Comment to editor

We hope that we answered all reviewers comments in a way that no open questions are left. Apart from the changes to the manuscript as result of the reviewers comments, we made some minor additional corrections and editorial improvements. All changes to the text are highlighted in the marked-up manuscript.

In figure 1, we had to correct a typo in the lower right box. *spectra*.. has been changed to *spectra*.

Figure 5 has been replaced by a slightly modified version. In the middle panel we used data that had minor issues resulting in three faulty values indicated by drops of the green line to NaN. These have been corrected and replaced. The description and conclusions related to this plot didn't change at all. The new figure 5 is included in the revised manuscript.

The biggest change made to the revised version, is the new appendices that have been included according to a comment of reviewer 2.

Anonymous reviewer 1

We thank the reviewer very much for her/his very useful comments and suggestions on the manuscript, and thereby the possibility to further improve it. In the following we will address the more detailed minor comments and list the changes we made in the manuscript. The technical corrections will be addressed afterwards. Text that has been revised or that has been added to the manuscript is written in italic letters.

Minor comments

1) I think more information are needed about the limitations of the model. I have noticed for examples the following points. What are the limitations due to the column independent approximation (P5 L12). What is the maximum range for the elevation angle? Would it be possible to quantify strong scattering (P5, L14) and strong precipitation . . . large radar footprint (P6, L26).

To address these points, subsection 2.1 has been revised. It now reads: For the passive part, the one dimensional, polarized, and monochromatic vector RT equation for an azimuthally symmetric scattering media in a plane-parallel atmosphere applying the independent column approximation is solved using the RT4 code of Evans and Stephens (1995). 3D effects can not be modeled but horizontal inhomogeneity can be taken into account by the independent column approximation by realistically describing atmospheric variations along the path (Meunier et al. 2013). The assumption of a plane-parallel geometry is sufficient for most RT problems in the microwave spectral range with the exception of strongly scattering precipitation situations where the radiation does not originate within the instruments field of view (Battaglia and Tanelli, 2011).

The simulation of the passive radiative transfer at high frequencies for very strong scattering might require that the number of angles to describe the scattering matrix has to be increased. This number is fixed to 16 at the moment in PAMTRA, which is sufficient for most of the applications the model has been applied for so far. For future versions we will give the user the opportunity to adapt this variable, as it is already implemented in the

solver backend RT4.

In respect to multiple scattering, we have to stress out again, that PAMTRA is not able to simulate multiple scattering. Whether multiple scattering occurs and whether it needs to be considered for specific situation depends on many different parameters like: considered frequency, beam width, observing geometry, particles, and particle size distribution present, etc.. This is described in more detail by the studies referred to in the manuscript (i.e., Matrosov and Battaglia (2009); Battaglia and Tanelli (2011)).

The radar simulator section has therefore been extended.

Currently, the simulation of multiple-scattering effects is not implemented in PAMTRA. Multiple-scattering generally increases with the amount of scatterers, with larger measurement volume, and with increasing radar frequency (Battaglia et al., 2010). For satellite radars, such as CloudSat, multiple scattering effects have to be accounted for in case of heavy precipitation events (Matrosov and Battaglia, 2009). Due to the smaller measurement volume of common ground-based cloud radars, multiple scattering can be usually be neglected for this application.

2) Sect. 2.1: What are the atmospheric input? (temperature, pressure, humidity, trace gas profiles?). I think that the radiative transfer equation solved by RT4 should be written. It is the core of PAMTRA for passive observations and it will help to better understand the model simplifications.

The module to solve the radiative transfer RT4 requires as input profiles of temperature and gaseous absorption at the specified frequency, and if present, profiles of the scattering properties of the hydrometeors for the same frequency. These are calculated (apart from the temperature) by appropriate methods as described in the manuscript. In addition, the type of the surface reflection and emissivity of the surface is needed by RT4. As minimum atmospheric input, PAMTRA needs profiles of temperature and pressure on a height grid. All other values can be either zero or are automatically set to reasonable default values. For these cases a warning is raised.

In our opinion the inclusion of the RT equation does not help the reader or potential user of the model. Including the equation would result in a lengthy explanation of the single terms and contributions which is beyond the scope of this manuscript. To help the interested reader to gain further understanding into the equations behind, we point to the formulation in a more detailed publication by Evans and Stephens (1993, Eq. 2.22) and in Evans and Stephens (1995, Eq.1). Furthermore, we extended the text by describing in more detailed what the assumptions and simplifications of the model are. The whole subsection 2.1 has been completely reformulated-

To make these points more clear in the manuscript, the subsection on the passive radiative transfer has been adapted in addition to what has been mentioned in the answer to comment 1.

