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- 30 **Text S1**.
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# 32 **1** Experimental methods

Experiments were performed at the University of Eastern Finland (Kuopio) in the aerosol
 laboratory of the Department of Applied Physics. Both gas phase and particle phase were
 monitored with a comprehensive suite of instrumentation including a Scanning Mobility
 Particle Sizer (SMPS), Aerosol Mass Spectrometer (AMS) and 2 CPCs for monitoring the
 particle phase, and Proton Transfer Reaction Time of Flight Mass Spectrometer (PTR-ToF-MS),
 O<sub>3</sub>-monitor and RH-probes for monitoring the gas phase. The basic experimental setup is
 presented in Fig. S1.

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# 1.1 Glass flow tube – Particle generation

42 SOA was generated via  $\alpha$ -pinene ozonolysis in a continuous flow reactor made of guartz 43 (4.5 m length, 0.056 m diameter). The  $\alpha$ -pinene (Sigma-Aldrich, 98%) was introduced to the 44 flow reactor by flowing 0.13 lpm of clean air through a diffusion source, which was constructed 45 from a glass bottle with a vial of  $\alpha$ -pinene inside. VOC flow was controlled with a needle valve. 46 Ozone was generated via photodissociation of oxygen in o.6 lpm of clean air with a stable 47 ozone generator (UVP, part # 97-0067-02). Dilution air was added at the inlet of the flow tube 48 to adjust particle mass loadings and particle size. Dilution flow ranged from 1-6 lpm. A vacuum flow was also added upstream of the flow reactor to enable additional dilution of the  $\alpha$ -49 50 pinene/ozone mixture while maintaining the low flow rates through the flow reactor that were 51 needed to grow the particles. Vacuum flows ranged from 0-3.5 lpm. Total flow through the flow 52 reactor 2.5-3.33 lpm (average residence times in the flow tube ranged from 3.3-4.4 min). All 53 flows were controlled with mass flow controllers unless otherwise noted. The polydisperse 54 mass loadings in the flow reactor was 100-130 µg m<sup>-3</sup>. High polydisperse mass loadings had to 55 be used to ensure there was enough monodisperse particle mass to conduct the evaporation 56 experiments. Geometric mean diameter was 51±2 nm. SOA was generated under dry condition 57 in all experiments except was one dry evaporation experiment (experiment 1) where particles 58 were formed at 30% RH.

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# 1.2 DMA – Particle sampling and size selection

61 The first Differential Mobility Analyzer (DMA1) selected a nearly monodisperse particle 62 population from the generated SOA population. The selected particle size was 80 nm. A nano-63 DMA (TSI, model 3085) or a short Vienna style DMA (custom made) was used with open loop 64 sheath flow configuration using sheath flow of 8.5 or 10 lpm and sample flow of 0.3 or 1.5 lpm, 65 respectively, depending on the desired residence time. The total flow in DMA1 leads to a 66 drifting time of 0.2-0.3 s (nano-DMA) or 0.8-0.9 s (Vienna-DMA) inside DMA1 [Li and Chen, 67 2005]. The nano-DMA was used in dry experiment 1 and the short Vienna style DMA was used 68 in all other experiments. Ozone was removed by passing the SOA through an ozone scrubber 69 (copper tubing coated with dried KI solution) before entering DMA1 to prevent oxidation 70 reactions in the evaporation chamber. Purified air was used for the sheath air and therefore the 71 gas phase of the sample at the exit from the DMA1 can be assumed to be clean air. This was 72 confirmed by PTR-ToF-MS and O<sub>3</sub>-monitor measurements. In case of 40% and 80% RH 73 experiments the sheath air of DMA1 was humidified with a Nafion humidifier. The RH was 74 measured with Vaisala sensors (HMP 110) just before the inlet and after the outlet of the 75 evaporation chamber. RH in the evaporation chamber was 39±0.5% (40% RH) and 78±1% (80% 76 RH).

