based and aspect-ratio based hypotheses, respectively.

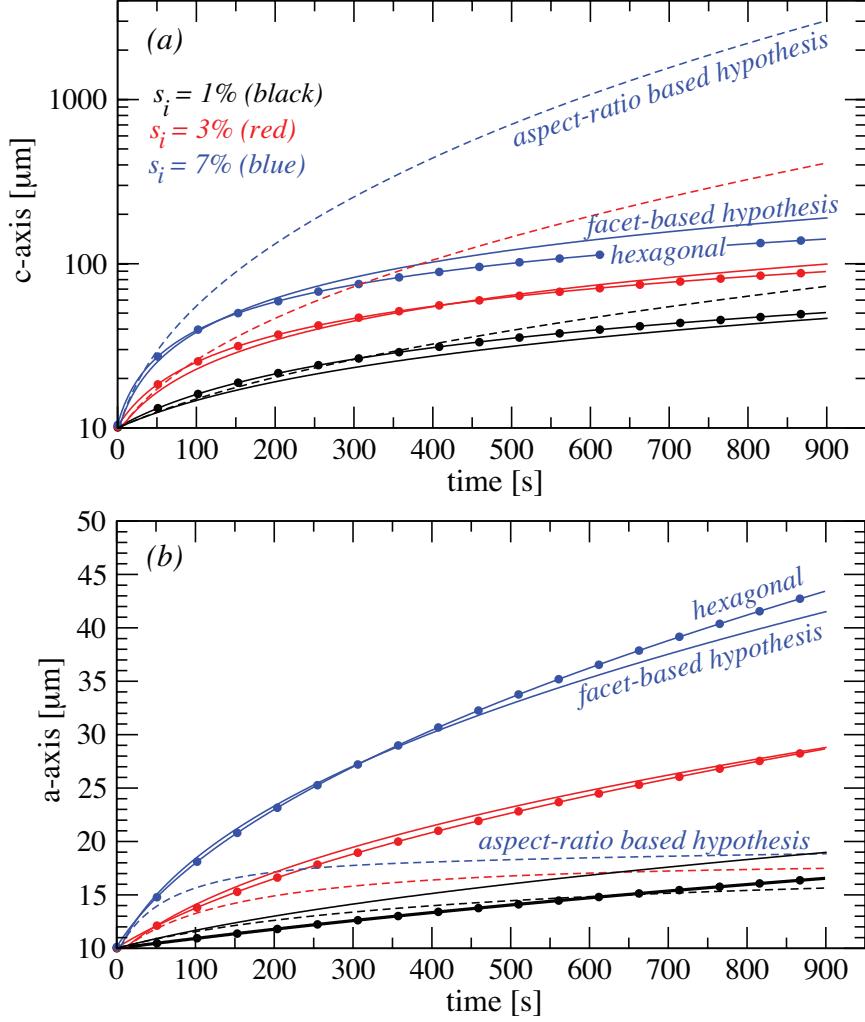


FIG. 7. Time-series of (a) c-axis and (b) a-axis lengths for the simulations shown in Fig. 5; three different ice supersaturations (colored commensurately with lines) are shown at $T = -7^\circ\text{C}$. Lines with circles indicate hexagonal model solutions, solid and dashed lines indicate DiSKICE solutions with the facet-based and aspect-ratio based hypotheses, respectively.