The RT equation is described by the formulation in Eq. 2.22 by Evans and Stephens (1993) or Eq. 1 in Evans and Stephens (1995). It is solved numerically by the doubling and adding method which is formulated and described in detail by several textbooks (i.e., Liou, 2002, p. 290). RT4 requires as input the vertical profiles of temperature and gaseous absorption coefficients and a lower and upper boundary condition. If hydrometeors are present, the profiles of the single scattering properties are required as well. Since a plane-parallel geometry with isotropic thermal emission is considered and all the particles are assumed to be azimuthally random oriented and mirror-symmetric, the radiation fluxes are also isotropic in azimuth. This symmetry in azimuth implies that the third and fourth Stokes components are zero and the RT problem simplifies to the first two components. RT4 does not make use of the Rayleigh-Jeans approximation which relates the Planck function linearly to the brightness temperature is widely used on the microwave regions.

3) Sect. 2.2: The pulse width is not discussed for radar simulations. Is-it a model parameter? I think it will have an effect on the spectral width and on the measurement vertical resolution? Is the latter computed? (I did not see any description of it in the manuscript)

PAMTRA only provides a relatively simple 1D radar simulator, so no beam geometry etc. is considered. Of course, the pulse width affects the vertical resolution. In the model this resolution is defined by the user and his choice of the vertical grid in the atmospheric input. This treatment is reasonable because pulse width and vertical resolution are not strictly tied when pulse compression is used.

In the manuscript we have added at the end of the first paragraph in radar simulator description: The vertical resolution of the simulated full radar Doppler spectrum is determined by the vertical resolution of the input profiles.

Technical corrections:

P7L21: .. dynamical and instrument effects such as attenuation . . . Atmospheric attenuation is not dynamics nor instrument. The sentence should be rephrased.

The sentence has been re-phrased to:

In reality, the idealized $\eta_v(v)$ spectrum is affected by attenuation, kinematic broadening, vertical air motion, and radar noise (Doviak and Zrnic, 1993).

P8, L15: To my knowledge, N_2 does not have resonant lines in the microwave domain but it contributes to the continuum absorption. This contribution is included in a dry continuum term in Liebe together with a contribution from O_2 . This should be corrected.

The reviewer is right. This was wrong in the manuscript. We re-phrased the sentence to state it in a correct manner.

Absorption by atmospheric gases in the microwave range can be separated into contributions by resonant line absorption (i.e., H_2O , O_2 , and O_3) and the water vapor and dry continuum.

P13, Fig.2: What are the spatial coverage and resolution of the maps?

We changed the figures so that meridians and parallels are included. Given that the ECMWF IFS cycle 41r2 has a resolution of 0.1° this shows that the resolution is 6 to 7 km. The spatial coverage is approx. 950 km in North-South direction and 800 and 950 km in Northern and Southern part, respectively.

We added to the manuscript: ...cycle 41r2 with a 0.1° grid (6 to 7 km) ...

P18,L10: correct [3] in up to [3]km

Corrected

P18L22: correct "model" in "Rosenkranz 98 m odel"

Corrected

Fig.6 caption: correct denotes in The white line denote . . .

Corrected

Anonymous reviewer 2

We thank the reviewer very much for her/his detailed thoughts, the very useful comments, and suggestions on the manuscript, and thereby the possibility to further improve it. In the following we will first address all "bigger minor comments" and list the changes we made in the manuscript. The minor review points will be answered afterwards. In general, the manuscript has been revised and thereby strengthened according to the reviewers comments. For the more extensive comments in 1a) we have added an appendix to the manuscript.

Text that has been revised or that has been added to the manuscript is written in italic letters.

Bigger minor comments

- 1) Throughout the examples in section 3, it would be good to have clearer documentation of the atmospheric model (ICON-LEM) and the exact settings of the radiative transfer model (as PAMTRA has a number of options, shown in Table 1).
- a) It would be useful to have a short section to centralise the description of, and give further details on, ICON-LEM. It is important to know the type of microphysics schemes being employed, and which prognostic and active variables are used (e.g. which hydrometeors are represented, and which moments?) Is there any possibility of a mismatch in assumptions (e.g. PSD, shape, fall-speed) between those in PAMTRA and in the model?

Indeed, a description of the Seifert and Beheng microphysical scheme was missing has been added to the manuscript in the subsection of the ground base example (Sec. 3.2) where it used the first time for simulations based on ICON-LEM runs.

The ICON-LEM model used here implements the 2-moments microphysical scheme from Seifert and Beheng (2006). The cloud scheme has six hydrom-

eteor classes (cloud drops, rain, cloud ice, snow, graupel, and hail) which are assumed to be distributed according to a modified gamma function (Petty and Huang, 2011). The model simulates the evolution of two moments of the hydrometeor distributions, namely the mass mixing ratio q and the total number concentration N. Details on the treatment of the ICON microphysical scheme are given in appendix A2.

