

Supplementary Information

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MICROFLUIDIC INVESTIGATION OF PORE SIZE DEPENDENCY OF BARITE NUCLEATION

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19 **Supplementary Note 1**

20 **Equation for pore size-controlled solubility (PCS)**

21 When a crystal grows from a solution in a pore of a porous material, its surface tension will give
22 rise to excess pressure p within the crystal¹. For cylindrical pores with radius r (**Fig. S1.1**) this
23 excess pressure can be approximated by

24
$$p = \frac{-2\gamma\cos\theta}{r} \quad (\text{S1.1})$$

25 where the factor of 2 results from the cylindrical pore geometry and γ is the surface tension
26 [N m⁻¹] and θ is the contact angle between the pore wall, crystal, and liquid interface. The pressure
27 increases for decreasing pore sizes.

28 The excess pressure p in a crystal can shift the solubility S of the solid from the bulk solubility S_0
29 and is given as^{2,3}:

30
$$p = (n^+ + n^-)RT \left[\frac{1}{v_s^c} \ln \left(\frac{S}{S_0} \right) - \frac{Xv_w^l}{v_w^c} (S - S_0) \right] \quad (\text{S1.2})$$

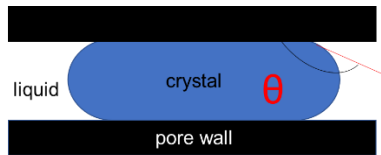
31 where n^+ and n^- are the numbers of positive and negative ions the solid dissociates into, R is the
32 universal gas constant [J mol⁻¹ K⁻¹], T is the temperature [K], v_w^c the molar volume of the solid
33 phase (m³ mol⁻¹), v_w^l the molar volume of water in the liquid phase [L mol⁻¹], and v_w^c the molar
34 volume of the water in the solid [L mol⁻¹]. The factor X (=1000) accounts for the conversion of
35 mol L⁻¹ to mol m⁻³. The ratio of v_w^l and v_w^c accounts for the hydration of the solid phase. In the
36 case of anhydrous crystals like barite the equation is written as:

37
$$p = nRT \frac{1}{v_s^c} \ln \left(\frac{S}{S_0} \right) \quad (\text{S1.3})$$

38 The solubility of a salt inside the pores can be different from that in a bulk solution and increases
39 when the stress on the crystal is compressive. Combining equations S1.1 and S1.2, the solubility
40 can be written as a function of the pore radius⁴:

41
$$\ln \frac{S}{S_0} = \frac{2\gamma v_s^c}{(n^+ + n^-)RT r} \quad (\text{S1.4})$$

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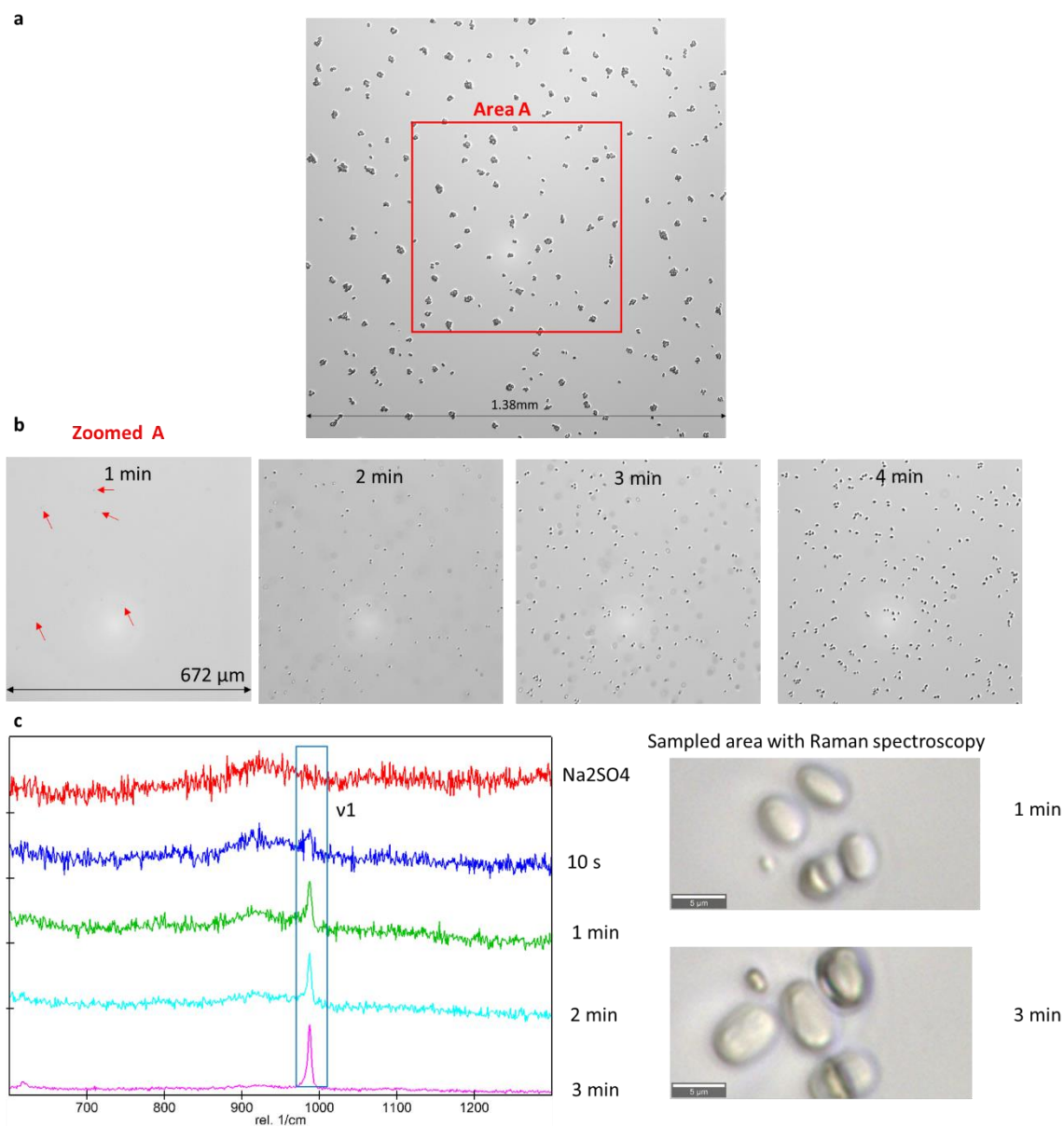
44 **Figure S1** Crystal growing in a cylindrical pore.