## 1.3 Evaporation chamber – Varying evaporation time

79 After DMA1 the nearly monodisperse particle population was fed into the evaporation 80 chamber. The evaporation chamber is a 100 L (length 1.5 m) cylindrical stainless steel chamber 81 with an inlet in the center of the upper end of the chamber and an outlet at center of the lower 82 end. To prevent turbulence in the flow, laminar nets were installed at the top and bottom of the 83 chamber. Particles were supplied continuously to the chamber until the desired concentration 84 in the chamber was reached. After this the voltage in DMA1 was set to zero to provide air with 85 the same gas phase composition but without particles. This was necessary to replenish the 86 sampled volume in the chamber. After filling, the instruments were disconnected and the 87 chamber closed. Sampling was resumed for approx. 20 min at a time to collect data points for 88 long residence times. Gas phase composition, particle size and composition were monitored 89 during filling of the chamber providing data points with 20-30 min residence time. Shorter 90 residence time data points (2.5-130 s) were achieved by increasing the tube length between the 91 DMA1 and SMPS (i.e. bypassing mode shown in the Fig. S1).

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## 1.3.1 SMPS – Measuring the size of the evaporated particles

94 Particle size distribution was measured with a Scanning Mobility Particle Sizer (SMPS, TSI 95 model 3080 platform accompanied with CPC TSI model 3775 or 3772). The DMA1 and SMPS 96 systems were calibrated with ammonium sulfate (AS) particles at dry conditions before and 97 after the campaign with multiple particle sizes. The TSI 3080 platform uses closed loop sheath 98 flow configuration. The flow circulation unit consist of HEPA filters and an air circulation pump. 99 For the filling of the evaporation chamber high sample flows were used in the SMPS to reduce 100 filling time (3775 high flow mode, 1.5 lpm, or 3772, 1.0 lpm). The later sampling from the 101 evaporation chamber was done with sample flows of 0.3 lpm (3775, low flow mode) or 1.0 lpm 102 (3772). For short residence time measurements in bypassing mode, low flows were used in the 103 SMPS. For the RH40% and RH80% experiments the SMPS was continuously sampling from air 104 at the target RH, thus ensuring that the RH in the closed loop stayed close to that in the 105 evaporation chamber.

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## 1.3.2 HR-ToF-AMS – Measuring the chemical composition of the particles

108 Particle chemical composition was measured with an Aerodyne High-resolution Time-of-109 flight Aerosol Mass Spectrometer (AMS, Aerodyne). Detailed description of the instrument, 110 measurement and data processing are presented elsewhere [DeCarlo et al., 2006]. Briefly, AMS 111 measures the non-refractory composition of sub-micrometer aerosols before and after the 112 evaporation chamber. The V-mode mass spectra were analyzed using standard TOF-AMS data 113 analysis toolkits (SQUIRREL 1.56A and PIKA 1.15). A particle collection efficiency factor of 0.5 114 was applied to account for the loss of particles in the aerodynamic transmission lens and 115 vaporizer. The relatively ionization efficiency values of 1.1, 1.2, 1.3, 1.4 and 3.7 were used for 116 nitrate, sulfate, chloride, organic and ammonium, respectively. An elemental analysis was 117 processed to determine the ratio of oxygen to carbon atoms (O:C) taking into account CHO<sup>+</sup> ion 118 correction [Aiken et al., 2007; Canagaratna et al., 2015]. The particle mass in the monodisperse 119 particle population evaporation experiments was too low for AMS experiments. Therefore, a 120 separate evaporation experiment was performed for particle composition measurements. In 121 this experiment, the polydisperse particle population was diluted after the  $O_3$  scrubber with 10 122 Ipm of clean air at the desired RH and led directly to the evaporation chamber, i.e. DMA1 was 123 bypassed. The clean air was added to reduce the gas phase concentrations of organics and to 124 set the desired RH.

## 126 **1.3.3 PTR-ToF-MS – Measuring the gas phase composition**

127 A proton-transfer-reaction time-of-flight mass spectrometer (PTR-ToF-MS, PTR-TOF 8000 128 Ionicon Analytik, Austria) was used to measure the gas-phase compounds during the 129 experiments. The PTR-ToF-MS measured from two points in the setup: after DMA1, and after 130 the evaporation chamber. After DMA1, the gas-phase was measured in order to verify the 131 DMA's effect of dilution on the gas phase compounds. After the evaporation chamber, the PTR-132 ToF-MS was used to monitor the gas-phase compounds formed due to evaporation. The PTR-133 TOF was operated under the following conditions: drift tube voltage and temperature were set 134 to 600 V and 60 °C respectively, drift pressure 2.30 mbar, and E/N was set to 130 Td. The 135 sampling rate of the PTR-TOF was 160 ml min<sup>-1</sup>through heated PEEK (I.D. 1 mm) tubing 136 (temperature 60 °C) to minimize wall losses in the sampling line.

