Hemispheric contrasts in ice formation in stratiform mixed-phase clouds: disentangling the role of aerosol and dynamics with ground-based remote sensing

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Received: 29 April 2021 – Discussion started: 17 May 2021
Revised: 30 September 2021 – Accepted: 17 October 2021 – Published: 8 December 2021

Abstract. Multi-year ground-based remote-sensing datasets were acquired with the Leipzig Aerosol and Cloud Remote Observations System (LACROS) at three sites. A highly polluted central European site (Leipzig, Germany), a polluted and strongly dust-influenced eastern Mediterranean site (Limassol, Cyprus), and a clean marine site in the southern mid-latitudes (Punta Arenas, Chile) are used to contrast ice formation in shallow stratiform liquid clouds. These unique, long-term datasets in key regions of aerosol–cloud interaction provide a deeper insight into cloud microphysics. The influence of temperature, aerosol load, boundary layer coupling, and gravity wave motion on ice formation is investigated. With respect to previous studies of regional contrasts in the properties of mixed-phase clouds, our study contributes the following new aspects: (1) sampling aerosol optical parameters as a function of temperature, the average backscatter coefficient at supercooled conditions is within a factor of 3 at all three sites. (2) Ice formation was found to be more frequent for cloud layers with cloud top temperatures above $-15^\circ C$ than indicated by prior lidar-only studies at all sites. A virtual lidar detection threshold of ice water content (IWC) needs to be considered in order to bring radar–lidar-based studies in agreement with lidar-only studies. (3) At similar temperatures, cloud layers which are coupled to the aerosol-laden boundary layer show more intense ice formation than decoupled clouds. (4) Liquid layers formed by gravity waves were found to bias the phase occurrence statistics below $-15^\circ C$.

By applying a novel gravity wave detection approach using vertical velocity observations within the liquid-dominated cloud top, wave clouds can be classified and excluded from the statistics. After considering boundary layer and gravity wave influences, Punta Arenas shows lower fractions of ice-containing clouds by 0.1 to 0.4 absolute difference at temperatures between $-24$ and $-8^\circ C$. These differences are potentially caused by the contrast in the ice-nucleating particle (INP) reservoir between the different sites.

1 Introduction

Clouds and aerosol are inseparably linked via complex pathways of interaction whose outcome manifests in the macroscopic properties of precipitation and radiation fields. In the first place, aerosol particles are required as cloud condensation nuclei (CCN) on which cloud droplets can form. On the other hand, primary ice formation in the heterogeneous freezing temperature range from 0 to about $-40^\circ C$ requires ice-nucleating particles (INPs) to be present in the aerosol reservoir. The ways in which aerosol and cloud particles interact are controlled by the dynamics and thermodynamics of the atmospheric environment, with temperature being the most important driver. Thermodynamic processes are considered to dominate the cloud microphysical properties over aerosol-related processes because the thermodynamics con-
control the amount of water vapor that is available for being transferred to either the liquid or the ice phase. This dominance makes it difficult to isolate aerosol-related effects in observations of cloud properties.

Despite the dominance of dynamics and thermodynamics, observations as well as aerosol-permitting model studies suggest a considerable influence of the aerosol conditions on the properties and evolution of clouds and precipitation (Seifert et al., 2012; Possner et al., 2017; Solomon et al., 2018; Zhang et al., 2018). Solomon et al. (2018) used high-resolution modeling of Arctic mixed-phase clouds to show that perturbations in the INP concentration dominate over changes in the CCN concentrations. Cloud chamber studies suggest that holding CCN constant, the ratio of ice to liquid water content (LWC) in the steady state is predominantly controlled by INP concentrations (Desai et al., 2019).

There is also a distinct spatiotemporal variability of the performance of weather and climate model simulations, which is attributed to the insufficient representation of aerosol–cloud–dynamics interaction in the models (Pan et al., 2016; Seinfeld et al., 2016). For instance, a less accurate treatment of the radiative balance in the southern hemispheric midlatitudes compared to their northern hemispheric counterparts was found (Trenberth and Fasullo, 2010; Grise et al., 2015). The atmosphere of the Southern Hemisphere’s midlatitudes is a unique component of the Earth’s climate system. It is the stormiest (e.g., Young, 1999) and one of the cloudiest places on Earth (Haynes et al., 2011; Naud et al., 2014), but process understanding of clouds in that region is still limited. The reported biases in the solar radiation budget are attributed to shallow supercooled liquid-topped clouds, which are insufficiently represented by current models (Bodas-Salcedo et al., 2014; Kay et al., 2016; Bodas-Salcedo et al., 2016; Kuma et al., 2020). With the frequency of liquid cloud tops too low, less short-wave radiation is reflected. Hence, the simulations show too strong a heating at the surface, leading to a bias in ocean heat uptake (Franklin et al., 2013; Hyder et al., 2018). There is an ongoing controversy about the reasons for the observed differences and prevailing model deficiencies, but indications are given that a combination of hemispheric contrasts in aerosol load and atmospheric dynamics plays a role.

Different causes for regional contrasts in the abundance of supercooled liquid cloud layers are proposed. On the one hand, a reason for the excess of supercooled liquid water in southern hemispheric cloud systems could be the inhibition of ice formation caused by the lack of ice-nucleating particles in the predominantly pristine environment of the Southern Ocean (Hamilton et al., 2014), where terrestrial sources, which are frequently considered a good source for INPs, are rare or far apart (Vergara-Temprado et al., 2017). In the heterogeneous freezing regime, suitable aerosol particles are a prerequisite for the formation of ice. Missing ice formation as a sink for cloud water, the liquid phase may be sustained for long periods of time. On the other hand, dynamical processes can lead to an enhancement of supercooled liquid water. Korolev (2007) showed that depending on the number and size of ice crystals, a threshold vertical velocity can be found, which allows for sufficient supersaturation to grow the ice as well as the liquid phase. Gravity waves have been suggested to play a role in the phase partitioning of Southern Ocean clouds (Alexander et al., 2017; Silber et al., 2020). Due to the orographic effects and the strong westerlies, gravity waves are a general feature in the vicinity of all landmasses in the middle and high latitudes of the Southern Hemisphere (Sato et al., 2012; Alexander et al., 2016). It is however also noteworthy that stronger turbulence increases the amount of ice formed in stratiform cloud layers (Bühl et al., 2019). Indications are thus given that it is necessary to also consider turbulence in studies of ice formation.

The undetermined contributions of dynamical and aerosol effects on the observed excess of supercooled liquid in the Southern Hemisphere require dedicated contrasting studies. In numerous previous studies, liquid-topped supercooled stratiform cloud layers have been proven to be suitable natural laboratories for the investigation of the relationships between aerosol properties, thermodynamics, and microphysical properties of clouds in the heterogeneous freezing regime. The temperature at which the ice formation occurs needs to be strongly constrained because the concentration of efficient INPs increases rapidly with decreasing temperature (e.g., Kanji et al., 2017). Accordingly, the amount of ice formed also increases by orders of magnitude for lower temperatures (Bühl et al., 2016). The lowest temperature in such clouds occurs on top of the liquid-dominated layer; hence the cloud top temperature (CTT) can be used to constrain the ice-formation temperature. Turbulence is usually confined to the liquid-dominated cloud top (Westbrook and Illingworth, 2013; de Boer et al., 2009), and due to the limited thickness of this layer, secondary ice formation or ice multiplication are strongly constrained (Fukuta and Takahashi, 1999; Myagkov et al., 2016). Contrasting microphysical properties observed in pristine, clean regions with observations from areas with higher aerosol load allows us to advance understanding of the impact of different aerosol loads (Choi et al., 2010; Kanitz et al., 2011; Seifert et al., 2015; Tan et al., 2014). Recent studies based on the A-Train satellite constellation suggest systematically lower ice amounts in the southern midlatitudes (Zhang et al., 2018) and a strong susceptibility to dust load (Villanueva et al., 2020). Supercooled liquid clouds are frequent over the Southern Ocean (Huang et al., 2015; Hu et al., 2010), and studies by Kanitz et al. (2011) and Choi et al. (2010) showed that – at similar temperatures – ice is formed less frequently by liquid layers in the Southern Hemisphere midlatitudes than in the Northern Hemisphere. The study of Kanitz et al. (2011) first used a ground-based lidar at Punta Arenas (53.1° S, 70.9° W, Chile) to assess the thermodynamic phase of stratiform mixed-phase clouds above the Southern Hemisphere midlatitudes. Major caveats of this study are the limited duration of the observations during...
found frequent long-lived liquid-topped clouds, also below sites with strongly contrasting aerosol conditions in order to cloud microphysics, and atmospheric dynamics from three based remote-sensing observations of aerosol properties, ern Ocean and on the coast of Antarctica. the recent past, ground-based remote-sensing observations of stratiform cloud layers of the Southern Hemispher midlatitudes based on long-term observations was not established. One reason is that, despite increased activity in the atmosphere.

Recent activities include shipborne, land-based, and aircraft campaigns targeting aerosols and clouds above the Southern Ocean between Australia and Antarctica (60 to 160° E). An overview is provided by McFarquhar et al. (2021). In terms of stratiform clouds, the large abundance of supercooled liquid water, occasionally down to −30°C, was confirmed. But, apart from the year-long lidar–radar dataset at Macquarie Island (54.6° S, 158.9° E, Australia) the cloud observations focused on austral summer.