45 N.B.: In the case of barite, the solubility, S_0 , is the square root of the solubility product constant
46 K_{sp} (cf. Supplementary Note 3).

47

48 **Supplementary Note 2**

49 **Temporal evolution of bulk experiment and associated Raman measurements**



50
51 **Figure S2 a** Micrograph of bulk solution in petri-dish after 15 minutes, with **b** a zoom in for the
52 first 4 minutes (with red arrows indicating first visible crystallites), and **c** associated Raman
53 spectra with a measuring time of 50 s (0.5 ms x100 integration and x100 objective).

54 These additional measurements are in agreement with the results reported in the main document
55 with a much shorter measuring time of 5 s.

56 Supplementary Note 3

57 Calculation of nucleation rates of barite based on CNT

58 The nucleation rate depends on supersaturation and can be calculated as follows

$$59 \quad J = r \exp\left(-\frac{\Delta G_c}{kT}\right) \quad (\text{S3.1})$$

60 where k is the Boltzmann constant, T is the absolute temperature (298.15K), r a pre-exponential
61 factor and ΔG_c is the energy required for the formation of a nucleus of critical size. ΔG_c is given
62 as:

$$63 \quad \Delta G_c = \frac{\beta v_0^2 \sigma^3}{(kT \ln \Omega)^2} \quad (\text{S3.2})$$

64 where v_0 is the molecular volume of the solid phase ($8.6 \times 10^{-29} \text{ m}^3$ for barite), β is a geometry factor
65 that depends on the shape of the nucleus and was set to 16.8 (value for spheres according to
66 Nielsen⁵), and σ (J m^{-2}) is the specific surface energy of the cluster/solution interface set to 0.134
67 Jm^{-2} for barite⁶. Ω is the saturation ratio given as $\Omega = \frac{\{Ba^{2+}\}\{SO_4^{2-}\}}{K_{sp}^0}$, with the saturation index SI
68 defined as $SI = \log_{10} \Omega$.

69 The pre-exponential factor, r , includes a volume diffusion step and is given as:

$$70 \quad r = 2\pi Z D N_0 N_1 d_c \quad (\text{S3.3})$$

71 where D is the diffusion coefficient of BaSO_4 monomers in free water⁷ set to $9.3 \times 10^{-10} \text{ m}^2 \text{ s}^{-1}$ and
72 $d_c = \frac{4\sigma v}{kT \ln \Omega}$. N_1 and N_0 are concentrations that represent the number of monomers per unit volume
73 of fluid and the number of nucleation sites, respectively.

74 The values for N_0 are usually fitted values. For the interpretation of our results, we used the values
75 reported for barite in Prieto⁷, given in Table S3 as case 0.

76 N_1 depends on the solute concentrations in the aqueous solution and refers to the amount of
77 aqueous BaSO_4 molecules per m^3 ; their concentration was determined from speciation calculations
78 using GEMS⁸.