Matching the hydrometeor assumptions in the atmospheric model and PAM-TRA is of greatest importance for the accurate simulation of remote sensing measurements. PAMTRA has been designed to be highly consistent with the model assumptions, but is still transparent and easy to use. However, as the user might employ different atmospheric model runs with different assumptions in the end the user needs to check that. For two prominent models (ICON-LEM and IFS) this procedure is described in detail. We have included a new appendix section where all these processing steps are described in greater detail for both the classical ICON and the IFS cycle 41r2 model used in the application examples.

Note that in the manuscript the appendix A1 is referenced for the IFS at the end of the 2nd paragraph of subsection 3.1 as

...(Forbes et al., 2011) as prognostic variables. More details on the treatment of the IFS microphysics in PAMTRA are given in appendix A1.

b) The PAMTRA settings used in the examples in section 3 need to be more clearly stated. One option might be to extend Table 1.

The information on settings can be found at different parts of the manuscript, very detailed in the accompanying code for the examples, and the general model documentation. In the subsections of section 2, where we describe the options available in PAMTRA for gaseous absorption, models for refractive indices, or surface treatment, we always give the default option if appropriate. None of these default options has been changed in the examples. To make this a bit more prominent, we used bold letters in table 1 for default options and changed the caption of the table to:

Main characteristics and features of PAMTRA. Default options are written in bold letters.

Settings important for the presented simulations are mentioned in the manuscript where these are described. More technical settings necessary to run the model for the specific example are shown and described in detail in the jupyter notebooks of each example available with the github project of PAMTRA. Therefore, we kindly like to point to these for answering the comment regarding the settings apart from the ones mentioned in the revised version of the manuscript.

- 2) Section 3.1 uses the IFS as input to PAMTRA and compares to AMSU-A and MHS. There are a few issues here:
- a) The inputs to PAMTRA likely only include the four prognostic hydrometeors from the large-scale cloud parametrisation (P12 L17). This is insufficient to replicate observed brightness temperatures. In the all-sky forward modelling of passive microwave data at ECMWF, the convective hydrometeors from the convection scheme are also included (see e.g. Geer and Baordo, 2014, section 2.2). However these fields are not available from the standard archived ECMWF products. If the convective hydrometeors were added, brightness temperature depressions in frontal areas (which often contain embedded convection) would likely be deeper.

The reviewer is absolutely right. The ECMWF IFS data used in the simulations only include the mass mixing ratios of the four hydrometeor categories as prognostic variables. Additional hydrometeor contents from convective rain and snow flux (personal communication with Richard Forbes) are not available. These, most likely, will influence the resulting brightness temperatures. To account for this insufficiency of the simulations to reflect the observations, this is now mentioned in the manuscript ...rain and snow (Forbes et al., 2011) as prognostic variables. Because the convective rain and snow flux profiles are not available in the standard output, we - in contrast to Geer and Baordo (2014) - can not consider their contribution which may modify the results.

b) This text is overly strong: is even stronger than in the observation of MHS for the north-eastern area. With the aforementioned capability of SSRGA to reproduce TB depressions in agreement with observations, this overestimation can be linked to an overestimation of snow water content of ECMWF IFS. The implication from more extensive comparisons in Geer and Baordo (2014) would be that snow water content in the IFS in frontal areas is consistent with observations, at least within the uncertainty on the assumed PSDs and particle shapes in the radiative transfer. The authors seem to be claiming that SSRGA is perfect and the IFS is wrong. It is unlikely that simple, especially given point (a) which, if addressed, would likely make the overestimation of simulated brightness temperature depressions look even worse when using SSRGA.

We agree with the reviewer that our statement was too strong. It is important to note the it is not our intention to provide an in-depth evaluation of the ECMWF IFS model. This is not possible with the data available and the methods we apply in the comparison. Our intention is more to present an application example of PAMTRA and the tools delivered with it. It is up to the user to make use the toolbox in an appropriate way for tasks like atmospheric model evaluation. To reflect this fact, we weakened our points and made them less conclusive, more speculative, and vague.

P14 L8 up to the end of subsection 3.1 now reads: For the simulations shown here based on IFS and PAMTRA using SSRGA for the frozen hydrometeors, the depression is much stronger than for Mie (Fig. 2r) and comparing it to the observation (Fig. 2o), it can be seen that it is even stronger than in the observation of MHS for the north-eastern area, although the contributions to the total precipitating hydrometeor amounts through convection are not included in the simulations. With the aforementioned capability of SSRGA to reproduce T_B depressions in agreement with observations, this overestimation might be either connected to an overestimation of snow water content of ECMWF IFS especially in the middle and upper troposphere or to an overestimation of the scattering by the SSRGA.

