

Dear Editor,

We received your comments and suggestions for our discussion paper “Can Semi-Volatile Organic Aerosols Lead to Less Cloud Particles?” on ACPD. We thank you very much for your efforts in evaluating our submission.

Of the final referee’s comment, we have modified this line in the Conclusions section, “*We would like to emphasize that our results do not imply that the Earth has less CCN than currently thought. Instead, they imply that if in a model semi-volatile organics will be simulated together with aerosol microphysics, a general decrease is to be expected.*”, which now reads, “*We would like to emphasize that our results do not imply that the Earth has less CCN than currently thought. Instead, they imply that if in a model semi-volatile organics will be simulated together with aerosol microphysics, a general decrease is to be expected, **assuming our model captures all relevant contributory processes.***” This change is highlighted in the updated revised manuscript.

We hereby resubmit the revised discussion paper to be considered for publication in *Atmospheric Chemistry and Physics*. We confirm that all authors listed on the manuscript concur with submission in its revised form. Should you have any remaining questions, we will be happy to address them.

Sincerely,

C. Y. Gao, S. E. Bauer and K. Tsigaridis

Can Semi-Volatile Organic Aerosols Lead to Less Cloud Particles?

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Abstract. The impact of condensing organic aerosols on activated cloud number concentration is examined in a new aerosol microphysics box model, MATRIX-VBS. The model includes the volatility-basis set (VBS) framework coupled with the aerosol microphysical scheme MATRIX (Multiconfiguration Aerosol TRacker of mIXing state) that resolves aerosol mass and number concentrations and aerosol mixing state. By including the condensation of organic aerosols, the new model produces less activated particles compared to the original model, which treats organic aerosols as non-volatile. Parameters such as aerosol chemical composition, mass and number concentrations, and particle sizes which affect activated cloud number concentration are thoroughly tested via a suite of Monte-Carlo simulations. Results show that by considering semi-volatile organics in MATRIX-VBS, there is lower activated particle number concentration, except in cases with low cloud updrafts, in clean environment at above freezing temperatures, and in polluted environments at high temperature (310K) and extremely low humidity conditions.

1 Introduction

Atmospheric aerosols influence climate mainly via two pathways: aerosol-radiation interactions (the aerosol direct effect; Charlson et al., 1992) which affect the Earth's radiative energy balance by absorbing and scattering terrestrial and solar radiation, and aerosol-cloud interactions (the aerosol indirect effect; Twomey, 1974; Albrecht, 1989) which affect cloud microphysics by activating and serving as seeds for cloud formation (Myhre et al., 2013; Seinfeld and Pandis, 2016). Aerosol activation as cloud condensation nuclei (CCN) is critical to the evolution and microphysics of clouds (Reutter et al., 2009). However, the relationship between aerosol mixing state and cloud microphysical properties remain a large uncertainty in aerosol-

31 cloud interactions (Ghan et al., 1998; McFiggans et al., 2006; Ervens et al., 2007; Gibson et al.,
32 2007; Medina et al., 2007; Cubison et al., 2008; Anttila, 2010).

33 Climate models calculate cloud droplet number concentration (CDNC) using aerosol
34 activation schemes, whose main governing parameters include aerosol number, size,
35 hygroscopicity, updraft velocity, as well as critical supersaturation. Physically-based aerosol
36 activation schemes (e.g. Abdul-Razzak and Ghan, 2000; Fountoukis and Nenes, 2005; Ming et al.,
37 2006; Shipway and Abel, 2010) are commonly used in global climate models for fast diagnostics
38 of nucleation and to estimate the aerosol indirect effect in long-term climate simulations (Ghan,
39 2011). Several studies examined the relationship between the fore-mentioned parameters and how
40 they play together to activate particles. Ghan et al. (1998) examined sea salt's influence on sulfate
41 particle activation and introduced the competition effect. Since all CCN have to compete for
42 available water vapor in order to activate, the competition limits the maximum supersaturation in
43 in-cloud updrafts (Storelvmo et al., 2006). Ghan et al. (1998) concluded that activated number
44 concentration increases with increasing sea salt when sulfate is low and updraft is strong, and it
45 decreases when sulfate is high and updraft is weak, because maximum supersaturation is reduced.
46 Another study (Reutter et al. 2009) explored how much CDNC depend on updraft velocity, size
47 distribution and hygroscopicity. They found that size distribution played a greater role than particle
48 hygroscopicity on CDNC and discovered different CCN activation and cloud droplet formation
49 regimes, which are determined by aerosol number concentration and updraft velocity.