79 Z is the Zeldovich factor given as:

$$80 \quad Z = \sqrt{\left(\frac{\Delta G_c}{3\pi kT (n_c)^2}\right)} \quad (\text{S3.4})$$

81 with the number of monomers in the critical nucleus, n_c , given as:

$$82 \quad n_c = \left(\frac{2\sigma a}{3kT \ln \Omega}\right)^3 \quad (\text{S3.5})$$

83 where $a = 9.42 \times 10^{-19} \text{ m}^2$ is the area occupied by a molecule of BaSO_4 ⁷.

84 In the case of 3D nucleation, σ is replaced by an effective specific surface energy σ_{eff} (J m^{-2}) given
85 as:

86

$$\gamma_{ef} = \Psi^{\frac{1}{3}}(\theta)\gamma \quad (\text{S3.6})$$

87 where $\Psi(\theta) = \left(\frac{1}{4}\right)(2 + \cos\theta)(1 - \cos\theta)^2$. n_c and the pre-exponential factor, δ , were re-evaluated
88 accordingly.

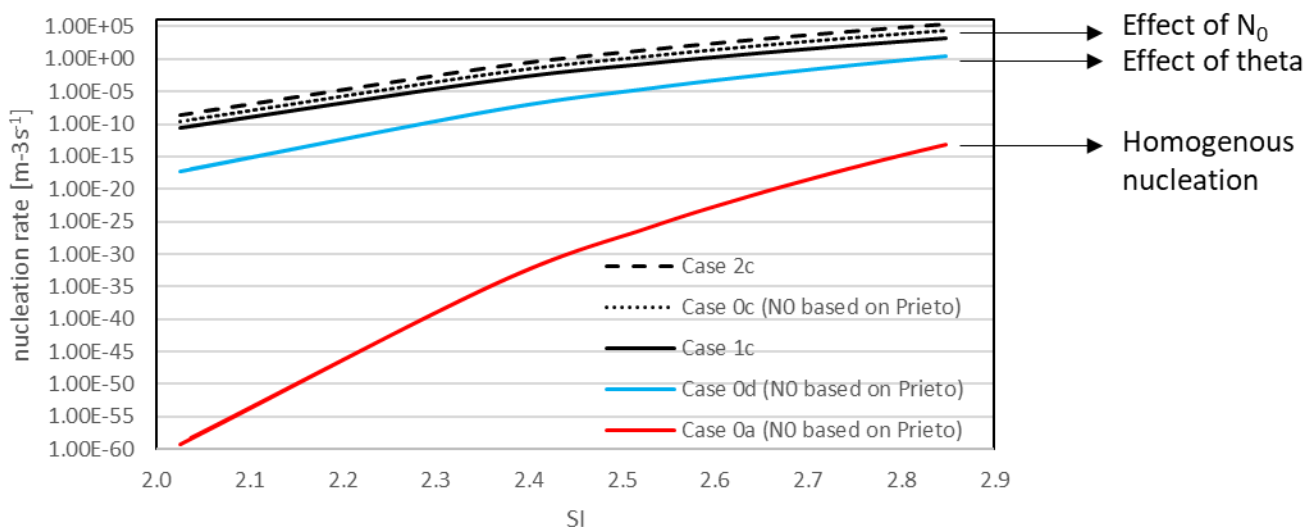
89 We used the values for N_0 as reported by Prieto⁷ for calculating the nucleation rate for the section
90 “Barite nucleation in confinement” i.e. (droplet experiment). In addition to values reported by
91 Prieto⁷, we conducted a sensitivity analysis (case 1 and 2) to quantify the impact of N_0 (HET) on
92 the nucleation rate using different contact angles: 30° for hydrophilic surfaces, 80° for hydrophobic
93 surfaces, and 72.9° as a fitted value to match our experimentally measured rates in the droplets.
94 As reported in the following Table S3, the nucleation rate scales proportionally with the values of
95 nucleation sites (N_0). This follows directly from eq. S3.1 and S3.3. In contrast, the nucleation rate
96 (J) has an exponential dependence on the contact angle, implying that small changes in the contact
97 angle change the nucleation rate (J) by several orders of magnitude. These effects including (i)
98 HOM proceeds at a much faster rate than HET (ii) the influence of N_0 on the computed nucleation
99 rate and (iii) the influence of θ on the nucleation rate, are also captured by Figure S3 which depicts
100 the nucleation rate as function of SI for various calculation cases presented in Table S3. Based on
101 the data reported in Table S3, in our experiments heterogeneous nucleation is most likely involved;
102 homogeneous nucleation can be excluded based on the vanishingly small, calculated values of J
103 compared to the experimentally observed rates.

104 **Table S1** Influence of N_0 and contact angle (θ) on the nucleation rate calculated for an SI of 2.8
105 using a fixed N_1 of $3.65 \times 10^{22} \text{ m}^{-3}$. (N.B the highlighted row indicates the contact angle value fitted
106 to the presented experimental data).