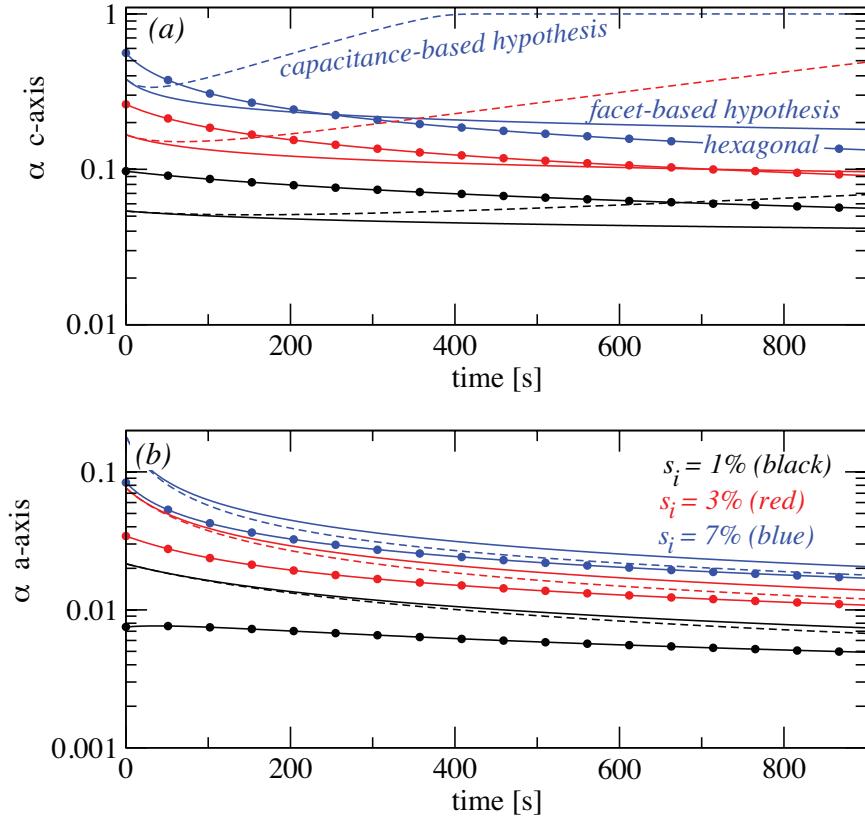


FIG. 8. Time-series of the (a) c-axis and (b) a-axis deposition coefficients for the simulations shown in Fig. 5 at $T = -7^\circ\text{C}$. Lines with circles indicate hexagonal model solutions, solid and dashed lines indicate DiSKICE solutions with the facet-based and aspect-ratio based hypotheses, respectively.

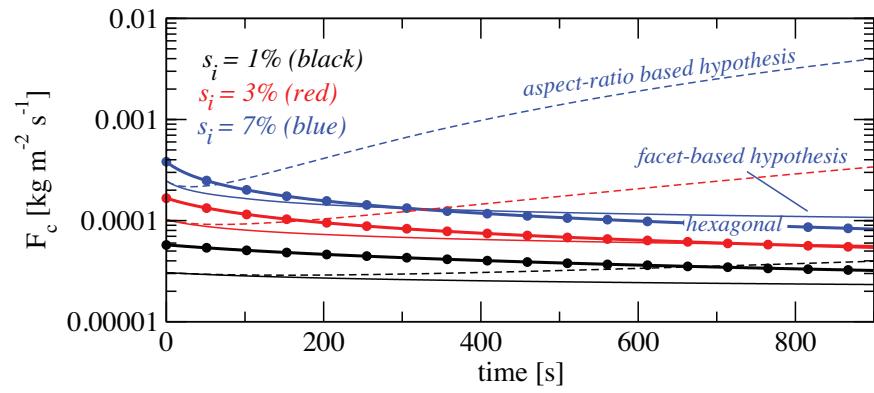


FIG. 9. Time-series of the vapor flux onto the c-axis for the simulations shown in Fig. 5 at $T = -7^\circ\text{C}$. Lines with circles indicate hexagonal model solutions, solid and dashed lines indicate DiSKICE solutions with the facet-based and aspect-ratio based hypotheses, respectively.