50 Semi-volatile organic aerosols contribute significantly to the growth of particles to CCN
51 sizes (Yu, 2011). More notably, as aerosol size increases, the range of organic volatilities involved
52 in aerosol growth increases (Pierce et al., 2011; Yu, 2011). The inclusion of semi-volatile organics
53 in models modifies CCN formation rates (Petters et al., 2006, Riipinen et al., 2011; Scott et al.,
54 2015) as well as hygroscopicity (Petters and Kreidenweis, 2007), in addition to bulk aerosol mass,
55 size distribution and composition. By adding semi-volatile organic partitioning to our existing
56 microphysics model MATRIX (Multiconfiguration Aerosol TRacker of mIXing state; Bauer et al.,
57 2008), which resolves aerosol mixing state, we were able to examine how they change bulk aerosol
58 mass, size distribution and composition. However, the effects of semi-volatile organic partitioning
59 combined with aerosol mixing state on particle activation remain unexplored.

60 In our previous work, we demonstrated that including semi-volatile organics would lead to
61 higher aerosol number concentration and smaller particles (Gao et al., 2017). As was the case for

62 the original aerosol microphysics model MATRIX, our further-developed box model MATRIX-
63 VBS (Gao et al., 2017) follows the same multi-modal aerosol activation approach by Abdul-
64 Razzak and Ghan (2000). The activation parameterization accounts for aerosol size distribution,
65 composition, mixing state, and in-cloud updraft velocity. Curious about the change in activation
66 with the newly-present semi-volatile organics and the governing parameters influencing it, we
67 investigated the difference in activated number concentration in two box model set ups: MATRIX
68 (Bauer et al., 2008) and MATRIX-VBS (Gao et al., 2017).

69 **2 Methods**

70 2.1 Model Description

71 MATRIX-VBS (Gao et al., 2017) is an aerosol microphysics model that includes organic
72 aerosol volatility in its calculations. It was developed by implementing VBS (volatility-basis set;
73 Donahue et al., 2006) in the aerosol microphysics model MATRIX (Bauer et al., 2008), which is
74 a box model that is also used in the NASA GISS ModelE Earth System Model (Bauer et al., 2008,
75 2012; Schmidt et al., 2014). Since the publication of Gao et al., 2017, which included organic
76 condensation on fine mode aerosols, we further developed the model which now allows semi-
77 volatile organics in the system to condense on coarse mode dust and sea salt as well. We have also
78 included nitrate radicals as an oxidant for organics in addition to the hydroxyl radical that was used
79 in the original VBS scheme, even though it is a very minor oxidation pathway in the model (rate
80 constant for the oxidation by NO_3^* is $1 \times 10^{-13} \text{ cm}^3 \text{ molecules}^{-1} \text{ s}^{-1}$; Atkinson, 1997). As previously
81 stated, we use Abdul-Razzak and Ghan (2000) activation parameterization, which calculates the
82 activated particle number concentration depending on chemically-resolved number concentrations
83 using Köhler Theory. The hygroscopicity parameters κ for each aerosol species presented in Table
84 1 were calculated from their solubility fraction. For organics, we assumed a linear increase of
85 solubility with decreasing volatility (Jimenez et al., 2009). Since we use Pankow type partitioning
86 (Pankow, 1994), water is not considered in the partitioning process. In addition, we do not use
87 different κ /RH relationships per organic species, which was found to be important for
88 biogenic SOA (Rastak et al., 2017).

89

90 2.2 Simulations

91 A Monte-Carlo analysis with a range of chemical and meteorological conditions (Table 2)
92 was performed, to pinpoint which processes affect organics and the mixed aerosol population in
93 general the most. Since global models need to resolve a wide range of conditions, from very clean
94 to very polluted and for a wealth of meteorological conditions, we simulated 630 possible
95 atmospheric scenarios on Earth across the whole parameter space, e.g. temperature, relative
96 humidity, latitude, emissions levels and updraft velocity, for 120 hours (5 days) simulations with
97 no deposition and dilution. Three types of environmental conditions were simulated: clean,
98 moderate and polluted, as defined by different levels of emissions which were determined using a
99 probability distribution of the gridded emission fields in GISS ModelE for January present-day
100 conditions. During this development phase, biogenic secondary organic aerosols from terpenes
101 oxidation in MATRIX-VBS are treated as nonvolatile, while only the anthropogenic aerosols are
102 treated as semi-volatile.