Calculation case	N_0 [m^{-3}]	θ [°]	J [$\text{m}^{-3}\text{s}^{-1}$]
Homogeneous nucleation			
Case 0a (N_0 based on Prieto)	$3.3 \cdot 10^{28}$	n.a.	6.69×10^{-14}
Case 1a	$3.3 \cdot 10^{27}$	n.a.	6.69×10^{-15}
Case 2a	$3.3 \cdot 10^{29}$	n.a.	6.69×10^{-13}
Heterogeneous nucleation			
Case 0b (N_0 based on Prieto)	$2.5 \cdot 10^{13}$	30	6.52×10^{15}
Case 1b	$2.5 \cdot 10^{12}$	30	6.52×10^{14}
Case 2b	$2.5 \cdot 10^{14}$	30	6.52×10^{16}
Case 0c (N_0 based on Prieto)	$2.5 \cdot 10^{13}$	80	2.01×10^4
Case 1c	$2.5 \cdot 10^{12}$	80	2.01×10^3
Case 2c	$2.5 \cdot 10^{14}$	80	2.01×10^5
Case 0d (N_0 based on Prieto)	$2.5 \cdot 10^{13}$	72.9	2.84×10^0
Case 1d	$2.5 \cdot 10^{12}$	72.9	2.84×10^{-1}
Case 2d	$2.5 \cdot 10^{14}$	72.9	2.84×10^1
Experimental results	n.a.	n.a.	$2.0 \times 10^4 - 2.0 \times 10^6$

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111 **Figure S3** Nucleation rate of barite as function of saturation index (SI) for various combinations
112 of N_0 and contact angle θ (cf. **Table S3**).

113 For the calculation of the nucleation rate and resulting induction time presented in **Figure 6a-b**,
114 i.e., the pore network experiment, the number of nucleation sites was set to $2.5 \times 10^{13} \text{ m}^{-3}$ (i.e., the
115 same value as in Prieto⁷) for the large pores. Since such a value depends on available surface area,
116 decreasing the pore radius from 20 to 6 μm involves a reduction of the surface area by a factor of
117 8.5. N_0 for the small pores was thus estimated to decrease by the same factor, i.e., $N_0 = 2.9 \times 10^{12}$
118 m^{-3} for the small pores. For the 2.5 mM reacting solutions, the initial SI lies between 2.8 and 3.7
119 resulting in N_{Imin} and N_{Imax} of 2.11×10^{13} and 1.5×10^{13} monomers per m^3 , respectively. The
120 calculated J_{min} and J_{max} provide a range of nucleation rates for the large and small pores based on
121 the maximum and minimum concentration of $\text{BaSO}_4(\text{aq})$ monomers. Nucleation starts on the
122 surface of PDMS and not on the plasma-treated glass⁸ and consequently the contact angle was set
123 to 81° (based on previous work and within the range expected for PDMS⁹). Using these values,
124 we calculated the apparent induction time (equation 6 in main document) for the 2.5 mM reacting
125 solutions ($2.8 < \text{SI} < 3.7$) and compared the results with the corresponding experimental datasets.

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127 **Supplementary Note 4**