Using a shipborne dataset, Mace and Protat (2018) also found frequent liquid-dominated clouds with low radar reflectivities and one-third of the liquid layers only observed with lidar. Comparing the observations with a Cloud–Aerosol Lidar and Infrared Pathfinder Satellite Observations (CALIPSO) dataset from Hu et al. (2010), they found an overestimation of supercooled liquid in the satellite dataset, especially strong at temperatures above −15°C. In a followup study, Mace et al. (2020) refined the CALIPSO classification scheme, leading to more frequent detections of the mixed phase, especially during wintertime and in the lower latitudes of the Southern Ocean. However, no CTT-resolved phase occurrence statistics is presented. Liquid layers in deeper clouds, observed during another shipborne campaign (McFarquhar et al., 2021; Alexander et al., 2021), could only be reproduced in regional model simulations, when INP parametrization was tuned to lower concentrations (Vignon et al., 2021). Zaremba et al. (2020) investigated airborne active remote-sensing observations of Southern Ocean clouds south of Tasmania. They also found widespread liquid cloud tops at temperatures down to −30°C. By investigating the ground-based remote-sensing dataset assembled at McMurdo (77.8° S, 166.7° E, Antarctica), Silber et al. (2018) found frequent long-lived liquid-topped clouds, also below −30°C.

Yet, a statistical analysis of the relationship between aerosol conditions, cloud vertical dynamics, and the phase partitioning in stratiform cloud layers of the Southern Hemisphere midlatitudes based on long-term observations was not established. One reason is that, despite increased activity in the recent past, ground-based remote-sensing observations of clouds and aerosol are still sparsely distributed in the Southern Ocean and on the coast of Antarctica.

The goal of this study is to analyze long-term groundbased remote-sensing observations of aerosol properties, cloud microphysics, and atmospheric dynamics from three sites with strongly contrasting aerosol conditions in order to attribute ice formation in the heterogeneous freezing regime to atmospheric dynamics and aerosol conditions.

For that contrasting approach, we utilized the recent campaigns of the Leipzig Aerosol and Cloud Remote Observations System (LACROS) at Leipzig (51.4° N, 12.4° E, Germany), Limassol (34.7° N, 33.0° E, Cyprus), and Punta Arenas (53.1° S, 70.9° W, Chile), which provide datasets that cover the aerosol conditions of a continental northern hemispheric background site, a hotspot of mineral dust, and the marine-dominated pristine Southern Ocean, respectively. Hence, these datasets collected with a single set of groundbased remote-sensing instrumentation provide an ideal basis for contrasting studies. The broad variety of instruments covers the decisive properties of aerosols, dynamics, clouds, and precipitation for a more comprehensive picture of aerosol–cloud interaction. The observations at Punta Arenas provide the first multi-year dataset of synergistic ground-based remote-sensing observations in the western half of the Southern Ocean and allows us to contextualize prior findings.

The paper is structured as follows: the instruments and field campaigns are described first (Sect. 2.1), followed by the retrievals of aerosol properties (Sect. 2.2), the automated cloud selection algorithm (Sect. 2.4), and the characterization of vertical velocity (Sect. 2.5). The results are presented in Sect. 3, including the lidar-derived average profiles of aerosol optical properties (Sect. 3.1) and frequency of iceformation statistics with special emphasis on instrument sensitivity (Sect. 3.2.1). Boundary layer coupling and vertical dynamics are discussed in Sects. 3.2.2 and 3.2.3, respectively. The amount and efficiency of ice production is assessed in Sect. 3.2.4. The study concludes with a discussion of the contrast identified in aerosol load and stratiform cloud properties (Sect. 4), followed by a summary and outlook (Sect. 5).

2 Data and methods

This section introduces the datasets and methods used in the remainder of this study, starting with the campaigns and instrumentation (Sect. 2.1), followed by a short description of the retrievals used (Sects. 2.2 and 2.3), and finally the methods for automated selection of shallow stratiform clouds (Sect. 2.4) and characterization of vertical motion (Sect. 2.5).

2.1 Datasets from Leipzig, Limassol, and Punta Arenas

The basis of the observational datasets is the Leipzig Aerosol and Cloud Remote Observations System (LACROS), the mobile ground-based remote-sensing supersite of the Leibniz Institute for Tropospheric Research (TROPOS), Leipzig, Germany. The instrumentation used for the synergistic approaches applied in this study comprises a MIRA-35 35GHz scanning cloud radar, a PollyXT multi-wavelength Raman polarization lidar, a Streamline XR 1.5 µm scanning

https://doi.org/10.5194/acp-21-17969-2021 Atmos. Chem. Phys., 21, 17969–17994, 2021
Doppler lidar, a HATPRO 14-channel microwave radiometer, a 1064 nm ceilometer, an optical disdrometer, and radiation sensors. The main properties of the sensors are summarized in Table 1.

LACROS was established in 2011 and was first introduced by Bühl et al. (2013b). After the initial setup phase, the basic set of instrumentation of LACROS has been kept unchanged since the year 2014. Hence, this comparative study uses data from the 6-year period since then. During that period, LACROS was, besides several 4–8 week short-term deployments at other sites, stationed at Leipzig, Limassol, and Punta Arenas.

Figure 1 provides an overview of the campaigns and instrument availability. From October 2016 to March 2018 LACROS was located at Limassol for the Cyprus Clouds, Aerosol, and Rain Experiment (CyCARE). In the context of the present study, this location in the eastern Mediterranean in the middle of the dust belt serves as a reference for a highly aerosol-laden environment. First results of the CyCARE campaign are described by Ansmann et al. (2019). For the second long-term campaign, LACROS was deployed at Punta Arenas for the 2-year Dynamics, Aerosol, Clouds, And Precipitation Observations in the Pristine Environment of the Southern Ocean (DACAPO-PESO) field campaign since November 2018. Located at 53.1°S, Punta Arenas is in the midst of the Southern Ocean, farther south than any other continental land mass or major island apart Antarctica. Under prevailing westerly flow, the next land mass upwind is more than 8000 km away and significantly to the north. Hence, the free tropospheric aerosol is predominantly of marine origin with low concentrations overall. First results of the DACAPO-PESO campaign regarding events of significant aerosol load (Ohneiser et al., 2020; Floutsi et al., 2021), liquid cloud microphysics (Jimenez et al., 2020), and integrated water vapor observations (Bromwich et al., 2020) are already published. Observations at Leipzig were performed before October 2016 and between the deployments to Limassol and Punta Arenas. Due to the location in central Europe, the Leipzig dataset serves as a reference for continental background in the Northern Hemisphere.

Data coverage for each of the sites is summarized in Table 2. Gaps in the observations were usually caused by periods of short-term deployments at other sites, failure of primary instruments, or power cuts (see also Fig. 1). Coverage with the core instrumentation is generally well above 80%, the only exception being Doppler lidar and PollyXT at Leipzig.

2.2 Lidar-based aerosol statistics

As a proxy for the aerosol conditions at the three measurement sites, we used vertically resolved observations from the PollyXT lidar system (Althausen et al., 2009; Engelmann et al., 2016). Quantities of interest are the aerosol backscatter coefficient, the extinction coefficient, the particle linear depolarization ratio, and the cloud-relevant concentration of INPs. Their retrieval is explained below.

The basis for the statistical analysis is profiles of particle backscatter coefficient $\beta_p$ at 532 nm wavelength. The profiles are computed with the Klett method (Fernald, 1984) by the PollyNET retrieval whenever atmospheric conditions are suitable (Baars et al., 2016, 2017; Yin and Baars, 2021). The PollyNET retrieval chain also ensures a homogenized analysis of the data from the three different PollyXT instruments which were utilized in the frame of this study (see Fig. 1 and Engelmann et al., 2016; Baars et al., 2016).

Profiles of the particle linear depolarization ratio (hereafter referred to as particle depolarization ratio) are only calculated when the ratio of molecular backscatter coefficient to $\beta_p$ is below a value of 18. Additionally, any particle depolarization ratios larger than 0.7 are masked, as they are indications for noise artifacts in the cross-polarized signal component. All profiles are then filtered with the co-located Cloudnet target classification (see Sect. 2.3) to exclude clouds, especially optically thin ice clouds that are only clearly classified by the cloud radar. Finally, a manual screening excluded fragments of thin liquid clouds, which would otherwise artificially increase $\beta_p$. For the averages, the optical data of each retrieved profile are binned to vertical intervals of 200 m or 3 K.

