Size-dependent ice nucleation by airborne particles during dust events in the eastern Mediterranean

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Abstract. The prediction of cloud ice formation in climate models remains a challenge, partly due to the complexity of ice-related processes. Mineral dust is a prominent aerosol in the troposphere and is an important contributor to ice nucleation in mixed-phase clouds, as dust can initiate ice heterogeneously at relatively low supercooling conditions. We characterized the ice nucleation properties of size-segregated mineral dust sampled during dust events in the eastern Mediterranean. The sampling site allowed us to compare the properties of airborne dust from several sources with diverse mineralogy that passed over different atmospheric paths. We focused on particles with six size classes determined by the Micro-Orifice Uniform Deposit Impactor (MOUDI) cutoff sizes: 5.6, 3.2, 1.8, 1.0, 0.6 and 0.3 µm. Ice nucleation experiments were conducted in the Weizmann Supercooled Droplets Observation on a Microarray (WISDOM) setup, whereby the particles are immersed in nanoliter droplets using a microfluidics technique. We observed that the activity of airborne particles depended on their size class; supermicron and submicron particles had different activities, possibly due to different composition. The concentrations of ice-nucleating particles and the density of active sites ($n_s$) increased with the particle size and particle concentration. The supermicron particles in different dust events showed similar activity, which may indicate that freezing was dominated by common mineralogical components. Combining recent data of airborne mineral dust, we show that current predictions, which are based on surface-sampled natural dust or standard mineral dust, overestimate the activity of airborne dust, especially for the submicron class. Therefore, we suggest including information on particle size in order to increase the accuracy of ice formation modeling and thus weather and climate predictions.

1 Introduction

Cloud droplets can supercool to 238 K before homogeneous freezing occurs (Koop and Murray, 2016; Rosenfeld and Woodley, 2000). At warmer temperatures, heterogeneous ice nucleation (HIN), whereby the presence of aerosol particles lowers the required energy barrier to form a stable ice nucleus, is the common pathway of ice formation (Murray et al., 2012; Pruppacher and Klett, 1997; Khvorostyanov and Curry, 2004; Hoose and Möhler, 2012). These ice-nucleating particles (INPs) can be activated at subzero temperatures and subsequently lower humidity conditions, mainly by interaction with supercooled droplets. INPs are relatively rare particles and comprise only about $10^{-5}$ of the total ambient particles in the free troposphere (Rogers et al., 1998). Yet, their interaction with clouds can greatly influence climate (Gettelman et al., 2012; Tan et al., 2016; Lohmann and Feichter, 2005). Therefore, it is important to represent them well in weather and climate models (DeMott et al., 2010). Currently, ice formation is a source of great uncertainty in cloud and climate models, partly due to the complexity of ice processes and the insufficient understanding of the key surface properties that determine an INP (IPCC, 2013). To improve the predictions of models, great effort is invested in the characterization of INPs and in the development of parameterizations.
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based on their physical and chemical properties (Cantrell and Heymsfield, 2005; Niemand et al., 2012; Ullrich et al., 2017).

One of the most abundant INPs in the atmosphere is mineral dust, which originates in dryland zones such as deserts (Middleton, 2017; DeMott et al., 2003b). Field observations have identified an increase in INP concentrations and ice cloud formation in the presence of mineral dust (Ansmann et al., 2008; Rosenfeld et al., 2001; DeMott et al., 2003b; Cziczo et al., 2004; Sassen et al., 2003). Ice residuals often contain mineral particles (Cziczo et al., 2013; Cziczo and Fryod, 2014; Twwoy and Poellot, 2005). Mineral dust has high spatial and temporal variability, impacting atmospheric, oceanic, biological, terrestrial and human systems (Garrison et al., 2003; Gat et al., 2017; Jickells et al., 2005; Mahowald et al., 2014; Mazar et al., 2016; Middleton and Goudie, 2001). Each year, gigatons of dust are transported globally over long distances, dominating the atmospheric aerosol mass and aerosol optical depth (AOD) (Chiapello et al., 1999; Tegen and Fung, 1994; Ben-Ami et al., 2010; Prospero, 1999; Koren et al., 2006). Though the exact property of an aerosol that determines its ice nucleation ability remains unclear, it was consistently shown that the mineral composition plays an important role (Kanji et al., 2017) and that for a certain mineral type, larger particles are more effective heterogeneous INPs than small ones (Archuleta et al., 2005; Lüönd et al., 2010; Welti et al., 2019). Local surface features such as steps, cracks and cavities, a close match of the surface lattice with that of ice, or surface hydroxyl groups (Freedman, 2015; Marcoll, 2014; ZIELKE et al., 2015; KISELEV et al., 2017; FLETCHER, 1969; Tunega et al., 2004; Anderson and Hallett, 1976; PRUPPACHER and KLETT, 1997) are believed to be the responsible factors for the ice nucleation ability of mineral surfaces.

Natural mineral dust particles are often chemically similar but differ in their mineralogy (Engelbrecht et al., 2009), and the particles are often composed of a mixture of minerals (internally mixed), such as clays, quartz, feldspars and calcites (Clauquin et al., 1999). Other common minerals are palygorskite, hematite, halite, gypsum, gibbsite and goethite (Ganor et al., 1991; Perlwitz et al., 2015; Kandler et al., 2007; Mahowald et al., 2014). The mineralogy of mineral dust is set by its source region and is considered to be an important factor that determines its freezing characteristics (Zimmermann et al., 2008; Augustin-Bauditz et al., 2014). Traditionally, clay minerals were thought to be responsible for atmospheric ice nucleation because they compose much of the dust fraction. However, using standard mineral particles, Atkinson et al. (2013) showed that K feldspar is the most efficient type and suggested that it could dominate atmospheric ice formation at relatively high temperatures above 258 K. This was further supported by measurements of natural mineral dust from desert surfaces worldwide, wherein the importance of quartz mineral was also indicated (Boose et al., 2016b).