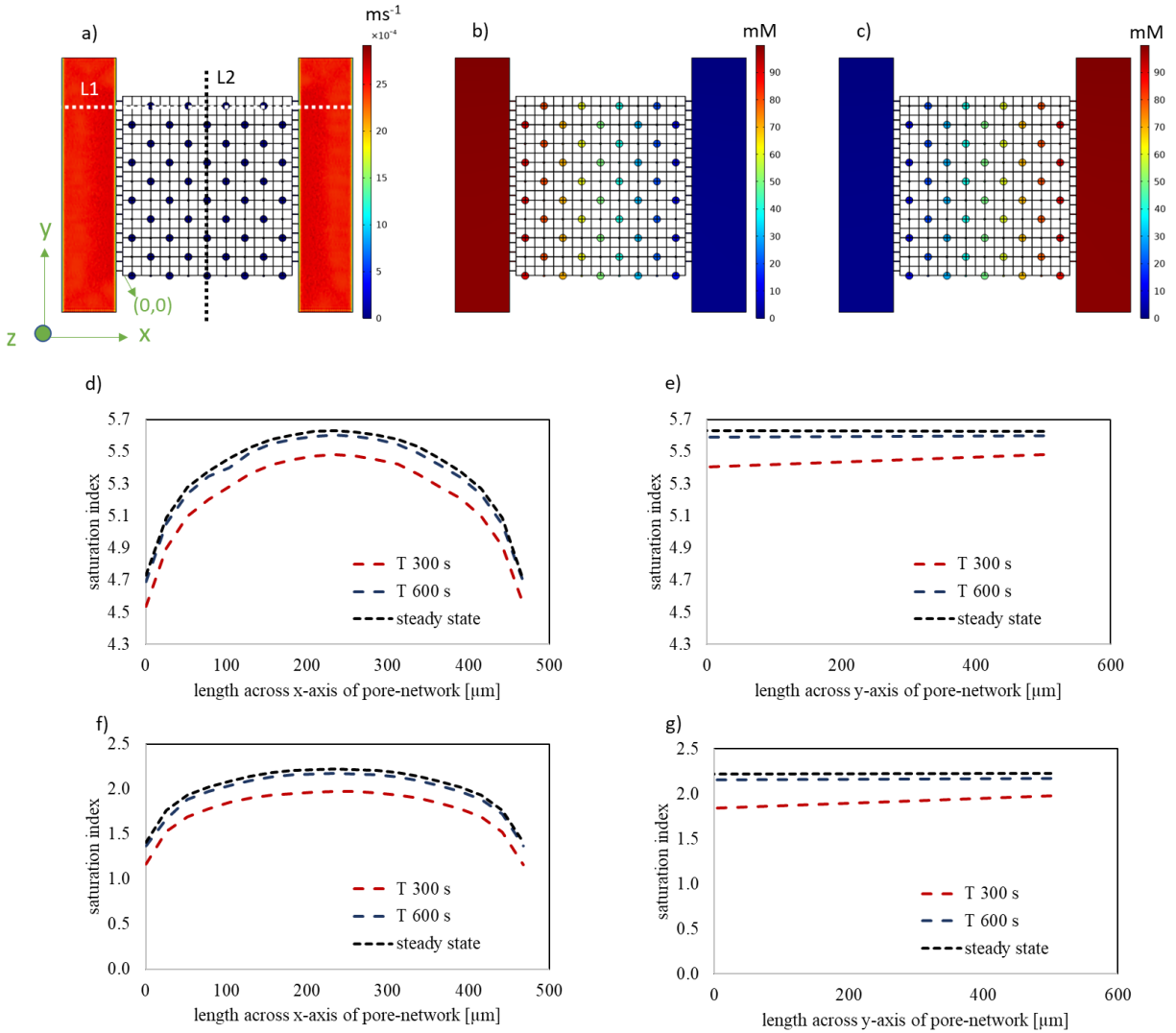
128 **Velocity fields and aqueous solute distributions in the pore network**

129 The velocity fields and the distributions of solutes were simulated using Comsol Multiphysics at
130 a transient state (i.e., the first 30 minutes of the experiment) and at steady state. The simulation
131 results for the 100 mM experiment are depicted in **Figure S4a-c**. A concentration gradient builds
132 up 30 s after the start of the injection of the reactants, with the highest concentrations at the
133 respective supply channel, which decreases to zero in the opposite channel. The compositions of
134 the aqueous solutions along L1 and L2 were exported, and the geochemistry (activities of solutes)
135 was solved using GEMS selector¹⁰ to determine the saturation indices (SI). The saturation indices
136 (SI) along lines L1 and L2 for the highest and lowest concentrations of solutes used in the
137 experiments are plotted in **Figure S4d-g**. These graphs show that there is a negligible difference
138 between the SI calculated for 600 s (10 minutes) after the start of the experiments and the steady
139 state. Consequently, for any nucleation event after 600 s, the calculated steady-state SI was used
140 in our analysis. The ranges of SI at steady state for the various experiments are given in **Table S4**.

141
142 **Table S2** Calculated SI ranges associated with experiments conducted at different concentrations
143 of reactants.

Equimolar reacting solutions of Na ₂ SO ₄ and BaCl ₂ [mM]	Range of saturation index (SI) at steady state
0.325	1.4 - 2.2
0.5	1.7 - 2.5
0.75	2.0 - 2.8
1	2.2 - 3.1
1.25	2.4 - 3.2
2.5	2.8 - 3.7
5	3.2 - 4.1
10	3.6 - 4.5
100	4.7 - 5.6

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146

147 **Figure S4 a** Velocity field at steady state across plane $Z=0.5 \mu\text{m}$ with numerical sampling lines
 148 L1 and L2, **b** and **c** simulated concentrations of Ba(aq) and $\text{SO}_4(\text{aq})$ across the pore network at
 149 steady state for the 100 mM solutions experiment (highest concentration injected); **d** and **e**
 150 temporal evolution of the saturation indices along lines L1 and L2, respectively, for the 100 mM
 151 solution experiment; **f**, and **g** temporal evolution of the saturation indices along lines L1 and L2,
 152 respectively, for the 0.325 mM solution experiment.