Other minor comments

1) The introduction motivates the idea of using remote sensing measurements for improving the atmospheric models (e.g. P2 L29). However (P2 L30) when describing the importance of these measurements in data assimilation and NWP, it would be possible to infer that model validation was still their main purpose. It would be worth making it more explicit that the main aim of using these

observations in NWP is to infer initial conditions for weather forecasts (the aim to improve models is not yet so well developed in NWP.)

To our knowledge there are quite some NWP evaluation studies that use are RT models. Therefore we think this is important to include. To make it more clear and to distinguish between these two application examples, we included "or".

On the other hand, the remote sensing measurements shall be used for improving the atmospheric models, or, most directly, measurements are used in data assimilation together with fast RT operators to infer the initial conditions for NWP models.

2) P3 L17 suggests that hydrometeor single scattering properties for fast R/T models are derived from line-by-line models, which is not correct. It would be best just to remove the mention of single scattering properties here.

This refers to something that has been addressed by the answer to comment 2 of reviewer 1.

3) P3 L21 I suggest to delete principally as it is not clear what this means in the context.

We deleted "principally".

4) P5 L12 → P6 L2 gives a discussion on horizontal homogeneity, suggesting it is not important in microwave radiative transfer. This is not correct, because the beamfilling effect (due to the nonlinear dependence of backscatter or brightness temperature on water content) means there is ambiguity between water mass and water inhomogeneity at scales below the model grid or sensor field of view. The importance of horizontal inhomogeneity in forward modelling for NWP is illustrated by, among others, Geer et al. (2009) and references therein. However, its easy to deal with horizontal inhomogeneity by using the independent column approximation. Presumably what the authors really mean is that full 3D radiative transfer with horizontal inhomogeneity is unnecessary.

We rephrased this subsection completely. The part where the statement regarding independent column approximation is included new reads the following:

For the passive part, the one dimensional, polarized, and monochromatic vector RT equation for an azimuthally symmetric scattering media in a plane-parallel atmosphere applying the independent column approximation is solved using the RT4 code of Evans and Stephens (1995). 3D effects can not be modeled but horizontal inhomogeneity can be taken into account by the independent column approximation by realistically describing atmospheric variations along the path (Meunier et al. 2013). The assumption of a plane-parallel geometry is sufficient for most RT problems in the microwave spectral range with the exception of strongly scattering precipitation situations where the radiation does not originate within the instruments field of view (Battaglia and Tanelli, 2011).

5) P6 L5 The description of the doubling-adding method in this paragraph is not particularly helpful, and it closely follows the description in Evans and Stephens (1995) which itself doesnt much help summarise the method or ideas like the interaction principle or initialisation. There might well be a textbook that can help the authors formulate a clearer and simpler description of the technique - is it covered in Petty (2006) or Thomas and Stamnes (2002), for example?

We agree with the reviewers comment, that the description of the doubling and adding method as it is done here is not of much help. Therefore, we decided to reformulate the whole subsection 2.1 to follow as well a suggestion of another reviewer. For the doubling an adding we now point to textbooks that describe the method and do not give anymore a very rough and non-understandable description. Adding a complete description of the doubling and adding here is beyond the scope of the manuscript.

6) P6 L25 The word However suggests a dependence between the first part of the paragraph (on the dielectric factor) and the second part (on multiple scattering). In practice these are two completely separate issues, Maybe the second part of the paragraph would be better introduced with Another issue rather than However?

The reviewer is right, this is not very well written. We changed P6 L25 and the following till the end of the paragraph. It now reads:

Currently, the simulation of multiple-scattering effects are not implemented in PAMTRA. Multiple-scattering generally increases with the intensity of precipitation, with larger measurement volume, and with increasing radar frequency (Battaglia et al., 2010). For satellite radars, such as CloudSat, multiple scattering effects have to be accounted for in case of heavy precipitation events (Matrosov and Battaglia, 2009). Due to the smaller measurement volume of common ground-based cloud radars, multiple scattering can be usually be neglected.

7) P6 L32 the minimal sensitivity - this is unclear and would still be unclear if what the authors mean is the minimum sensitivity. Is it rather the radar noise that is being referred to?

After re-reading that part we agree with the reviewer that is actually the noise that is important here. We have rephrased the sentence specifying that what affects the radar measurements is

...the intensity and variance of the spectral noise.