Airborne mineral dust (AMD) can experience chemical and physical modifications during its atmospheric transport that may alter dust’s ability to nucleate ice (Kanji et al., 2013). It was shown that atmospheric aging processes can change the size, morphology and surface chemistry of the particles. This includes, for example, the adsorption of organic components on AMD (Murphy et al., 2006; DeMott et al., 2003a; Falkovich et al., 2004) or coatings of nitrates, chlorides and sulfates that enhance the hygroscopicity of the particles (Krueger et al., 2004; Laskin et al., 2005; Li and Shao, 2009). Levin et al. (1996) found that AMD particles transported over the Mediterranean Sea were often coated with sulfate and other soluble materials, which affect cloud microphysical properties and can eventually result in enhanced ice nucleation. In addition, mineral dust carries biological components, such as bacteria and fungi, which are known to have the ability to induce ice nucleation at relatively high temperatures (Gat et al., 2017; Mazar et al., 2016; Pratt et al., 2009; O’Sullivan et al., 2016). Further modifications that can occur during AMD atmospheric transport are the differentiation of size and mineralogy. These can occur due to gravitational sedimentation, for example, whereby larger particles sediment faster than smaller ones. Near source regions, dust samples were richer in components that are more abundant in the coarse fractions, such as quartz and potassium feldspars, while in remote locations, higher amounts of clay minerals and sodium–calcium feldspar were observed (Murray et al., 2012; Schepanski, 2018).

While there are few measurements of AMD close to source regions (Price et al., 2018; Boose et al., 2016a; Ardon-Dryer and Levin, 2014; Schrod et al., 2017), parameterizations of ice formation in climate models are often based on the freezing properties of natural dust or soil samples collected from deserts or standard dust particles (Niemand et al., 2012; Connolly et al., 2009; Ullrich et al., 2017; Atkinson et al., 2013; Broadley et al., 2012) that may not sufficiently represent AMD (Boose et al., 2016b; Spichtinger and Cziczo, 2010). Natural dust samples showed higher ice nucleation ability than AMD samples, possibly due to atmospheric processing of AMD that may lead to deactivation and possibly due to laboratory processes, such as milling or sieving, that were applied to the natural dust samples and may have enhanced its activity (Boose et al., 2016b).

In this study, we sampled airborne particles during dust events in the eastern Mediterranean and investigated their ice nucleation abilities. The eastern Mediterranean is located in the strip of the world’s main deserts and experiences the transport of desert dust from different sources. The main source is the Sahara in North Africa. It is estimated that about 100 million tons of dust per year is lifted from the Sahara towards the eastern Mediterranean during late winter and spring (Ganor, 1994; Ganor and Mamane, 1982; Ganor et al., 2010). In autumn, local dust is transported, commonly from the Arabian Peninsula and the Syrian Desert (Dayan et al., 1991; Ganor, 1994). The dust events are often associated with the regional eastern Mediterranean synoptic systems, such as winter lows and Red Sea troughs (Ganor et al., 2010). Our sampling site was located in Israel, where Saharan dust
transported over North Africa and/or the Mediterranean Sea and Syrian and Arabian dust is transported over land from the east (Ganor et al., 1991). These distinct sources and paths allow for the investigation of the ice nucleation properties of AMD with diverse origins and transport paths.

The ability of the collected particles to initiate immersion freezing was studied using the Weizmann Supercooled Droplets Observation on a Microarray (WISDOM) instrument (Reicher et al., 2018), and one of the dust events was studied using the Bielefeld Ice Nucleation ARraY (BINARY) instrument (Budke and Koop, 2015). We characterized the concentrations and the density of ice-nucleation-active sites (INASs) of AMD in different size classes for several dust cases, as well as combining recent literature and available AMD data to understand how well AMD is represented in models based on recent parameterizations.

2 Data and methods

2.1 Sampling

Airborne particles were sampled during six dust events in 2016 and 2017, detailed in Table 1. Sampling started when visibility was reduced due to increasing concentrations of particulate matter (PM). The sampling site is located on the roof of a three-story building in Rehovot, Israel (31.9°N, 34.8°E; about 80 m a.s.l.). The location is often impacted by mineral dust storms, transported from nearby and distant geographical locations, mainly from the Sahara and Arabia deserts and less frequently from the Syrian Desert depending on the season and the synoptic conditions (Dayan, 1986; Ganor et al., 2010; Kalderon-Asael et al., 2009).

Particles were collected on polycarbonate filters (47 mm Cyclopore, 0.1 µm Isopore; Whatman) using the Micro-Orifice Uniform Deposit Impactor (MOUDI) (model 110-R). The MOUDI is a 10-stage impactor with an 18 µm cut-point inlet stage followed by size-segregating stages with cut points ($D_{50}$) between 0.056 and 10 µm in aerodynamic diameter (Marple et al., 1991). The particles are collected on the different stages as a function of their aerodynamic diameter. The collection efficiency for each particle size is described in Marple et al. (1991). Sampling time ranged between 17 and 48 h with a 30 L min$^{-1}$ sample flow rate, similar to previous studies (Huffman et al., 2013; Mason et al., 2015).