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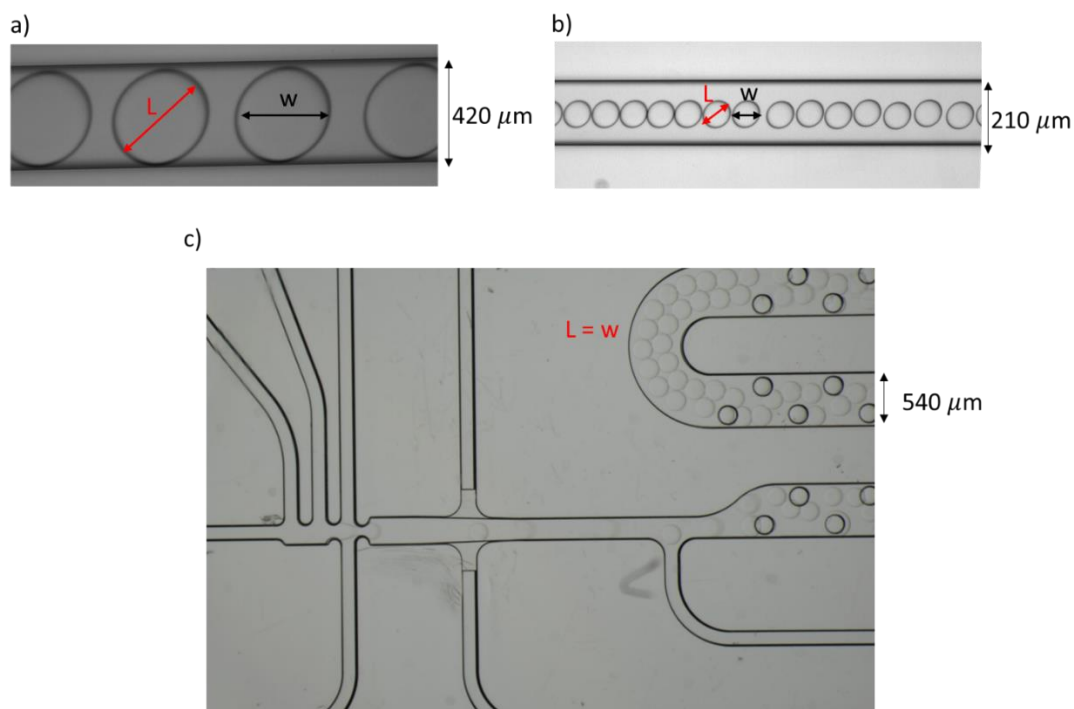
155 **Supplementary Note 5**

156 **Determination of the volume of droplets**

157 Droplets formed in microfluidic channels can either fill the cross-sectional area where they can be
 158 assumed to be rectangular cuboids or can be smaller than the channel depth and therefore assumed
 159 to be spheres. According to the images collected from the droplets in motion, they tend to adopt a
 160 cylindroid shape with rounded edges. The volume of these droplets can be calculated using the
 161 following equation given by dos Santos et al.¹³:

162
$$V = \frac{\pi}{4}hwL - \frac{\pi}{12}wh^2 \quad (S5.1)$$

163 where h is the depth of the microfluidic channel, and w and L are the width and length of the
 164 cylindroid (**Figure S5**). The dimensions of the droplets and the calculated number of BaSO₄(aq)
 165 monomers (for reactant concentrations of 0.75 mM) for the different droplet microfluidic
 166 experiments are provided in **Table S5**.



167
 168 **Figure S5** Illustration of the width and length of cylindroid shaped droplets; **a** for experiment A;
 169 **b** for experiment B; **c** for experiment C.

170 **Table S3** Summary of size of droplets and calculated number of BaSO₄(aq) monomers (for
 171 reactant concentrations of 0.75 mM) for the different droplet microfluidic experiments A-C.

Design	L [μm]	W [μm]	h [μm]	Volume [nL]	Number of monomers
A	390	360	140	15	5.5×10^8
B	86	77	70	0.3	1.1×10^7
C	173	173	100	1.9	6.9×10^7

172 **Supplementary Note 6**

173 **Image processing of droplet experiments**

174 Given the increasing interest in studying nucleation, it is important to establish a robust and
175 automated image analysis procedure to produce reliable and reproducible results that can be easily
176 stored, analyzed, and compared with other experimental or simulated results.

177 **Python packages**

178 The analysis methodology was developed and implemented in the open-source Python
179 programming language version 3.9.13¹⁴, using the OpenCV-Python version 4.0.0.66¹⁵, numpy
180 version 1.21.5¹⁶, and pandas version 1.4.4¹⁷ libraries.

181 **Datasets**

182 Two datasets were analyzed: The first dataset comprised 20 images captured over 10 hours, with
183 the first image taken 30 minutes after starting the experiment (experiment B). The second dataset
184 comprised a single image taken 30 minutes after the start of experiment A. All the images had a
185 size of 11934 x 6490 pixels. The original images were converted to 8-bit grayscale.