8) P7 L16 $v_{\rm nyq}$ - is it worth explaining why this parameter is called nyq or giving it a simpler notation? (since Nyquist is not mentioned in the text here)

We changed the beginning of the sentence mentioning the Nyquist frequency to

The maximum/minimum Doppler velocity of a real radar Doppler spectra is determined by the Nyquist velocity

9) Section 2.4 describes the Stokes reflection matrix but is insufficiently clear on how this is being set up, particularly for the components that describe non-specular reflection. For example TELSEM, TESSEM and FASTEM are all emissivity schemes that assume specular and non-polarised reflection at the surface, and provide a simple emissivity to describe this. Yet the text implies they provide a full reflection matrix. There is also an ambigu-

ity as to whether TESSEM is providing just the roughness and foam coverage corrections, or the entire emissivity and reflectivity calculation (P9 L8). A much clearer description is needed here, given that determining the full Stokes reflection matrix (including polarisation changes and non-specular reflections) is not at all straightforward.

TESSEM2 provides the emissivity in the frequency range of 10 to 700 GHz for horizontal and vertical polarization and arbitrary angle from nadir. The parameter range of TESSEM2 is valid for is 10 m windspeed between 0 and 25 m/s, sea surface temperature of 27 to 310K, and salinity between 0 and 40 /. As stated by the reviewer, P9 L8 is a misleading and gives the impression, the TESSEM² only provides a correction. Therefore, we re-phrased the corresponding part, which should as well address the first comment related to emissivity.

...have to be applied. PAMTRA utilizes the Tool to Estimate Sea-Surface Emissivity from Microwaves to sub-Millimeter waves (TESSEM²; Prigent et al., 2017) for the calculation of polarized and foam and roughness corrected emissivities. TESSEM² is based on the community model FAST microwave Emissivity Model (FASTEM; Liu et al., 2011) and is designed for frequencies up to 700 GHz. The resulting emissivities and thereby reflectivities, are used to calculate the elements R_{ij} with i, j = 1, 2 of the 4-by-4 reflection and radiance matrix needed for solving the radiative transfer by RT4. All other values are set to 0.

10) P9 L23 particle maximum extend should have a clearer definition (and extent, not extend, is probably intended)

We have specified that the maximum extent is the particle 3D maximum dimension in P9 L23.

11) P10 L7 consider defining M_k with an equation so that its easier to understand why $q=aM_b$.

We have included the definition of a distribution moment as suggested.

12) P11 L9-10 suggests that the reason to choose Mie or T-Matrix is simply speed - surely its whether you have a sphere or a spheroid?

We agree with the reviewer and changed these lines to make it more specific and clear.

Dependent on the requirements on computational speed and particle properties, the scattering properties can be calculated using Mie (Mie, 1908) or T-matrix theory (Mishchenko and Travis, 1994); for the latter, also the orientation and aspect ratio of the particles have to be defined.

13) P14 L2 - the surface is very often visible in satellite 157 GHz observations, outside the humid conditions of the tropics, so errors in the surface representation could very well be suspected here. If the authors want to claim that the surface influence can be neglected this would need to be backed up by a map of the surface-to-space transmittances at 157 GHz for this case study.

The reviewer is right. The argumentation for the influence of the surface signal at 157 GHz is not very good. However, the signal in the microwave region from ocean surfaces is quite well understood. In addition, the IFS SST should not differ that much from observations. Therefore, differences between model and observation that can be attributed to the surface should be rather small.

P13 L33 now reads: Since the ocean surface signal in the microwave region can be model quite well by $TESSEM^2$ and the sea surface temperature in the model and reality should not differ to much, the differences in the T_B can be most likely attributed to an underestimation of the liquid water contents (Fig. 2s) or to the water vapor field (not shown). At the higher frequency of 157 GHz (Fig. 2l,o), the scattering at larger frozen hydrometeors ...

14) P15 L32-33 - the Arctic is very far from a measurement void, since polar orbiting operational meteorological satellites cover it with very high temporal frequency. The authors should be more specific on this point.

We changed the text to:

Airborne campaigns can provide unique information in this area where ground based observations are made at very few stations and where polar orbiting satellites have rather coarse spatial resolution.

15) P20 L26 Using adiabatic to describe the droplet size variation with height is loose terminology and should be improved - adiabatic of course refers to thermodynamic processes, and the radius of water droplets wont change much under a true adiabatic assumption (water being incompressible).

We removed the "near-adiabatic".

16) P20 L32 and surrounding discussion is initially confusing. It could be more clearly stated in the text that the Doppler spectrum is only simulated, not observed. The suggestion that the larger droplets (secondary peaks in the Doppler spectrum) are invisible in the in-situ measurements is confusing as they must be present in the data, just not visible on the colour scale chosen for this plot, or possibly hidden under the white line.

We changed that part and hope it is more clear now. It reads:

However, the radar Doppler spectra reveal Doppler velocities up to 1m s1 in certain heights sometimes showing a clear bi-modality of the spectra, e.g. 800 and 1000m. These can clearly be attributed to the high impact of low-concentration drizzle droplets on radar observations. These drizzle droplets are not visible in Fig. 6a despite the logarithmic color scale as they are rare. The small sampling volume of the optical probes used during ACME-V leads to poor statistics for drizzle drops which can explain the presence of inhomogeneities in the spectra forward modeled with PAMTRA (Fig. 6.b).