2.2 Air mass back trajectories

Back trajectories were calculated by a Lagrangian method using LAGRANTO 2.0 (Sprenger and Wernli, 2015). The calculation of air mass trajectories was based on wind data from the European Centre for Medium-Range Weather Forecasts ERA-Interim reanalysis (Dee et al., 2011), available every 6 h at 1° × 1° horizontal grid with 60 vertical hybrid levels. For each 6 h time step during each event, 72 h back trajectories were calculated from all available data grid points with pressure larger than 850 hPa, resulting in 11 trajectories, which end their path in the lower troposphere for each calculation. In a second step, the Eulerian densities of the resulting trajectories were computed by gridding the trajectories for each event, smoothed by using a radius of 100 km and interpolated to 1 h. Finally, the trajectory density was summed over the entire event duration and normalized by the maximum trajectory count.

2.3 Dust column mass density maps

Time-averaged maps of dust column mass density (hourly 0.5° × 0.625°) reanalysis data were obtained from the Modern-Era Retrospective analysis for Research and Applications (MERRA-2). Maps were produced using NASA’s Global Modeling and Assimilation Office (GMAO) (Gelaro et al., 2017) for a period of up to 72 h prior to the sampled event.

2.4 Particulate matter data

Particulate matter mass data were obtained from the Israeli Ministry of Environment website. Concentrations of particles with aerodynamic diameters smaller than 10 µm (PM$_{10}$) were measured in the Rehovot station, located about 1 km from our sampling site. The 5 min mean data were used to calculate peak and mean concentrations of the sampled dust events.

2.5 Particle number size and surface area distributions

Particle size distribution and concentrations between 0.25 and 32 µm were measured on site by an optical particle counter (OPC; GRIMM Technologies model 1.109) in parallel to the MOUDI sampling. In order to estimate the total surface area that was collected on the different stages, we assumed that the particles are spheres and used the diameter of the GRIMM midpoint of the different GRIMM channels as the particle diameter.

2.6 Conversion of GRIMM channels to MOUDI stages

To determine the total surface area collected on MOUDI’s filter, a conversion matrix between the GRIMM channels and the MOUDI stages was applied. The conversion was based on the particle collection efficiency curves of MOUDI and interstage particle losses reported in Marple et al. (1991). Figure 1 demonstrates the fraction of particles that are collected on the stages based on their aerodynamic diameter. Freezing analyses focused on stage no. 2 ($D_{50} = 5.6$ µm), stage no. 3 ($D_{50} = 3.2$ µm), stage no. 4 ($D_{50} = 1.8$ µm), stage no. 5 ($D_{50} = 1.0$ µm), stage no. 6 ($D_{50} = 0.6$ µm) and stage no. 7 ($D_{50} = 0.3$ µm). For example, most of the particles with an optical diameter > 8.5 µm will be collected on stage no. 2 ($D_{50} = 5.6$ µm), whereas all the particles with an optical diameter > 17.5 µm are assumed to be collected on former stages (inlet and stage no. 1). In some cases, particles in a
Table 1. Summary of the investigated dust storm events. The events are denoted by their geographic origin: Saharan dust storm (SDS), Syrian dust storm (SyDS), mixed contribution of the two (MDS) and mix of dust event with a dust-free period (CSDS).

<table>
<thead>
<tr>
<th>Event</th>
<th>Start (UTC)</th>
<th>Sampling period (h)</th>
<th>PM$_{10}$ peak ($\mu$g m$^{-3}$)</th>
<th>PM$_{10}$ mean ($\mu$g m$^{-3}$)</th>
<th>Freezing analysis technique</th>
</tr>
</thead>
<tbody>
<tr>
<td>SyDS1, 19 April 2016</td>
<td>07:30</td>
<td>25</td>
<td>132</td>
<td>76 ± 20</td>
<td>WISDOM</td>
</tr>
<tr>
<td>CSDS, 27 April 2016</td>
<td>07:30</td>
<td>24</td>
<td>67</td>
<td>30 ± 13</td>
<td>WISDOM, BINARY</td>
</tr>
<tr>
<td>SyDS2, 23 November 2016</td>
<td>15:30</td>
<td>18</td>
<td>332</td>
<td>184 ± 68</td>
<td>WISDOM</td>
</tr>
<tr>
<td>SDS1, 9 March 2017</td>
<td>11:00</td>
<td>48</td>
<td>387</td>
<td>96 ± 66</td>
<td>WISDOM</td>
</tr>
<tr>
<td>SDS2, 12 March 2017</td>
<td>12:00</td>
<td>24</td>
<td>717</td>
<td>206 ± 120</td>
<td>WISDOM</td>
</tr>
<tr>
<td>MDS, 12 April 2017</td>
<td>13:30</td>
<td>25</td>
<td>409</td>
<td>141 ± 106</td>
<td>WISDOM</td>
</tr>
</tbody>
</table>

