

Retrieval of terahertz ice cloud properties from airborne measurements based on the irregularly shaped Voronoi ice scattering models

Ming Li^{1,2}, Husi Letu¹, Hiroshi Ishimoto³, Shulei Li⁴, Lei Liu⁴, Takashi Y. Nakajima⁵, Dabin Ji¹, Huazhe Shang¹, and Chong Shi¹

¹State Key Laboratory of Remote Sensing Science, Aerospace Information Research Institute,

Chinese Academy of Sciences, Beijing, 100101, China

²University of Chinese Academy of Sciences, Beijing, 100049, China

³Meteorological Research Institute, Japan Meteorological Agency (JMA), Nagamine 1-1, Tsukuba, 305-0052, Japan
 ⁴College of Meteorology and Oceanography, National University of Defense Technology, Changsha, 410073, China
 ⁵Research and Information Center (TRIC), Tokai University, 4-1-1 Kitakaname Hiratsuka, Kanagawa, 259-1292, Japan

Correspondence: Husi Letu (husiletuw@hotmail.com)

Received: 27 August 2022 – Discussion started: 8 September 2022 Revised: 13 December 2022 – Accepted: 14 December 2022 – Published: 24 January 2023

Abstract. Currently, terahertz remote sensing technology is one of the best ways to detect the microphysical properties of ice clouds. Influenced by the representativeness of the ice crystal scattering (ICS) model, the existing terahertz ice cloud remote sensing inversion algorithms still have significant uncertainties. In this study, based on the Voronoi ICS model, we developed a terahertz remote sensing inversion algorithm of the ice water path (IWP) and median mass diameter (D_{me}) of ice clouds. This study utilized the singlescattering properties (extinction efficiency, single-scattering albedo, and asymmetry factor) of the Voronoi, sphere, and hexagonal column ICS models in the terahertz region. Combined with 14408 groups of particle size distributions obtained from aircraft-based measurements, we developed the Voronoi, sphere, and column ICS schemes based on the Voronoi, sphere, and column ICS models. The three schemes were applied to the radiative transfer model to carry out the sensitivity analysis of the top-of-cloud (TOC) terahertz brightness temperature differences between cloudy and clear skies (BTDs) on the IWP and D_{me} . The sensitivity results showed that the TOC BTDs between 640 and 874 GHz are functions of the IWP, and the TOC BTDs of 380, 640, and 874 GHz are functions of the $D_{\rm me}$. The Voronoi ICS scheme possesses stronger sensitivity to the D_{me} than the sphere and column ICS schemes. Based on the sensitivity results, we

built a multi-channel look-up table for BTDs. The IWP and $D_{\rm me}$ were searched from the look-up table using an optimal estimation algorithm. We used 2000 BTD test data randomly generated by the RSTAR model to assess the algorithm's accuracy. Test results showed that the correlation coefficients of the retrieved IWP and Dme reached 0.99 and 0.98, respectively. As an application, we used the inversion algorithm to retrieve the ice cloud IWP and D_{me} based on the Compact Scanning Submillimeter-wave Imaging Radiometer (CoS-SIR) airborne terahertz radiation measurements. Validation against the retrievals of the Bayesian algorithm reveals that the Voronoi ICS model performs better than the sphere and hexagonal column ICS models, with enhancement of the mean absolute errors of 5.0 % and 12.8 % for IWP and D_{me} , respectively. In summary, the results of this study confirmed the practicality and effectiveness of the Voronoi ICS model in the terahertz remote sensing inversion of ice cloud microphysical properties.

1 Introduction

Ice clouds account for about 20 %–30 % of the total global cloud mass (Liou, 1992; Rossow and Schiffer, 1991). They play an essential regulatory role in the global radiation bal-

ance and the water cycle (Hong et al., 2016; Stephens et al., 2012). Microphysical properties such as the ice water content, ice particle size, orientation, and shape are the main influencing factors of the scattering and radiative properties of ice clouds (Li et al., 2022; Yi et al., 2017; Zhao et al., 2018; Chen and Zhang, 2018) and, in turn, affect the radiation budget (Heymsfield et al., 2013, 2017; Rossow, 2014). The latest report of the 6th Intergovernmental Panel on Climate Change (IPCC6) (Forster et al., 2021) identifies cloud radiative properties and their feedback effects as the largest source of uncertainty in the overall climate feedback, with errors in ice clouds being one of the most significant factors (Zhang et al., 2021). Therefore, the accurate acquisition of microphysical properties of ice clouds is of great importance for studying global climate change and improving the accuracy of numerical model simulations (Baran, 2009, 2012). Remote sensing techniques are one of the most effective means of obtaining microphysical and radiative properties of ice clouds, mainly including ground-based (Cimini et al., 2006), aircraft-based (Fox et al., 2017), and satellite-based remote sensing observation technologies (Yang et al., 2015, 2018). Currently, large numbers of passive sensors (visible, infrared, and microwave detectors) have been developed, and related ice cloud retrieval algorithms have been reported in a substantial body of literature (Nakajima and King, 1990; Nakajima et al., 1991, 2019; Nakajima and Nakajima, 1995; Platnick et al., 2003, 2017; Fox et al., 2019; Brath et al., 2018). The visible and infrared wavelengths are sensitive to the visible optical depth and cloud top (Minnis et al., 1993a, b). The millimetre-wave ice cloud remote sensing technique is more suited to detect vertical cloud properties. Sensors such as the Millimeter-wave Imaging Radiometer (MIR) (Racette et al., 1992) and Special Sensor Microwave Water Vapor Sounder (SSM/T-2) have been used in several studies of ice water path (IWP) and particle size retrievals (Lin and Rossow, 1996; Liu and Curry, 1998, 1999). MIR channels at 89, 150, and 220 GHz have been used by Deeter and Evans (2000) and Liu and Curry (2000) to retrieve IWP and particle size in cirrus anvils over the tropical ocean. Compared to passive sensors, the cloud satellite radar (CloudSat), with an onboard millimetre-wavelength (94.05 GHz) radar, and the radar-lidar cloud product (DARDAR) (Ceccaldi et al., 2013) present new opportunities to infer the microphysical properties of ice clouds on a global scale. In practice, the current detectors and approaches are confined to a limited range of ice particle sizes (Cho et al., 2015). For example, visible and infrared sensors are only sensitive to particles smaller than 50 µm (Evans et al., 2005). Additionally, microwave detectors are mainly useful for particles larger than 500 µm (Fox, 2020; Fox et al., 2019). There is a pressing need to develop an effective frequency region for obtaining comprehensive information about ice particles. To bridge the gap between the technologies of visible-infrared and microwave measurement of ice clouds, the terahertz (THz) wavelength between the infrared and microwave regions has the potential advantage of complementing existing visible-infrared and microwave techniques (Gasiewski, 1992).

The terahertz wave is the sub-millimetre wave with frequencies in the range of 0.1-10 THz and corresponding wavelengths of 0.03-3 mm, comparable to the size of ice particles in ice clouds. Many theoretical studies (Evans and Stephens, 1995a, b; Evans et al., 1998) have shown that the passive terahertz wave has a higher detection capability and sensitivity to the IWP and particle size of ice clouds (Liu et al., 2020). Although terahertz waves were proposed for ice cloud remote sensing in the 1960s (Gao et al., 2016), the technology of terahertz radiometry of ice clouds lagged behind the theory (Evans et al., 2005). It is only within the last decade that terahertz radiometry has been applied to aircraftbased and satellite-based ice cloud remote sensing. With advances in terahertz detectors, researchers have successively developed aircraft-based terahertz radiometers (Gao et al., 2016), such as the Sub-millimeter Wave Cloud Ice Radiometer (SWCIR) (Evans et al., 2002), the Compact Scanning Submillimeter-wave Imaging Radiometer (CoSSIR) (Evans et al., 2012), and the International SubMillimeter Airborne Radiometer (ISMAR) (Fox et al., 2017). Also, research institutions have developed satellite-based terahertz radiometers, including the Superconducting Submillimeter-Wave Limb Emission Sounder (SMILES) (Inatani et al., 2000), Ice Cloud Imager (ICI) (Eriksson et al., 2020; Kangas et al., 2014), and IceCube (Wu et al., 2014), which were proposed in 2013 and are still under development.

Several methods have been reported to retrieve the ice clouds' IWP and particle size from terahertz brightness temperature. Inversion methods can be divided into physical and linear regression methods (Weng et al., 2019). The linear regression method is efficient and convenient, but it lacks a definite physical mechanism and is heavily dependent on the accuracy of the pre-calculated database. The physical method includes the Bayesian algorithm (Evans et al., 2005, 2002), the look-up table (LUT) method (Li et al., 2017, 2018; Liu et al., 2021), and the neural network algorithm (Jimenez et al., 2003). Evans et al. (1998) modelled terahertz brightness temperature using a polarized radiative transfer model based on eight ice particle shapes calculated by the discrete dipole approximation (DDA) method. The study found that the ice particle shape plays a vital role in modelling ice cloud scattering in the terahertz region. Evans et al. (2002) developed a Monte Carlo Bayesian integration (MCBI) algorithm to retrieve ice clouds' IWP and median mass diameter (D_{me}) from simulated SWCIR brightness temperatures. According to the validation results, the total median retrieval error for clouds with IWP of more than 5 gm^{-2} is roughly 30% for IWP and 15 % for D_{me} (Evans et al., 2002). Moreover, Evans et al. (2005) utilized the MCBI method to retrieve the IWP and $D_{\rm me}$ of ice clouds from the CoSSIR brightness temperatures (referred to as the CoSSIR-MCBI hereafter). During the retrieval procedure, one of five particle shapes (spherical snow, aggregates of frozen droplets, aggregates of hexagonal

plates, and aggregates of plates and hexagonal columns) developed by the DDA method (Evans and Stephens, 1995a) was selected for each ice cloud retrieval. The CoSSIR-MCBI results are validated by the Cloud Radar System data and showed a good agreement of radar backscattering with errors smaller than 5 dB. Later, Jimenez et al. (2007) used the neural network method combined with the radiative transfer code to retrieve the IWP and D_{me} of ice clouds. The ice cloud microphysical input is based on one of three randomly oriented particle shapes (solid columns, hexagonal plates, and four-bullet rosettes) simulated with the DDA method (Evans and Stephens, 1995a; Evans et al., 1998) in the microwave region. Results showed overall median relative errors of around 20 % for IWP and 33 μ m for D_{me} for a mid-latitude winter scenario and 17 % for IWP and 30 μ m for $D_{\rm me}$ for a tropical scenario. Based on early studies and the background, Buehler et al. (2007) proposed a formal scientific requirement for a passive sub-millimetre-wave cloud ice mission. The requirements are that the low IWP should be less than 10 g m^{-2} , the high IWP should be less than 50 %, and the particle diameter should be less than 50 µm. Li et al. (2016) investigated the effects of five ice crystal scattering (ICS) models from the single-scattering property database of Hong et al. (2009) at frequencies ranging from 100 to 1000 GHz, namely aggregates, hollow columns, flat plates, rosettes, and spheres (Van de Hulst, 1957), on the transmission characteristics of terahertz radiation. Results showed that the ice particle shape is one of the dominant factors affecting terahertz radiation. Lately, Fox (2020) evaluated seven ice particle habits from the ARTS scattering database (Eriksson et al., 2018). They showed that the randomly oriented large column aggregate performs best when simulating brightness temperatures between 183 and 664 GHz.

In summary, the physical method is based on the radiative transfer principle and relies on the forward physical model simulation and effective ice crystal scattering (ICS) model. Different assumptions of ice cloud microphysical properties (shape, size, orientation, and particle size distribution of ice particles) in the forward physical model significantly affect the retrieval of the IWP and particle size of ice clouds. Therefore, a practical and representative ICS model is essential for the ice cloud remote sensing in the terahertz region.