103 **3 Results and discussion**

104 We found that activated number concentration is lower for most cases in the MATRIX-
105 VBS model, which considers semi-volatile organic aerosols, as compared to the MATRIX model.
106 However, under low updrafts, in clean environment at above freezing temperatures, and in polluted
107 environments at high temperature (310K) and extremely low humidity conditions (0% RH) during
108 aerosol formation, activated number concentration is higher in MATRIX-VBS than in MATRIX.

109 As an example, the activated number concentration for a case with temperature at 290°K,
110 relative humidity at 40%, medium emission levels and an updraft of 0.5 m/s at 30°N latitude is
111 shown in Figure 1 for the two models. Mixing states of aerosols in MATRIX and MATRIX-VBS
112 are represented as aerosol populations, which all contain SO₄, NO₃, NH₄ and H₂O, in addition to
113 the species that define the populations (Bauer et al., 2008, 2013). The four most dominant aerosol
114 populations for the activated number concentration in MATRIX are ACC (SO₄, NO₃, NH₄), OCS
115 (organics, SO₄, NO₃, NH₄), BOC (black carbon, organic carbon, SO₄, NO₃, NH₄) and BCS (black
116 carbon, SO₄, NO₃, NH₄). Only two dominant populations are calculated in MATRIX-VBS, OCS
117 and BOC, as in Gao et al., 2017, since OCC evaporates and re-condenses on all particles, based on
118 their calculated surface area and mass concentration. Since OCS and BOC have the largest surface
119 area, they are calculated to have the strongest growth via organics condensation. Additionally, the

120 competition between sulfate, organics and black carbon, determines the loss of ACC and the
121 formation of BCS: OCC coagulates with ACC to form OCS, and this coagulation increases in
122 MATRIX-VBS due to smaller OCC particles; therefore, there are less ACC particles left to
123 coagulate with black carbon to form BCS. At the end of the 5-day simulation (Figure 1), MATRIX-
124 VBS has approximately a total of 30 activated particles/cm³, whereas MATRIX has approximately
125 60 activated particles/cm³ under the same conditions.

126 Figure 2 shows a more comprehensive look across all temperature and relative humidity
127 scenarios studied. The results show that for most scenarios, MATRIX-VBS has lower (red circles)
128 activated number concentration compared to MATRIX. However, some rare cases show the
129 opposite behavior. These are for above freezing temperatures in the low emission level under low
130 updraft (top left) scenarios, high temperature (310K) and extremely low humidity (0% RH) in the
131 medium emission level under low updraft (middle left) scenarios, as well as the high emission
132 level under low (bottom left) and medium (bottom middle) updraft scenarios. Note that low RH
133 values do not mean that these correspond to cloud conditions. Aerosols form outside of clouds in
134 our model, where RH can be very low. Activation though will occur after aerosol formation, when
135 an air parcel starts rising with a given updraft velocity, in which air parcel supersaturation will
136 develop and will cause aerosol activation.

137 Across all scenarios, the changes in activated number concentration between MATRIX-
138 VBS and MATRIX range from a -56% to +31% (Table 3). The range of the difference becomes
139 more significant as emission levels increase, yet less significant as updraft velocity increases.
140 Within most emission level-updraft velocity scenarios, as temperature increases, the fractional
141 change in activated number concentration between the two models decreases. Also within most
142 emission level-updraft velocity scenarios (Figure 3, Table 4), as temperature increases, there are
143 less activated particles in MATRIX. We also observed the same behavior in MATRIX-VBS,
144 higher temperature, less activated particles.

145 In order to understand the cause of the difference in activation, we traced back to the key
146 difference between the two models: partitioning of organics. The inclusion of organics partitioning
147 leads to changes in aerosol mixing state and size distribution, as discussed in Gao et al. (2017).
148 Therefore, the change in activated number concentration could only be caused by changes in mass
149 concentration, number concentration and particle size. Since we use the Abdul-Razzak and Ghan
150 (2000) parameterization, the activated number concentration is mainly a function of number

151 concentration and dry particle diameter in our model. The parameterization is also a function of
152 geometric standard deviation, which is constant per population in our model as it did in MATRIX
153 (Bauer et al., 2008), as well as a function of aerosol composition and hygroscopicity, as mentioned
154 in the model description, for which we assume a linear increase of solubility with decreasing
155 volatility. The hygroscopicity of the aerosol populations changes with time, as the internal mixing
156 of aerosol populations is altered by aerosol microphysics.