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## 2 Modelling

139 The evaporation was simulated using two models. One of the models simulates the evaporation of a well-mixed liquid-like particle and the other model can simulate the evaporation of semi-solid particles since the particle phase is represented with a layered structure.

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#### 2.1 Well-mixed particle evaporation model

The "traditional" well-mixed particle evaporation model assumes liquid-like particles for which particle phase molecular transport kinetics do not limit the evaporation, i.e. the particle phase mixing time scale is shorter that the time scale of evaporation. The composition and size evolution of the particles and the evolution gas phase concentrations are obtained by solving the differential equations for the number of organic molecules in a particle and in the gas phase [Vesala et al., 1997; Lehtinen and Kulmala, 2003]

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$$\frac{dN_{p,i}}{dt} = \frac{2\pi (d_i + d_p) (D_{air,i} + D_{air,p}) \beta_i}{k_B T} (p_{\infty,i} - p_{eq,i})$$
(S1)

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 $\frac{dN_{gas,i}}{dt} = -N_{tot} \cdot \frac{dN_{p,i}}{dt} - N_{gas,i} \cdot \text{WLC}$ (S2)

154 155

156 where  $N_{p,i}$  and  $N_{qas,i}$  are the number of molecules *i* in a particle and in the gas phase,  $d_i$  and  $d_p$  are 157 the diameters of molecule *i* and the particle,  $D_{air,i}$  and  $D_{air,p}$  are the gas phase diffusion 158 coefficients of vapor molecule i and a particle,  $\beta_i$  is the transition regime correction factor for 159 mass transport for vapor i and it's calculated based on the version by Lehtinen and Kulmala, 160 [2003],  $k_B$  is the Boltzmann constant, T is the temperature,  $p_{\infty,i}$  and  $p_{eq,i}$  are the partial vapor pressure of *i* in the gas phase and the equilibrium vapor pressure of *i* at the particle surface,  $N_{tot}$ 161 162 is the number concentration of particles and WLC is the vapor wall loss coefficient. 163 The model assumes ideal solution and the equilibrium concentration for compound *i* is

thus  $p_{eq,i} = X_i \cdot p_{sat,i}$ . Ke, where  $X_i$  is the molar fraction in the particle phase,  $p_{sat,i}$  is the saturation vapor pressure and Ke is the Kelvin factor of *i*. In terms of the saturation concentration ( $C_i^*$ ) the equilibrium concentration can be written as

$$p_{eq,i} = X_i C_i^* \frac{RT}{M_i} \exp\left(\frac{4\sigma v_i}{RTd_p}\right) \cdot 10^{-9}$$
(S3)

170 where *R* is the gas constant,  $M_i$  and  $v_i$  are the molar mass and molar volume of *i* and  $\sigma$  is the 171 surface tension of the particle. The exponential term in Eq. (S<sub>3</sub>) is the Kelvin factor. Eq. (S<sub>3</sub>) 172 gives  $p_{eq,i}$  in pascals when  $C_i^*$  is given in  $\mu$ g m<sup>-3</sup> (thus the factor 10<sup>-9</sup>) and other variables in SI 173 units.

The vapor wall loss coefficients (WLC) for the organic vapors in the evaporation chamber 174 175 are not known. The effect of vapor wall losses was tested by carrying out the evaporation 176 experiments in dry conditions with two different particle number concentrations differing 177 approximately by one order of magnitude. The measured evaporation rates in these two cases 178 did not differ much. Based on model simulations, such a similarity in the evaporation rates 179 despite the difference in particle concentrations suggest that the vapor wall losses were fast 180 with approximately WLC  $\ge 10^{-2} \text{ s}^{-1}$  (Fig. S2). Therefore, in the model simulations we assumed 181 that all of the evaporated vapors were lost on walls immediately (infinite WLC) unless otherwise 182 stated.

Only organic compounds were included in the model system when simulating the particle
 evaporation in the dry conditions. For simulating the evaporation in humid conditions water
 was included in the model system. The amount of water in a particle was calculated assuming
 constant equilibrium between gas and particle phase and two different methods were used:

187 Method 1 (basic version of the model): The number of water molecules in a particle was188 calculated at each time step based on the criteria

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$$p_{eq,water} = \frac{\text{RH}}{100\%} \cdot p_{sat,water}$$
(S4)

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192where  $p_{eq,water}$  is the equilibrium vapor pressure of water above the particle surface and  $p_{sat,water}$  is193the saturation vapor pressure of pure water. Here the ideal solution assumption was used also194for water (unity activity coefficient) and the Kelvin effect was taken into account.