Many aircraft observations have demonstrated that ice clouds comprise a complex and diverse range of nonspherical ice particles (Lawson et al., 2006, 2019; Liou, 1992). It is still challenging for one specific light-scattering method to precisely calculate the single-scattering properties for non-spherical particles with different size parameters (SZPs). The SZP is the ratio of the equivalent-volume sphere's circumference dimension (or π times particle maximum diameters) to the incident wavelength (Baran, 2012; Nakajima et al., 2009). So far, the light-scattering methods can be roughly divided into the approximation method (AM) and the numerical simulation method (NM). The AM method is based on ray-tracing techniques, including the geometrical optics method (GOM) (Macke et al., 1996; Takano and Liou, 1989), which is suitable for large particles. The NM includes the finite-difference time domain (FDTD) (Yang and Liou, 1996a; Yee, 1966) and the DDA (Draine and Flatau, 1994; Yurkin and Hoekstra, 2007) methods, which are appropriate for small particles. Combined with the advantages of the AM and NM methods, several improved geometrical optics approximation (GOA) methods, including the geometric optics integral equation (GOIE) method (Yang and Liou, 1996b), have been developed. Moreover, the boundary element method (Groth et al., 2015; Kleanthous et al., 2022) has been recently applied to complex ice particles. Based on the above-mentioned light-scattering calculation methods, a series of regularly shaped ICS models have been designed, including hexagonal columns, hexagonal plates, bullet rosettes, droxtals, and so on (Yang et al., 2000a, 2013; Fu et al., 1999; Takano and Liou, 1989). Since the regularly shaped ICS models are not fully representative of the scattering properties of natural ice particles, researchers have developed complex ICS models. Yang et al. (2013) developed the surface-roughened non-spherical ICS models and applied them to the production of MODIS C6 ice cloud products (Platnick et al., 2017). C.-Labonnote et al. (2000, 2001) and Doutriaux-Boucher et al. (2000) developed an ICS database for the inhomogeneous hexagonal monocrystal (IHM) containing embedded inclusions (air bubbles and aerosols). The IHM database was applied for the ice cloud retrievals from the French satellite Polarization and Directionality of the Earth's Reflectance (POLDER) measurements (Deschamps et al., 1994). Furthermore, Baran and Labonnote (2007) and Baran et al. (2014a) developed an ensemble ice particle model made of hexagonal column ice particles for use in the Met Office Unified Model Global Atmosphere 5.0 (GA5) configuration (Baran et al., 2014b). Unlike the above-mentioned regularly shaped ICS models, Letu et al. (2012) and Ishimoto et al. (2013) analysed ice particles of different shapes and sizes from many NASA aircraft observations and developed an irregularly shaped complex Voronoi ICS model. The single-scattering property database of the Voronoi ICS model in the visible and infrared regions uses a combined FDTD, GOIE, and GOM approach. The Voronoi ICS model has been adopted for generating official ice cloud products for the Second Generation gLobal Imager (SGLI)/Global Change Observation Mission - Climate (GCOM-C) (Letu et al., 2012, 2016; Nakajima et al., 2019); AHI/Himawari-8 (Letu et al., 2018); and the Multi-Spectral Imager (MSI)/Earth Cloud Aerosol and Radiation Explorer (EarthCARE) satellite programmes (Illingworth et al., 2015), which will be launched in 2023. Furthermore, Li et al. (2022) showed the effectiveness of the Voronoi ICS model in describing ice clouds' global visible and infrared radiative properties in the Community Integrated Earth System Model (CIESM). As a result, researchers have proven the effectiveness of the database of the Voronoi ICS model in the visible and infrared regions in the ice cloud satellite remote sensing and climate model applications (Letu et al., 2016). Recently, Letu et al. (2012) and Ishimoto et al. (2013) have successfully extended the spectral range of the single-scattering property database of the Voronoi ICS model to the terahertz wave region. The database of the Voronoi ICS model in the terahertz region was adopted by Baran et al. (2018) as standard data for the modelling and evaluation of the ISMAR (International SubMillimeter Airborne Radiometer), which the European Space Agency (ESA) and Met Office jointly developed (Kangas et al., 2014). The results showed good evaluation performance of the Voronoi ICS model. However, the effectiveness of the Voronoi ICS model in the terahertz region in a practical retrieval of ice cloud microphysical properties from terahertz radiation has yet to be discovered.

Motivated by the above-mentioned situations, this study aims to apply the Voronoi ICS model to the ice cloud remote sensing retrieval of IWP and particle size from aircraftbased terahertz radiation measurements. To achieve this goal, we use the Voronoi ICS model to create a parameterization scheme (referred to as the Voronoi ICS scheme hereafter) for the ice cloud scattering property in the terahertz region. The Voronoi ICS scheme is employed in the RSTAR radiative transfer model (Nakajima and Tanaka, 1986, 1988) to build the LUT for upward terahertz brightness temperature differences between cloudy and clear skies (BTDs) at the top of the ice cloud (TOC) between multi-channel frequencies. The LUT based on the Voronoi ICS model is compared with that of the sphere and hexagonal column ICS models. Finally, the CoSSIR-MCBI evaluates the retrieval results searched iteratively from the LUTs for the Voronoi, sphere, and hexagonal column models. This paper is organized as follows: Sect. 2 introduces the data and radiative transfer model used in this study. Section 3 describes the ice cloud parameterization scheme, retrieval algorithm, and validation with the CoSSIR-MCBI algorithm. Section 4 presents the results of the retrieved IWP and particle size; a comparison of the Voronoi, sphere, and hexagonal column ICS models; and the validation of the retrieval algorithm. Section 5 presents the conclusions of this study.

2 Data and model

2.1 Single-scattering property database for the Voronoi, sphere, and column ICS models

This study used the single-scattering property database of three ice crystal habits (Voronoi, sphere, and hexagonal column) in the terahertz radiative transfer forward simulation. For the Voronoi ICS model, the single-scattering property database contains 31 ice particle sizes ranging from 0.25 to 9300 μ m and covers 20 terahertz channels with frequencies ranging from 10 to 874 GHz (see Table 1), corresponding to wavelengths from 0.03 to 3 cm. The hexagonal column ICS model (referred to as the column ICS model here-

after) is randomly oriented and was defined by Yang et al. (2000b). For the column ICS model, the aspect ratios a/L (defined as the ratio of the semiwidth a of a particle to its length L) are defined as 0.35 and 3.48, respectively, when L is less than 100 µm and greater than or equal to 100 µm. The single-scattering property database of the column ICS model used in the study was developed by Hong (2007) and Hong et al. (2009) using the DDA method. For the column ICS model, the single-scattering property database includes 38 maximum particle dimensions ranging from 2 to 2000 µm and covers 21 terahertz channels with frequencies ranging from 90 to 874 GHz (see Table 1).

The single-scattering properties - including the extinction efficiency, single-scattering albedo, and asymmetry factor of the Voronoi, sphere, and column ICS models in the terahertz region - are used to calculate the terahertz scattering properties of ice clouds. For the Voronoi ICS model, the single-scattering properties are derived from the singlescattering property database developed by Letu et al. (2012, 2016) and Ishimoto et al. (2012) using a combination of FDTD and DDA methods. The DDA method is used to calculate the single-scattering properties of moderate ice particles (SZP > 30). The FDTD method is used for small ice particles (SZP < 30). As the particle size increases, the shape of the Voronoi ICS model changes and becomes complicated. The geometrical characteristics of Voronoi ICS model shape with increasing ice crystal size have been shown in Fig. 3 and Table 1 in Ishimoto et al. (2012). The mass-dimension and area-dimension power law relationships of the Voronoi ICS model are defined by Ishimoto et al. (2012) and are described in Eqs. (1)–(3) as shown below.

$$m = 0.00528D^{2.1} \tag{1}$$

(in centimetre-gram-second system of units, cgs)

$$A_{\rm r} = \frac{4G}{\pi D^2} \tag{2}$$

$$A_{\rm r} = 0.20 D^{-0.29},\tag{3}$$

where *m* is the mass, *G* is the cross-sectional area, A_r is the area ratio, and D is the maximum particle dimension of the Voronoi ICS model. The single-scattering property database of the sphere ICS model is developed based on the exact solution of the Lorentz-Mie theory (Van de Hulst, 1957). The sphere ICS database contains the same ice particle sizes and terahertz frequencies as the Voronoi ICS database. For the Voronoi and sphere ICS models, calculations of their singlescattering properties utilized the real and imaginary parts of ice from the newest library of the refractive index provided by Warren and Brandt (2008). The refractive indices of the Voronoi and sphere ICS models at frequencies of 10-874 GHz are computed at a temperature of 266 K, according to Warren and Brandt (2008). The refractive indices of the column ICS model at frequencies of 90-340 GHz are derived from Warren (1984) at a temperature of $-30 \,^{\circ}$ C.

Voronoi ICS model		Column ICS model	
Maximum particle dimension (μm)	Frequency (GHz)	Maximum particle dimension (μm)	Frequency (GHz)
0.400	0.10000×10^2	0.200×10^{1}	0.900×10^{2}
0.100×10^{1}	0.15000×10^2	0.400×10^{1}	0.118×10^{3}
0.200×10^{1}	0.18700×10^2	0.600×10^{1}	0.157×10^{3}
0.300×10^{1}	0.23800×10^{2}	0.800×10^{1}	0.166×10^{3}
0.500×10^{1}	0.31400×10^2	0.100×10^{2}	0.183×10^{3}
0.750×10^{1}	0.35000×10^{2}	0.120×10^{2}	0.190×10^{3}
0.150×10^2	0.50300×10^2	0.150×10^{2}	0.203×10^{3}
0.250×10^2	0.53750×10^2	0.200×10^2	0.220×10^{3}
0.350×10^{2}	0.55000×10^{2}	0.250×10^{2}	0.243×10^{3}
0.450×10^{2}	0.89000×10^2	0.300×10^{2}	0.325×10^{3}
0.600×10^{2}	0.94000×10^{2}	0.400×10^2	0.340×10^{3}
0.700×10^{2}	0.11875×10^{3}	0.500×10^{2}	0.380×10^{3}
0.147×10^{3}	0.16550×10^3	0.600×10^2	0.425×10^{3}
0.225×10^3	0.18331×10^3	0.700×10^2	0.448×10^{3}
0.314×10^3	0.22900×10^3	0.800×10^2	0.463×10^{3}
0.419×10^{3}	0.24300×10^3	0.900×10^{2}	0.487×10^{3}
0.500×10^{3}	0.32500×10^{3}	0.100×10^{3}	0.500×10^{3}
0.623×10^{3}	0.44800×10^3	0.125×10^{3}	0.640×10^{3}
0.752×10^{3}	0.66400×10^3	0.150×10^{3}	0.664×10^{3}
0.867×10^3	0.87400×10^3	0.175×10^3	0.683×10^3
0.964×10^3	0107 100 7 10	0.200×10^3	0.874×10^3
0.108×10^4		0.250×10^3	0.0717.10
0.140×10^4		0.300×10^3	
0.175×10^4		0.350×10^3	
0.256×10^4		0.400×10^3	
0.350×10^4		0.500×10^3	
0.500×10^4		0.600×10^3	
0.300×10^{4}		0.000×10^{3}	
0.100×10^5		0.800×10^3	
0.120×10^5		0.900×10^{3}	
0.120×10^{5} 0.150 × 10 ⁵		0.100×10^4	
0.120 / 10		0.100×10^{4}	
		0.120×10^4	
		0.130×10^4	
		0.140×10^4	
		0.160×10^4	
		0.180×10^4	
		0.200×10^4	
		0.200 × 10	

Table 1. The frequency channels and maximum particle dimensions of ice particles included in the single-scattering property database of the Voronoi and column ICS models.

2.2 Airborne measurements and auxiliary data

The measured terahertz brightness temperature from CoSSIR/ER-2 aircraft during the TC4 mission on 17 and 19 July 2007 is utilized in this study (available at https: //espoarchive.nasa.gov/archive/browse/tc4/ER2, last access: 26 December 2021). During the TC4 field campaign, the CoSSIR measured brightness temperatures in channels from 183.3 to 873.6 GHz (183.3 \pm 1.0, 3.0, 6.6, 220,

 380.2 ± 1.8 , 3.3, 6.2, and 640 V and 874 GHz), all with matched beamwidths of about 4° (Evans et al., 2012). According to the studies by Li et al. (2016) and Liu et al. (2021), the atmospheric windows are near 640 and 874 GHz, and the atmospheric absorption peaks are near 380 GHz. The leading absorbing gases are water vapour and ozone. Therefore, both the 640 and 874 GHz brightness temperature are affected by ice clouds, while the brightness temperature of 380 GHz is insensitive to ice cloud microphysical properties. Hence, the 380 GHz minus 640 GHz brightness temperature differences can highlight the brightness temperature depression caused by ice clouds. And the 640 GHz minus 874 GHz brightness temperature differences can reflect the difference in the scattering properties of differently shaped ice clouds. According to Li et al. (2016), the differences between 640 and 874 GHz also can offset the regional errors due to different latitudes and atmospheric profiles. In this study, we choose the centre frequencies of 380, 640, and 874 GHz.