157 As was the case in Gao et al., (2017), MATRIX-VBS has higher aerosol number
158 concentration (Figure 4 left) but smaller particles (Figure 4 right) compared to MATRIX in the
159 case presented in Figure 1. At first we expected that smaller particles would less likely activate, so
160 we performed a simple sensitivity test to confirm it. By changing dry particle diameter of the
161 particles in the activation scheme, the decreasing dry particle diameter indeed led to lower
162 activated number concentration. However, a second sensitivity test with changing only number
163 concentration showed that higher number concentration would actually lead to lower activated
164 number concentration as well.

165 In the Abdul-Razzak and Ghan (2000) scheme, increasing number concentration decreases
166 critical supersaturation, and lower critical supersaturation leads to higher minimum dry particle
167 radius that is able to activate. Therefore, activation is suppressed, since less particles exceed the
168 threshold radius. The activated number concentration is calculated from the activation fraction and
169 the number concentration. When the fraction is greater than the increase in number concentration,
170 lower activated number concentration is achieved, as shown here.

171 As mentioned previously, within most of the scenarios, there is a decrease in fractional
172 change as temperature increases, while both models experience decrease in activated number
173 concentration with increased temperature. This means the decrease in activated number
174 concentration for MATRIX-VBS is not as significant as that for MATRIX. There are two factors
175 that contribute to such change. First, the heat and moisture diffusion term is dependent on
176 temperature in the activation scheme (Abdul-Razzak and Ghan, 2000). Second, volatility of
177 organics is temperature dependent. In MATRIX-VBS, when organic volatility is considered, the
178 change is dampened. In other words, its number of activated particles is less sensitive to
179 temperature change as compared to MATRIX, leading to what we see in the circle plots that the
180 greater change at lower temperatures.

181 The length of day and season changes the duration and intensity of gas phase oxidation of
182 semi-volatile gases, which is why we also looked at aerosol evolution driven by photochemistry
183 at different latitudes. Since the model uses January emissions, different seasons are simulated at
184 the different hemispheres, while different day lengths are simulated at higher latitudes of the
185 southern hemisphere compared to tropical and high latitude northern hemisphere ones. As we
186 inspected results across latitudes in the two hemispheres, we found varying activated number
187 concentration in MATRIX-VBS compared to MATRIX and observed no evident trend. Such
188 inconclusive and complex results may be due to gas-phase chemistry and photochemical ageing
189 of semi-volatile organic vapors, which would require further examination in a separate dedicated
190 study.

191 **4 Conclusions**

192 With the inclusion of organic partitioning in an aerosol microphysics model, activated
193 aerosol number concentration is decreased under most temperature and relative humidity
194 conditions, except when under low updrafts, in clean environments at most temperatures and
195 relative humidities, and in polluted environments at high temperatures and extremely low humidity
196 conditions. Such changes are due to increased aerosol number concentration and smaller particles
197 in the new model, as well as how number concentration and size are calculated in the chosen
198 aerosol activation scheme, which determines how many particles are activated. Additionally, the
199 temperature dependence of activated number concentration is decreased for most scenarios.

200 Our conclusion that fewer particles are activated at higher updrafts is in contrast to
201 Connolly et al. (2014), who found that fewer particles activated at low updrafts, using a different
202 geometric standard deviation in the same parameterization of aerosol activation as the one we use.
203 Such a difference can be due to the fact that the Abdul-Razzak and Ghan (2000) activation
204 parameterization produces a different response when multiple modes are used, as shown by
205 Connolly et al. (2014b) and Simpson et al. (2014). Additionally, in our study, the geometric
206 standard deviation remained constant per aerosol population. However, it is worth exploring in the
207 future to use reduced geometric standard deviation in our calculations to directly compare with
208 values used by Connolly et al. (2014) and Crooks et al. (2017). In fact, in a comparison study,
209 Ghan et al. (2011) found that the Abdul-Razzak and Ghan (2000) scheme tend to have lower

210 activation fractions and droplet concentrations compared to the Fountoukis and Nenes (2005)
211 activation scheme.

212 Topping et al. (2013) showed that co-condensing organics lead to enhanced cloud droplet
213 number concentration, which seems to contradict our results. However, it is important to note that
214 contrary to Topping et al. (2013), our study is performed in a box model that does not resolve
215 cloud droplet growth as the air mass rises and cools, which leads to additional condensation of
216 organic vapors and water due to the temperature decline, and contributes to cloud droplet growth
217 due to additional water uptake. The simulations in this study, however comprehensive, are still
218 highly idealized.