The derived average optical properties can be used to estimate aerosol microphysical properties, such as concentrations of ice-nucleating particles (INPs). This is an important step in order to evaluate the datasets of the three sites with respect to contrasts in the potential contribution of aerosol effects on heterogeneous ice-formation efficiency. Conversion from optical properties as observed by lidar to microphysical aerosol properties is based on the parametrizations described by Mamouri and Ansmann (2016). By means of this approach, the lidar-measured aerosol extinction coefficient is converted to the number and surface concentration $N_{500}$ and $S_{500}$ of aerosol particles larger than 500 nm in diameter. These quantities are applied in available in situ-based parametrizations for the retrieval of INP concentrations. The prerequisite for the retrieval is a correct aerosol typing, as different types of particles differ by orders of magnitude in their ice-forming efficiency. In order to do so, the average backscatter profile is separated into the categories marine, continental, and mineral dust, based on air mass source (see Appendix B and Radenz et al., 2021) and particle depolarization ratio (one-step POLIPHON; Mamouri and Ansmann, 2017). The average extinction is calculated from the profiles of $\beta_p$ by assuming a typical lidar ratio of 20 sr for marine, 50 sr for continental, and 45 sr for dust aerosol (Müller et al., 2007; Baars et al., 2017; Bohlmann et al., 2018). In a next step, the extinction coefficient is converted to $N_{500}$ and $S_{500}$, using sun-photometer-based conversion factors (Mamouri and Ansmann, 2016). Then, the above-mentioned INP parametrizations are applied for each aerosol class, such as DeMott et al. (2015) for mineral dust, DeMott...
Table 1. Specifications of the LACROS instruments and measured quantities used in this study.

<table>
<thead>
<tr>
<th>Instrument (reference)</th>
<th>Frequency $\nu$ Wavelength $\lambda$</th>
<th>Measured quantity</th>
<th>Temporal resolution</th>
<th>Vertical range</th>
<th>Vertical resolution</th>
</tr>
</thead>
<tbody>
<tr>
<td>Doppler cloud radar</td>
<td>$\nu = 35$ GHz</td>
<td>Radar reflectivity factor</td>
<td>3.5 s</td>
<td>150–13000 m</td>
<td>30 m</td>
</tr>
<tr>
<td>METEK MIRA-35 (Görsdorf et al., 2015)</td>
<td></td>
<td>Vertical velocity</td>
<td>3.5 s</td>
<td>150–13000 m</td>
<td>30 m</td>
</tr>
<tr>
<td>Raman polarization lidar PollyXT (Engelmann et al., 2016)</td>
<td>$\lambda = 355, 532, 1064$ nm</td>
<td>Attenuated backscatter coeff.</td>
<td>30 s</td>
<td>100–15000 m</td>
<td>7.5 m</td>
</tr>
<tr>
<td>Microwave radiometer</td>
<td>$\nu = 22.24–31.4$ GHz</td>
<td>Brightness temperatures</td>
<td>1 s</td>
<td>column integral</td>
<td></td>
</tr>
<tr>
<td>RPG HATPRO-G2 (Rose et al., 2005)</td>
<td></td>
<td>Brightness temperatures</td>
<td>1 s</td>
<td>column integral</td>
<td></td>
</tr>
<tr>
<td>Doppler lidar HALO</td>
<td>$\lambda = 1.5$ $\mu$m</td>
<td>Attenuated backscatter coeff.</td>
<td>2 s</td>
<td>48–12000 m</td>
<td>48 m</td>
</tr>
<tr>
<td>Photonics Streamline XR (Pearson et al., 2009)</td>
<td></td>
<td>Vertical velocity</td>
<td>2 s</td>
<td>48–12000 m</td>
<td>48 m</td>
</tr>
<tr>
<td>Ceilometer Jenoptik CHM15kx</td>
<td>$\lambda = 1064$ nm</td>
<td>Attenuated backscatter coeff.</td>
<td>30 s</td>
<td>15–15300 m</td>
<td>15 m</td>
</tr>
<tr>
<td>Optical disdrometer Ott Parsivel2 (Lößler-Mang and Joss, 2000)</td>
<td>$\lambda = 650$ nm</td>
<td>Hydrometeor size distribution</td>
<td>30 s</td>
<td>4 m</td>
<td>–</td>
</tr>
<tr>
<td>Sun photometer Cimel CE318-T (Barreto et al., 2016)</td>
<td>$\lambda = 340–1064$ nm</td>
<td>Aerosol optical thickness</td>
<td>variable</td>
<td>column integral</td>
<td></td>
</tr>
</tbody>
</table>

Figure 1. Overview of instrument availability and campaign durations of the LACROS deployments covered in this study. Characteristics of the three different PollyXT instruments used at Leipzig and named IFT, Arielle, and LACROS are described in Engelmann et al. (2016).

et al. (2010) for continental aerosol, and McCluskey et al. (2018b, a) for marine aerosol.

2.3 Cloudnet processing

Synergies between lidar, cloud radar, microwave radiometer, and meteorological data are utilized for the determination of cloud macro- and microphysical properties and as a basis for the stratiform cloud identification scheme. State-of-the-art routines for achieving this synergy are comprised in the Cloudnet retrieval (Illingworth et al., 2007). Cloudnet re-grids the observations to a common resolution (30 s and 31.18 m, determined by the vertical resolution of MIRA-35 with a pulse length of 208 ns) and provides products, such as a target classification and microphysical cloud properties. Re-gridding is a crucial step, as each of the instruments has different temporal and vertical resolution (Table 1). Regular scans of the MIRA-35 cloud radar (hourly) and the Streamline Doppler lidar (twice per hour) are not used within the Cloudnet processing scheme. Profiles of temperature, pres-
Table 2. Overview of location, data availability, climate, aerosol load, and related studies for the LACROS datasets used.

<table>
<thead>
<tr>
<th>Location</th>
<th>Leipzig, Germany</th>
<th>Limassol, Cyprus</th>
<th>Punta Arenas, Chile</th>
</tr>
</thead>
<tbody>
<tr>
<td>Station altitude</td>
<td>51.4°N, 12.4°E</td>
<td>34.7°N, 33.0°E</td>
<td>53.1°S, 70.9°W</td>
</tr>
<tr>
<td>Campaign name</td>
<td>125 m.a.s.l.</td>
<td>11 m.a.s.l.</td>
<td>9 m.a.s.l.</td>
</tr>
<tr>
<td>Duration</td>
<td>976 d</td>
<td>524 d</td>
<td>765 d</td>
</tr>
<tr>
<td>Cloudnet data</td>
<td>771 d</td>
<td>520 d</td>
<td>674 d</td>
</tr>
<tr>
<td>Doppler lidar data</td>
<td>513 d</td>
<td>523 d</td>
<td>717 d</td>
</tr>
<tr>
<td>PollyXT data</td>
<td>633 d</td>
<td>460 d</td>
<td>702 d</td>
</tr>
<tr>
<td>Climate</td>
<td>Northern midlatitudes</td>
<td>Northern subtropics</td>
<td>Southern midlatitudes</td>
</tr>
<tr>
<td>Typical aerosol load</td>
<td>Continental background, occasionally dust</td>
<td>Dust, marine, continental</td>
<td>Marine, occasionally continental</td>
</tr>
<tr>
<td></td>
<td>Seifert et al. (2010)</td>
<td></td>
<td>Bromwich et al. (2020)</td>
</tr>
<tr>
<td></td>
<td>Bühl et al. (2013a)</td>
<td></td>
<td>Jimenez et al. (2020)</td>
</tr>
<tr>
<td></td>
<td>Bühl et al. (2016)</td>
<td></td>
<td>Floutsi et al. (2021)</td>
</tr>
</tbody>
</table>

2.4 Automated cloud selection and characterization

While the Cloudnet algorithm provides a pixel-by-pixel classification of cloud phase and microphysical properties, information on temporal coherence and cloud evolution is not readily available. Hence, an automatic reproducible filtering algorithm for the selection of targeted stratiform, supercooled cloud systems is necessary. Based on the data cube LARDA\textsuperscript{3} (Bühl et al., 2021), the approach of Bühl et al. (2016) is implemented into an automated selection algorithm.

An example of the Cloudnet processing of measurement data and the application of the cloud selection scheme is shown in Fig. 2. Starting with a profile of the Cloudnet target classification mask (Hogan and O’Connor, 2004), consecutive pixels classified as cloud pixels (liquid droplets, ice, ice and supercooled droplets) are grouped together and defined as features. In case similar types of hydrometeors were observed in matching heights, single features in neighboring time steps are connected to coherent cloud cases. For the analysis, the cloud cases are filtered for shallow stratiform clouds, which are liquid-topped and either have an ice virga or not (rectangles in Fig. 2d). An overview of the microphysical parameters sampled for each cloud case is provided in Table 3. It is assumed that if ice is formed in a liquid layer, it also sediments out of the cloud. This is required, as the signal in the top layer is dominated by return from liquid droplets, and Cloudnet provides no reliable mixed-phase classification there.

To pinpoint the potential effects of aerosol load, thermodynamic and dynamic drivers of ice formation have to be constrained. This is especially important, as the sites are situated in different climate zones. We presume for our study that thin stratiform clouds serve as a natural laboratory, with only a limited number of microphysical processes being possible.

The cloud cases are filtered for a length of more than 20 min and smooth cloud top heights (standard deviation < 150 m) to exclude convective clouds. Laboratory studies of Fukuta and Takahashi (1999) and the cloud radar observations by Myagkov et al. (2016) indicate that the average thickness of the liquid-dominated cloud top layer has to be less than 350 m in order to restrict the dataset to the regime of pristine ice formation and avoid strong effects of secondary ice formation. Seeding by ice clouds above is avoided by excluding clouds with ice pixel above the liquid layer. Only cloud cases with top temperatures between −38 and 0 °C are considered for the statistics. For the phase occurrence frequency statistics, a cloud is classified as ice-producing if ice pixels were observed 180 m below the liquid-dominated layer within at least 5% of the duration. On the other hand, cloud cases are classified as liquid-only if ice pixels are observed to sediment out of the liquid-dominated layer less than 5% of the time. This threshold is simpler than the decision tree used by Bühl et al. (2013a) but provides similar results. The resulting fraction is weighted by the length of the cloud. Sim...
Figure 2. Stratiform liquid-topped clouds as observed with LACROS. (a) Cloud radar reflectivity overlaid with the temperature, (b) lidar attenuated backscatter, (c) Doppler lidar vertical velocity, and (d) Cloudnet target categorization on 28 November 2018, 00:20 to 09:00 UTC. The temperature information in (a) is obtained from ECMWF IFS model data. Red rectangles in (d) indicate the automatically detected liquid-topped shallow stratiform clouds. The inset in (c) shows the autocorrelation function of the vertical velocities in the liquid-dominated layer of each cloud. The height of the velocity sampling is indicated with dashed lines. Further details are provided in Sect. 2.5.