2.7 Ice-freezing experiments and quantification

2.7.1 WISDOM

The immersion freezing activity of the sampled ambient mineral dust was measured using suspensions of the collected particles that were extracted from the filters by dry sonication (VialTweeter, model UP200St; Hielcher). This type of sonication method is more effective than an ultrasonic bath in which most of the energy dissipates in the surrounding water. A quarter filter was inserted into a 1.5 mL Eppendorf vial with 0.3 mL of deionized water and sonicated in three 30 s cycles to avoid heating produced during intense sonication. The suspension was immediately used for droplet production and freezing experiments in WISDOM as detailed in Reicher et al. (2018). Briefly, an array of 0.5 mL monodispersed droplets (~100 µm diameter, suspended in an oil mixture) was generated in a microfluidic device that was cooled by a commercial cooling stage (THMS600, Linkam) under a microscope (BX-51 with 10× magnification, Olympus) coupled to a charge-couple device (CCD) camera. The device was first cooled at a faster constant rate of 10 K min$^{-1}$ from room temperature to 263 K since freezing events were not expected and indeed were never observed in that temperature range. Then a constant cooling rate of 1 K min$^{-1}$ was used until all the droplets froze. The temperature uncertainty was ±0.3 K based on error propagation between the calibrated droplet temperature and the uncertainty of the temperature sensor that is located in the cooling stage (see Reicher et al., 2018, for more details).

2.7.2 BINARY

The Bielefeld Ice Nucleation ARraY (BINARY) is an optical freezing array of droplets pipetted on a hydrophobic substrate in separated sealed compartments and cooled in a Linkam cooling stage (LTS120) (Budke and Koop, 2015). In the present study an array of 64 droplets of 0.6 µL was employed. Suspensions were prepared by extracting a quarter filter in 1.5 mL of double-distilled water (that is, 5 times more diluted than WISDOM suspensions) using a bath sonicator (Elma Transsonic Digital, TP 670/H) for 30 min. The
bath temperature increased during sonication from about 288 to 308 K. The obtained suspensions were used directly and further diluted (1:10) for another set of measurements with reduced surface area of the particles in the droplets. For the freezing experiments, the droplets were cooled at a rate of 1 K min\(^{-1}\). Temperature uncertainty was ±0.3 K.

### 2.7.3 Quantification of freezing properties

The cumulative concentration of INPs present in a volume of solvent, \( V \), at temperature \( T \) was derived using the fraction of frozen droplets (\( f_{\text{ice}}(T) \)) that was obtained directly from the freezing experiments (Vali, 1971):

\[
K(T) = \frac{-\ln(1-f_{\text{ice}}(T))}{V} \quad \text{[cm}^{-3}\text{ of water]}.
\]  

For control experiments, a quarter of blank filter was immersed in pure water, similar to freezing experiments for the airborne samples, and the concentration of the background impurities (\( K_{\text{imp}}(T) \)) was subtracted from the concentrations that were detected for airborne samples.

The atmospheric concentrations of INPs per unit volume of air as a function of temperature, INP(\( T \)), were determined by incorporating the sampling and solvent parameters into Eq. (2) (Hader et al., 2014):

\[
\text{INP}(T) = \left( K(T) - K_{\text{imp}}(T) \right) \frac{V_{\text{solvent}}}{f \cdot V_{\text{air}}} \quad \text{[L}^{-1}\text{ air]},
\]  

where \( V_{\text{solvent}} \) is the volume of the water used for extraction, \( V_{\text{air}} \) is the total sampled air volume and \( f \) is the fraction of filter that was used in the extraction.

For a comparison of the ice nucleation activity of the different dust events, the INP concentration in the liquid was converted to the number of active sites per unit surface area of INPs, i.e., the surface density of sites \( n_s \) active above temperature \( T \) (Vali, 1971):

\[
n_s(T) = \frac{-\ln(1-f_{\text{ice}}(T))}{A} \quad \text{[m}^{-2}\text{]},
\]  

where \( A \) is the surface area immersed in a single droplet of the experiment based on the total surface area of particles in the suspension.

### 2.8 Scanning electron microscopy

A quarter of selected filters were coated with iridium for analyzing the chemical composition of airborne particles using a scanning electron microscope (SEM; Supra 55VP, LEO) equipped with an energy-dispersive X-ray spectroscopy (EDX) detector for elemental microanalysis. The analysis was done at a voltage of 5 kV using the Quantax software (Bruker).

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**Figure 2.** Air mass origin and atmospheric paths of the dust events. Colors represent the density of 72 h backward air mass trajectories (normalized to the total trajectory counts). The green contours represent the geographic locations where a high mass of dust occurred during the air mass transition, which is defined as the potential origin of the dust. Abbreviations in the top right of each panel indicate the particular dust event.

### 3 Results and discussion

#### 3.1 Air mass back trajectories and the origin of the dust storms

The density of air mass back trajectories for a 72 h period prior to the sampling for all events is shown in Fig. 2. The sampling site and the surrounding main deserts are shown as well. During the sampled events, the air mass trajectories were diverse. In some cases, the air masses traveled directly to the sampling site from the source region, while in other cases, they traveled a longer distance. In most events, the air mass had either an easterly or westerly component and was often concentrated in the same geographical area.