219 We would like to emphasize that our results do not imply that the Earth has less CCN than
220 currently thought. Instead, they imply that if in a model semi-volatile organics will be simulated
221 together with aerosol microphysics, a general decrease is to be expected, [assuming our model](#)
222 [captures all relevant contributory processes](#). We will investigate the effects of condensing organics
223 in a global climate model in the future. The results presented here implicate that in the new model,
224 most areas on Earth would experience less CCN on a typical day, but clean environments with
225 above freezing temperatures, or polluted environments on an extremely dry and hot day, would
226 form more CCN under low updraft velocity conditions, as compared to the old model. We expect
227 that implementing the improved box model in the global scale that includes a two moment cloud
228 microphysical scheme (Morrison and Gettelman, 2008; Gettelman and Morrison, 2015) would
229 more accurately represent aerosol-cloud interactions, which will be our focus on a follow up study.
230 Thus it would offer us valuable insights on how the addition of process level phenomena in aerosol
231 microphysics, as applied here for the organics partitioning, would affect cloud microphysics in the
232 global atmosphere and its implications for climate.

233
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239 The GISS ModelE Earth system model is publicly available. The box model code used here is
240 available upon request and will be publicly available in the future as part of GISS ModelE. The
241 data from all model simulations will be available upon request.
242 The authors declare that they have no conflict of interest.

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431

432 **Table 1. Hygroscopicity κ used for each organic aerosol volatility bin.**

	$\log_{10}C^*$ [$\mu\text{g m}^{-3}$]	soluble fraction [%]	κ
Sulfate	/	100	0.507
Black carbon	/	0	$5 \cdot 10^{-7}$
Non-volatile organic carbon	/	78	0.141
	-2	100	0.180
	-1	87.5	0.158
	0	75	0.135
Semi-volatile organic carbon	1	62.5	0.113
	2	50	0.090
	3	37.5	0.068
	4	25	0.045
	5	12.5	0.023
	6	0	0.000
Dust	/	13	0.14
Sea salt	/	100	1.335

433

434

435 **Table 2. Parameters used in the Monte-Carlo simulations.**

Parameter		Range	
	T [K]	270, 280, 290, 300, 310	
	RH [%]	0.1, 20, 40, 60, 80, 100	
	Latitude	0, 30N/S, 60N/S, 90N/S	
	Updraft velocity [m/s]	0.5, 1, 2	
Emissions of aerosols [$\mu\text{g}/\text{m}^3/\text{s}$]	Sulfate (SO_2 in molecules/ cm^3)	$10^5, 10^6, 5 \cdot 10^6$	
	Primary organics	$5 \cdot 10^{-6}, 5 \cdot 10^{-5}, 5 \cdot 10^{-4}$	
	Nonvolatile biogenic organics from terpene source	$1 \cdot 10^{-8}, 5 \cdot 10^{-6}, 1 \cdot 10^{-5}$	
	Black Carbon	$10^{-6}, 10^{-5}, 10^{-4}$	
Emissions of gases [molecules/ cm^3]	VOCs (in sets)	Alkenes	$5 \cdot 10^2, 5 \cdot 10^3, 5 \cdot 10^4$
		Paraffin	$5 \cdot 10^3, 10^4, 5 \cdot 10^4$
		Terpenes	$10^4, 10^5, 10^6$
		Isoprene	$10^4, 10^5, 50^6$
		NO_x	$10^5, 10^6, 10^7$