The particle size distributions (PSDs) describe the relationship between ice particle size and particle number concentration and are essential in the calculation of the average scattering properties of ice clouds. In this study, we select 14408 groups of PSDs from aircraft observation sampling data (available at https://www.ssec.wisc.edu/ ice models/microphysical data.html, last access: 18 February 2022) (Heymsfield et al., 2013), which are based on 11 field flight observation experiments covering ice cloud observations in mid-latitude and tropical regions. The 11 field programmes span a wide range of locations (ranging from 12° S to 70° N and from 148° W to 130° E) and encompass the temperature range 0 to $-86 \,^{\circ}$ C, with altitudes from near the surface to 18.7 km. This dataset is representative of the wide range of conditions where ice clouds are found in the troposphere and lower stratosphere on a near-global scale (Li et al., 2022; Heymsfield et al., 2013). For the fitting of PSDs for the Voronoi, sphere, and column ICS models, we adopted the gamma distribution following Heymsfield et al. (2013) in the form of Eq. (4):

$$n(L) = N_0 L^{\mu} e^{-\lambda L},\tag{4}$$

where *L* is the maximum particle dimension, n(L) is the particle concentration per unit volume (e.g. cm⁻³), N_0 is the intercept, λ is the slope, and μ is the dispersion. The physical meaning of the PSDs is that $n(L) \times dL$ is the number of particles per unit area.

The total water vapour and ozone column data provided by the ERA5 reanalysis data are used as input data for the radiative transfer model to simulate clear-sky brightness temperature. The ERA5 reanalysis data are the fifth-generation reanalysis data product developed by the European Centre for Medium-Range Weather Forecasts (ECMWF) (Hans et al., 2019). The total water vapour and ozone column data used have a horizontal resolution of $0.25^{\circ} \times 0.25^{\circ}$ and a temporal resolution of 1 h. The retrieved results are validated against the IWP and D_{me} (Evans et al., 2005) from the CoSSIR-MCBI algorithm during the same period. The IWP and D_{me} are available at https://espoarchive.nasa.gov/archive/browse/ tc4/ER2 (last access: 20 May 2022).

2.3 Radiative transfer model

The RSTAR radiative transfer model is a set of numerical radiative transfer models developed by Nakajima and Tanaka (1986, 1988) for the plane-parallel atmosphere. The calculated wavelengths can cover from 0.17 to 1000 µm, and the assumed plane-parallel atmosphere could be divided into 50 layers from sea level to a maximum altitude of 120 km. The RSTAR model contains six atmospheric profiles (tropical, mid-latitude summer, mid-latitude winter, highlatitude summer, high-latitude winter, and US standard atmosphere), including vertical profiles of temperature, pressure, water vapour, and ozone. Calculating gas absorption in RSTAR is based on the K-distribution method developed by Sekiguchi and Nakajima (2008), which considers important atmospheric radiative gases (H₂O, CO₂, O₃, N₂O, CO CH₄, etc.) and trace gases. The K-distribution method parameters were obtained from the HITRAN-2004 database. The RSTAR radiative transfer model assumes the simulated scene is composed of a homogeneous ice cloud layer.

3 Method

Figure 1 shows the general flowchart for the inversion of the IWP and D_{me} of ice clouds using the CoSSIR brightness temperature measurements. For convenience, the difference between the 640 GHz BTD and the 874 GHz BTD is simplified to BTD₂₋₃, and the difference between the 380 GHz BTD and the 640 GHz BTD is simplified to BTD_{1-2} . We refer to the difference between the BTD_{1-2} and BTD_{2-3} as BTD_{1-3} . Firstly, based on the single-scattering property database of the Voronoi, sphere, and column ICS models, the atmospheric radiative transfer model RSTAR is used to build LUTs of top-of-cloud (TOC) BTD₂₋₃ and BTD₁₋₃ for three ICS models, respectively. With the assumptions of an a priori value of IWP and D_{me} , the initial TOC BTD₂₋₃ and BTD₁₋₃ simulations can be searched from the LUTs for three ICS models. Secondly, we use the RSTAR to construct a clearsky LUT of TOC brightness temperature (BT) for 380, 640, and 874 GHz, respectively. The inputs are different ozone and water vapour column values under clear-sky conditions (see Sect. 3.2 for details). Then, the ERA5 reanalysis data are used to estimate the clear-sky TOC BT based on the clearsky LUT. The measured cloudy-sky TOC BTs of 380, 640, and 874 GHz are obtained from the CoSSIR measurements. Consequently, the TOC BTD₂₋₃ and BTD₁₋₃ measurements can be calculated using the measured cloudy-sky TOC BT minus the interpolated clear-sky TOC BT. Finally, the cost function will provide an estimate of the coherence between the measured and simulated TOC BTD₂₋₃ and BTD₁₋₃. The IWP and D_{me} are obtained by using the Gaussian–Newtonian non-linear optimization estimation method. The retrieved results are validated against the IWP and D_{me} retrieved from the CoSSIR-Evans algorithm. In the RSTAR forward simulation, the input parameter of the ice particle size is R_e , which is defined by Eq. (5):

$$R_{\rm e} = \frac{\int_0^\infty r^3 n(r) dr}{\int_0^\infty r^2 n(r) dr},$$
(5)

where *r* is the equivalent-volume sphere's particle radius, and n(r) is the particle size distribution. According to Sieron et al. (2017), a mass-weighted size D_{me} would be a more appropriate parameter size than the area-weighted size for describing the PSDs. The D_{me} is given by Eq. (6):

$$D_{\rm me} = \frac{\int_0^\infty Dm(D)n(D){\rm d}D}{\int_0^\infty m(D)n(D){\rm d}D},\tag{6}$$

where *D* is the maximum particle dimension of ice particles, *m* is the particle mass, and n(D) is the particle size distribution. To validate the retrieved D_{me} using the D_{me} from the CoSSIR-Evans algorithm, the transformation from the R_e to the D_{me} is necessary due to the different definitions of R_e and D_{me} . Hence, we developed a conversion relationship between the R_e and D_{me} combined with different particle size distributions. Based on the integration of *r* and *D* over 14 408 PSDs, multiple groups of R_e and D_{me} are calculated according to Eqs. (5) and (6). Numerical fitting was used to build a relationship between the R_e (independent variable) and D_{me} (dependent variable) over 14 408 PSDs. The conversion formula between R_e and D_{me} is demonstrated as follows:

$$D_{\rm me} = a + bR_{\rm e} + cR_{\rm e}^2,\tag{7}$$

where *a*, *b*, and *c* are fitting coefficients obtained by numerical fitting and provided as input. Based on this relationship, we have unified all the R_e into D_{me} for comparability for the Voronoi, sphere, and column ICS models.

3.1 Parameterization of ice cloud optical properties

To apply the Voronoi, sphere, and column ICS models to the RSTAR radiative transfer model for forwarding simulation, three parameterization schemes (referred to as the Voronoi, sphere, and column ICS schemes hereafter) for the scattering properties of ice clouds in the terahertz spectrum need to be constructed. The ice cloud optical properties depend on the single-scattering properties of the ICS and the PSDs. The effective ice cloud particle size measures the average ice cloud particle size for a given PSD. Different methods have been used to determine the effective ice cloud size, and according to Baum et al. (2005a, b), the effective particle diameter for a given PSD is determined by Eq. (8):

$$D_{\rm e} = \frac{3}{2} \frac{\int_{L_{\rm min}}^{L_{\rm max}} V(L)n(L) \mathrm{d}L}{\int_{L_{\rm min}}^{L_{\rm max}} A(L)n(L) \mathrm{d}L},\tag{8}$$

where D_e is the effective particle diameter, and V and A are the volume and projected area of Voronoi and sphere models. Then, the spectral ice cloud optical properties (massaveraged extinction coefficients, single-scattering albedo, asymmetry factor, and mass-averaged absorption coefficients) for the Voronoi and sphere ICS schemes are calculated for all PSDs given by Eqs. (9)–(12):

$$K_{\text{ext}}(\lambda) = \frac{\int_{L_{\min}}^{L_{\max}} Q_{\text{ext}}(\lambda, L) A(L) n(L) dL}{\rho_{\text{ice}} \int_{L_{\min}}^{L_{\max}} V(L) n(L) dL},$$
(9)

$$\varpi(\lambda) = \frac{\int_{L_{\min}}^{L_{\max}} Q_{\text{sca}}(\lambda, L) A(L) n(L) dL}{\int_{L_{\min}}^{L_{\max}} Q_{\text{ext}}(\lambda, L) A(L) n(L) dL},$$
(10)

$$g(\lambda) = \frac{\int_{L_{\min}}^{L_{\max}} g(\lambda, L) \sigma_{\rm sca}(\lambda, L) n(L) dL}{\int_{L_{\min}}^{L_{\max}} \sigma_{\rm sca}(\lambda, L) n(L) dL},$$
(11)

$$K_{\rm abs}(\lambda) = \frac{\int_{L_{\rm min}}^{L_{\rm max}} Q_{\rm abs}(\lambda, L) A(L) n(L) dL}{\rho_{\rm ice} \int_{L_{\rm min}}^{L_{\rm max}} V(L) n(L) dL},$$
(12)

where $K_{\text{ext}}(\lambda)$ represents spectral-mass-averaged extinction coefficients (m²g⁻¹), $\varpi(\lambda)$ is spectral single-scattering albedo, $g(\lambda)$ is a spectral asymmetry factor, and $K_{\text{abs}}(\lambda)$ represents spectral-mass-averaged absorption coefficients (m²g⁻¹). Q_{ext} , g, Q_{sca} , and Q_{abs} are extinction efficiency, asymmetry factor, scattering efficiency, and absorption efficiency for Voronoi, sphere, and column models. Based on the parameterized scattering properties of ice clouds and the effective particle diameter of ice clouds, we developed a parameterization scheme for the scattering properties of ice clouds by establishing the spectral bulk-scattering properties of ice clouds. The least squares method is used to obtain firstorder and third-order polynomial fitting equations according to Eqs. (13)–(16):

$$K_{\text{ext}}(\lambda) = a_0 + \frac{a_1}{D_{\text{e}}},\tag{13}$$

$$\varpi(\lambda) = b_0 + b_1 D_e + b_2 D_e^2 + b_3 D_e^3, \tag{14}$$

$$g(\lambda) = c_0 + c_1 D_e + c_2 D_e^2 + c_3 D_e^3,$$
(15)

$$K_{\rm abs}(\lambda) = d_0 + d_1 D_{\rm e} + d_2 D_{\rm e}^2 + d_3 D_{\rm e}^3, \tag{16}$$

where a_i , b_j , c_j , and d_j (i = 0, 1; j = 0, 1, 2, 3) are fitting coefficients and are spectral functions of the terahertz wavelength. Values of the above coefficients for the Voronoi ICS scheme are listed in Appendix A (Tables A.1–A.4).

3.2 Look-up table

Before constructing the LUT, the sensitivity of different terahertz brightness temperatures to the IWP and D_{me} of ice clouds needs to be analysed to build a representative and efficient LUT. According to the sensitivity results (see Sect. 4.3), the BTDs between the cloudy and clear-sky conditions at 380, 640, and 874 GHz frequencies are simplified to BTD₁, BTD₂, and BTD₃, respectively, and the differences between the two BTDs are represented by a dash. Based on the Voronoi, sphere, and column ICS schemes, the RSTAR is used to construct three LUTs (Voronoi, sphere, and column LUTs) of BTD₂₋₃ and BTD₁₋₃ for cloudy- and clear-sky conditions. For the cloudy-sky LUT, the number of the IWP is set



Figure 1. The overall flowchart of the retrieval of the IWP and D_{me} of ice clouds based on the Voronoi, column, and sphere ICS models.

to 12, and the range of values is defined in log10 space from 0 to 3.36 in steps of 0.28. The number of R_e 's is set to six, and the range of values is defined in log10 space from 1.6 to 3.1 in steps of 0.25. For the clear-sky LUT, the number of total ozone columns is set to seven, and the range of values is from 200 to 500 in steps of 50. To be consistent with aircraft observations of terahertz brightness temperatures, BTD₂₋₃ and BTD₁₋₃ are simulated at the mean altitude (20 km). The US standard atmospheric profile is used in the simulation, and the cloud top temperature is assumed to be the same as the atmospheric temperature at that level. The surface is assumed to be a black body (emissivity equals 1) with a temperature of 288.15 K.