The dust origins were identified based on back trajectory analysis, integrated with reanalysis data from remote sensing of atmospheric dust. We followed the dust mass concentration prior to the sampling period, as detailed in Fig. S1 in the Supplement. Locations that contained high levels of suspended dust that overlapped with the air mass trajectories were identified as possible sources of dust. The green contours in Fig. 2 represent the assigned dust origin for each sampled event based on the reanalysis data. Note that in two
events, there was no overlap between the dust origin and air mass trajectories. These events will be further discussed below. Two events, denoted by SDS1 and SDS2, originated in the northern Sahara. The source of SDS1 was near the border of Egypt and Libya, and the source of SDS2 was in Egypt, east of SDS1. The dust traveled over the Mediterranean Sea and was potentially affected by the marine environment, possibly obtaining a sea salt or anthropogenic sulfate coating (Levin et al., 1996). Two other events, denoted by SyDS1 and SyDS2, originated from the Syrian Desert, from western Iraq and southern Syria. Compared to the Saharan events, the dust mass density in the Syrian Desert events was relatively low.

Another event was defined as a “mixed dust” event (MDS) because it was more complicated and included contributions of different sources: the analysis indicates that there is one possible dust origin east of the sampling site in the Syrian Desert and another one southwest of the sampling site in the Sahara. However, the air mass trajectories did not overlap the Saharan dust origin but indicated that the air mass was transported from the Red Sea. Further analysis of the air mass trajectories prior to the sampling period in the Red Sea showed that both Sahara and Arabia dusts were transported to the Red Sea (see the Supplement, Fig. S2a). Another event did not show overlap between the air mass trajectories and the dust origin. Further analysis of air mass back trajectories in the days prior to the sampling period showed that dust was transported to the Mediterranean Sea from the region of Libya in the Sahara, towards Turkey, and was deflected eastward by westerly winds to the sampling site (see the Supplement, Fig. S2b). The dusty air masses rapidly cleared up, and relatively non-dusty air masses arrived at the sampling site, as inferred from PM$_{10}$ concentrations and the OPC size distributions; see Sect. 3.2. This event was defined as a “clean and Saharan dust storm” and is denoted by CSDS. Table 1 summarizes the sampled events, their sampling periods, and the peak and mean PM$_{10}$ concentrations during sampling. Peak values ranged from 67 $\mu$g m$^{-3}$ in CSDS and 132 $\mu$g m$^{-3}$ in SyDS1 to 717 $\mu$g m$^{-3}$ in SDS2, which was the strongest dust event in this study. In SDS1, MDS and SyDS2, the values ranged between $\sim$300 and 400 $\mu$g m$^{-3}$. When comparing the mean PM$_{10}$ concentrations during the entire sampling periods, CSDS was categorized as a non-dusty event, with the lowest concentrations of 30 $\pm$ 13 $\mu$g m$^{-3}$, i.e., below the threshold of 42 $\mu$g m$^{-3}$ for dusty conditions (Krasnov et al., 2014). The mean values in the rest of the events ranged from 76 to 206 $\mu$g m$^{-3}$, and they were therefore categorized as dust storms.

3.2 Particle number size distributions

Figure 3a describes the mean particle number size distributions of sampled air during the dust events, as detected by the GRIMM OPC. The lowest channel of the GRIMM includes particles that are larger than 0.25 $\mu$m. This channel possibly underestimates the total particle count since the counting efficiency is less than 100%.

The number size distributions had similar patterns in all the events. The highest particle number concentrations were in the submicron size range, decreasing towards larger particles. Events SDS1, SDS2 and MDS had a rather similar particle concentration distribution. Event SyDS1 showed similar particle concentrations in the submicron range, but the particle concentrations in the supermicron range were about an order of magnitude lower, which was also apparent in the PM$_{10}$ data. CSDS, a predominantly non-dusty event, had the lowest particle concentrations in comparison to the rest of the sampled events, as also indicated by the PM$_{10}$ data. In the SyDS2 event, exceptionally high concentrations in the supermicron range above 3 $\mu$m were observed, and the peak extended towards larger particle sizes combined with relatively high particle concentrations. Note that prior to and during this event, a series of biomass burning events occurred in Israel extending to about 100 km north and 50 km east of the sampling site. Therefore, this peak may also include contributions from biomass burning particles. This is further supported by the SEM-EDX analysis of the filters from this event, which in comparison with the other events contained super-aggregates in the supermicron range, typically observed in biomass burning emissions (Chakrabarty et al., 2014), with distinct morphologies and elemental composition (shown in the Supplement in Fig. S3).
The surface area size distributions shown in Fig. 3b compare the contribution of supermicron and submicron particles to the available ambient surface area. Ice nucleation initiated on the surface of the particles, and therefore their surface area concentration is an important parameter in addition to number concentrations. Here it is clearly seen that the potential contribution of supermicron particles to ice nucleation may be significant when compared to the submicron particles, although their number concentrations were up to 2 orders of magnitude lower.

3.3 Airborne INP concentrations

The cumulative INP concentration spectra for the six dust events are shown in Fig. 4. In each event, different particle size classes are marked by different colors. Freezing was observed between 255 and 238 K, and the INP concentrations spanned 4 orders of magnitude from $10^{-1}$ to $10^3$ L$^{-1}$ of air.