Table 3. Parameters sampled from the automatically identified cloud cases. A more detailed description of the automated spatiotemporal selection method is provided in Bühl et al. (2016). The IWC and ice extinction are retrieved with the parametrization of Hogan et al. (2006).

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Time series</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cloud top temperature</td>
<td>yes</td>
<td>Temperature of the topmost liquid pixel</td>
</tr>
<tr>
<td>Cloud top height</td>
<td>yes</td>
<td>Geometrical height of the topmost liquid pixel</td>
</tr>
<tr>
<td>Liquid layer thickness</td>
<td>yes</td>
<td>Geometrical depth of the liquid-dominated cloud top layer</td>
</tr>
<tr>
<td>Phase classification</td>
<td>no</td>
<td>Ice pixel observed below liquid-dominated layer &gt; 5% of duration</td>
</tr>
<tr>
<td>In-cloud vertical velocity</td>
<td>yes</td>
<td>Doppler lidar vertical velocity in the middle of the cloud top layer</td>
</tr>
<tr>
<td>LWC</td>
<td>yes</td>
<td>LWC of the liquid-dominated layer</td>
</tr>
<tr>
<td>Reflectivity virga</td>
<td>yes</td>
<td>Radar reflectivity 180 m below the base of the liquid-dominated layer</td>
</tr>
<tr>
<td>IWC in virga</td>
<td>yes</td>
<td>Reflectivity-derived IWC 180 m below the base of the liquid-dominated layer</td>
</tr>
<tr>
<td>Extinction in virga</td>
<td>yes</td>
<td>Optical extinction of ice at 180 m below the base of the liquid-dominated layer</td>
</tr>
<tr>
<td>Ice-to-liquid content ratio</td>
<td>yes</td>
<td>Ratio of IWC virga to LWC in the liquid-dominated layer</td>
</tr>
</tbody>
</table>

Similarly to Bühl et al. (2016), the properties of the ice in the virga are taken 180 m below the liquid-dominated cloud top layer to avoid uncertainties in the liquid base estimate and sublimation within the virga. The ice water content (IWC) is derived from the radar reflectivity and the temperature using the Hogan et al. (2006) retrieval. With this automated cloud selection scheme, multi-year datasets can be analyzed using objective criteria while yielding statistics similar to manual cloud selection (e.g., Seifert et al., 2010; Kanitz et al., 2011; Seifert et al., 2015).

2.5 Gravity wave detection

In order to pinpoint effects of aerosol and dynamics on the phase partitioning in the stratiform cloud dataset, an approach is required to assign vertical motion regimes to each cloud case. Here we focus on the temporal structure of vertical velocity, to constrain the dynamics forcing on a cloud. Usually, shallow clouds are characterized by a fully developed turbulence in the liquid-dominated cloud top (Bühl et al., 2019), where the vertical motion is driven by cloud top cooling (e.g., Shao et al., 1997; Fang et al., 2014; Sim-
However, orographic gravity waves can trigger vertical motion and associated cloud formation, as well. Microphysical processes in these wave clouds are governed by large-scale dynamics, where vigorous up- and downdrafts may appear stationary. Due to this dynamics, the mixed phase and the ice phase are horizontally separated, with the liquid drops predominantly in the ascending branch and the ice particles in the descending branch (Heymsfield and Miloshevich, 1993; Baker and Lawson, 2006). The properties of the horizontal wind field determine the regions of the up- and downdraft in such orographic clouds. Observing these clouds by stationary ground-based remote sensing might thus not sample the full horizontal extent of the cloud, which causes a misclassification of the cloud case in terms of liquid-only and ice-producing. An illustrative example of such a wave cloud is shown in Appendix A.

Also, the flow is highly laminar, as opposed to the confined, fully developed turbulence found in layered mixed-phase clouds. Cloud microphysics in wave conditions cannot directly be compared to layered clouds. For clouds in the heterogeneous freezing regime, Cotton and Field (2002) found that only rapid evaporation freezing once the downdraft commences could explain their observations, where evaporation freezing can be better characterized as inside-out contact freezing of shrinking particles (Durant, 2005). A more recent study by Field et al. (2012) found that condensation and immersion freezing are needed together with deposition and evaporation freezing to explain their aircraft observations of ice formation in wave clouds.

On the other hand, the frequent occurrence of atmospheric gravity waves in a specific region might increase the frequency of thermodynamic conditions that favor the presence of a sustained liquid phase. As Korolev (2007) demonstrates, long-lasting steady updrafts are required in order to make the liquid phase dominate over the ice phase. Probing an observational dataset for the presence of long-lasting updrafts could therefore provide a hint on the atmospheric gravity waves in the occurrence of enhanced concentration of supercooled liquid.

Turbulence properties of the liquid layer in the clouds subject to our study can be derived from the vertical velocity observation by Doppler lidar (Bühl et al., 2019). The small size of droplets in the liquid-dominated cloud-top layer and their negligible terminal velocity makes them tracers of air motion. From the Doppler lidar observations, the vertical velocity is sampled at the average height of the center of the layer identified as liquid-containing in the Cloudnet classification. The height of this sampling is indicated by dashed lines in Figs. 2c and A1d. The temporal resolution of the resulting time series is 2 s, i.e., equal to the Doppler lidar raw data. This time series is then used to get insights into the turbulent properties of the cloud-top layer. To separate both regimes in the large dataset, an characteristic parameter is needed. The autocorrelation function was found to be a pragmatic choice for such a parameter.

The autocorrelation function \( \Psi(\tau) \) for a time series of vertical velocities \( v_t \) is defined as

\[
\Psi(\tau) = \sum_{t} v_t v_{t+\tau},
\]

with the temporal shift \( \tau \) and the vertical velocity \( v \) at time \( t \). To compare different cloud cases, the autocorrelation function \( \Psi(\tau) \) is normalized with \( \Psi(0) \). The temporal shift \( \tau \) from the observations is converted into a horizontal shift or autocorrelation length \( l \) with the horizontal wind velocity \( v_{\text{hor}} \) from the model analysis included in the Cloudnet product:

\[
\Psi(l) = \Psi(\tau)v_{\text{hor}}.
\]

Similarly, the vertical-velocity spectral power density is calculated by a fast Fourier transform of the vertical-velocity time series. High autocorrelation coefficients for large shifts and low power density are indications for wave-driven, low-turbulent flow. The inset in Fig. 2c shows the autocorrelation function for each of the identified cloud cases. All of them are weakly affected by gravity waves, but small-scale turbulence dominates. In contrast, the wave cloud in Fig. A1d shows high autocorrelation coefficients for longer shifts. As a characteristic value of the autocorrelation function, the shift at which the coefficient drops below 0.8 was chosen after visually inspecting the whole dataset. When the ice-formation frequency statistics is investigated for the influence of gravity waves (Sect. 3.2.3), the shift threshold is reduced step by step. As the autocorrelation function is generally decreasing, a lower characteristic value than 0.8 would result in larger shift thresholds. For shifts larger than about 500 m, random fluctuations appear for clouds with a rapid drop in autocorrelation coefficients. Thus, 0.8 is a robust choice for the characteristic value.

3 Results

In this section, the thermodynamic phase partitioning and quantitative ice mass production in ice-forming shallow cloud layers are presented and compared between all measurement sites under study. First, the average profiles of aerosol optical and microphysical properties are presented. Then, instrumental detection thresholds, boundary layer effects, and gravity wave activity are all analyzed as potential influencing factors on the retrieved ice-formation characteristics.

3.1 Aerosol conditions at Leipzig, Limassol, and Punta Arenas

To provide a general insight into the aerosol conditions at the three sites, the average optical and microphysical aerosol
properties as derived from the lidar observations (Sect. 2.2) are shown in Figs. 3 and 4. The impact of aerosol on clouds is controlled more strongly by temperature than geometrical height; hence the averages are also calculated with temperature as a vertical coordinate.

3.1.1 Optical properties

The average aerosol backscatter coefficient $\beta_p$ at 532 nm and the particle depolarization ratio derived from the PollyXT observations with the Klett method (see Sect. 2.2) are investigated in this section. The evaluation of the air mass sources is covered more quantitatively in the next section and Appendix B.

The central European site of Leipzig is characterized by predominantly continental aerosol mixed with anthropogenic pollution (Baars et al., 2016). Long-range transport of dust may occur periodically, especially during spring and autumn (Ansmann et al., 2003), as well as lofted smoke layers from wildfires (Haarig et al., 2018; Baars et al., 2021). Aerosol optical thickness (AOT) is derived from sun-photometer observations with the Klett method (see Sect. 2.2) are investigated in this section. The evaluation of the air mass sources is covered more quantitatively in the next section and Appendix B.