3.3 Optimal estimation inversion method

Based on the terahertz BTD LUTs for the Voronoi, sphere, and column ICS schemes established in the previous section, the IWP and D_{me} are retrieved using an optimization method with Gaussian–Newtonian non-linear iterations. Based on the atmospheric radiative transfer transmission in the terahertz spectrum, if the background field under clear-sky conditions is subtracted from the cloudy-sky condition, the TOC BTDs are only functions of the cloud microphysical property parameters, which are given by Eqs. (17) and (18):

$$Y = F(X) + \epsilon, \tag{17}$$

$$X = \begin{pmatrix} \text{IWP,} \\ D_{\text{me}} \end{pmatrix}, \quad Y = \begin{pmatrix} \text{BTD}_{2-3}, \\ \text{BTD}_{1-3} \end{pmatrix}, \quad (18)$$

where X is the vector matrix composed of the variables of the IWP and D_{me} to be solved. Y is the vector composed of the two BTDs and the uncertainty vector. The vector ϵ represents the uncertainties that are attached to the measurements (i.e. instrumental accuracy) and to the radiative transfer forward model (i.e. approximation errors in the radiative transfer model). Following Marks and Rodgers (1993), a good convergence can be obtained when the value of the cost function is lower than the size of the measurement vector. Since there is no robust a priori value for IWP and D_{me} , we selected an average value as the initial value for an a priori value of the IWP and D_{me} . Assuming that Y is the value of two BTDs measured by the aircraft, the inverse X is the minimum value for solving Eq. (19):

$$J(X) = \min\{\|y - F(X)\| + \gamma \|X - X_a\|\},$$
(19)

where X_a is a vector consisting of the prior estimates of the IWP and R_e , γ is a Lagrangian operator, min denotes the minimal value of solving this function, and the value of X at the minimal value is considered to be the retrieved result. For the solution method of the non-linear problem in the inversion process, the Newton non-linear iterative method is usually used. It has a faster inversion speed and higher solution accuracy, and its iterative form follows Eq. (20):

$$X_{i+1} = X_i - J''(X_i)^{-1} J'(X_i), (20)$$

where the i represents the number of iterations, and the superscripts denote the first-order and second-order derivations. Then the iterations start from the a priori initial estimate until the convergence criterion is satisfied or when the number of iterations satisfies the required number of iterations. Then the iteration is stopped, and the solution results are obtained. The final inversion results are validated using the results from the CoSSIR-MCBI algorithm.

4 Results

4.1 Single-scattering property database for the Voronoi, sphere, and column ICS models

Figure 2 compares the extinction efficiency, single-scattering albedo, and asymmetry factor, varying with the SZP for the Voronoi, column, and sphere ICS models at 325 and 874 GHz. For small ice particles with SZPs less than 0.1, the single-scattering properties are small and barely influenced by the shape of the ice particles. This is because the singlescattering properties of ice particles are close to Rayleigh scattering when the SZP is small. As particle size increases, particle scattering is predominantly Mie scattering, and the sensitivity of the single-scattering properties to the ice crystal habits becomes pronounced so that the ice crystal shape contributes to the large differences for large particle sizes. The single-scattering properties of the Voronoi and column models vary more smoothly than those of the sphere model. As shown in Fig. 2, the smooth curve of the Voronoi model reflects the notion that the FDTD and DDA methods provide good continuity for the transition between different size parameters of the Voronoi model. For large particles (SZP > 3) at 874 GHz, the Voronoi model has a higher extinction efficiency than the other two models. At 325 GHz, the singlescattering albedo of the Voronoi model is close to the sphere model and is higher than the column model. At 874 GHz, the single-scattering albedo of the Voronoi model is close to the column model and is higher than the sphere model. The Voronoi model has a higher asymmetry factor than the other two models at 325 and 874 GHz. On the one hand, the higher extinction efficiency and single-scattering albedo of the Voronoi ICS model for large particles are possibly due to the multifaceted shapes of the Voronoi ICS model, which can result in significant side and backward scattering and increase the scattered energy. On the other hand, the higher asymmetry factor of the Voronoi ICS model for large particles is possible because the scattered energy is dominated by diffraction. The diffracted energy is concentrated in the forward direction, leading to a large asymmetry factor of the Voronoi ICS model.

Figure 3 shows the contours of the single-scattering properties for the Voronoi, sphere, and column ICS models over 20 terahertz wavelengths (0.03-0.3 cm) and 31 particle sizes $(1-270 \,\mu\text{m})$. The sharp changes in the single-scattering albedo for the Voronoi, sphere, and column ICS models can be seen from small to large ice particles. For small particles in the low-frequency channels (long wavelengths), their scattering is mainly Rayleigh scattering with significant absorption effects. The absorption energy is proportional to the volume of ice particles and is barely affected by the ice particle shape. In the high-frequency channels (short wavelengths), the Mie scattering plays a leading role, and the scattering function plays a dominant role. Especially for large ice particles, the single-scattering albedo is close to one, and the influence of the ice particle shape becomes obvious. Figure 4 shows the differences in extinction efficiency, singlescattering albedo, and asymmetry factor between the Voronoi and the other two ICS models as functions of the effective radius and terahertz wavelength. Overall, the extinction efficiency of the Voronoi model is higher than the other two ICS models. The asymmetry factor of the Voronoi model is higher than the other two ICS models in the low-frequency channel for large ice particles.

Figure 5 displays the scattering phase functions of the Voronoi, sphere, and column ICS models. The variation in the scattering phase function for the Voronoi model tends to be more dramatic compared to the sphere and column ICS models. The scattering phase function is axisymmetric about the 90° scattering angle for small ice particles, which can be approximated as Rayleigh scattering. For small ice particles, the scattering phase function shows extreme values in the forward (0°) and backward (180°) directions and shows minimal values in both side directions (90 and 270°). As the ice particle size increases, the forward scattering is significantly larger than the side and backward scattering and gradually tends to be more Mie scattering. For the sphere model in the high-frequency channels, the forward scattering increases with the increase in particle size. In the low-frequency channels, the forward scattering remains almost constant with the increase in particle size. Compared to the Voronoi and sphere ICS models, the column ICS model shows smaller phase functions. For the Voronoi model, the low values of the scattering phase function move towards large scattering angles with the increase in particle size.



Figure 2. The extinction efficiency, single-scattering albedo, and asymmetry factor as functions of the SZP for the Voronoi (solid blue line), sphere (dashed red line), and column (dashed green line) ICS models with a refractive index of 1.78 + 0.005i in the (**a**, **c**, **e**) 325 GHz and 1.78 + 0.015i in the (**b**, **d**, **f**) 874 GHz frequencies.



Figure 3. The comparison of the extinction efficiency, single-scattering albedo, and asymmetry factor as functions of 31 particle sizes $(1-270 \,\mu\text{m})$ and 20 terahertz wavelengths $(0.03-0.3 \,\text{cm})$ for the (**a**, **d**, **g**) Voronoi, (**b**, **e**, **h**) sphere, and (**c**, **f**, **i**) column ICS models.



Figure 4. The (**a**, **c**, **e**) Voronoi minus sphere and the (**b**, **d**, **f**) Voronoi minus column ICS model differences in (**a**, **b**) extinction efficiency, (**c**, **d**) single-scattering albedo, and (**e**, **f**) asymmetry factor as functions of 31 particle sizes $(1-270 \,\mu\text{m})$ and 20 terahertz wavelengths (0.03–0.3 cm).



Figure 5. The scattering phase functions for ice particles with four sizes ($R_e = 30, 71, 107, and 153 \mu m$) for the (a, d) Voronoi, (b, e) sphere, and (c, f) column ICS models with a refractive index of 1.78 + 0.005i in the 325 GHz and 1.78 + 0.015i in the 874 GHz, respectively.

4.2 Bulk-scattering properties of ice clouds in the terahertz region

Based on the single-scattering properties of the Voronoi, sphere, and column ICS models, the Voronoi, sphere, and column ICS schemes are developed in the terahertz region with the integration over both PSDs and terahertz frequencies. The calculated bulk-scattering properties of ice clouds include the mass extinction coefficients, single-scattering albedo, asymmetry factor, and mass absorption coefficients. Figure 6 compares the calculated bulk-scattering properties of the Voronoi, sphere, and column ICS schemes over 14 408 PSDs at four terahertz frequencies (325, 448, 664, and 874 GHz). The bulk-scattering properties of ice clouds depend on the effective diameter and the terahertz frequency. The single-scattering albedo increases linearly with the effective diameter for the sphere ICS scheme. For the Voronoi ICS scheme, the single-scattering albedo increases for the effective diameter smaller than 100 µm and approaches 0.95 with effective diameters exceeding 100 µm. The column ICS scheme has the lowest single-scattering albedo compared to the other two schemes for small effective diameters and gradually exceeds the other two schemes with the increase in the frequency and effective diameter. The mass extinction coefficients obtained from the three schemes show a uniformly positive correlation with the effective diameter and the terahertz frequency. At four terahertz frequencies, the mass extinction coefficients of the Voronoi ICS scheme are close to the column ICS scheme for all effective diameters. The sphere ICS scheme has the highest mass extinction coefficients compared to the other two schemes for all effective diameters and four terahertz frequencies. The Voronoi ICS scheme has a similar asymmetry factor to the sphere ICS scheme at low frequency and becomes the highest compared to the other two schemes at high frequency. The column ICS scheme has the lowest asymmetry factor compared to the other two schemes for all effective diameters. For all effective diameters, the mass absorption coefficients of the sphere ICS scheme have a stronger absorption effect than the other two schemes and decrease with increasing effective diameter. The Voronoi ICS scheme has the lowest mass absorption coefficients compared to the other two schemes for effective diameter smaller than 100 µm and becomes higher than the column ICS scheme with increasing effective diameters. Overall, the bulk-scattering properties for all schemes increase with increasing frequencies. This is mainly because the scattering properties of ice clouds are more significant as the wavelength decreases.

4.3 Sensitivity results

We discuss the sensitivity of the TOC BTD_{2-3} and BTD_{1-3} to the IWP and D_{me} based on the Voronoi, sphere, and column ICS schemes, respectively, in the RSTAR model. Figure 7 shows that the BTD_{2-3} and BTD_{1-3} are positively cor-

M. Li et al.: Retrieval of terahertz ice cloud properties

related with the IWP and D_{me} . For the Voronoi and column ICS models, the BTD₂₋₃ and BTD₁₋₃ show monotonically increasing relationships with the increase in IWP. For ice clouds with the D_{me} larger than 200 µm, the 50 to 1000 µm particle sizes can lead to approximately 0–20 K BTD₂₋₃ and 0–30 K BTD₂₋₃. The BTD₂₋₃ and BTD₁₋₃ increase with the increase in the D_{me} and are close to a constant value for D_{me} larger than 600 µm. As shown in Fig. 8, the Voronoi ICS scheme has higher BTD₂₋₃ and BTD₁₋₃ compared to the sphere and column ICS scheme. In summary, the TOC BTD₂₋₃ and BTD₁₋₃ have a strong sensitivity to the IWP and D_{me} , especially for moderate to large ice particles and large IWP. The different ICS schemes vary considerably in their modelling of terahertz brightness temperatures.

Figure 9 exhibits the variation in the TOC BTDs at different IWPs and D_{me} for 380, 640, and 874 GHz frequencies based on the RSTAR model. The BTD₁₋₃ has a strong sensitivity to the IWP and increases with the increase in the IWP, and the slope shows that the increase is pronounced when the IWP is less than 600 g m^{-2} . The BTD₂₋₃ has a strong sensitivity to the $D_{\rm me}$ and increases with the increasing $D_{\rm me}$. By comparing the LUTs of the Voronoi, sphere, and column ICS models, it is found that at the same BTD_{2-3} , the sphere LUT has the highest IWP compared to the column and Voronoi LUT, while the Voronoi LUT has the lowest. The D_{me} of the sphere LUT is smaller than that of the column and Voronoi LUT for the same BTD_{1-3} . This is mainly due to the higher single-scattering albedo and asymmetry factor of the Voronoi ICS model at low frequencies, resulting in stronger forwardscattering energy and a larger BTD than the sphere and column ICS model for the same IWP and D_{me} .