A particle size dependence of the freezing temperature and INP concentration was observed. Larger particles froze at warmer temperatures with a higher number of INPs. The variation between the six size classes ranged from 1 to 2 orders of magnitude, and in some cases the smallest particles had behavior similar to the large ones. For example, in event SDS2, size classes $D_{50} = 0.6 \mu m$ and $D_{50} = 0.3 \mu m$ were less ice-active than the rest of the size classes, while in MDS, all size classes showed similar activity. As an exception, event SyDS2 showed a weaker size dependence in comparison to the other dust events and in some size classes lower INP concentrations. In comparison, in the relatively non-dusty event CSDS, the variability between the different size classes was higher, especially at lower temperatures. In Fig. 5, similar to Fig. 4, INP concentrations are presented but arranged according to the different size classes. The variability within each size class was relatively high and spans over 2 orders of magnitude; for example, at size class $D_{50} = 0.3 \mu m$ near 245 K, the INP concentration ranged from about 1 to almost $10^2$ L$^{-1}$ of air. It is clearly seen that INP concentrations in dusty conditions (SDS1, SDS2, MDS and SyDS1) were higher than in non-dusty conditions (CSDS) for the supermicron range but similar in the submicron range. Previous studies also pointed out the significant contribution of supermicron particles to the INP population. Mason et al. (2016) studied the immersion freezing abilities of airborne particles in North America and Europe and found that supermicron particles dominated the freezing, especially at relatively high temperature (258 K). Recent measurements in a coastal tropical site conducted by Ladino et al. (2019) also found high concentrations of INPs at relatively high temperatures (> 258 K) due to supermicron particles. In these studies, however, mineral dust is not expected to dominate the samples, and bioaerosol particles are thought to dominate the freezing at high temperatures (> 258 K). At lower temperatures (below 253 K), Ladino et al. (2019) suggested that mineral dust dominated the freezing. Moreover, DeMott et al. (2010) found that INP concentrations are correlated with particles > 0.5 \mu m. Other studies, such as Rosinski et al. (1986) and Huffman et al. (2013), also found that supermicron particles were responsible for most of the INP population in some cases, while when changing the freezing mode that was analyzed or the measurement meteorological conditions, their contribution was reduced. Vali (1966), in contrast, found that submicron particles dominate freezing in hail melt samples.

3.4 Size dependence of ice active site density ($n_s(T)$)

Figure 6 presents the $n_s(T)$ curves for the different dust events spanning a range of $10^6$ m$^{-2}$ at 253 K to $10^{11}$ m$^{-2}$ at 238 K. In general, $n_s(T)$ increased with particle size. The highest $n_s$ values were observed in the supermicron range $D_{50} = 5.6 \mu m$, followed by $D_{50} = 3.2$, 1.8 and 1.0 \mu m. The activity of the latter three classes was similar within measurement uncertainties. In the submicron range, stages $D_{50} = 0.6$ and 0.3 \mu m, the $n_s(T)$ values were lower than in the supermicron range and showed higher variability between the dif-
Figure 5. Airborne INP concentrations for various size classes. INP concentrations per liter of air as a function of temperature are presented in different colors for the different dust events that were sampled. Uncertainty in temperature is 0.3 K. The grey diagonal line is presented for orientation only.

Figure 6. Ice-active site density as a function of temperature, \( n_s(T) \), for airborne particles dominated by mineral dust are presented individually for each dust event. The different colors represent the different size classes that were investigated. SDS, SyDS and MDS represent Saharan, Syrian and mixed dust events, respectively (see text for more details). The linear grey line is identical in each panel to facilitate comparison.

Figure 7 displays the same \( n_s(T) \) curves as Fig. 6 but now arranged according to the different size classes. It is observed that in the supermicron range, all \( n_s(T) \) curves from the different events merge (with the exception of SyDS2), suggesting that freezing was dominated by a common component. While the freezing activity decreases with decreasing particle size, the abundance of this common component decreases with particle size. One possible explanation for this observation may be mineralogy segregation, known to occur with particle size: larger particles contain more primary minerals, such as K feldspar, whereas smaller particles contain more secondary minerals, such as clays and quartz that are common in all particle sizes (Perlwitz et al., 2015; Claquin et al., 1999). Therefore, the reduced activity in the submicron range and the higher variability between the dust events, especially at \( D_{50} = 0.3 \mu m \), may be attributed to a different mineralogical composition of the particles or to the lack of the important ice-inducing component. Alternatively, it is also possible that the submicron particles are mixed with other particle types that are more common in this size range, such as urban pollution (Li et al., 2016), and therefore freezing may not be dominated exclusively by mineral dust. Moreover, due to their larger surface-to-volume ratio, submicron particles are more sensitive to atmospheric processing than supermicron particles, which can lead to further deactivation of their ice-active
sites (Boose et al., 2016a). These considerations may explain the variability in the activity between different events. For example, we propose that the passage of SDS1 and SDS2 over the Mediterranean Sea can contribute to their reduced activity in the submicron range, while for the MDS event, a shorter and relatively direct transport path resulted in less atmospheric processing. Although speculative, these considerations may possibly explain why the freezing activity of submicron particles converged with those of the supermicron particles, but we acknowledge that further measurements are needed to confirm these suggestions.

In Fig. 7, we also compare a few relevant \( n_s(T) \) curves of standard minerals, as derived by Atkinson et al. (2013) and Niedermeier et al. (2015), together with our measured \( n_s(T) \) curves. The standard curves of K feldspar, Na–Ca feldspar and quartz were scaled to the estimated fraction of these minerals in AMD (see Table S1) and are typically used for the prediction of AMD ice-nucleating activity. A good agreement of the absolute \( n_s \) values was observed in the relevant temperature range, and the slopes of the curves were similar to those of the feldspars, especially for the supermicron range. A good agreement was also observed with the standard \( n_s(T) \) curve of quartz, suggesting that it contributes to the freezing of submicron particles in the lower temperature range. Note that the standard \( n_s(T) \) curves of clay minerals and calcite were not plotted here despite their large abundance in AMD because there was no overlap with the ice nucleation activity in this study. Only the freezing activity of the largest particles (\( D_{50} = 5.6 \mu m \)) overlapped the K feldspar prediction of Atkinson et al. (2013), indicating that this prediction possibly overestimates the freezing activity of the entire size distribution of AMD. For the particles in the size range of \( 3.2 < D_{50} < 1.0 \mu m \), there is an overlap in activity with the K feldspar prediction of Niedermeier et al. (2015) and Na–Ca feldspar of Atkinson et al. (2013). However, in all cases, the feldspar predictions overestimate the freezing activity of AMD in the submicron range.