436

437

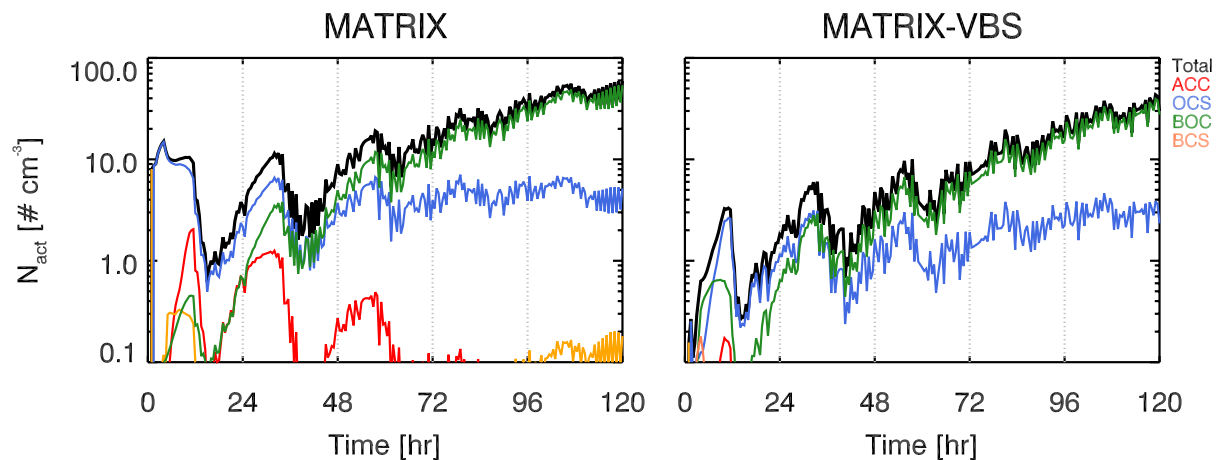
438 **Table 3. Minimum and maximum of fractional change in average activated number**
 439 **concentration over the last 24 hours between the two models with low, medium and high**
 440 **level emissions at updraft velocities of 0.5, 1 and 2 m/s.**

Updraft velocity (m/s)	Fractional change in activated number concentration					
	0.5		1		2	
	min	max	min	max	min	max
Low emission level	-9%	+21%	-16%	+2%	-14%	+5%
Medium emission level	-51%	+14%	-42%	-5%	-36%	-13%
High emission level	-56%	+31%	-48%	+9%	-43%	-9%

441
442