Typos

P4 L14 plan parallel \rightarrow plane parallel Changed

P18 L11 and L22 extend \rightarrow extent Changed

P18 L24 resulting → resulting simulated Changed (but it is P20 L24)

PAMTRA 1.0: A Passive and Active Microwave radiative TRAnsfer tool for simulating radiometer and radar measurements of the cloudy atmosphere

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Abstract. Forward models are a key tool to generate synthetic observations given the knowledge of the atmospheric state. In this way, they are an integral part of inversion algorithms that aim to retrieve geophysical variables from observations or in data assimilation. Their application for the exploitation of the full information content of remote sensing observations becomes increasingly important when these are used to evaluate the performance of cloud resolving models (CRMs). Herein, CRMs CRM profiles or fields provide the input to the forward model whose simulation results are subsequently compared to the observations. This paper introduces the freely available comprehensive microwave forward model PAMTRA (Passive and Active Microwave TRAnsfer), demonstrates its capabilities to simulate passive and active measurements across the microwave spectral region for up- and downward looking geometries, and illustrates how the forward simulations can be used to evaluate CRMs and to interpret measurements to improve our understanding of cloud processes.

PAMTRA is unique as it treats passive and active radiative transfer (RT) in a consistent way with the passive forward model providing up- and down-welling polarized brightness temperatures and radiances for arbitrary observation angles. The active part is capable of simulating the full radar Doppler spectrum and its moments. PAMTRA is designed to be flexible with respect to instrument specifications, interfaces to many different formats of in- and output type, especially CRMs, spanning the range from bin-resolved microphysical output to one- and two-moment schemes, and to in situ measured hydrometeor properties. A specific highlight is the incorporation of the self-similar Rayleigh–Gans Approximation (SSRGA) both for active and passive applications, which becomes especially important for the investigation of frozen hydrometeors.

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1 Introduction

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The use of passive and active microwave sensors in atmospheric research has experienced rapid growth in the last decades due to their unique ability to provide information on clouds and precipitation as well as for thermodynamic profiling even under cloudy conditions. Passive and active microwave sensors are highly complementary and are therefore often collocated on space-, airborne, or ground-based observing platforms. The strength of this combination is based on the ability of the radar to provide very detailed information about the vertical structure of hydrometeors, in-cloud dynamics (Borque et al., 2016), as well as microphysical processes (Kalesse et al., 2016), while the passive sensors add information on thermodynamic profiles, constrain column integrated hydrometeor quantities, and provide wide swath information from satellitesatellites. Prominent examples for combined satellite sensors are the Tropical Rainfall Measurement Mission (TRMM; Kummerow et al., 1998), the Global Precipitation Mission (GPM; Hou et al., 2014), and the Afternoon-Train (A-Train; L'Ecuyer and Jiang, 2011). Passive and active microwave instruments are also commonly combined in airborne observatories such as the High Altitude and LOng range research aircraft (HALO) Microwave Package (HAMP; Mech et al., 2014) or the remote sensing package of the Wyoming King Air (Wang et al., 2012). From the ground, the standard configuration needed to determine detailed cloud vertical profile information (e.g., -the CloudNet algorithm (Illingworth et al., 2015)) at several ground-based super-sites (Mather and Voyles, 2013; Löhnert et al., 2015) includes a cloud radar, a microwave radiometer, and a ceilometer.

To fully exploit remote sensing measurements, radiative transfer (RT) models are needed which convert an atmospheric state into a synthetic measurement. They are key tools for the design of new sensors, for the development of retrieval algorithms and the improvement of atmospheric models both for numerical weather prediction (NWP) and climate applications. Herein, a particular challenge is the realistic description of hydrometeors —and their particle size distributions (PSD) as well as their respective single scattering properties (Petty, 2001), which are required when solving the RT equation. Specifically, an accurate but also computationally efficient description of the scattering properties of ice and snow particles for global applications is needed (Geer and Baordo, 2014). Closure studies in which detailed in –situ measurements of hydrometeor properties and corresponding remote sensing measurements are connected with the help of an RT model can improve our knowledge of these interaction processes (e.g., Tridon et al., 2019)(e.g., Tridon et al., 2019).