The \( n_s(T) \) curves of SyDS2 display moderate slopes and lower ice nucleation activity in comparison with the other dust events in all size classes, except for the smallest particles with \( D_{50} = 0.3 \mu m \). As already mentioned, these particles were most likely mixed with smoke particles from biomass burning events that occurred during the same period, and the filters from this event were covered with super-aggregate particles in the supermicron size, rich with potassium, similar to particles seen in other biomass burning events (Chakrabarty et al., 2014).

### 3.5 Comparison of WISDOM and BINARY measurements for event CSDS

A complementary analysis for the CSDS event using BINARY is shown in Fig. 8. BINARY probes droplets with larger volumes and it is thus more sensitive to less common ice-nucleating sites that may not show a signal in WISDOM. In the BINARY experiments, two suspensions were tested with different dilution factors to extend our sensitivity. The higher total dust surface area per droplet sample that was investigated in the BINARY experiments, the yellow markers in Fig. 8, demonstrates the warmest freezing temperatures ranging from 255 to 246 K, and the \( n_s(T) \) values ranged from \( 10^6 \) to \( 10^9 \) m\(^{-2}\). The 1 : 10 diluted samples (purple markers) showed freezing at lower temperatures, ranging from about 251 to 244 K, with higher \( n_s(T) \) values ranging from \( 10^6 \) to \( 10^{11} \) m\(^{-2}\). In some of the dilute cases of the BINARY experiments, the data were at the limit of the background impurities (see the Supplement, Fig. S5). In order to include only data that are significantly different from the background, a criterion was set in which only data points that are larger by at least 2 standard deviations than the mean background impurities were further considered in Fig. 8. If data were below that threshold, they were considered not significant and were thus removed (e.g., the data for \( D_{50} = 0.6 \) and 0.3 \( \mu m \) for the diluted BINARY samples).

Figure 8 shows a very good agreement between the BINARY and WISDOM data because the \( n_s(T) \) curves merged nicely onto each other for each size class. Whereas BINARY was more sensitive than WISDOM to the warmer and relatively rare active sites, WISDOM detected the more common active sites in the low temperature range. Overall, the dependence of the freezing activity temperature range on the immersed surface area per droplet is well demonstrated here, whereby a reduction in the surface area of the different experiments (WISDOM < BINARY diluted < BINARY) decreased the probability to observe freezing at higher temperatures. This was also previously demonstrated in studies of standard mineral dust (Broadley et al., 2012; Marcoll et al., 2007; Reicher et al., 2018). Overall, the data shown in Fig. 8 indicate the added value when using experimental techniques of different sensitivity for the purpose of measuring the concentration and active site density of INPs in field studies (e.g., Atkinson et al., 2013; Chen et al., 2018; Harrison et al., 2018).

### 3.6 Comparison of supermicron and submicron ranges with AMD measurements and predictions

The particle surface area that was used to derive \( n_s(T) \) represents the total airborne particles that were collected for each sample, regardless of particle composition. When mineral dust dominated the composition, as in a dust event case (see, for example, Fig. S4 in the Supplement), we treat \( n_s(T) \) as representative for AMD freezing. Figure 9a compiles the \( n_s(T) \) results of AMD from a few recent studies that focused on airborne particles (albeit not size-selected) during dust events. Results from our current study, excluding the events SyDS2 and CSDS that were not dominated by AMD, are presented alongside those of Price et al. (2018) and Boose et al. (2016b). Price et al. (2018) collected airborne particles in flights west of the Sahara over the tropical Atlantic at alti-
Figure 7. Ice-active site density during dust events in different particle size classes. Dust events from the Sahara (SDS), Syrian Desert (SyDS) or both (MDS) are marked by the different colors. Data for $D_{50} = 3.2$, 1.8 and 1.0 µm of SDS no. 2 adopted from Reicher et al. (2018). Relevant standard minerals scaled to ambient values are shown: K feldspar, Na–Ca feldspar and quartz from Atkinson et al. (2013), as well as K feldspar from Niedermeier et al. (2015).
geographical sources. Due to the different behavior of submicron and supermicron particles, we also suggest that accounting for the particle size class will improve the prediction of ice cloud formation. For that purpose, we derived two basic parameterizations (Eq. 4), for supermicron and submicron particles, based on the combined AMD data (including data from this study, Price et al., 2018, and Boose et al., 2016b, and excluding SyDS2), which cover a wide range of temperatures and spread more than 5 orders of magnitude in $n_s(T)$ values. These parameterizations are the best mathematical fit for a Hill-type equation, which is normally used for fitting S-shaped data as they are observed in this compilation:

$$n_s(T) = \exp\left[\gamma_0 + a/(b + \exp[(T - 248) / c])\right] \text{[m}^{-2}]$$

where the coefficients (95% confidence bounds) for supermicron range particles are set to

$\gamma_0 = 11.4 \pm (10.97, 11.98)$, $a = 24.00 \pm (22.01, 25.99)$,

$b = 1.53 \pm (1.35, 1.70)$, and $c = 4.54 \pm (4.06, 5.02)$,

$T \in [236 \text{ K}, 266 \text{ K}]$ ($R^2 = 0.93$).