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Limassol is characterized by a distinct dry season with no precipitation and very few clouds during the summer. Generally, Limassol is frequently affected by aerosol transport from Africa, the Middle East, and Europe, with aerosol characteristics including dust (mineral and soil), marine (organics and sea salt), and anthropogenic pollution as well as mixtures of these (Nisantzi et al., 2015). Mean AOT at 500 nm is 0.198 at TROPOS, Leipzig, between 2014 and 2018. When only the periods with co-located PollyXT observations (used in this study) are considered, the AOT is 0.216. Mean $\beta_p$ at 532 nm drops below 0.2 Mm$^{-1}$ sr$^{-1}$ only above 4 km height, which corresponds to an extinction coefficient of 1.0 Mm$^{-1}$, assuming continental aerosol conditions and a corresponding lidar ratio of 50 sr.

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The aerosol load at Punta Arenas can be separated into two distinct layers, with an aerosol-rich boundary layer and pristine conditions aloft. The free troposphere is dominated by marine aerosol from the Southern Ocean and was reported to show no changes compared to pre-industrial conditions (Hamilton et al., 2014). Nevertheless, events of aerosol long-range transport also occur occasionally (Foth et al., 2019; Floutsi et al., 2021). The boundary layer is laden with a mixture of marine and continental aerosol, as Punta Arenas is located 230 km inland from the Pacific coast. Mean AOT at 500 nm is 0.055 during the whole campaign but drops to 0.047 when excluding the period of long-range wildfire smoke transport in early 2020 (Ohneiser et al., 2020). Average boundary layer height is around 1.5 km (Foth et al., 2019) with negligible $\beta_p$ above 2.0 km height (90% percentile of $\beta_p$ at 532 nm dropping below 0.2 Mm$^{-1}$ sr$^{-1}$). Comparing the backscatter at Punta Arenas and Limassol, the 90% percentile at Punta Arenas is more than 30% below the mean of Limassol and Leipzig at similar heights above ground. Both European sites show quite some variability, as well, with the 90% percentile $\beta_p$ twice as large as the mean.

At temperatures above 5°C, $\beta_p$ shows the strongest difference between the three sites (Fig. 3b), also explaining the larger AOT at Limassol and Leipzig. Between 0°C and +12°C, the mean $\beta_p$ at Limassol and Punta Arenas is almost equal. This counterintuitive behavior can be explained by different temperature regimes. The −5°C isotherm at Punta Arenas varies between 1.1 and 3.3 km height, whereas at Limassol it varies between 3.0 and 4.8 km height and at Leipzig between 2.1 and 4.8 km. At Punta Arenas, corresponding to the height of 2.0 km, only very low values of $\beta_p$ are observed at heights below a temperature of −10°C. The pronounced decrease of backscatter at Leipzig and Limassol is observed at slightly higher temperatures, which agrees with the, on average, higher boundary layer temperature there. Assuming typical lidar ratios of 50 (continental), 45 (dust), and 20 sr (marine) at Leipzig, Limassol, and Punta Arenas, respectively, typical aerosol extinction coefficients can be estimated from the median $\beta_p$ profile (Sect. 2.2). As shown in Fig. 3c, the extinction coefficients at Punta Arenas are a factor 2–3 lower than over Limassol for temperatures below −10°C. This difference decreases to a factor of 1.5 for slightly supercooled conditions of above −10°C. Comparing Punta Arenas and Leipzig, the extinction is a factor of 3–4 higher at the latter site for all temperatures below freezing. These optical properties serve as a proxy for the background reservoir of aerosol particles that could act as cloud condensation nuclei and ice-nucleating particles in the free troposphere.

The particle depolarization ratio provides hints on the aerosol particle types (Fig. 3d). At Leipzig, the depolarization ratio is approximately 0.05 at all temperatures, typical for a continental aerosol with a slight contribution of mineral dust. The depolarization ratio at Limassol is bimodal, with one peak above 20°C, a second one at −14°C, and a minimum at 4°C. The first peak can be ascribed to mineral dust in the boundary layer and originating from local sources. The second one at +8°C is caused by long-range transport of mineral dust. At Punta Arenas, a maximum at +8°C is caused by occasional events of dried sea salt aerosol at the top of the atmospheric boundary layer (Haarig et al., 2017; Bohlmann et al., 2018). Going to lower temperatures, the depolarization ratio has a minimum of 0.01 at −20°C and a slight increase afterwards.

3.1.2 Contrasts in INP concentrations

An important aspect for the discussion of microphysical contrasts in clouds over the three sites is how the differences in aerosol optical profiles are linked to contrasts in the INP
Figure 3. Profiles of average aerosol optical properties at 532 nm wavelength derived by the PollyNET retrieval using the Klett method. (a) particle backscatter coefficient over height and (b) temperature for Leipzig, Limassol, and Punta Arenas. (c) Extinction coefficient and (d) particle depolarization ratio for the same locations. For the extinction coefficient, only the median values based on the typical lidar ratio at each site are shown. Variability caused by different aerosol mixtures (see Sect. 3.1.2) is denoted by shading.

Figure 4. Average INP concentrations derived from optical properties for Leipzig (green), Limassol (orange), and Punta Arenas (blue). Shading shows the spread covered by different aerosol compositions, similar to Fig. 3c. Temperature is used as a vertical coordinate. The fraction of continental aerosol is abbreviated “conti.”. The used parametrizations are described in Sect. 2.2.

An estimate of average INP concentrations covering the whole troposphere can be derived with the parametrizations described in Sect. 2.2. The decision for the aerosol typing required in the INP retrieval is based on the particle depolarization ratio and air mass source estimates. The air mass source methodology is based on Radenz et al. (2021) and shown in Appendix B.

At Punta Arenas marine sources are by far the most frequent, contributing 90% to the residence time throughout the troposphere and peaking 95% at 2.7 km height. Only below 2.0 km height (or above −10°C) do local terrestrial sources make up to 10% of the air mass. Above 5 km height (below −20°C) sparsely vegetated areas in Australia contribute up to 4% to the air mass source. Mean particle depolarization ratios below 0.02 in the free troposphere exclude frequent presence of mineral dust (Fig. 3d). Contrarily, at Limassol, air masses with a marine source contribute 50% to 80% to the mixture. But these air masses are not pristine marine, as the eastern Mediterranean is enclosed by strongly populated landmasses. Below 5 km height (above −16°C), continental Europe is the second strongest source, responsible for up to 45% at 0.5 km height. In the upper troposphere, barren ground from the Sahara is the strongest terrestrial source, with contributions around 15%. POLIPHON (see Sect. 2.2) shows a peak at −12°C with mean dust fractions of 0.3 (90% percentile 0.87). Hence, the backscatter is divided into dust and non-dust according to the dust fraction. The non-dust portion is then split up into continental and marine, with 40% contribution of the continent above −12°C and 20% below. The aerosol mixture at Leipzig is dominated by continental aerosol, with an average dust fraction of 0.1.

The derived INP concentrations are depicted in Fig. 4. Note that the INP concentration at a certain atmospheric temperature (as a proxy for height) is shown based on the optical properties, aerosol types, and the parametrization, not a freezing spectrum obtained from sampling a single air parcel and varying temperature.

At temperatures above −10°C, average INP concentrations between $4 \times 10^{-3}$ and $6 \times 10^{-2} L^{-1}$ can be expected at all three locations. With decreasing temperatures, the concentration increases to $0.1-1 L^{-1}$ at −25°C at the northern hemispheric sites of Leipzig and Limassol. A strong in-
crease of ice-nucleating efficiency with decreasing temperature is counterbalanced by a decreasing aerosol concentration at lower temperatures, which are tied to greater heights. INP concentrations at Punta Arenas are strongly controlled by the fraction of continental aerosol in the free troposphere. Continental sources can contribute up to 1% between by the fraction of continental aerosol in the free troposphere. INP concentrations at Punta Arenas are strongly controlled by the fraction of continental aerosol in the free troposphere. Equally low concentrations of INPs were also found by McFarquhar et al. (2021) in the remote Southern Ocean, south of Australia. As soon as very small fractions of continental aerosol are present, the INP concentration increases significantly. However, even if a few percent of continental aerosol are assumed, throughout the heterogeneous freezing regime, INP concentrations remain a factor of 2–6 lower at Punta Arenas, compared to Leipzig and Limassol. Due to the absence of suitable remote-sensing or in situ measurements, the actual contribution of continental aerosol to the free tropospheric aerosol load over Punta Arenas cannot be obtained. The range given in Fig. 4 is thus a solid estimate of the expectable range of possible INP concentrations. At temperatures above −10°C, which refers to a height range that is frequently within the boundary layer, INP concentrations at Punta Arenas are high and within an order of magnitude of the concentrations at Leipzig and Limassol at the same temperature. Concentrations that are equally high as retrieved from the remote-sensing observations above −10°C were also found in situ at the upwind hilltop station on Cerro Mirador at an altitude of 622 m a.s.l. (Gong et al., 2021).