4.4 Inversion and validation results

To develop the inversion method from the CoSSIR measurements of the terahertz brightness temperature, we need to perform a quantitative simulation to test the accuracy of the algorithm. Figure 10 shows the validation results of the inversion algorithm using 2000 groups of BTD_{2-3} and BTD_{1-3} simulated by RSTAR using 2000 randomly generated IWP and D_{me} . The results show that most of the validation results of the red density lie on the 1 : 1 line. The correlation coefficients of the IWP and D_{me} are 0.94 and 0.99, and the mean absolute errors (MAEs) are 35.46 g m⁻² and 8.56 µm, respectively. The test results meet the formal scientific mission requirement of ice cloud remote sensing retrieval, according to Buehler et al. (2007). The high accuracy of the validation results proves the effectiveness and accuracy of the inversion algorithm.

We adopt the CoSSIR-MCBI results as the benchmark for evaluating the retrieved IWP and D_{me} based on three schemes using the new inversion algorithm. To quantify the differences in the retrieval results between the three schemes and the CoSSIR-MCBI, we use the CoSSIR-MCBI results minus the retrieved results of the Voronoi, column, and



Figure 6. The comparison of the parameterized single-scattering albedo, mass extinction coefficients, asymmetry factor, and mass absorption coefficients as functions of effective diameters for 325, 448, 664, and 874 GHz for the sphere (blue line), Voronoi (orange line), and column (green line) ICS models.

sphere ICS schemes, respectively, to analyse their differences. Figure 11a and b show the CoSSIR-MCBI results on 19 July 2007 overlaid with the retrieval results of the IWP and $D_{\rm me}$ based on the Voronoi scheme. Red and blue dots, respectively, represent the Voronoi ICS scheme and CoSSIR-MCBI results. Figure 11a and b show that the overall performance shows good agreement for the Voronoi ICS scheme compared to the CoSSIR-MCBI results. The matching rates for the IWP and $D_{\rm me}$ are 80.3 % and 83.2 %, respectively. Figure 11c and d show the joint histogram of differences in the retrieved IWP and $D_{\rm me}$ between the Voronoi, sphere, and column schemes and the CoSSIR-MCBI results, respectively. Overall, the retrieved IWP and D_{me} parameters from the Voronoi ICS scheme agree well with the CoSSIR-MCBI results for all cases. Figure 12 shows the inversion results of IWP and D_{me} based on the optimal estimation algorithm from the CoSSIR aircraft-measured brightness temperature. For the Voronoi, sphere, and column ICS models, the corre-

lation coefficients of the retrieved IWP are 0.87, 0.67, and 0.74, with MAEs of 22.38, 23.55, and 22.91 g m⁻² and RM-SEs of 30.45, 35.22, and 32.64 g m^{-2} , respectively. The correlation coefficients of the D_{me} are 0.83, 0.64, and 0.76, with MAEs of 18.46, 21.19, and 19.91 µm and RMSEs of 24.57, 26.51, and 25.04 µm, respectively. Overall, the accuracy for the IWP and D_{me} based on the Voronoi ICS model is better than the other two ICS models compared to the CoSSIR-MCBI results. The sphere ICS model overestimates IWP and $D_{\rm me}$ compared with the validation data. According to the sensitivity results of Figs. 7 and 8, the Voronoi ICS scheme has higher BTD_{2-3} and BTD_{1-3} compared to the sphere and column ICS schemes, especially for large particles and IWP. This characteristic can be explicitly explained by the higher asymmetry factor of the Voronoi ICS model compared to the sphere and column ICS models. Thus, stronger forward-scattering energy can be detected for the Voronoi ICS model than for the other two models. The look-up ta-



Figure 7. The BTD₂₋₃ and BTD₁₋₃ for the (**a**, **d**) Voronoi, (**b**, **e**) sphere, and (**c**, **f**) column ICS models as functions of the IWP and D_{me} , respectively.



Figure 8. The BTD₂₋₃ and BTD₁₋₃ differences for the (a, c) Voronoi-minus-sphere and (b, d) Voronoi-minus-column ICS models as functions of the IWP and D_{me} , respectively.



Figure 9. The LUT of BTD_{2-3} and BTD_{1-3} for the (a) Voronoi, (b) column, and (c) sphere ICS models varying with the logarithm of IWP and D_{me} . Grey dots in circles represent the randomly generated 2000 test data from the RSTAR model.



Figure 10. The scatterplots of the randomly generated 2000 test data and the retrieved results. Panels (a) and (b) are for the IWP and D_{me} , respectively.

ble of the Voronoi ICS model can cover more IWP and $D_{\rm me}$. The brightness temperature variations in the Voronoi-shaped ice clouds are more prominent and sensitive to the IWP and $D_{\rm me}$. Therefore, the Voronoi ICS model results are better than the other two models. According to the look-up table as shown in Fig. 9, there are overlapping lines when $D_{\rm me}$ is small ($D_{\rm me} < 40 \,\mu$ m) and large ($D_{\rm me} > 140 \,\mu$ m). When BTD₂₋₃ and BTD₁₋₃ data fall under such overlapping lines of the look-up table, this overlapping region can lead to obtaining the same IWP and D_{me} when searching the look-up table.



Figure 11. The scattered red dots are the retrieved (a) IWP and (b) D_{me} (bottom row) validated by the results from the CoSSIR-MCBI algorithm (blue scattered dots). The joint histogram of differences in the retrieved (c) IWP and (d) D_{me} between the Voronoi, sphere, and column schemes and the CoSSIR-MCBI results, respectively.



Figure 12. The scatterplots of the retrieved IWP (**a**–**c**) and D_{me} (**d**–**f**) against the CoSSIR-MCBI results for the sphere (**c**, **f**), column (**b**, **e**), and Voronoi (**a**, **d**) ICS models.

5 Conclusions

In this study, we applied the irregularly shaped Voronoi ice crystal scattering (ICS) model to the ice cloud remote sensing retrieval of the ice water path (IWP) and particle size based on the Compact Scanning Submillimeter-wave Imaging Radiometer (CoSSIR) terahertz radiation measurements. The bulk-scattering property parameterization (Voronoi ICS scheme) in the terahertz region was developed based on the single-scattering properties of the Voronoi ICS database and 14 408 groups of particle size distributions from in situ observations. The Voronoi ICS scheme was applied to the atmo-

spheric radiative transfer model RSTAR and compared with the sphere and column ICS schemes to carry out the sensitivity analysis. We analysed the sensitivity of brightness temperature differences between cloudy and clear skies (BTDs) at 380, 640, and 874 GHz to the IWP and particle size. Based on the sensitivity analysis results, we built three terahertz multi-channel BTD look-up tables (LUTs) for the Voronoi, sphere, and column ICS schemes using the RSTAR atmospheric radiative transfer model. Based on the three LUTs, we utilized the Gaussian–Newtonian non-linear optimization estimation method to retrieve the IWP and particle size from the CoSSIR terahertz radiation measurements. Finally, the retrieval results were evaluated by the IWP and median mass diameter (D_{me}) derived from the CoSSIR-MCBI algorithm. The main conclusions were obtained as follows.

The bulk-scattering properties of ice clouds in the terahertz region, including the mass extinction coefficients, singlescattering albedo, asymmetry factor, and mass absorption coefficients of the Voronoi ICS scheme, were applied to the RSTAR model and were compared with the sphere and column ICS schemes. The results showed that the Voronoi ICS scheme has a distinct feature of lower-absorption properties and higher asymmetry factors for larger particle sizes in the terahertz region. This feature could be related to the complex, multifaceted shape of the Voronoi ICS model and suggests that the Voronoi ICS scheme can produce relatively stronger forward scattering and fewer absorption effects compared with the sphere and column ICS schemes.

The sensitivity analysis showed that the BTDs between 640 and 874 GHz are sensitive to the IWP variation, and the BTDs of 380, 640, and 874 GHz are sensitive to the particle size. The atmospheric absorption peak near 380 GHz and the atmospheric window near 640 and 874 GHz can be effectively used for the IWP in the range of $50-200 \text{ gm}^{-2}$ and particle size of 50-300 µm.

A comparison of the results from the Voronoi, sphere, and column ICS models shows that the results from the Voronoi ICS model are better than the sphere and column ICS models. The IWP and D_{me} retrieved from the Voronoi ICS scheme showed higher consistency with CoSSIR-MCBI results than the other two schemes, with correlation coefficients of 0.87 and 0.83 for IWP and D_{me} , respectively. For the Voronoi ICS model, the mean absolute errors (MAEs) of the IWP and $D_{\rm me}$ are improved by 5.0% and 12.8%, and RMSE is improved by 13.5 % and 7.3 %, respectively. The sphere ICS scheme overestimates the IWP and D_{me} by up to the MAEs of $23.55\,g\,m^{-2}$ and $21.19\,\mu m,$ respectively. This is mainly due to the differences in the absorption efficiency and asymmetry factor in the single-scattering properties of ice particles, which have a significant impact on the description of the scattering and radiative properties of ice clouds in the terahertz region.

In conclusion, the analysis of terahertz BTDs of 380, 640, and 874 GHz exhibits obvious sensitivity to the IWP and particle size of ice clouds, which could complement visible– infrared and microwave spectra. The present work provides the potential utility of inferring the IWP and particle size of ice clouds using BTDs of 380, 640, and 874 GHz. With the LUT for BTDs of 380, 640, and 874 GHz, the retrieval of terahertz ice cloud properties from airborne measurements based on the irregularly shaped Voronoi ICS models is newly developed. We find that the retrieval results based on the Voronoi ICS scheme present a better agreement with the CoSSIR-MCBI algorithm than the sphere and column ICS schemes. This study confirmed that the Voronoi ICS model has ice cloud inversion capabilities in the terahertz region, which may provide a reference for future use in aircraftbased and satellite-based terahertz ice cloud remote sensing applications.

Appendix A: Coefficients for Voronoi ICS scheme used in this study are tabulated in the Appendix

Table A1. Coefficients in the fitting of terahertz mass extinction coefficients $(m^2 g^{-1})$.

Frequency (GHz)	$a_0 (\mathrm{m}^2\mathrm{g}^{-1})$	$a_1 (\mathrm{m}^3\mathrm{g}^{-1})$
325	7.0891×10^{-1}	-1.6965×10^{1}
448	2.1347	-5.0405×10^{1}
664	7.5009	-1.6770×10^{2}
874	1.5790×10^{1}	-3.2850×10^{2}

Frequency (GHz)	b_0	<i>b</i> ₁	<i>b</i> ₂	<i>b</i> ₃
325	-3.1317×10^{-1}	2.7448×10^{-2}	-2.0449×10^{-4}	5.0815×10^{-7}
448	$-2.3947 imes 10^{-1}$	2.9461×10^{-2}	$-2.4145 imes 10^{-4}$	6.4366×10^{-7}
664	$-8.2857 imes 10^{-2}$	$2.7985 imes 10^{-2}$	$-2.4357 imes 10^{-4}$	$6.7691 imes 10^{-7}$
874	4.7425×10^{-2}	2.5164×10^{-2}	$-2.2395 imes 10^{-4}$	6.3152×10^{-7}

Table A2. Coefficients in the fitting of terahertz single-scattering albedo.

Table A3. Coefficients in the fitting of terahertz asymmetry factor.

Frequency (GHz)	<i>c</i> ₀	<i>c</i> ₁	<i>c</i> ₂	<i>c</i> ₃
325	2.2045×10^{-2}	$-8.2487 imes 10^{-4}$	2.5764×10^{-5}	-4.7767×10^{-8}
448	1.0168×10^{-2}	-5.1223×10^{-5}	3.0599×10^{-5}	$-8.0591 imes 10^{-8}$
664	$-4.4704 imes 10^{-2}$	3.5331×10^{-3}	1.2997×10^{-5}	-7.2297×10^{-8}
874	-1.1685×10^{-1}	8.8403×10^{-3}	-3.0410×10^{-5}	2.6790×10^{-8}

Table A4. Coefficients in the fitting of terahertz mass-averaged absorption coefficients (m² g⁻¹).