Those for the submicron range are

$\gamma_0 = 9.48 \pm (8.19, 10.76)$, $a = 23.00 \pm (20.23, 25.77)$,

$b = 1.34 \pm (1.10, 1.57)$, and $c = 7.38 \pm (5.84, 8.92)$,

$T \in [238 \text{ K}, 266 \text{ K}]$ ($R^2 = 0.93$).

Parameterizations for each individual size class can be found in Table S2 in the Supplement.

In Fig. 9b, the parameterizations derived here are presented next to the recent parameterizations of the ice nucleation of desert dust by Ullrich et al. (2017) and Niemand et al. (2012). These parameterizations are predominantly based on natural surface-collected dust samples, but they also contained one sample of AMD from Israel and agree within an order of magnitude with our supermicron data in the low temperature range (243–247 K). However, they overpredict $n_s(T)$ by more than an order of magnitude when compared to our submicron data and to the Price et al. (2018) data at warmer temperatures (247–259 K). This emphasizes that AMD ice nucleation may not be correctly represented when based on desert dust sampled from the surface, consistent with the conclusions of Boose et al. (2016b), who showed that the average freezing activity of AMD is reduced when compared to the activity of surface-collected desert dust. Key feldspar parameterizations by Atkinson et al. (2013) and Niedermeier et al. (2015) are also shown here and, as mentioned before, overpredict the freezing activity of AMD at temperatures lower than about 255 K.

4 Conclusions

We characterized the INP activity of particles collected during several mineral dust events in the eastern Mediterranean. Dust from the Sahara, the major source for atmospheric dust, together with dust from the Arabian and Syrian deserts were included. Six size classes were studied that cover both the supermicron and submicron size ranges. The INP concentrations ranged from $10^{-1} \text{ L}^{-1}$ of air in the relatively weak dust events to $10^3 \text{ L}^{-1}$ of air in the strongest event. The $n_s$ values ranged from $10^6$ to $10^{11} \text{ m}^{-2}$ in the temperature range of 238–255 K. A size dependence was observed in both the INP concentration and $n_s$ values. Larger particles were more active INPs, exhibiting higher INP concentrations and a higher number of nucleating sites per surface area at higher temperatures. Comparison between freezing results of WISDOM with BINARY showed good agreement and strengthened previous studies that observed how freezing activity could depend on the technical properties and limitations of the instrumentation used, thereby emphasizing the importance of using complementary instruments.

The dust events studied here represent a range of dust loads, different dust origins and atmospheric paths. Yet, the supermicron particles in these events exhibited similar freezing abilities. This may indicate that there is a unique component that is responsible for freezing activity, as was previously suggested (Atkinson et al., 2013; Boose et al., 2016b; Kaufmann et al., 2016; Price et al., 2018). Our measurements showed that the activity of the supermicron particles was in the range of standard particles of feldspar mineral and that the activity of the submicron particles was in the range of standard quartz. Therefore, we suggest that these may be the two most important components that dominate freezing by atmospheric mineral dust (AMD) and may therefore be important for heterogeneous ice nucleation in atmospheric clouds. The submicron particles showed higher variability.
between events, possibly due to different composition of the particles or higher sensitivity to atmospheric processing during long-range transport. In general, supermicron particles contributed the most to the INP concentration, in agreement with other previous studies (Mason et al., 2016; Huffman et al., 2013; Ladino et al., 2019). However, our current study is probably the only case in which mineral dust dominated the samples. Nevertheless, all of these studies highlight the importance of the supermicron size class of AMD for atmospheric ice nucleation.

Mineral dust is important both on a regional scale near its source region and on a global scale, since it remains ice-active even after long transport in the atmosphere and thus over considerable distances (DeMott et al., 2003b; Chou et al., 2011). With distance from the dust source, supermicron particles will settle, and submicron particles may then dominate ice nucleation on the global scale (Ryder et al., 2013; Murray et al., 2012). However, recent airborne measurements found coarse and giant particles in the vicinity of and also far from source regions (Ryder et al., 2013). Therefore, including the particle size class in INP parameterizations can improve predictions of ice formation in clouds. Moreover, information on airborne INP size distributions may be helpful in identifying the dominant INP sources (Mason et al., 2016). The overprediction of AMD freezing ability demonstrated in this study, as well as by the Atkinson et al. (2013), Niedermeier et al. (2015), Niemand et al. (2012) and Ullrich et al. (2017) parameterizations, especially for submicron particles, emphasizes the importance of future studies to better quantify the changes in the ice-nucleating properties of AMD by atmospheric processing.

Data availability. Data are available upon request to the first author.

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Author contributions. NR and YR designed the experiments, carried out the field measurements, conducted freezing experiments in WISDOM and wrote the paper. CB, LE and TK designed and performed freezing experiments in BINARY. SRR performed back trajectory analyses. NR and IKA performed the chemical analyses of filters. All authors contributed to the discussion, analysis of data and the writing of the paper.

Competing interests. The authors declare that they have no conflict of interest.

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