3.2 Mixed-phase stratiform cloud properties

The automated stratiform cloud selection algorithm introduced in Sect. 2.4 is applied to the Cloudnet datasets of all three locations. An overview over identified clouds and temperature-resolved phase occurrence frequency is provided in Sect. 3.2.1. Afterwards, it is shown how boundary layer aerosol load (Sect. 3.2.2) and gravity waves affect the phase occurrence (Sect. 3.2.3). In Sect. 3.2.4 contrasts of ice content in the virga are briefly discussed.

3.2.1 Phase occurrence frequency and detection thresholds

Figure 5 shows all the cloud cases that fulfill the selection criteria introduced in Sect. 2.4. The time share of ice virga sedimenting out of the liquid-dominated top is given by the fraction of profiles for which ice was classified below the liquid layer. Values range from 0 (no ice virga observed at all) to 1 (ice virga observed all the time). At all three locations, the fraction of profiles (i.e., fraction of time) in which a virga originated from the liquid layer increases with decreasing temperature. Similarly, the ice-to-liquid content ratio (Bühl et al., 2016) increases for lower temperatures. Liquid-only clouds (fraction of profiles where no ice was produced from the liquid layer) were observed at Leipzig and Limassol down to −16°C, whereas at Punta Arenas, such clouds were observed even at temperatures as low as −38°C.

First, the temperature-resolved fraction of occurrence of ice-forming clouds is depicted in Fig. 6. For the phase occurrence frequency, a cloud is classified as ice-producing, if an ice virga was observed during at least 5% of the duration of a cloud case (Sect. 2.4). Comparing the frequency of ice-containing clouds at different locations provides insights into differences of primary ice formation (Choi et al., 2010; Kanitz et al., 2011; Seifert et al., 2015; Tan et al., 2014; Zhang et al., 2018). Generally, clouds contain ice more frequently for decreasing temperature (Fig. 6, solid lines). While at Leipzig and Limassol nearly all clouds with CTTs below −16°C contained ice, at Punta Arenas a fraction of 0.4 ± 0.1 of shallow stratiform clouds at these temperatures were classified as liquid-only. This behavior is discussed further in Sect. 3.2.3.

Compared to prior lidar-based studies (e.g., Kanitz et al., 2011; Choi et al., 2010; Seifert et al., 2010), the fraction of ice-containing clouds in the synergistic dataset is higher at temperatures above −10°C. At these temperatures, the amounts of ice produced and, hence, the radar reflectivity and optical extinction are usually very low and stay undetected for lidar (Bühl et al., 2013a) and spaceborne radars (Bühl et al., 2016). To quantify a lidar detection threshold in terms of optical extinction, the reflectivity-to-IWC and the reflectivity-to-extinction relationships by Hogan et al. (2006) are used. The only additional information needed for both parametrizations is the ambient temperature. Using these relationships, the response of the occurrence statistics to arbitrary extinction detection thresholds can be tested. As shown in Fig. 6 (dashed lines), for the lidar data used in the Cloudnet classification, an extinction threshold of 12 Mm−1 had to be applied in order to best match the lidar-only statistics from Punta Arenas and Leipzig presented by Kanitz et al. (2011).

3.2.2 Effect of boundary layer aerosol load on phase occurrence

As discussed in Sect. 3.1, the aerosol load at Punta Arenas is confined to the lowermost 2 km, and the aerosol load at temperatures above −10°C is similar to Limassol. To check for possible impact of this boundary layer aerosol on the ice-formation efficiency, the basic temperature-resolved phase occurrence frequency (Fig. 6) is split into two subsets, one containing cloud cases with bases below 2 km height and one with cloud cases having bases above that threshold, in the following denoted as coupled and uncoupled clouds, respectively. The resulting ice-formation frequency is shown in Fig. 7. At any height, temperatures vary by more than 11°C (10% to 90% percentile), providing ample coverage for the height threshold. Generally, coupled clouds show
higher fractions of ice (absolute difference increases by 0.3 at all locations at $-8^\circ$C). The temperature-resolved phase occurrence frequency for the coupled clouds shows rapid increase in fraction of ice-containing clouds, reaching 1.0 at temperatures of only $-15^\circ$C. The absolute difference between the fractions is less than 0.20, with the lowest fractions still being observed at Punta Arenas. Below $-15^\circ$C, almost no clouds were observed at such low heights, especially at Limassol; hence no meaningful comparison can be done in this temperature interval.

Considering only uncoupled clouds (cloud base above 2 km height; Fig. 7b), stronger contrasts become evident. The wave clouds discussed in the following section are still included and impact the frequency at temperatures below $-18^\circ$C at Punta Arenas. The frequencies of ice-forming clouds are similar at Leipzig and Limassol but 0.15 to 0.5 at $-10^\circ$C lower at Punta Arenas. Comparing the coupled and uncoupled state at Leipzig and Limassol, ice formation is more frequent in the coupled case by 0.15 between $-5$ and $-12^\circ$C.

### 3.2.3 Gravity wave influence on phase occurrence at low temperatures

The lower frequency of ice-containing cloud layers at temperatures below $-18^\circ$C over Punta Arenas is a prominent feature of both the lidar–radar and the lidar-only-equivalent datasets shown in Fig. 6. The associated high abundance of supercooled liquid water is frequently reported as a general phenomenon of stratiform clouds in the Southern Hemisphere midlatitudes and higher latitudes. Within this subsection, the reasons for the found behavior over Punta Arenas will be elaborated in more detail.

Frequently, the stratiform liquid-only cloud layers observed over Punta Arenas at temperatures below $-18^\circ$C are embedded in orographic gravity waves. Following the wave detection methodology introduced in Sect. 2.5, Fig. 8 shows the autocorrelation and power spectra of the Doppler lidar vertical velocity for clouds classified as liquid-only over Leipzig, Limassol, and Punta Arenas, respectively. These supercooled liquid-only clouds at Punta Arenas show high autocorrelation coefficients at long shifts (Fig. 8c1), whereas liquid-only clouds with similar characteristics are absent at Limassol.
Figure 7. Fraction of ice-containing clouds over temperature for Leipzig, Limassol, and Punta Arenas, (a) below 2 km and (b) above 2 km. Total duration of the clouds in each bin is given by the numbers above in hours.

Limassol (Fig. 8b1) and Leipzig (Fig. 8a1). In terms of spectral power density (Fig. 8c2), the strongly supercooled clouds at Punta Arenas show only low turbulence.

In a next step, the autocorrelation is used to filter the dataset for clouds affected by gravity waves. As described in Sect. 2.5, the length at which the autocorrelation coefficient drops below 0.8 is used as a characteristic value. Decreasing the threshold value for this characteristic length will remove gravity-wave-influenced clouds from the dataset. For large values of the length threshold, e.g., larger than 1000 m, only clouds that are strongly forced by gravity waves will be removed, whereas going to shorter thresholds (< 500 m) will also remove weakly gravity-wave-influenced clouds. Figure 9 shows an increase in the fraction of ice-containing clouds below $-12^\circ$C with decreasing correlation-length thresholds from 30 000 to 300 m. At Punta Arenas, the fraction of ice-containing clouds increases from 0.5 to 0.85. In the temperature interval between $-15$ and $-12^\circ$C, the fraction is 0.05 to 0.1 lower at Punta Arenas compared to Leipzig and Limassol. Hence, clouds with fully developed turbulence show similar ice-formation frequencies, independent of the location, with indications for a still slightly reduced ice-formation efficiency over Punta Arenas.

3.2.4 **Comparison of radar reflectivity factor of the ice virga**

Prior studies of Zhang et al. (2018) identified a strong contrast in radar reflectivity factor between the different 30° latitude bands of the globe, with the Southern Hemisphere mid-latitudes (30°-60° S) showing the lowest mean reflectivity of all regions. They concluded that this difference in reflectivity factor is associated with a respective difference in ice crystal mass and number concentration. In the following we provide a similar representation of regional contrasts of ice-virga reflectivity from ground-based perspective that is based on a single radar instrument. As described in Sect. 2.4, the amount of ice formed in the mixed-phase layer is measured at six height bins (180 m) below the base of the liquid-dominated cloud top and hence at the top of the virga (Bühl et al., 2016). Figure 10a shows the cloud top temperature-resolved statistics of reflectivity for the three stations, which is based on the full cloud dataset, including the wave-influenced clouds (as these are also included in the study by Zhang et al., 2018). From $-28$ to $-16^\circ$C and again above $-12^\circ$C, Punta Arenas shows the lowest reflectivity and Limassol the highest. For most temperatures, Punta Arenas is 5–8 dB below the northern hemispheric stations; the only exception is the interval between $-4$ and $-8^\circ$C, where the reflectivity for Punta Arenas and Leipzig is almost equal. At temperatures below $-28^\circ$C, Limassol and Punta Arenas show equal reflectivity, with values at Leipzig being slightly higher.

We further investigated the properties of the ice-forming liquid-dominated cloud top layers and found that cloud thickness agrees within 40 m above $-30^\circ$C. The ice-to-liquid content ratio (Fig. 10b) is smaller at Punta Arenas than at the northern hemispheric locations, especially (factor 3) between $-24$ and $-20^\circ$C but also above $-10^\circ$C. Hence, in these temperature regimes, the liquid phase is less efficiently converted into ice in stratiform clouds above Punta Arenas.

https://doi.org/10.5194/acp-21-17969-2021
4 Discussion of observed contrasts in properties of shallow clouds

In the previous section, a comprehensive overview of the aerosol conditions and stratiform cloud properties at the strongly contrasting sites of Leipzig, Limassol, and Punta Arenas was presented. Several aerosol-, temperature-, surface-coupling-, and dynamics-related differences have been identified and will be discussed in the following.