195 Method 2: Number of water molecules in a particle was calculated based on measured 196 hygroscopic growth factors (HGF) of  $\alpha$ -pinene SOA particles. The hygroscopic growth factor is 197 defined as the ratio between the dry  $(d_{p,dry})$  and wet  $(d_{p,wet})$  particle diameters (HGF = 198  $d_{p,wet}/d_{p,dry}$ ). We assumed that the molecular volumes (v) in a mixture equal those of pure 199 compounds in which case the number of water molecules in a particle is 200

$$N_{p,water} = \left(\mathrm{HGF}^{3} - 1\right) \frac{\sum_{i} N_{p,i} v_{i}}{v_{water}}$$
(S5)

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where the summation goes over the organic compounds. The HGF was assumed to be sizeindependent, i.e. HGF value stayed the same along the evaporation.

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#### 2.2 Multi-layer particle evaporation model

207The multi-layer particle evaporation model is based on the Kinetic multi-layer model for208gas-particle interactions in aerosols and clouds (KM-GAP) [Shiraiwa et al., 2012; Shiraiwa et al.,2092013]. In KM-GAP the gas-particle system is divided into multiple layers: gas phase, near-

210 surface gas, sorption layer at the particle surface, particle surface layer and multiple particle 211 bulk layers. KM-GAP simulates particle evaporation/condensation based on molecular fluxes in 212 and between gas, particle surface and particle bulk. The model was modified regarding the 213 treatment of the surface layer. In the modified version a new surface layer (ss layer in Shiraiwa 214 et al. [2013]) is formed by combining the surface layer with the first particle bulk layer when the 215 thickness of the surface layer reduces to less than a prescribed value. In our simulations this 216 value was 0.5 nm. As a result, the number of layers is decreasing along the particle evaporation. 217 This modification allowed for simulating the evaporation of particles where there is no non-218 volatile compound (e.g. seed particle). The particle bulk was initially divided into 30 layers. 219 Effect of the number of particle bulk layers was tested and for  $\geq$  30 layers the model results 220 were stable with respect to the change in number of layers.

221 The multi-layer model was applied for dry and 40% RH experiments. For the 40% RH 222 experiment water was included in the model by assuming ideal solution and constant 223 instantaneous equilibration between gas phase and particle surface for water. Transport of 224 water molecules between surface and particle bulk phase and within particle bulk was 225 calculated similarly as for the organic compounds. Organics were treated as an ideal solution 226 (i.e. activity coefficients were unity), desorption lifetimes of all organic compounds were 227 assumed to be 10<sup>-6</sup> s and other properties were the same as in the well-mixed particle 228 evaporation model.

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# 2.3 Fitting of VBS using the genetic algorithm

231 The well-mixed particle evaporation model calculates the particle diameter as a function 232 of time based on the initial particle phase mass fractions and the properties of each compound 233 provided as inputs. We presented the multi-component SOA particles using an 8-bin volatility 234 basis set (VBS). In order to find a satisfactory set of eight initial mass fractions that produce an 235 evaporation similar to the observations, the parameter space was explored heuristically using 236 the genetic algorithm (GA) [Goldberg, 1989], which mimics natural selection.

Ultimately, the goal of the GA in this work is to minimize the fitness function

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 $F(y_{i}; y_{m,i}) = \sum_{i=1}^{n} [y_{i}(t_{i}) - y_{m,i}(t_{i})]^{2}$ 

- 239 (S6) 240 241 where  $y_i$  and  $y_{m,i}$  are the measured and modelled particle size normalized by the initial size 242  $d_p/d_{p,o}$  at time  $t_i$ , respectively. The summation in equation (S6) is over the measured data points. 243 The working principle of the GA in this study can be broken down into 5 steps: 244 1) Create the first generation of initial mass fractions. 245 2) Perform evaporation simulation for each member of the generation. 246 3) Calculate the fitness of each member in the first generation. 4) Create the next generation. 247 248 a. Select two parents. 249 b. Create a child from the parents. 250 c. Perform mutation to the child. 251 d. Perform evaporation simulation to the child. 252 e. Calculate the fitness of the child. 253 f. If the fitness of the child is better than the worst individual in the previous
- 254 generation accept the child to the new generation. Otherwise accept one of the 255 parents.