The connection between atmospheric models and RT is twofold. On the one hand, realistic representations of the atmospheric state with an emphasis on hydrometeors are needed as input for the RT together with instrument models to yield synthetic measurements (e.g., sampling geometry, noise characteristics). Thus, cloud resolving models (CRMs) are frequently used as RT input in design studies and retrieval development (Chaboureau et al., 2008; Matsui et al., 2013). On the other hand, the remote sensing measurements shall be used for improving to improve the atmospheric models. Most, or, most directly, measurements are used in data assimilation together with fast RT operators to infer the initial conditions for NWP models. This is especially demanding under cloudy conditions but of growing importance for NWP (Geer et al., 2017). To improve the representation of clouds and precipitation in atmospheric models in general, microphysical schemes are under development with a tendency towards increasing the number of hydrometeor categories and PSD moments incorporated. Microwave measurements are well suited for evaluating the performance of these schemes but a thorough matching of the predicted hydrometeor proper-

ties and the assumptions in the RT needs to be realized (e.g., Han et al., 2013; Matsui et al., 2013; Schemann and Ebell, 2020) (e.g. Han et al., 2013; Matsui et al., 2013; Schemann and Ebell, 2020).

In summary, various applications require RT models to be flexible in respect to the adaptation of the given hydrometeor information and the different instrument specifics. Here we present the Passive and Active Microwave radiative TRAnsfer operator (PAMTRA) which has been developed exactly for this purpose. Along with the increasing use of microwave remote sensing several RT models have already been developed in the past. In the following, we give examples of important RT models to provide context to our motivation to develop a new RT framework.

The Radiative Transfer for the TIROS Operational Vertical Sounder (TOVS) (RTTOV; Saunders et al., 1999, 2018) has been developed for the specific application of NWP data assimilation to respond to the requirement of high computational performance. For this purpose, RTTOV employs parameterizations tailored to specific microwave satellite radiometers. It provides the tangent linear, adjoint and Jacobian matrix to enable all-sky data assimilation. Recently, RTTOV-gb has been released which also allows to simulate simulating ground-based sensors (De Angelis et al., 2016). Similar to RTTOV the Community Radiative Transfer Model (CRTM; Ding et al., 2011) has been developed to efficiently simulate specific sensors (e.g., satellite instrumentation).

For developing the parameterizations for the fast RT models, i.e., determining sensor specific coefficients, reference RT simulations with line-by-line models are needed. These are typically one-dimensional models that assume a plane-parallel atmosphereand need information on gaseous absorption and hydrometeor single scattering properties for each vertical layer. For example, AMSUTRAN (Turner et al., 2019) calculates profiles of layer-to-space transmittances as the basis for the training of RTTOV. It includes absorption routines based on the Millimeter-wave Propagation Model (MPM; Liebe et al., 1991, 1993) with subsequent spectroscopic modifications.

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The RT can be principally solved if gas absorption and single scattering properties for hydrometeors are specified. Especially, a realistic representation of single-scattering properties of frozen particles is still a challenge for any RT. The number of databases including scattering properties of various habits, densities, orientations, and composition is rapidly increasing during recent years (Kneifel et al., 2018). However, many RT models are still using spheroidal approximations due to their low computational costs and flexibility to account for particle properties such as mass—size relation. New approximations, which take the fractal properties of aggregates better into account became recently available (Hogan and Westbrook, 2014; Hogan et al., 2017).

Two widely used codes for polarized microwave radiation are the RT3 and RT4 models provided by Evans and Stephens (1991); ?, 2010) Evans and Stephens (1991, 1995, 2010). RT3 solves RT for atmospheres with randomly oriented particles; RT4 is an extension of RT3 and also accepts azimuthal symmetric oriented particles. Several RT models implemented these codes as the RT solving algorithm (Deiveegan et al., 2008; Buehler et al., 2018) with different options for information on gaseous absorption, or single scattering properties. RT4 is also used for the passive component of the RT framework presented in this manuscript.

For active microwave sensors, QuickBeam (Haynes et al., 2007) is able to simulate radar reflectivity profiles for bottom-up and top-down perspective and is part of the CFMIP (Cloud Feedback Model Intercomparison Project) Observation Simulator (COSP; Bodas-Salcedo et al., 2011). Higher Doppler spectral moments (Kollias et al., 2007) and radar polarimetry which are

often provided by ground-based Doppler cloud radars can be calculated with the Cloud Resolving Model Radar Simulator (CR-SIM; Oue et al., 2019). The Polarimetric radar simulator (POLARRIS) recently presented by (Matsui et al., 2019) is a forward and inverse model for polarimetric radar observables.

Only very few RT models provide simultaneous passive/active simulations. Two examples are the Passive and Active Microwave-Vector Radiative Transfer (PAM-VRT; Yang and Min, 2015) and the Atmospheric Radiative Transfer Simulator (ARTS; Eriksson et al., 2011; Buehler et al., 2018). Both RT are suited to simulate spaceborne and ground-based sensors including more complex (non-spheroidal) single scattering databases for frozen particles. However, both RT models they do not provide simulations of the full Doppler spectrum.