The analysis of the profiles of aerosol optical properties obtained from the PollyXT lidar observations reveals almost similar aerosol load at Punta Arenas and Limassol between 0 and $-10^\circ$C. Leipzig shows a higher aerosol load in this temperature interval. For lower temperatures, the lowest average extinction was observed at Punta Arenas. Low values of particle depolarization ratio indicate that non-spherical particles, such as mineral dust, are completely absent at Punta Arenas below $-10^\circ$C. The absence of continental aerosol species in the free troposphere is also consistent with the Southern Ocean, south of Australia (e.g., Alexander and Protat, 2019).

Additionally, Uetake et al. (2020) showed for this region that INPs are related to sea spray emissions. Together with an air mass source estimate, the average profiles of optical parameters were used as an input for INP parametrizations for a first estimate of long-term averages of INP concentrations. The lidar-based estimate of average INP concentrations identifies the strongest differences in INP concentrations between $-12$ and $-35^\circ$C. However, it is still subject to discussion as to which fraction of marine particles is transported from the boundary layer to the free troposphere. This insufficient knowledge propagates further into uncertainties in the fraction of continental aerosol, which leads to considerable uncertainties in the estimated concentrations of INPs. Bourgeois et al. (2018) argue that due to efficient removal processes, marine particles predominantly remain in the boundary layer (globally 80% of their AOT). Murphy et al. (2019) argue in a similar direction, especially when sea salt is regarded a tracer of marine origin.

Opposed to prior lidar-only studies (e.g., Kanitz et al., 2011; Alexander and Protat, 2018), more than half of the clouds observed at $-8^\circ$C at the three sites were found to contain ice. This is in line with the lidar detection threshold introduced by Bühl et al. (2013a) but extending their results to Limassol and Punta Arenas. This result has significance for
be identified based on their autocorrelation function, as pre-
servations, clouds driven by orographic wave dynamics can
quency is biased. Using Doppler lidar vertical velocity ob-
parts of the clouds are covered, and the phase occurrence fre-
rapid flow and weak turbulence (Heymsfield and Miloshe-
phase is separated horizontally from the ice virga, due to the
in boundary layer clouds observed during Arctic
summer. (Achtert et al., 2020). Griesche et al. (2021) also
served for Arctic boundary layer clouds during summer and
layer, where presumably INPs of biological origin are present
nas were influenced by gravity waves. Similar frequencies of
occurrence of non-turbulent liquid layers were also found at
Utqiagvik (Alaska) and McMurdo (Antarctica), deriving sta-
bility criteria from radiosoundings (Silber et al., 2020).

In a final step, the separation techniques for coupling and
orographic waves can be combined to assess contrasts in ice
frequency for free tropospheric and fully turbulent clouds. The
resulting occurrence frequency is shown in Fig. 11. The
fraction of ice-containing clouds at temperatures below
−15 °C is above 0.85 at all three sites. However, in com-
parison to Leipzig and Limassol, the fraction of ice-forming
clouds with CTT between −25 and −15 °C remains 0.1
lower at Punta Arenas. Mineral dust is known to be an ef-
ficient INP at these temperatures (Kanji et al., 2017) but was
not observed at the respective temperatures at Punta Arenas
(Fig. 3). This difference in the non-wave cloud phase occur-
ence is also in agreement with satellite-based studies of Vil-
lanueva et al. (2020), who also attributed latitudinal differ-
ces in the ice occurrence to associated differences in the
dust load. Figure 11 also depicts the low frequency of ice for-
mation in free tropospheric fully turbulent clouds at temper-
atures above −15 °C. With less coupling to the near-surface aerosol reservoir, ice formation is strongly suppressed com-
pared to Leipzig and Limassol.

The ice mass formed by stratiform liquid layers is lowest at
Punta Arenas, as was found by investigating the radar reflect-
vity factor in the ice virga as a proxy. Especially between
−28 and −16 °C, the observed radar reflectivity factor is up
to 7 dB lower, compared to both sites in the Northern Hemi-
sphere. This result is consistent with estimates from space-
borne sensors covering the full Southern Ocean (Zhang et al.,
2018). In contrast, Arctic mixed-phase clouds were found to
respond with lower IWC to increased loads of anthropogenic
pollution (Norgren et al., 2018). As for the lower frequency
of ice formation discussed above, the difference in ice mass
coincides with a lack of dust INPs at these temperatures. Slight
differences in IWC or radar reflectivity, respectively, might be explained by a slower mass growth rate, caused by a
smaller vapor diffusion coefficient at higher ambient pressure
(Hall and Pruppacher, 1976). When temperature and particle
size are considered similarly, the stratiform clouds subject
to this study experience 10% to 20% larger growth rates at
Leipzig and Limassol than at Punta Arenas. Hence, the dif-
ference of a factor 3–6 larger ice mass cannot be explained
solely by this effect alone.

Frequent occurrences of ice-forming clouds above −10 °C
were found, which were not covered by studies based on
spaceborne active remote sensing. Similar to their colder
counterparts, they show a lower ice amount in the virga above
Punta Arenas but with a smaller difference compared to the
sites in the Northern Hemisphere. With an average reflectiv-
ity of −36 dBZ, they are usually well below the detection
limit of the CloudSat satellite in the A-Train constellation

Figure 9. Fraction of ice-containing clouds over temperature for Leipzig, Limassol, and Punta Arenas. Dashed–dotted lines show the fractions for Leipzig, Limassol, and Punta Arenas with an autocorrelation coefficient smaller than 0.8 for a horizontal shift of 300 m. The fractions for cloud cases with longer autocorrelation are only shown for Punta Arenas. The total duration of the clouds in each bin is given by the numbers above in hours.
Figure 10. Amount of ice observed in the virga, (a) radar reflectivity factor observed at the top of the virga (180 m below the liquid-dominated layer base) and (b) ice-to-liquid content ratio binned by CTT. Temperature bins of 4 °C.

(Bühl et al., 2013a, 2016). Occurrence of such clouds in other parts of the Southern Ocean cannot be ruled out. A misclassification of supercooled drizzle clouds as ice-containing is unlikely, as they exceed −30 dBZ neither at cloud top, nor in the virga. From previous studies it is known that the onset of drizzle formation is usually associated with higher reflectivities, either above approximately −20 dBZ (Liu et al., 2008; Acquistapace et al., 2019) or at least above −30 dBZ (Wu et al., 2020).

5 Summary and outlook

This study investigated contrasts in aerosol–cloud interactions in shallow supercooled stratiform clouds observed with the ground-based remote-sensing supersite LACROS at Leipzig, Limassol, and Punta Arenas.

Sampling the profiles of optical properties with temperature as a vertical coordinate revealed aerosol load at temperatures between −15 and 0 °C being within a factor of 2 at Punta Arenas and Limassol. This finding is related to the cold and (compared to the free troposphere) aerosol-laden boundary layer at Punta Arenas. At lower temperatures, the lowest $\beta_p$ and extinction were observed over Punta Arenas, the highest over Leipzig. The very low particle depolarization ratio at Punta Arenas between −25 and −10 °C suggests the absence of mineral dust in a temperature regime, where dust is known to be an efficient INP. An estimate of INP concentrations at the respective temperatures based on the optical properties reveals differences of 1–4 orders of magnitude between Punta Arenas and the two northern hemispheric sites. In absence of abundant INPs from marine sources, the contribution of terrestrial sources causes strong variability.

The phase occurrence frequencies showed a higher fraction of ice-containing clouds at weakly supercooling temperatures of above −10 °C compared to prior lidar-only studies. A cloud radar with a sensitivity better than −40 dBZ is needed to sufficiently characterize low ice water contents in the virga formed by shallow stratiform clouds in this temperature regime. Coupling to the boundary layer increases the frequency of ice formation at slightly supercooling tempera-
tures at all sites. The strongest contrasts in the ice-formation frequency between free tropospheric and surface-coupled conditions were found for Punta Arenas. This finding is in compliance to the found contrasts in the INP profiles at the three sites and further indicates that the free tropospheric INP reservoir over the Southern Ocean is limited.

Frequent liquid-only layers below $-20^\circ$C at Punta Arenas were found to be associated with orographic gravity waves, causing two implications: (1) potential phase misclassification by stationary observers due to horizontal separation of ice and liquid phase and (2) sustained liquid water in updrafts because the associated vertical velocities allow for supersaturation over water; the newly developed Doppler lidar autocorrelation approach helps to address “[…] the difficulty in characterizing gravity waves from existing single-site measurements […]” as posed by Silber et al. (2020). Ice mass in the virga and the ice-to-liquid content ratio were found to be lowest at Punta Arenas, especially at temperatures where dust serves as an efficient INP at the northern hemispheric sites.

From the results of this study, two issues for further investigations arise: firstly, the importance of terrestrial sources of INPs in the Southern Ocean. Observations of free tropospheric aerosols upwind and downwind of major landmasses are required to quantify the terrestrial emission and identify regions where the emitted INPs impact cloud microphysics the strongest. Secondly, how important are gravity waves in the formation of clouds in other regions of the Southern Ocean? How frequently do gravity waves occur above the open ocean, which covers the vast majority of the Earth’s surface between 30 and $70^\circ$S? Such measurements are required in order to further constrain the role of INPs in the evident excess of liquid water in clouds of the southern hemispheric midlatitudes.