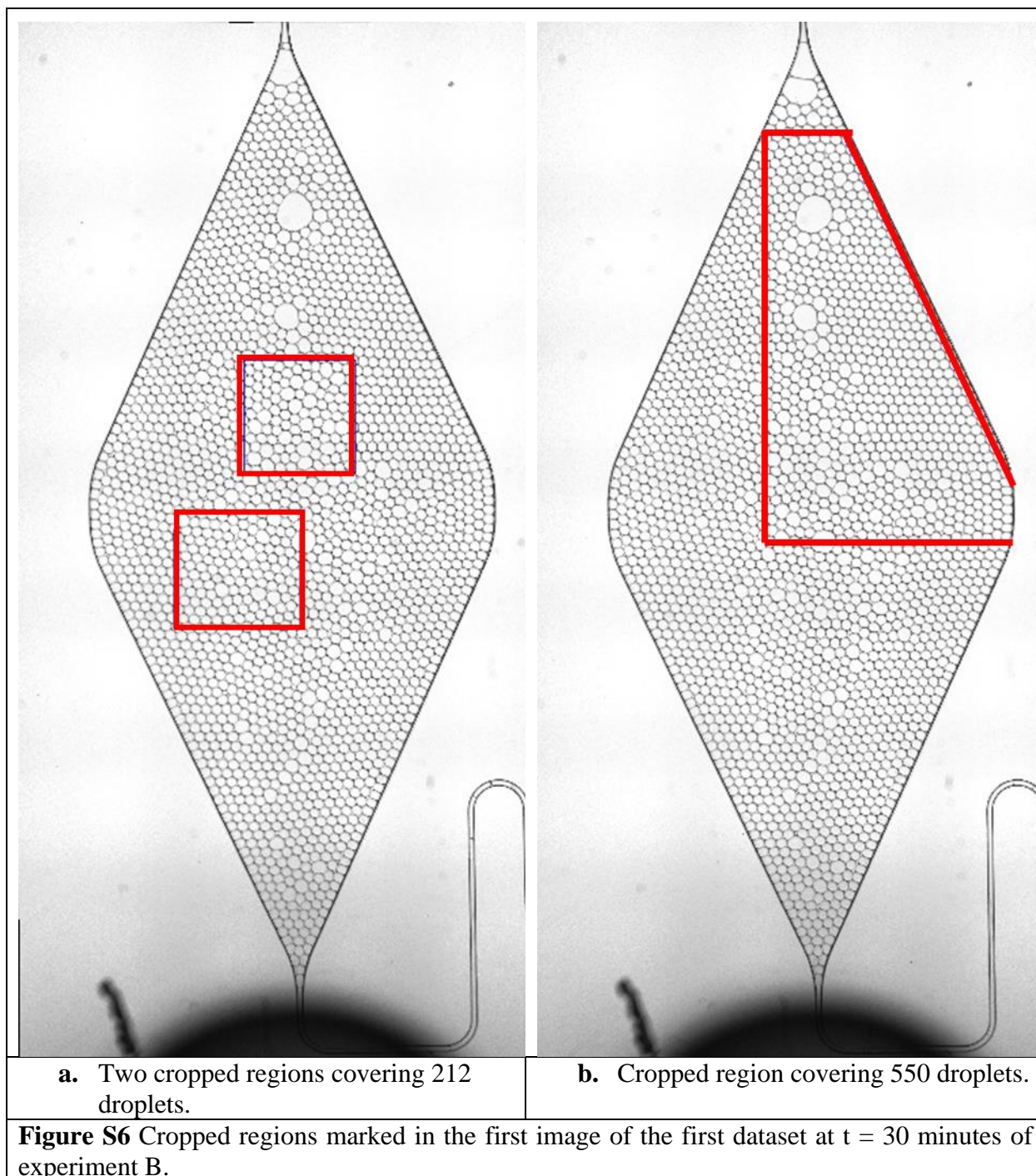
186 **Cropping**

187 For the first dataset, it was noticed that certain regions of the images could not be used because
188 the storage chips (experiment B) started to crack after long exposure to fluorinated oil. For this
189 reason, it was decided to crop two different regions from each image to be used for further analysis
190 covering almost 212 droplets (see **Figure S6a**). To check the validity of this decision, the ratio of
191 droplets containing crystals (N_0) to the total number of droplets (N) was compared between the
192 selected cropped regions and a much bigger portion of the image, covering almost 550 droplets, at
193 $t=30$ min (see **Figure S6b**). The results were found to be very compatible with the ratio at the
194 selected cropped regions in **Figure S6a** being $\frac{N_0}{N} = 0.971$, and the ratio from the bigger cropped
195 portion of the image (**Figure S6b**) being $\frac{N_0}{N} = 0.970$.