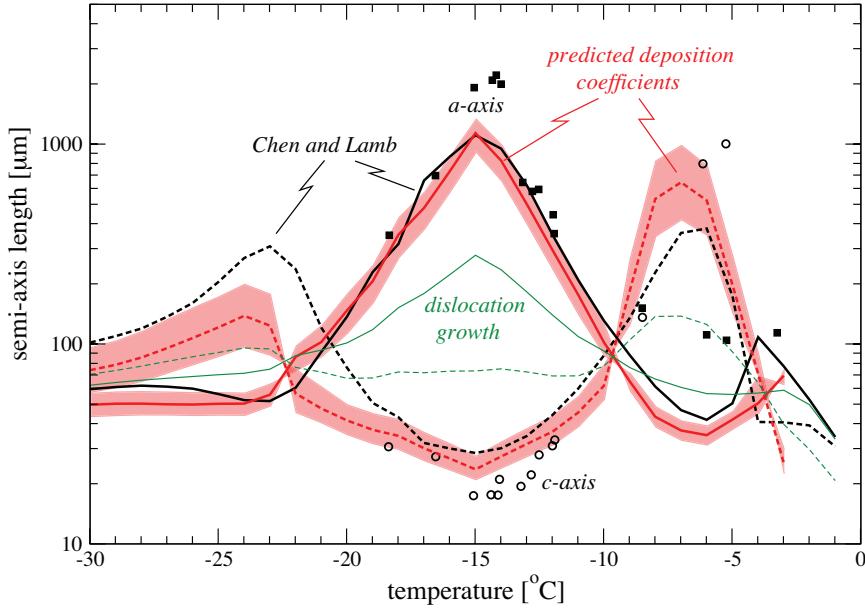


FIG. 10. Axis length after 15 minutes of growth at liquid saturation and 1000 hPa pressure as derived from wind tunnel data of Fukuta and Takahashi (1999) (a-axis, solid circles; c-axis, open circles) and from model simulations (a-axis, solid lines; c-axis, dashed lines). Simulations using the parameterization of Chen and Lamb (1994) are given by the black lines whereas simulations using predicted deposition coefficients (ledge nucleation, facet-based hypothesis) are given by the red lines. The red shaded region indicates the range of uncertainty in the characteristic supersaturation (s_{char}). Simulations with dislocation growth are indicated by the green lines.