Frequency (GHz)	$d_0 (\mathrm{m}^2 \mathrm{g}^{-1})$	$d_1 ({\rm mg^{-1}})$	$d_2 ({\rm g}^{-1})$	$d_3 (\mathrm{m}^{-1}\mathrm{g}^{-1})$
325	4.4262×10^{-2}	1.5585×10^{-4}	9.6647×10^{-7}	-5.1271×10^{-9}
448	8.2110×10^{-2}	5.0544×10^{-4}	2.0336×10^{-6}	-1.2945×10^{-8}
664	1.6909×10^{-1}	2.4299×10^{-3}	1.2784×10^{-6}	-3.3930×10^{-8}
874	2.6509×10^{-1}	7.6295×10^{-3}	-1.4488×10^{-5}	$-3.9275 imes 10^{-8}$

Data availability. The CoSSIR/ER-2 aircraft data during the TC4 mission on 17 and 19 July 2007 are available at https://espoarchive.nasa.gov/archive/browse/tc4/ER2 (NASA, 2022). The IWP and $D_{\rm me}$ from the CoSSIR-MCBI algorithm are available at https:// espoarchive.nasa.gov/archive/browse/tc4/ER2 (NASA, 2022). The 14408 groups of PSDs from 11 field flight observation experiments are available at https://www.ssec.wisc.edu/ice_models/microphysical_data.html (SSEC, 2022).

Author contributions. ML developed the terahertz ice cloud remote sensing inversion algorithm for the IWP and particle size of ice clouds and evaluated the retrieval result by validating it against the results from the CoSSIR-MCBI algorithm. ML is also responsible for downloading auxiliary data and writing the initial draft of this paper. HL provided the single-scattering property database of Voronoi models in the terahertz region and assisted in developing the parameterization of ice cloud scattering properties in the RSTAR model. HL also designed the aims and structures of this study and guided the writing and revisions of the manuscript.

HI developed the single-scattering property database of Voronoi models in the terahertz region and helped in the writing and revisions of the manuscript.

TYN provided the atmospheric radiative transfer model RSTAR and is responsible for the optimization of the RSTAR model and assisted in the parameterization of ice cloud scattering properties.

SL and LL assisted in developing the ice cloud remote sensing retrieval algorithm of the IWP and particle size of ice clouds, provided the CoSSIR terahertz radiation measurement data, and helped with reviewing the manuscript.

DJ guided the development of the ice cloud remote sensing retrieval algorithm and helped with the review of the manuscript.

HS assisted in analysing the results, guided the flowchart of the study, and reviewed the manuscript. CS assisted in designing the structures of this study, guided the writing of the paper, and helped review the manuscript.

Competing interests. The contact author has declared that none of the authors has any competing interests.

Disclaimer. Publisher's note: Copernicus Publications remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

Review statement. This paper was edited by Alexander Kokhanovsky and reviewed by five anonymous referees.

References

Baran, A., Ishimoto, H., Sourdeval, O., Hesse, E., and Harlow, C.: The applicability of physical optics in the millimetre and sub-millimetre spectral region. Part II: Application to a three-component model of ice cloud and its evaluation against the bulk single-scattering properties of various other aggregate models, J. Quant. Spectrosc. Ra., 206, 83–100, https://doi.org/10.1016/j.jqsrt.2017.10.027, 2018.

- Baran, A. J.: A review of the light scattering properties of cirrus, J. Quant. Spectrosc. Ra., 110, 1239–1260, https://doi.org/10.1016/j.jqsrt.2009.02.026, 2009.
- Baran, A. J.: From the single-scattering properties of ice crystals to climate prediction: A way forward, Atmos. Res., 112, 45–69, https://doi.org/10.1016/j.atmosres.2012.04.010, 2012.
- Baran, A. J. and Labonnote, L. C.: A self-consistent scattering model for cirrus. I: The solar region, Q. J. Roy. Meteor. Soc., 133, 1899–1912, https://doi.org/10.1002/qj.164, 2007.
- Baran, A. J., Cotton, R., Furtado, K., Havemann, S., Labonnote, L. C., Marenco, F., Smith, A., and Thelen, J. C.: A self-consistent scattering model for cirrus. II: The high and low frequencies, Q. J. Roy. Meteor. Soc., 140, 1039–1057, https://doi.org/10.1002/qj.2193, 2014a.
- Baran, A. J., Hill, P., Furtado, K., Field, P., and Manners, J.: A Coupled Cloud Physics Radiation Parameterization of the Bulk Optical Properties of Cirrus and Its Impact on the Met Office Unified Model Global Atmosphere 5.0 Configuration, J. Climate, 27, 7725–7752, 2014b.
- Baum, B. A., Heymsfield, A. J., Yang, P., and Bedka, S. T.: Bulk scattering properties for the remote sensing of ice clouds. Part I: Microphysical data and models, J. Appl. Meteorol., 44, 1885– 1895, https://doi.org/10.1175/JAM2308.1, 2005a.
- Baum, B. A., Yang, P., Heymsfield, A. J., Platnick, S., King, M. D., Hu, Y. X., and Bedka, S. T.: Bulk scattering properties for the remote sensing of ice clouds. Part II: Narrowband models, J. Appl. Meteorol., 44, 1896–1911, https://doi.org/10.1175/JAM2309.1, 2005b.
- Brath, M., Fox, S., Eriksson, P., Harlow, R. C., Burgdorf, M., and Buehler, S. A.: Retrieval of an ice water path over the ocean from ISMAR and MARSS millimeter and submillimeter brightness temperatures, Atmos. Meas. Tech., 11, 611–632, https://doi.org/10.5194/amt-11-611-2018, 2018.
- Buehler, S., Jimenez, C., Evans, K., Eriksson, P., Rydberg, B., Heymsfield, A., Stubenrauch, C., Lohmann, U., Emde, C., John, V., Tr, S., and Davis, C.: A concept for a satellite mission to measure cloud ice water path and ice particle size, Q. J. Roy. Meteor. Soc., 133, 109–128, https://doi.org/10.1002/qj.143, 2007.
- C.-Labonnote, L., Brogniez, G., Doutriaux-Boucher, M., Buriez, J.-C., Gayet, J.-F., and Chepfer, H.: Modeling of light scattering in cirrus clouds with inhomogeneous hexagonal monocrystals. Comparison with in-situ and ADEOS-POLDER measurements, Geophys. Res. Lett., 27, 113–116, https://doi.org/10.1029/1999GL010839, 2000.
- C.-Labonnote, L., Brogniez, G., Buriez, J.-C., Doutriaux-Boucher, M., Gayet, J.-F., and Macke, A.: Polarized light scattering by inhomogeneous hexagonal monocrystals: Validation with ADEOS-POLDER measurements, J. Geophys. Res.-Atmos., 106, 12139–12153, https://doi.org/10.1029/2000JD900642, 2001.
- Ceccaldi, M., Delanoë, J., Hogan, R., Pounder, N., Protat, A., and Pelon, J.: From CloudSat-CALIPSO to EarthCare: Evolution of the DARDAR cloud classification and its comparison to airborne radar-lidar observations, J. Geophys. Res.-Atmos., 118, 7962– 7981, https://doi.org/10.1002/jgrd.50579, 2013.

- Chen, Q. and Zhang, H.: Effects of ice crystal habit weight on ice cloud optical properties and radiation, Acta Meteorol. Sin., 76, 279–288, https://doi.org/10.11676/qxxb2017.088, 2018.
- Cho, H.-M., Zhang, Z., Meyer, K., Lebsock, M., Platnick, S., Ackerman, A. S., Di Girolamo, L., C.-Labonnote, L., Cornet, C., Riedi, J., and Holz, R. E.: Frequency and causes of failed MODIS cloud property retrievals for liquid phase clouds over global oceans, J. Geophys. Res.-Atmos., 120, 4132–4154, https://doi.org/10.1002/2015JD023161, 2015.
- Cimini, D., Westwater, E., Gasiewski, A., Klein, M., Leuski, V., Mattioli, V., Dowlatshahi, S., and Liljegren, J.: Ground-Based Millimeter- and Submillimiter Wave Observations of the Arctic Atmosphere, in: 2006 IEEE MicroRad, San Juan, PR, USA, 28 February–3 March 2006, 247–251, https://doi.org/10.1109/MICRAD.2006.1677097, 2006.
- Deeter, M. and Evans, K.: A Novel Ice-Cloud Retrieval Algorithm Based on the Millimeter-Wave Imaging Radiometer (MIR) 150- and 220-GHz Channels, J. Appl. Meteorol., 39, 623–633, https://doi.org/10.1175/1520-0450-39.5.623, 2000.
- Deschamps, P. Y., Breon, F. M., Leroy, M., Podaire, A., Bricaud, A., Buriez, J. C., and Seze, G.: The POLDER Mission: Instrument Characteristics and Scientific Objectives, IEEE T. Geosci. Remote, 32, 598–615, https://doi.org/10.1109/36.297978, 1994.
- Doutriaux-Boucher, M., Buriez, J.-C., Brogniez, G., C.-Labonnote, L., and Baran, A.: Sensitivity of retrieved POLDER directional cloud optical thickness to various ice particle models, Geophys. Res. Lett., 27, 109–112, https://doi.org/10.1029/1999GL010870, 2000.
- Draine, B. T. and Flatau, P. J.: Discrete-Dipole Approximation for Scattering Calculations, J. Opt. Soc. Am. A, 11, 1491–1499, https://doi.org/10.1364/JOSAA.11.001491, 1994.
- Eriksson, P., Ekelund, R., Mendrok, J., Brath, M., Lemke, O., and Buehler, S. A.: A general database of hydrometeor single scattering properties at microwave and submillimetre wavelengths, Earth Syst. Sci. Data, 10, 1301–1326, https://doi.org/10.5194/essd-10-1301-2018, 2018.
- Eriksson, P., Rydberg, B., Mattioli, V., Thoss, A., Accadia, C., Klein, U., and Buehler, S. A.: Towards an operational Ice Cloud Imager (ICI) retrieval product, Atmos. Meas. Tech., 13, 53–71, https://doi.org/10.5194/amt-13-53-2020, 2020.
- Evans, K., Wang, J., Racette, P., Heymsfield, G., and Li, L.: Ice Cloud Retrievals and Analysis with the Compact Scanning Submillimeter Imaging Radiometer and the Cloud Radar System during CRYSTAL FACE, J. Appl. Meteorol., 44, 839–859, https://doi.org/10.1175/JAM2250.1, 2005.
- Evans, K. F. and Stephens, G. L.: Microwave Radiative Transfer through Clouds Composed of Realistically Shaped Ice Crystals. Part I. Single Scattering Properties, J. Atmos. Sci., 52, 2041–2057, https://doi.org/10.1175/1520-0469(1995)052<2041:MRTTCC>2.0.CO;2, 1995a.
- Evans, K. F. and Stephens, G. L.: Microwave Radiative Transfer through Clouds Composed of Realistically Shaped Ice Crystals. Part II. Remote Sensing of Ice Clouds, J. Atmos. Sci., 52, 2058–2072, https://doi.org/10.1175/1520-0469(1995)052<2058:MRTTCC>2.0.CO;2, 1995b.
- Evans, K. F., Walter, S. J., Heymsfield, A. J., and Deeter, M. N.: Modeling of Submillimeter Passive Remote Sensing of Cirrus Clouds, J. Appl. Me-

teorol., 37, 184–205, https://doi.org/10.1175/1520-0450(1998)037<0184:MOSPRS>2.0.CO;2, 1998.