196 **Preprocessing**

197 The first step of preprocessing the cropped regions of the images was to treat the white noise,
198 which, if left untreated, would make it difficult to identify small crystals in the droplets. The image
199 denoising was performed using a non-local mean denoising algorithm¹⁸. This method proved to be
200 effective in treating the white noise in the images as it scans different portions of the image to
201 search for a pixel that resembles the pixel it wants to denoise. Additionally, to correctly identify
202 existing droplets and the crystals formed inside them, it is necessary to convert the grayscale
203 images into binary images. It is important to mention that applying conventional thresholding
204 techniques, like Otsu's thresholding method, which uses a global thresholding value proved
205 ineffective when treating the images in this case, due to existing variations in lighting throughout
206 the analyzed regions in the images. For this reason, an adaptive thresholding technique¹⁹ was
207 utilized as it can analyze a small set of neighboring pixels at a time and can compute the threshold
208 value for that specific local region before performing the binarization. Furthermore, two additional

209 morphological operations were performed on the binarized images to ensure that the droplet
210 perimeters were completely enclosed to allow for accurate analysis of the images. With the use of
211 a structuring element, a 3x3 kernel in this case, each morphological operation can identify the
212 region to be examined around each pixel in the image. The dilation and erosion operations were
213 applied consecutively to connect any parts of the droplet's outer perimeters that might have been
214 removed by the previous preprocessing steps. These two morphological preprocessing operations
215 were applied using the OpenCV-Python library¹⁵.



216 **Image analysis**

217 The first step after preprocessing the images was to identify the droplets, which was done by using
218 the find Contours operation in the OpenCV-Python library¹⁵. Additional limitations were
219 introduced to check if the detected contours were of circular shape. Based on the areas of the
220 identified droplets, it was necessary to statistically identify and remove the droplets that were
221 outliers. After testing the data using the q-q plot²⁰ and the Shapiro–Wilk test²¹ to check its
222 normality, it was determined that the data was not normally distributed, the 15th and the 95th
223 percentile were determined to be the limits of valid data in our assessment. Afterwards, the
224 remaining droplets were masked on the binarized image to remove the background to allow for a
225 better analysis of the formed crystals. Finally, the find Contours operation was used again in the
226 final image to determine the number of formed crystals. The final analysis for the cropped region
227 # 1 at t = 30 minutes can be seen in **Figure S5**, where the formed crystals are identified and
228 numbered.

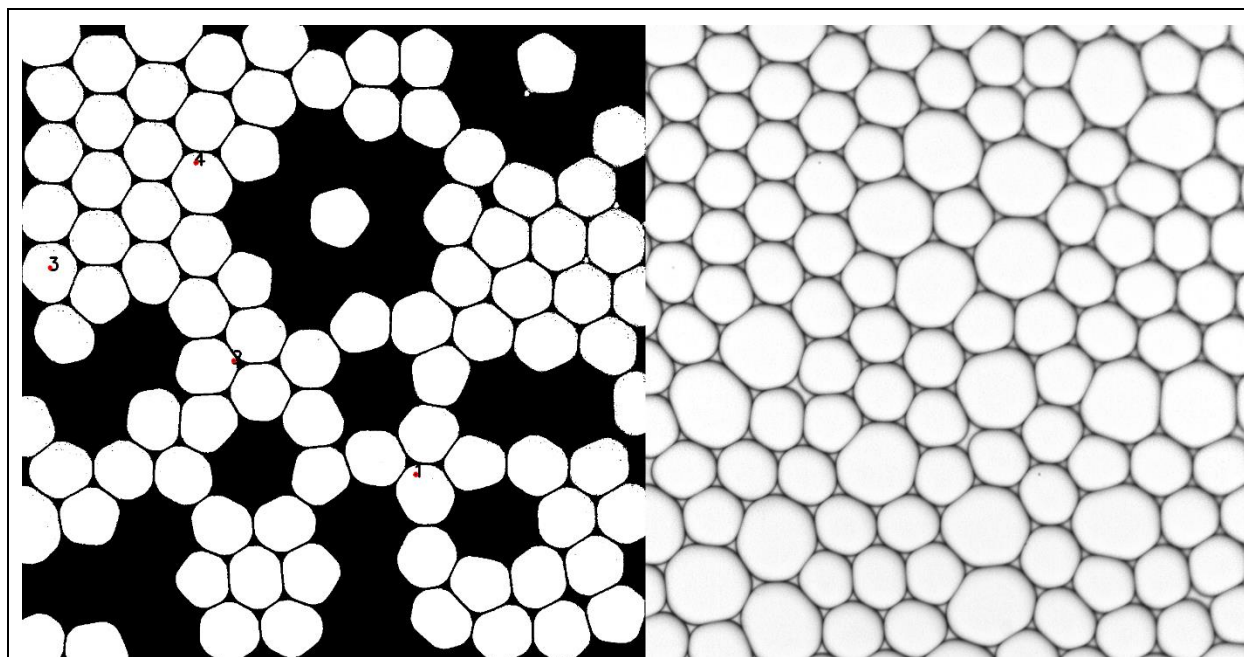


Figure S7 Final analysis of cropped region # 1 at t = 30 minutes.

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