- 256 257
- g. Go to (a) if the size of the new generation is smaller than the previous generation.
- 258 5) Repeat step 4 until the absolute difference between the fitnesses of the best members 259 of the current and previous generation is smaller than  $\varepsilon = 10^{-5}$  or until a predetermined 260 amount of generations has been created.

In our simulations, each member of a generation consists of eight mass fractions which represent the relative amount of the eight C\* bins (10<sup>-3</sup>-10<sup>4</sup> µg m<sup>-3</sup>) in a particle at the beginning of the evaporation. In the first generation the mass fractions of each member are created randomly. For the creation of pseudorandom numbers, we used the Mersenne twister algorithm.

The most complicated parts of the GA are steps 4a and 4b. Choosing parents from an entire population is not a straightforward task. If only the best members are constantly chosen as parents the parameter space might not be explored thoroughly. If the members are chosen to be parents with equal probability the algorithm might not converge to a minimum of the fitness function. In this study, we used the exponential function to calculate the probability of each member to be chosen as a parent

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$$P(\text{parent}) \propto \exp\left(\frac{F_m}{T}\right)$$
 (S7)

where *F<sub>m</sub>* is the value of the fitness function (S6) of that particular member and *T* is determined
as the difference between the worst and best fitness of the members of the previous
generation. This way when *T* is large, members with larger values of the fitness function have
higher probability to be chosen as parents compared with the situation where *T* is small.
Equation (S7) also ensures that the better members are chosen as parents more often than the
worse members.

After the parents are selected the mass fractions of the child are calculated from theparents' mass fractions as

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$$X_{mass,i,child} = X_{mass,i,l} + \frac{F_l}{F_h} \left( X_{mass,i,l} - X_{mass,i,h} \right)$$
(58)

285

286 where  $X_{mass,i}$  is the mass fraction of VBS bin *i* and subscripts *l* and *h* refer to the parents which 287 have lower and higher value of the fitness function, respectively. Using equation (S8) means 288 that one compares the parents' mass fractions and assumes that, if parent 1 has smaller value 289 of the fitness function than parent 2 it is due to the fact that the mass fractions differ between 290 the parents by some amounts  $\Delta X_{mass,z}$ ...  $\Delta X_{mass,s}$ . The algorithm then tries to change each mass 291 fraction of the child in the same direction where the difference between the parents points. In 292 order to not take too large steps in one generation this change is scaled by the factor 293 proportional to the values of the fitness functions of the parents.

The use of equation (S8) ensures that the sum of child's mass fractions equal unity but allows negative mass fractions to occur. That is why every child is checked for negative values of the mass fractions and, in the event of finding those, the mass fraction with the negative value is changed to zero and an amount equal to the negative value is removed from the next, less volatile bin(s).

Finally, a mutation is performed on the child by randomly changing two mass fractions with each other with probability  $p_{mutation} = 5\%$ . The child is accepted to the new generation if the 301 value of the fitness function is lower than the worst value in the previous generation. If the 302 fitness function has higher value, one of the parents is selected to the next generation with 303 equal probability.

In addition, in step 4 the child is not accepted to the new generation if the volatility
 distribution suggests unreasonably high oxidized organic concentration compared to the
 reacted α-pinene (200 ppb). The limit is set as

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$$\sum_{i} \left( X_{i} C_{i}^{*} \right) < C_{reacted} \tag{S9}$$

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310 where  $X_i$  is the dry particle molar fraction of compound i and  $C_{reacted}$  is the mass concentration of 311 the reacted  $\alpha$ -pinene. The left hand side of eq. (S9) describes the gas phase concentration of 312 the condensing organics at the end of the flow tube assuming that gas-particle equilibrium is 313 reached. In the cases where the oxidized organic mass concentration suggested by the volatility 314 distribution is higher than the allowed, new parents are chosen and a new child calculated until 315 this mass balance criterion is met. The mass balance criterion is approximate as it doesn't take 316 into account the increase in organic mass with oxidation (addition of O), vapor wall losses or 317 particle phase organic mass (< 10% of reacted  $\alpha$ -pinene mass). These factors are expected to 318 cause negligible effect on the results as applying the mass balance criterion described in eq. 319 (S9) caused only a minor effect on the results compared to applying no such mass balance limit.