- Evans, K. F., Walter, S. J., Heymsfield, A. J., and McFarquhar, G. M.: Submillimeter-Wave Cloud Ice Radiometer: Simulations of retrieval algorithm performance, J. Geophys. Res.-Atmos., 107, AAC 2-1–AAC 2-21, https://doi.org/10.1029/2001JD000709, 2002.
- Evans, K. F., Wang, J. R., O'C Starr, D., Heymsfield, G., Li, L., Tian, L., Lawson, R. P., Heymsfield, A. J., and Bansemer, A.: Ice hydrometeor profile retrieval algorithm for highfrequency microwave radiometers: application to the CoSSIR instrument during TC4, Atmos. Meas. Tech., 5, 2277–2306, https://doi.org/10.5194/amt-5-2277-2012, 2012.
- Forster, P., Storelvmo, T., Armour, K., Collins, W., Dufresne, J. L., Frame, D., Lunt, D. J., Mauritsen, T., Palmer, M. D., Watanabe, M., Wild, M., and Zhang, H.: The Earth's Energy Budget, Climate Feedbacks, and Climate Sensitivity, in: Climate Change 2021: The Physical Science Basis. Contribution of Working Group I to the Sixth Assessment Report of the Intergovernmental Panel on Climate Change, edited by: Masson-Delmotte, V., Zhai, P., Pirani, A., Connors, S. L., Péan, C., Berger, S., Caud, N., Chen, Y., Goldfarb, L., Gomis, M. I., Huang, M., Leitzell, K., Lonnoy, E., Matthews, J. B. R., Maycock, T. K., Waterfield, T., Yelekçi, O., Yu, R., and Zhou, B., Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA, 923– 1054, 2021.
- Fox, S.: An Evaluation of Radiative Transfer Simulations of Cloudy Scenes from a Numerical Weather Prediction Model at Sub-Millimetre Frequencies Using Airborne Observations, Remote Sens.-Basel, 12, 2758, https://doi.org/10.3390/rs12172758, 2020.
- Fox, S., Lee, C., Moyna, B., Philipp, M., Rule, I., Rogers, S., King, R., Oldfield, M., Rea, S., Henry, M., Wang, H., and Harlow, R. C.: ISMAR: an airborne submillimetre radiometer, Atmos. Meas. Tech., 10, 477–490, https://doi.org/10.5194/amt-10-477-2017, 2017.
- Fox, S., Mendrok, J., Eriksson, P., Ekelund, R., O'Shea, S. J., Bower, K. N., Baran, A. J., Harlow, R. C., and Pickering, J. C.: Airborne validation of radiative transfer modelling of ice clouds at millimetre and sub-millimetre wavelengths, Atmos. Meas. Tech., 12, 1599–1617, https://doi.org/10.5194/amt-12-1599-2019, 2019.
- Fu, Q., Sun, W. B., and Yang, P.: Modeling of scattering and absorption by nonspherical cirrus ice particles at thermal infrared wavelengths, J. Atmos. Sci., 56, 2937–2947, https://doi.org/10.1175/1520-0469(1999)056<2937:MOSAAB>2.0.CO;2, 1999.
- Gao, T., Li, S., Liu, L., and Huang, W.: Development study of THz instruments for atmospheric sounding, Infrared and Laser Engineering, 45, 56–67, https://doi.org/10.3788/IRLA201645.0425002, 2016.
- Gasiewski, A. J.: Numerical sensitivity analysis of passive EHF and SMMW channels to tropospheric water vapor, clouds, and precipitation, IEEE T. Geosci. Remote, 30, 859–870, https://doi.org/10.1109/36.175320, 1992.
- Groth, S. P., Baran, A. J., Betcke, T., Havemann, S., and Smigaj, W.: The boundary element method for light scattering by ice crystals and its implementation in BEM plus, J. Quant. Spectrosc. Ra., 167, 40–52, 2015.

- Hans, H., Bell, W., Berrisford, P., Andras, H., Muñoz-Sabater, J., Nicolas, J., Raluca, R., Dinand, S., Adrian, S., Cornel, S., and Dick, D.: Global reanalysis: goodbye ERA-Interim, hello ERA5, ECMWF Newsletter, 159, 17–24, https://doi.org/10.1002/qj.3803, 2019.
- Heymsfield, A. J., Schmitt, C., and Bansemer, A.: Ice Cloud Particle Size Distributions and Pressure-Dependent Terminal Velocities from In Situ Observations at Temperatures from 0° to -86 °C, J. Atmos. Sci., 70, 4123–4154, https://doi.org/10.1175/JAS-D-12-0124.1, 2013.
- Heymsfield, A. J., Krämer, M., Luebke, A., Brown, P., Cziczo, D. J., Franklin, C., Lawson, P., Lohmann, U., McFarquhar, G., Ulanowski, Z., and Van Tricht, K.: Cirrus Clouds, Meteor. Mon., 58, 2.1–2.26, 2017.
- Hong, G.: Parameterization of scattering and absorption properties of nonspherical ice crystals at microwave frequencies, J. Geophys. Res., 112, D11208, https://doi.org/10.1029/2006JD008364, 2007.
- Hong, G., Yang, P., Baum, B. A., Heymsfield, A. J., Weng, F., Liu, Q., Heygster, G., and Buehler, S. A.: Scattering database in the millimeter and submillimeter wave range of 100–1000 GHz for nonspherical ice particles, J. Geophys. Res.-Atmos., 114, D06201, https://doi.org/10.1029/2008JD010451, 2009.
- Hong, Y., Liu, G., and Li, J. L. F.: Assessing the Radiative Effects of Global Ice Clouds Based on CloudSat and CALIPSO Measurements, J. Climate, 29, 7651–7674, https://doi.org/10.1175/JCLI-D-15-0799.1, 2016.
- Illingworth, A. J., Barker, H. W., Beljaars, A., Ceccaldi, M., Chepfer, H., Clerbaux, N., Cole, J., Delanoe, J., Domenech, C., Donovan, D. P., Fukuda, S., Hirakata, M., Hogan, R. J., Huenerbein, A., Kollias, P., Kubota, T., Nakajima, T., Nakajima, T. Y., Nishizawa, T., Ohno, Y., Okamoto, H., Oki, R., Sato, K., Satoh, M., Shephard, M. W., Velazquez-Blazquez, A., Wandinger, U., Wehr, T., and van Zadelhoff, G.-J.: The Earth-CARE Satellite: The Next Step Forward in Global Measurements of Clouds, Aerosols, Precipitation, and Radiation, B. Am. Meteorol. Soc., 96, 1311–1332, https://doi.org/10.1175/BAMS-D-12-00227.1, 2015.
- Inatani, J., Ozeki, H., Satoh, R., Nishibori, T., Ikeda, N., Fujii, Y., Nakajima, T., Iida, Y., Iida, T., Kikuchi, K., Miura, T., and Masuko, H.: Submillimeter limb-emission sounder JEM/S-MILES aboard the Space Station, Proc. SPIE, 4152, 243–254, https://doi.org/10.1117/12.410604, 2000.
- Ishimoto, H., Masuda, K., Mano, Y., Orikasa, N., and Uchiyama, A.: Irregularly shaped ice aggregates in optical modeling of convectively generated ice clouds, J. Quant. Spectrosc. Ra., 113, 632–643, https://doi.org/10.1016/j.jqsrt.2012.01.017, 2012.
- Ishimoto, H., Masuda, K., Mano, Y., Orikasa, N., and Uchiyama, A.: Optical Modeling of Irregularly Shaped Ice Particles in Convective Cirrus, AIP Conf. Proc., 1531, 184–187, https://doi.org/10.1063/1.4804737, 2013.
- Jimenez, C., Eriksson, P., and Murtagh, D.: First inversions of observed submillimeter limb sounding radiances by neural networks, J. Geophys. Res., 108, 4791, https://doi.org/10.1029/2003JD003826, 2003.
- Jimenez, C., Buehler, S., Rydberg, B., Eriksson, P., and Evans, K.: Performance simulations for a submillimetre wave cloud ice satellite instrument, Q. J. Roy. Meteor. Soc., 133, 129–149, https://doi.org/10.1002/qj.134, 2007.

- Kangas, V., D'Addio, S., Klein, U., Loiselet, M., Mason, G., Orlhac, J.-C., Gonzalez, R., Bergada, M., Brandt, M., and Thomas, B.: Ice cloud imager instrument for MetOp Second Generation, 2014 13th Specialist Meeting on Microwave Radiometry and Remote Sensing of the Environment (MicroRad), Pasadena, CA, 24–27 March 2014, 228–231, https://doi.org/10.1109/MicroRad.2014.6878946, 2014.
- Kleanthous, A., Betcke, T., Hewett, D., Escapil-Inchauspé, P., Jerez-Hanckes, C., and Baran, A.: Accelerated Calderón preconditioning for Maxwell transmission problems, J. Comput. Phys., 458, 111099, https://doi.org/10.1016/j.jcp.2022.111099, 2022.
- Lawson, R. P., Baker, B., Pilson, B., and Mo, Q. X.: In situ observations of the microphysical properties of wave, cirrus, and anvil clouds. Part II: Cirrus clouds, J. Atmos. Sci., 63, 3186–3203, https://doi.org/10.1175/JAS3803.1, 2006.
- Lawson, R. P., Woods, S., Jensen, E., Erfani, E., Gurganus, C., Gallagher, M., Connolly, P., Whiteway, J., Baran, A. J., May, P., Heymsfield, A., Schmitt, C. G., McFarquhar, G., Um, J., Protat, A., Bailey, M., Lance, S., Muehlbauer, A., Stith, J., Korolev, A., Toon, O. B., and Kramer, M.: A Review of Ice Particle Shapes in Cirrus formed In Situ and in Anvils, J. Geophys. Res.-Atmos., 124, 10049–10090, https://doi.org/10.1029/2018JD030122, 2019.
- Letu, H., Nakajima, T. Y., and Matsui, T. N.: Development of an ice crystal scattering database for the global change observation mission/second generation global imager satellite mission: investigating the refractive index grid system and potential retrieval error, Appl. Optics, 51, 6172–6178, https://doi.org/10.1364/AO.51.006172, 2012.
- Letu, H., Ishimoto, H., Riedi, J., Nakajima, T. Y., C.-Labonnote, L., Baran, A. J., Nagao, T. M., and Sekiguchi, M.: Investigation of ice particle habits to be used for ice cloud remote sensing for the GCOM-C satellite mission, Atmos. Chem. Phys., 16, 12287– 12303, https://doi.org/10.5194/acp-16-12287-2016, 2016.
- Letu, H. S., Nagao, T. M., Nakajima, T. Y., Riedi, J., Ishimoto, H., Baran, A. J., Shang, H. Z., Sekiguchi, M., and Kikuchi, M.: Ice Cloud Properties From Himawari-8/AHI Next-Generation Geostationary Satellite: Capability of the AHI to Monitor the DC Cloud Generation Process, IEEE T. Geosci. Remote, 57, 3229– 3239, https://doi.org/10.1109/TGRS.2018.2882803, 2018.
- Li, M., Letu, H., Peng, Y., Ishimoto, H., Lin, Y., Nakajima, T. Y., Baran, A. J., Guo, Z., Lei, Y., and Shi, J.: Investigation of ice cloud modeling capabilities for the irregularly shaped Voronoi ice scattering models in climate simulations, Atmos. Chem. Phys., 22, 4809–4825, https://doi.org/10.5194/acp-22-4809-2022, 2022.
- Li, S., Liu, L., Gao, T., Huang, W., and Hu, S.: Sensitivity analysis of terahertz wave passive remote sensing of cirrus microphysical parameters, Acta Physica Sinica, 65, 100–110, https://doi.org/10.7498/aps.65.134102, 2016.
- Li, S., Liu, L., Gao, T., Hu, S., and Huang, W.: Retrieval method of cirrus microphysical parameters at terahertz wave based on multiple lookup tables, Acta Physica Sinica, 66, 78–87, https://doi.org/10.7498/aps.66.054102, 2017.
- Li, S., Liu, L., Gao, T., Shi, L., Qiu, S., and Hu, S.: Radiation characteristics of the selected channels for cirrus remote sensing in terahertz waveband and the influence factors for the retrieval method, J. Infrared Millim. W., 37, 60–66, https://doi.org/10.11972/j.issn.1001-9014.2018.01.012, 2018.