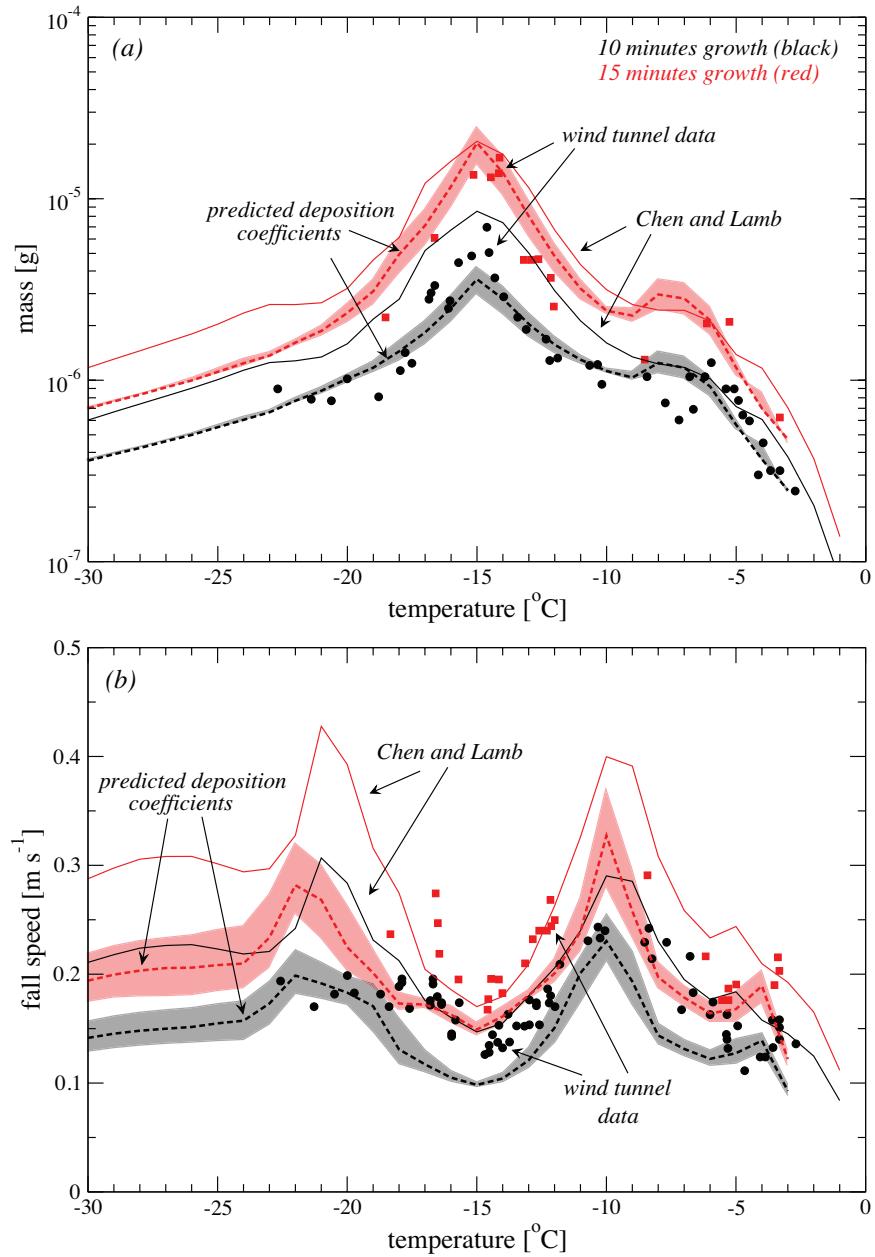


FIG. 11. Ice mass (a) and fall speed (b) after 10 (black) and 15 (red) minutes of growth at liquid saturation and 1000 hPa pressure. Wind tunnel data (Fukuta and Takahashi 1999) are indicated by the symbols and model simulations by the lines. Simulations using the Chen and Lamb (1994) parameterization are given by the solid lines whereas simulations using predicted deposition coefficients (ledge nucleation, facet-based hypothesis) are shown by the dashed lines. The shaded regions indicate the range of uncertainty in the characteristic supersaturation (s_{char}).

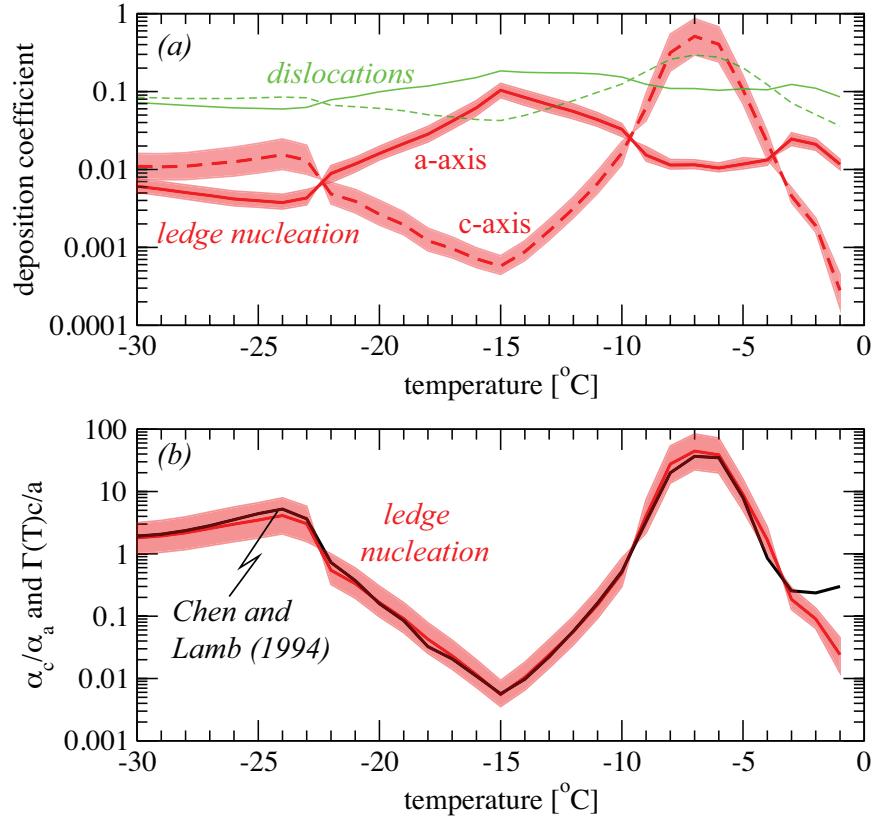


FIG. 12. (a) Deposition coefficients (a-axis, solid lines; c-axis, dashed lines) after 15 minutes of growth at liquid saturation and 1000 hPa pressure for the simulations shown in Fig. 10. Simulations using ledge nucleation are given by the red lines and the red shaded region indicates the range of uncertainty in the characteristic supersaturation (s_{char}). Simulations with dislocation growth are indicated by the green lines. (b) Ratio of the deposition coefficients (α_c/α_a) for ledge nucleation growth (red line and shaded region) using the results and uncertainty from (a). The combination of the inherent growth ratio and the aspect ratio ($\Gamma(T)c/a$) from the theory of Chen and Lamb (1994) is given by the black line.