443 **Table 4. Minimum and maximum of average activated number concentration over the last**
 444 **24 hours of MATRIX and MATRIX-VBS with low, medium and high level emissions at**
 445 **updraft velocities of 0.5, 1 and 2 m/s.**

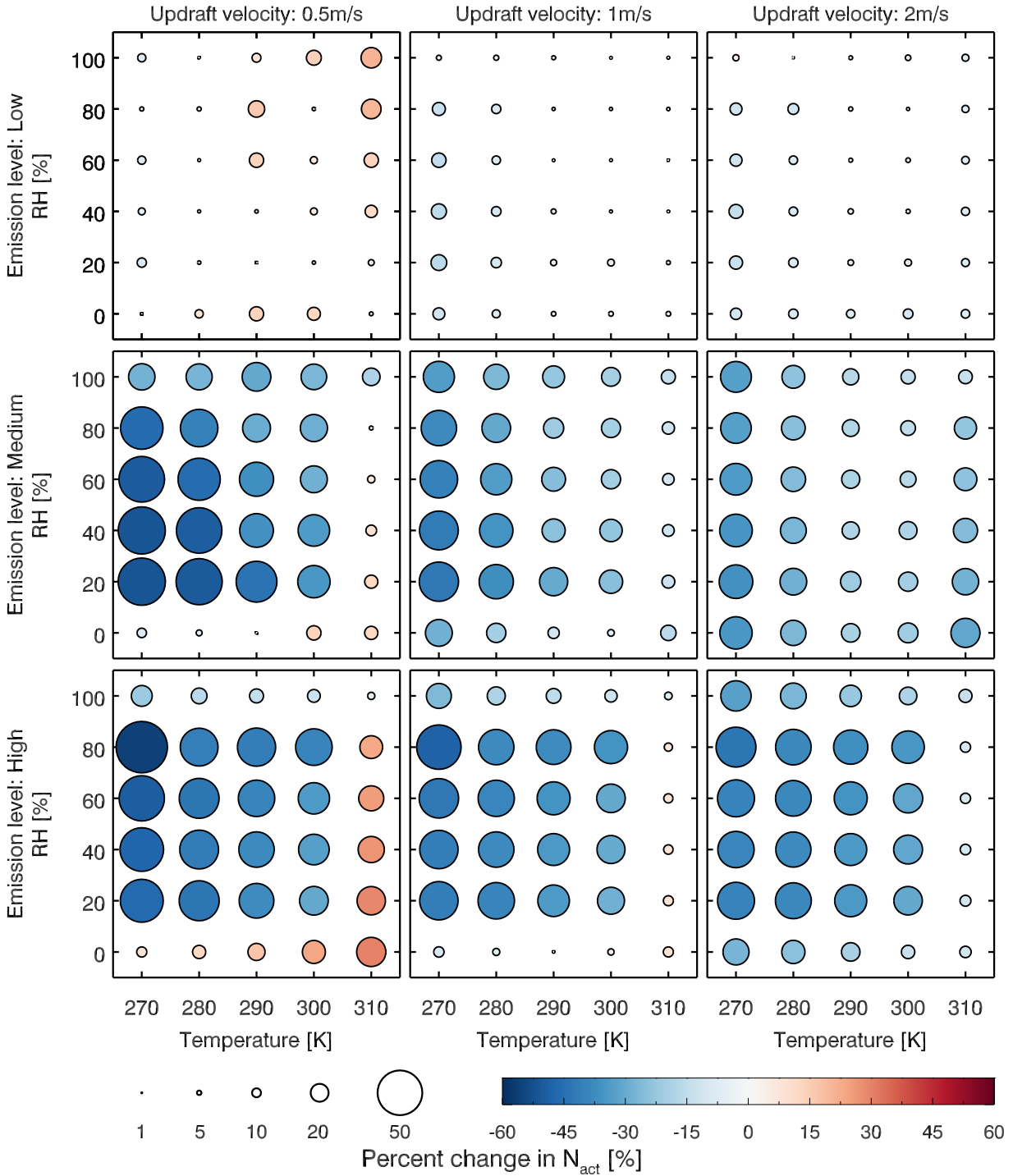
Updraft velocity (m/s)		Activated number concentration					
		0.5		1		2	
		min	max	min	max	min	max
Low emission level	MATRIX	23	305	351	1160	963	2799
	MATRIX-VBS	24	283	338	1026	887	2473
Medium emission level	MATRIX	19	152	359	1233	1476	3711
	MATRIX-VBS	16	139	304	884	1021	2498
High emission level	MATRIX	3	60	199	1280	1925	5703
	MATRIX-VBS	3	63	185	1150	1677	4142