Appendix A: Wave-cloud example

Figure A1 shows a typical wave cloud at Punta Arenas. Updrafts in stationary or slowly propagating waves can sustain liquid layers without showing signatures of sedimenting ice crystals to the fixed observer. The ice phase is only visible downwind as soon as the droplets have evaporated. Such a wave cloud was observed at Punta Arenas on 27 September 2019. A mid-tropospheric long-wave ridge caused a strong, zonal westerly flow at Punta Arenas. Velocity of the horizontal wind exceeded 25 m s$^{-1}$ at 4000 m height. Layered clouds developed within the orographic gravity waves triggered by this flow. One of the cloud layers with a CTT between $-34$ and $-31^\circ$C is depicted in Fig. A1. It illustrates how the liquid and ice phase are horizontally separated. Embedded in a slowly propagating gravity wave, the cloud was advected over the site, and the full life cycle could be observed. Based on the Doppler lidar vertical velocity (Fig. A1d), the updraft was present until 05:32 UTC. Afterwards, downward motion with increasing velocity set in. Only after the liquid droplets evaporated at 06:17 UTC can the ice particles be clearly identified in the remote-sensing observations (Fig. A1b, c).
Figure A1. Wave cloud observed at Punta Arenas on 27 September 2019, 05:00–06:45 UTC. (a) Lidar attenuated backscatter, (b) Cloudnet target categorization, (c) lidar volume linear depolarization ratio, and (d) Doppler lidar vertical velocity. The temperature information in (b) is obtained from ECMWF IFS model data. The red rectangle in (d) indicates the automatically detected liquid-topped shallow stratiform cloud. The inset in (d) shows the autocorrelation function of the vertical velocities in the liquid-dominated layer. The height of the velocity sampling is indicated with a dashed line.
Appendix B: Air mass source estimate

Figure B1. Mean normalized residence time profile at (a) Leipzig, (b) Limassol, and (c) Punta Arenas for a reception height of 2 km. Residence times are shown for the surface cover classification (1), named areas (2), and latitude bands (3).
Figure B2. Mean normalized residence time binned by temperature at (a) Leipzig, (b) Limassol, and (c) Punta Arenas for a reception height of 2 km. Residence times are shown for the surface cover classification (1), named areas (2), and latitude bands (3). Hatching indicates bins with less than 1% of the values.
The automated time–height-resolved air mass source attribution described by Radenz et al. (2021) is used to characterize air mass origin for the LACROS observations. As in the original publication, 10 d HYSPLIT ensemble backward trajectories are calculated in intervals of 3 h and 500 m throughout the period of the deployment. For the calculation of the residence times, a reception height of 2 km is used. Additionally to the MODIS-based surface cover classification and the named geography (individually for Leipzig, Limassol, and Punta Arenas), 30° wide bands of latitude are used to characterize meridional transport. The assignment of the surface cover classed and named geography is depicted in Figs. 2 and 3 of Radenz et al. (2021). The average residence time for each set of surface types is shown in Fig. B1. However, as for the aerosol optical properties, geometrical height is less insightful than temperature. Thus, the residence time at each 3 h profile is binned by ambient temperature, using analysis profiles. Then, the average residence time is calculated per temperature bin (Fig. B2).

Code and data availability. The Cloudnet datasets are provided by the ACTRIS Data Centre node for cloud profiling via the following links: https://hdl.handle.net/21.12132/2.e130b236928a4932 (Leipzig, last access: 20 October 2021), https://hdl.handle.net/21.12132/2.a056a828a6b94d1f (Limassol, last access: 20 October 2021), and https://hdl.handle.net/21.12132/2.b6c194d7d33b448e (Punta Arenas, last access: 20 October 2021). In the upcoming future, the Doppler lidar and PollyNET datasets will also be available via ACTRIS. Meanwhile, they can be obtained upon request from polly@tropos.de. The Lidar data was processed with the PollyNET processing chain (Yin and Baars, 2021). The source code of the cloud sniffer, the database of identified clouds, and analysis scripts are available at https://doi.org/10.5281/zenodo.5574997 (Radenz and Bühl, 2021) and require a setup of pyLARDA (https://doi.org/10.5281/zenodo.4721311; Bühl et al., 2021). AERONET photometer observations at Leipzig, Limassol, and Punta Arenas are available from the AERONET database (http://aeronet.gsfc.nasa.gov/, last access: 20 October 2021).

Author contributions. MR analyzed the data and drafted the manuscript. MR, PS, JB, HB, RE, BBG, REM, and FZ conducted the campaigns and operated LACROS. PS, MR, and JB generated the Cloudnet dataset. HB processed the PollyXT lidar data. PS, JB, and AA supervised the work and revised the manuscript. All authors jointly contributed to the paper and the scientific discussion.

Competing interests. The contact author has declared that neither they nor their co-authors have any competing interests.

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financial support. This research has been supported by the European Commission, H2020 Research Infrastructures (grant nos. 654109 and 857510), the EU Seventh Framework Programme (grant no. 603445), the German Bundesministerium für Bildung und Forschung funded projects “High Definition Clouds and Precipitation for Climate Prediction – HD(CP)2” (grant nos. 01LK1503F, 01LK1502I, 01LK1209C, and 01LK1212C). Parts of the CyCARE and DACAPO-PESO campaigns were funded by the Deutsche Forschungsgemeinschaft (DFG – German Research Foundation) project PICNICC (SE2464/1-1 and KA4162/2-1). Johannes Bühl acknowledges funding by DFG project COARSEMX (398285025). Rodanthi-Elisabeth Mamouri has been financially supported by the SIROCCO project (grant no. EXCELLENCE/1216/0217), co-funded by the Republic of Cyprus and the structural funds of the EU for Cyprus through the Research and Innovation Foundation. We also acknowledge the EXCELSIOR EU H2020 Widespread Teaming programme, grant agreement no. 857510 (http://www.exceLSIOR2020.eu, last access: 20 October 2021) and the Republic of Cyprus for the support of the establishment the ERATOSTHENES Centre of Excellence. Borís Barja González acknowledges funding by the ANID/CONICYT/FONDECYT Iniciación 11181335. Holger Baars acknowledges funding by the AEOUL Cal/Val activities (German Bundesministeriums für Wirtschaft und Energie; grant no. 50EE1721C). We thank AERONET Europe for providing the calibration service of the photometer. The authors gratefully acknowledge the NOAA Air Resources Laboratory (ARL) for the provision of the HYSPLIT transport and dispersion model. We thank Timothy J. Dunkerton for serving as an editor and gratefully acknowledge the constructive comments of the two referees.

Acknowledgements. The authors wish to thank TROPOS, Cyprus University of Technology and University of Magallanes, for their logistic and infrastructural support during the LACROS deployments. We furthermore thank Teresa Vogl, Willi Schimmel, Audrey Teisseire, Cristofer Jimenez, Heike Kalesse-Los, and Roland Schrödner for keeping LACROS operational at Punta Arenas. We gratefully acknowledge the ACTRIS Cloud Remote Sensing Unit for making the Cloudnet datasets publicly available. LACROS operations were supported by the European Union (EU) Horizon 2020 (ACTRIS; grant no. 654109) and the Seventh Framework Programme (BACCHUS; grant no. 603445). Observations at Leipzig were supported by the German Bundesministerium für Bildung und Forschung funded projects “High Definition Clouds and Precipitation for Climate Prediction – HD(CP)2” (grant nos. 01LK1503F, 01LK1502I, 01LK1209C, and 01LK1212C). Parts of the CyCARE and DACAPO-PESO campaigns were funded by the Deutsche Forschungsgemeinschaft (DFG – German Research Foundation) project PICNICC (SE2464/1-1 and KA4162/2-1). Johannes Bühl acknowledges funding by DFG project COARSEMX (398285025). Rodanthi-Elisabeth Mamouri has been financially supported by the SIROCCO project (grant no. EXCELLENCE/1216/0217), co-funded by the Republic of Cyprus and the structural funds of the EU for Cyprus through the Research and Innovation Foundation. We also acknowledge the EXCELSIOR EU H2020 Widespread Teaming programme, grant agreement no. 857510 (http://www.exceLSIOR2020.eu, last access: 20 October 2021) and the Republic of Cyprus for the support of the establishment the ERATOSTHENES Centre of Excellence. Borís Barja González acknowledges funding by the ANID/CONICYT/FONDECYT Iniciación 11181335. Holger Baars acknowledges funding by the AEOUL Cal/Val activities (German Bundesministeriums für Wirtschaft und Energie; grant no. 50EE1721C). We thank AERONET Europe for providing the calibration service of the photometer. The authors gratefully acknowledge the NOAA Air Resources Laboratory (ARL) for the provision of the HYSPLIT transport and dispersion model. We thank Timothy J. Dunkerton for serving as an editor and gratefully acknowledge the constructive comments of the two referees.

Review statement. This paper was edited by Timothy J. Dunkerton and reviewed by two anonymous referees.
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https://doi.org/10.5194/acp-21-17969-2021


https://doi.org/10.5194/acp-21-17969-2021