We performed 20 set of genetic algorithm simulations, each with 15 generations and 200 members per generation, for the dry, 40% and 80% RH experiment. The model simulation with the smallest fitness value from each simulation set was selected giving 20 different initial compositions. The best fit initial composition was chosen from these 20 cases as the one with the smallest fitness function value. Out of the 20 cases we considered only those with fitness < 0.002 (16 for dry, 20for 40% RH, 20 for 80% RH) when reporting the variability in the initial composition (Fig. 1 A, 2 A-B).

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## 2.4 Effect of theoretical treatment of water uptake

329 The basic version of the well-mixed particle evaporation model calculated water content 330 of a particle assuming an ideal solution. The initial particle composition optimized for 80% RH 331 with this basic version is shown in Fig. 1 B. When modelling the evaporation at 40% RH we 332 tested how the theoretical treatment of particle water uptake affected the model predictions. 333 This was done by using the experimentally determined hygroscopic growth factors (HGF) of  $\alpha$ -334 pinene SOA to calculate water uptake. The best fit initial VBS determined at 80% RH using HGF 335 = 0.05 [Pajunoja et al., 2015; Varutbangul et al., 2006] and the corresponding size evolution are 336 shown in Fig. S<sub>3</sub> A-B. This initial VBS was used for simulating the evaporation with the well-337 mixed particle evaporation model at 40% RH using HGF = 0.01 [Varutbangul et al., 2006]. 338 Results for RH40% are shown in Fig. 2B and also in Fig. S3 (blue dashed line). The best fit initial 339 VBS and the variability within the genetic algorithm simulations in the HGF simulations at 80% 340 RH show a similar general pattern as compared to the ideal solution simulations. Also with HGF 341 representation of water uptake, 20 genetic algorithm simulation sets were performed (all 20 342 resulting initial compositions had fitness < 0.002).

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- 349 **Figure S1.** Schematics of particle size measurement system. Pink line represents sample line
- 350 from polydisperse sample to the evaporated monodisperse sample. The SMPS system on the
- right hand side was TSI model 3080 and both high and low flow modes were used depending on
- the target residence time. The humidity control unit in the DMA sheat line was optional and
- only in use in 40% and 80% RH experiments.
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- 355



Figure S2. Time evolution of particle diameter normalized with the initial diameter (80 nm) at dry conditions. Circles: measured size evolution. Lines: Size evolution simulated with the wellmixed particle evaporation model with different assumptions of the vapor wall loss coefficient (WLC) and for different particle concentrations (indicated with the line color). The initial particle composition was the same in all simulations.

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365	Figure S3. Measured evaporation of particles with initial size 80 nm and model simulations at
366	40% and 80% RH with different treatment of particle water uptake. Time evolution of particle
367	diameter normalized with the initial diameter (a). Squares and triangles show measured
368	evaporation at 40% and 80% RH, respectively. The error bars in time originate from the
369	chamber filling time. Well-mixed particle model simulations at 40% and 80% RH with water
370	uptake calculated based on ideal solution assumption: Simulation with the best fit initial
371	volatility distribution at 80% RH (red solid line) and simulation with the corresponding initial
372	composition (shown in c) at 40% RH (red dashed line). Well-mixed particle model simulations at
373	40% and 80% RH with water uptake calculated based on HGF: Simulation with the best fit initial
374	volatility distribution at 80% RH (blue solid line) and simulation with the corresponding initial
375	composition (shown in b) at 40% RH (blue dashed line). Error bars in b-c show the variability in
376	the initial volatility distributions (minimum and maximum) within the best fits from 20 genetic
377	algorithm simulations. The corresponding size evolutions for HGF case at 80% RH are shown in
378	a (gray lines). The corresponding size evolutions for ideal solution case are shown in Fig. 1 a.
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measured in this study under dry (circles), 40% RH (squares) and 80% RH (triangles) conditions.

384 The error bars in time originate from the chamber filling time. Black line shows the

biexponential fit by Vaden et al. [2011] on their measured evaporation of particles (initially 125-

386 251 in diameter) under dry conditions.

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- 389 **Movie S1.** Modelled evolution of particle size and composition at 80% RH. Time evolution of
- 390 particle diameter normalized with the initial diameter: triangles show measured evaporation
- 391 with error bars in time originating from the chamber filling time and line shows the well-mixed
- 392 particle evaporation model simulation with the best fit initial particle composition. The
- 393 corresponding particle composition development is shown in the inset.
- 394
- 395