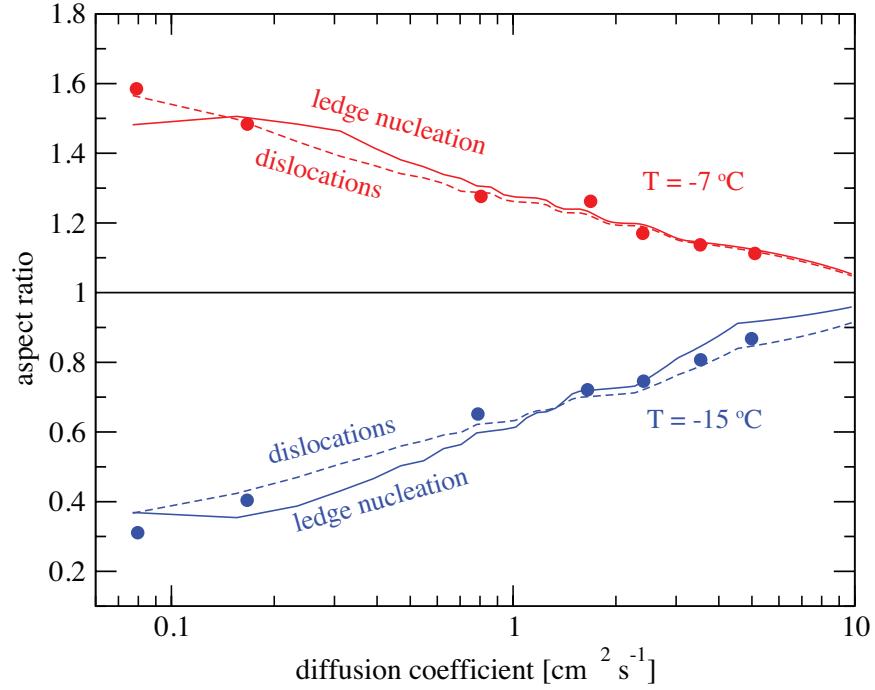


FIG. 13. Aspect ratio of crystals as a function of the vapor diffusivity, D_v , at -7 and -15°C (red and blue colors, respectively). Solid circles indicate the free-fall chamber measurements of Gonda (1980) made after crystal vapor growth at liquid saturation. Simulated crystals began as spheres with an initial radius of $2 \mu\text{m}$ and grew during free-fall over a distance of 10 cm . Dislocation growth (dashed line) used a and c -axis s_{char} , respectively, of 0.48% and 0.18% at -7°C , and 0.04% and 1.9% at -15°C . Ledge nucleation growth (solid line) used a and c -axis s_{char} , respectively, of 0.48% and 0.35% at -7°C , and 0.59% and 1.5% at -15°C .

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887	1	Polynomial fits to s_{char} (used in Eq. 1) as a function of temperature where $\Delta T =$	
888		$T - T_0$ and $T_0 = 273.15$ K for the c-axis (basal facet) and a-axis (prism facet). . .	60

TABLE 1. Polynomial fits to s_{char} (used in Eq. 1) as a function of temperature where $\Delta T = T - T_{\circ}$ and $T_{\circ} = 273.15$ K for the c-axis (basal facet) and a-axis (prism facet).

$c\text{-axis fit coefficients: } s_{char} = \sum_{n=0}^6 a_n \Delta T^n$							
temperature	a_0	a_1	a_2	a_3	a_4	a_5	a_6
$-22 < T \leq -1^{\circ}\text{C}$	1.1217	0.0381	-0.08375	-0.01573	-0.001011	-2.915×10^{-5}	-3.182×10^{-7}
$-30 \leq T \leq -22^{\circ}\text{C}$	753.63	105.97	5.553	0.1281	0.0011	0	0
$T < -30^{\circ}\text{C}$	3.7955	0.10614	0.00753	0	0	0	0
$a\text{-axis fit coefficients: } s_{char} = \sum_{n=0}^6 a_n \Delta T^n$							
temperature	a_0	a_1	a_2	a_3	a_4	a_5	a_6
$-15 < T \leq -1^{\circ}\text{C}$	0.3457	-0.0093	0.000308	0	0	0	0
$-22 < T \leq -15^{\circ}\text{C}$	-5.2367	-1.3184	-0.1107	-0.00323	0	0	0
$-30 \leq T \leq -22^{\circ}\text{C}$	-0.7106	-0.1478	0.00423	0	0	0	0
$T < -30^{\circ}\text{C}$	3.7955	0.10614	0.00753	0	0	0	0