446

447 **Figure 1. Activated number concentration of aerosol populations (see main text for details)**
 448 **for MATRIX (left) and MATRIX-VBS (right) for 290 K and 40% RH at 30°N latitude with**
 449 **medium emission levels and 0.5 m/s updraft velocity.**

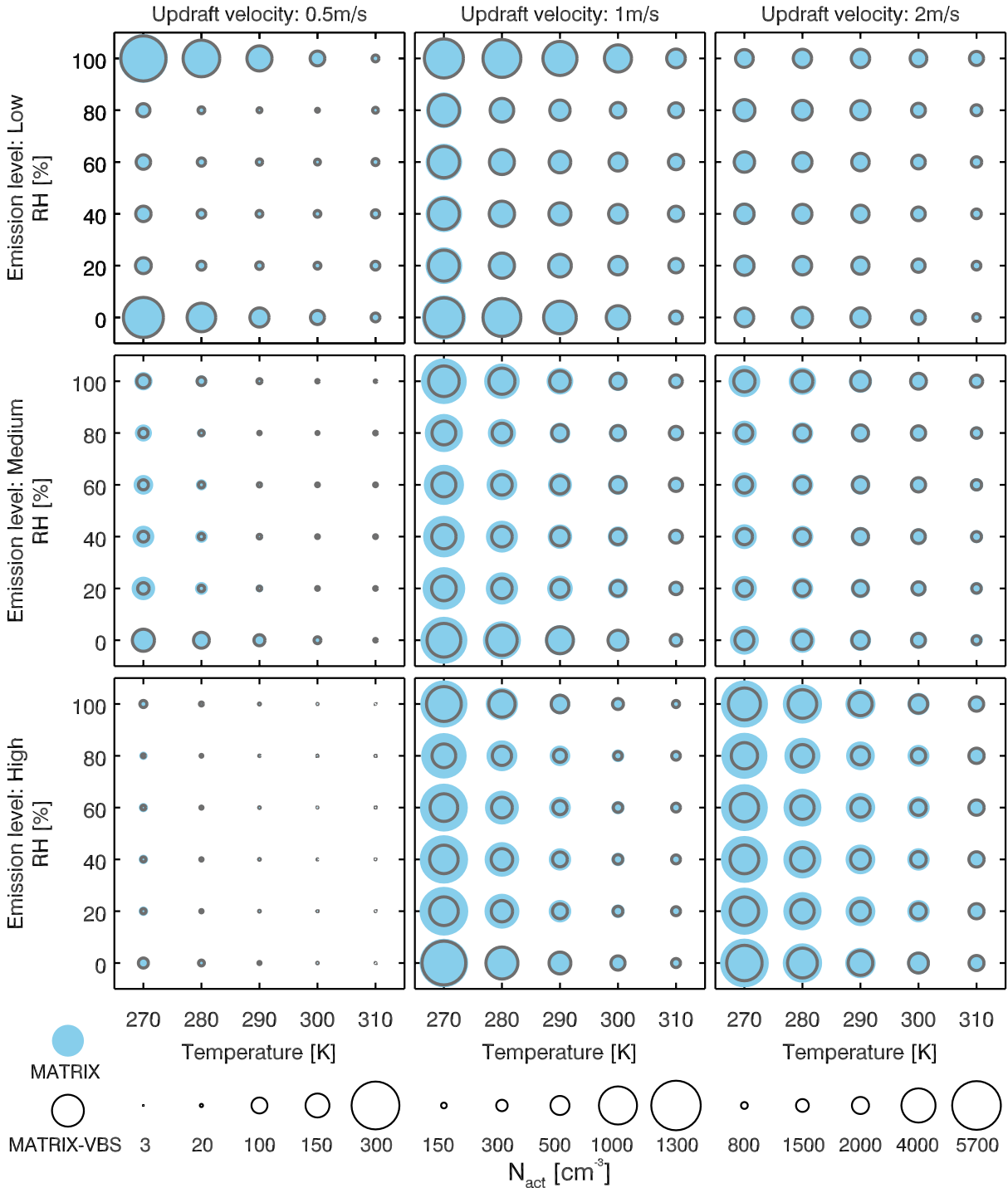
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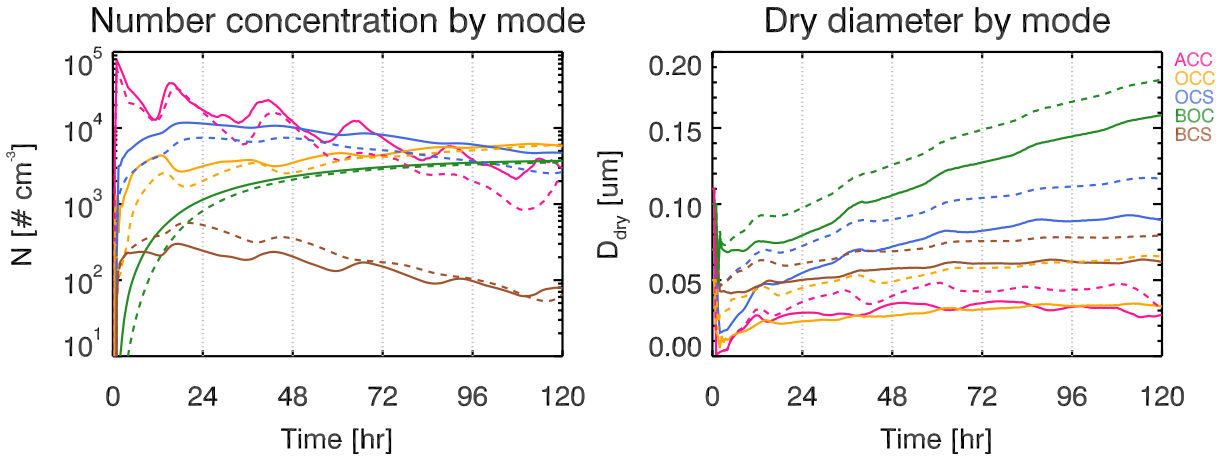
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452 **Figure 2. Fractional change of average activated number concentration (size and color of the**
 453 **circles) over the last 24 hours of a 5-day simulation between the two models with low (top**
 454 **row), medium (middle row) and high (bottom row) level emissions at updraft velocities of 0.5**
 455 **(left column), 1 (middle column) and 2 (right column) m/s.**

456



457
 458 **Figure 3. Average activated number concentration (circle size) during the last 24 hours of a**
 459 **5-day simulation in MATRIX and MATRIX-VBS with low (top row), medium (middle row)**
 460 **and high (bottom row) emission levels at updraft velocities of 0.5 (left column), 1 (middle**
 461 **column) and 2 (right column) m/s. Note difference in scales per column.**



462

463 **Figure 4. Number concentration (left column) and dry particle diameter (right column) by**
 464 **mode (color lines) for MATRIX (dashed lines) and MATRIX-VBS (solid lines) for the**
 465 **experiments with the same conditions as Figure 1.**