- Lin, B. and Rossow, W.: Seasonal Variation of Liquid and Ice Water Path in Nonprecipitating Clouds over Oceans, J. Climate, 9, 2890–2902, https://doi.org/10.1175/1520-0442(1996)009<2890:SVOLAI>2.0.CO;2, 1996.
- Liou, K.-N.: Radiation and cloud processes in the atmosphere: theory, observation and modeling, Oxford monographs on geology and geophysics, 20, Oxford University Press, New York, ix, 487 pp., https://doi.org/10.1063/1.2809044, 1992.
- Liu, G. and Curry, J.: Remote Sensing of Ice Water Characteristics in Tropical Clouds Using Aircraft Microwave Measurements, J. Appl. Meteorol., 37, 337–355, https://doi.org/10.1175/1520-0450(1998)037<0337:RSOIWC>2.0.CO;2, 1998.
- Liu, G. and Curry, J.: Tropical Ice Water Amount and Its Relations to Other Atmospheric Hydrological Parameters as Inferred from Satellite Data, J. Appl. Meteorol., 38, 1182–1194, https://doi.org/10.1175/1520-0450(1999)038<1182:TIWAAI>2.0.CO;2, 1999.
- Liu, G. and Curry, J.: Determination of Ice Water Path and Mass Median Particle Size Using Multichannel Microwave Measurements, J. Appl. Meteorol., 39, 1318–1329, https://doi.org/10.1175/1520-0450(2000)039<1318:DOIWPA>2.0.CO;2, 2000.
- Liu, L., Weng, C., Li, S., Hu, S., Ye, J., Dou, F., and Shang, J.: Review of terahertz passive remote sensing of ice clouds, Advances in Earth Science, 35, 1211–1221, https://doi.org/10.11867/j.issn.1001-8166.2020.103, 2020.
- Liu, L., Weng, C., Li, S., Letu, H., Hu, S., and Dong, P.: Passive Remote Sensing of Ice Cloud Properties at Terahertz Wavelengths Based on Genetic Algorithm, Remote Sens.-Basel, 13, 735, https://doi.org/10.3390/rs13040735, 2021.
- Macke, A., Mishchenko, M. I., and Cairns, B.: The influence of inclusions on light scattering by large ice particles, J. Geophys. Res.-Atmos., 101, 23311–23316, https://doi.org/10.1029/96jd02364, 1996.
- Marks, C. and Rodgers, C.: A retrieval method for atmospheric composition from limb emission measurements, J. Geophys. Res., 98, 14939–14953, https://doi.org/10.1029/93JD01195, 1993.
- Minnis, P., Heck, P., and Young, D.: Inference of Cirrus Cloud Properties Using Satellite-observed Visible and Infrared Radiances.
 Part II: Verification of Theoretical Cirrus Radiative Properties, J. Atmos. Sci., 50, 1305–1322, https://doi.org/10.1175/1520-0469(1993)050<1305:IOCCPU>2.0.CO;2, 1993a.
- Minnis, P., Liou, K.-N., and Takano, Y.: Inference of Cirrus Cloud Properties Using Satellite-observed Visible and Infrared Radiances. Part I: Parameterization of Radiance Fields, J. Atmos. Sci., 50, 1279–1304, https://doi.org/10.1175/1520-0469(1993)050<1279:IOCCPU>2.0.CO;2, 1993b.
- Nakajima, T. and King, M. D.: Determination of the Optical-Thickness and Effective Particle Radius of Clouds from Reflected Solar-Radiation Measurements. 1. Theory, J. Atmos. Sci., 47, 1878–1893, https://doi.org/10.1175/1520-0469(1990)047<1878:Dotota>2.0.Co;2, 1990.
- Nakajima, T. and Tanaka, M.: Matrix formulation for the transfer of solar radiation in a plane-parallel scattering atmosphere, J. Quant. Spectrosc. Ra., 35, 13–21, https://doi.org/10.1016/0022-4073(86)90088-9, 1986.
- Nakajima, T. and Tanaka, M.: Algorithms for radiative intensity calculations in moderately thick atmospheres using trun-

cation approximation, J. Quant. Spectrosc. Ra., 40, 51–69, https://doi.org/10.1016/0022-4073(88)90031-3, 1988.

- Nakajima, T., King, M. D., Spinhirne, J. D., and Radke, L. F.: Determination of the Optical-Thickness and Effective Particle Radius of Clouds from Reflected Solar-Radiation Measurements. 2. Marine Stratocumulus Observations, J. Atmos. Sci., 48, 728–750, https://doi.org/10.1175/1520-0469(1991)048<0728:Dotota>2.0.Co;2, 1991.
- Nakajima, T., Nakajima, T., Yoshimori, K., Mishra, S., and Tripathi, S.: Development of a light scattering solver applicable to particles of arbitrary shape on the basis of the surface-integral equations method of Müller type. I. Methodology, accuracy of calculation, and electromagnetic current on the particle surface, Appl. Optics, 48, 3526–3536, https://doi.org/10.1364/AO.48.003526, 2009.
- Nakajima, T. Y. and Nakajima, T.: Wide-area determination of cloud microphysical properties from NOAA AVHRR measurements for FIRE and ASTEX regions, J. Atmos. Sci., 52, 4043–4059, https://doi.org/10.1175/1520-0469(1995)052<4043:Wadocm>2.0.Co;2, 1995.
- Nakajima, T. Y., Ishida, H., Nagao, T. M., Hori, M., Letu, H., Higuchi, R., Tamaru, N., Imoto, N., and Yamazaki, A.: Theoretical basis of the algorithms and early phase results of the GCOM-C (Shikisai) SGLI cloud products, Prog. Earth Planet. Sc., 6, 1– 25, https://doi.org/10.1186/s40645-019-0295-9, 2019.
- NASA: ESPO Data Archive, NASA [data set], https://espoarchive. nasa.gov/archive/browse/tc4/ER2, last access: 26 December 2022.
- Platnick, S., King, M. D., Ackerman, S. A., Menzel, W. P., Baum, B. A., Riedi, J. C., and Frey, R. A.: The MODIS cloud products: algorithms and examples from Terra, IEEE T. Geosci. Remote, 41, 459–473, https://doi.org/10.1109/TGRS.2002.808301, 2003.
- Platnick, S., Meyer, K. G., King, M. D., Wind, G., Amarasinghe, N., Marchant, B., Arnold, G. T., Zhang, Z. B., Hubanks, P. A., Holz, R. E., Yang, P., Ridgway, W. L., and Riedi, J.: The MODIS Cloud Optical and Microphysical Products: Collection 6 Updates and Examples From Terra and Aqua, IEEE T. Geosci. Remote, 55, 502–525, https://doi.org/10.1109/Tgrs.2016.2610522, 2017.
- Racette, P., Dod, L., Shiue, J., Adler, R., Jackson, D., Gasiewski, A., and Zacharias, D.: An Airborne Millimeter-Wave Imaging Radiometer for Cloud, Precipitation, and Atmospheric Water Vapor Studies, J. Atmos. Ocean. Tech., 13, 610–619, https://doi.org/10.1175/1520-0426(1996)013<0610:AAMWIR>2.0.CO;2, 1992.
- Rossow, W.: Clouds in Weather and Climate: The International Satellite Cloud Climatology Project at 30: What Do We Know and What Do We Still Need to Know?, B. Am. Meteorol. Soc., 95, 441–443, https://doi.org/10.1175/BAMS-D-13-00138.1, 2014.
- Rossow, W. B. and Schiffer, R. A.: Isccp Cloud Data Products, B. Am. Meteorol. Soc., 72, 2–20, https://doi.org/10.1175/1520-0477(1991)072<0002:ICDP>2.0.CO;2, 1991.
- Sekiguchi, M. and Nakajima, T.: A k-distribution-based radiation code and its computational optimization for an atmospheric general circulation model, J. Quant. Spectrosc. Ra., 109, 2779–2793, https://doi.org/10.1016/j.jqsrt.2008.07.013, 2008.
- Sieron, S., Clothiaux, E., Zhang, F., Lu, Y., and Otkin, J.: Comparison of using distribution-specific versus effective radius methods for hydrometeor single-scattering properties for all-

sky microwave satellite radiance simulations with different microphysics parameterization schemes, J. Geophys. Res.-Atmos., 122, 7027–7046, https://doi.org/10.1002/2017JD026494, 2017.

- SSEC: The 14408 groups of PSDs from 11 field flight observation experiments, SSEC [data set], https://www.ssec.wisc.edu/ ice_models/microphysical_data.html, last access: 18 February 2022.
- Stephens, G. L., Li, J., Wild, M., Clayson, C. A., Loeb, N., Kato, S., L'Ecuyer, T., Stackhouse, P. W., Lebsock, M., and Andrews, T.: An update on Earth's energy balance in light of the latest global observations, Nat. Geosci., 5, 691–696, https://doi.org/10.1038/ngeo1580, 2012.
- Takano, Y. and Liou, K. N.: Solar Radiative-Transfer in Cirrus Clouds. 1. Single-Scattering and Optical-Properties of Hexagonal Ice Crystals, J. Atmos. Sci., 46, 3–19, https://doi.org/10.1175/1520-0469(1989)046<0003:SRTICC>2.0.CO;2, 1989.
- Van de Hulst, H. C.: Light scattering by small particles, Wiley, New York,, xiii, 470 pp., https://doi.org/10.1063/1.3060205, 1957.
- Warren, S.: Optical Constants of Ice from the Ultraviolet to the Microwave, Appl. Optics, 23, 1206, https://doi.org/10.1364/AO.23.001206, 1984.
- Warren, S. G. and Brandt, R. E.: Optical constants of ice from the ultraviolet to the microwave: A revised compilation, J. Geophys. Res.-Atmos., 113, D14220, https://doi.org/10.1029/2007JD009744, 2008.
- Weng, C., Liu, L., Gao, T., Hu, S., Li, S., Dou, F., and Shang, J.: Multi-Channel Regression Inversion Method for Passive Remote Sensing of Ice Water Path in the Terahertz Band, Atmosphere-Basel, 10, 437, https://doi.org/10.3390/atmos10080437, 2019.
- Wu, D. L., Esper, J., Ehsan, N., Piepmeier, J. R., and Racette, P.: Icecube: Spaceflight Validation of an 874 GHz Submillimeter Wave Radiometer for Ice Cloud Remote Sensing, 2014/12/1, 2014 Earth Science Technology Forum. Location: Leesburg, Virginia, USA, 28–30 October 2014, A22F-02, https://doi.org/10.1117/12.2530589.
- Yang, P. and Liou, K. N.: Finite-difference time domain method for light scattering by small ice crystals in threedimensional space, J. Opt. Soc. Am. A, 13, 2072–2085, https://doi.org/10.1364/JOSAA.13.002072, 1996a.
- Yang, P. and Liou, K. N.: Geometric-optics-integral-equation method for light scattering by nonspherical ice crystals, Appl. Optics, 35, 6568–6584, https://doi.org/10.1364/AO.35.006568, 1996b.

- Yang, P., Liou, K. N., Mishchenko, M. I., and Gao, B. C.: Efficient finite-difference time-domain scheme for light scattering by dielectric particles: application to aerosols, Appl. Optics, 39, 3727– 3737, https://doi.org/10.1364/Ao.39.003727, 2000a.
- Yang, P., Liou, K. N., Wyser, K., and Mitchell, D.: Parameterization of the scattering and absorption properties of individual ice crystals, J. Geophys. Res.-Atmos., 105, 4699–4718, 2000b.
- Yang, P., Bi, L., Baum, B. A., Liou, K. N., Kattawar, G. W., Mishchenko, M. I., and Cole, B.: Spectrally Consistent Scattering, Absorption, and Polarization Properties of Atmospheric Ice Crystals at Wavelengths from 0.2 to 100 µm, J. Atmos. Sci., 70, 330–347, https://doi.org/10.1175/JAS-D-12-039.1, 2013.
- Yang, P., Liou, K. N., Bi, L., Liu, C., Yi, B. Q., and Baum, B. A.: On the Radiative Properties of Ice Clouds: Light Scattering, Remote Sensing, and Radiation Parameterization, Adv. Atmos. Sci., 32, 32–63, https://doi.org/10.1007/s00376-014-0011-z, 2015.
- Yang, P., Hioki, S., Saito, M., Kuo, C. P., Baum, B. A., and Liou, K. N.: A Review of Ice Cloud Optical Property Models for Passive Satellite Remote Sensing, Atmosphere-Basel, 9, 499, https://doi.org/10.3390/atmos9120499, 2018.
- Yee, K. S.: Numerical solution of initial boundary value problems involving maxwell's equations in isotropic media, IEEE T. Antenn. Propag., 14, 302–307, https://doi.org/10.1109/TAP.1966.1138693, 1966.
- Yi, B. Q., Rapp, A. D., Yang, P., Baum, B. A., and King, M. D.: A comparison of Aqua MODIS ice and liquid water cloud physical and optical properties between collection 6 and collection 5.1: Cloud radiative effects, J. Geophys. Res.-Atmos., 122, 4550– 4564, https://doi.org/10.1002/2016JD025654, 2017.
- Yurkin, M. A. and Hoekstra, A. G.: The discrete dipole approximation: An overview and recent developments, J. Quant. Spectrosc. Ra., 106, 558–589, https://doi.org/10.1016/j.jqsrt.2007.01.034, 2007.
- Zhang, H., Wang, F., Zhao, S., and Xie, B.: Earth's energy budget, climate feedbacks, and climate sensitivity, Adv. Clim. Chang. Res., 17, 691–698, 2021.
- Zhao, W. J., Peng, Y. R., Wang, B., Yi, B. Q., Lin, Y. L., and Li, J. N.: Comparison of three ice cloud optical schemes in climate simulations with community atmospheric model version 5, Atmos. Res., 204, 37–53, https://doi.org/10.1016/j.atmosres.2018.01.004, 2018.