Though several studies in the past have already used parts of the Passive and Active Microwave TRAnsfer (PAMTRA) tool, e.g., Acquistapace et al. (2017) for optimized drizzle detection, Cadeddu et al. (2020) for ground-based radiometer retrieval in raining conditions, Maahn and Löhnert (2017) for simulations of in situ aircraft measurements, Heinze et al. (2017) and Schemann and Ebell (2020) for CRM evaluation, it has now been converted into a versatile, freely available tool. Herein the main motivation was the need to have a RT tool which can simulate microwave as well as Doppler radars for ground-based, airborne, or spaceborne platforms using state-of-the-art scattering models. PAMTRA provides passive and active RT simulations in a consistent way for a plan-parallel plane-parallel, one-dimensional, horizontally homogeneous atmosphere with hydrometeors for up- and down-welling microwave radiances. A particular focus in the design of PAMTRA was put on providing maximum flexibility to various model output outputs (one-moment, two-moment, or full-bin schemes) or in situ measured hydrometeor properties. It also was We also intended to allow the user to select a number of scattering and absorption models for maximum flexibility in the assumptions made in the microphysical parameterizations.

This paper provides a description of the first comprehensive PAMTRA version 1.0 and advocates its use with a range of examples demonstrating its value in investigating cloud and precipitation processes. Section 2 gives an overview of the general architecture of PAMTRA including the description of the passive and active RT. This general part of the RT is followed by descriptions of how atmospheric properties such as gas absorption, particle size distribution, scattering and absorption of hydrometeors, as well as boundary conditions are treated in PAMTRA. It also provides an overview of the wide range of selectable user options, e.g. —scattering/absorption models, databases. Application examples (Sect. 3) include ground-based, airborne, and satellite perspectives for passive and active microwave sensors. In Sect. 4 a summary as well as future perspectives are given.

2 Model framework

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PAMTRA is a FORTRAN/Python model framework for the simulation of passive and active RT (including radar Doppler spectra) in a plane-parallel, one-dimensional, and horizontally homogeneous atmosphere for the microwave frequency range. Figure 1 shows a flow diagram of the various steps performed in the FORTRAN core of the present model setup. For the simulation, the model needs various inputs (shown in reddish colors) that describe the atmospheric state, the assumption on absorption, scattering, and surface emissivity, and instrument specifications. Depending on this input, the interaction parameters

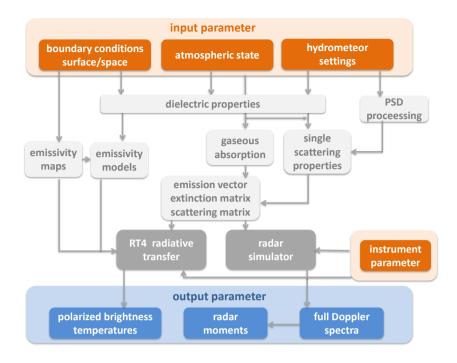


Figure 1. Flowchart of the various steps performed during a PAMTRA simulation. Orange areas describe input parameters given by the user with the Python interface or text files. Gray boxes are the FORTRAN model parts where the various interaction parameters are generated and the radiative transfer or the radar simulator get processed. Blue boxes describe the model output.

within various modules (white boxes) are generated. These parameters serve as input for the solving routines for the passive and active part (shown in gray). The simulations produce polarized radiances or brightness temperatures (T_B) for the passive part and radar polarimetric Doppler spectra (and derived moments such as reflectivity, mean Doppler velocity, skewness, and kurtosis, as well as left and right slopes) for the active part. The simulation is performed at any observation geometry (zenith/nadir looking or slanted). Table 1 summarizes the main features of PAMTRA.

pyPAMTRA adds a Python framework around the FORTRAN core which allows to eall calling PAMTRA directly from Python without using the FORTRAN I/O routines. Consequently, pyPAMTRA is a more user-friendly way to access the PAMTRA model and is the common way to use the model framework. It includes a collection of supporting routines, e.g., for importing model data or producing file or graphical output of the simulation results. With pyPAMTRA, parallel execution of PAMTRA on multi-core processor machines and clusters is possible. Furthermore, by using Python for I/O and flow control, it is easier to interface PAMTRA to instrument instruments or atmospheric models, as well as post-processing routines.

2.1 Microwave radiometer simulator

For the passive part, the one dimensional, polarized, and monochromatic vector RT equation for an azimuthally symmetric scattering media in a plane-parallel atmosphere applying the independent column approximation and plane-parallel atmosphere

is solved using the RT4 code of Evans and Stephens (1995). 3D effects cannot be modeled, but horizontal inhomogeneity can be taken into account by the independent column approximation by realistically describing atmospheric variations along the path (Meunier et al., 2013). The assumption of a plane-parallel geometry and horizontal homogeneity is sufficient for most RT problems in the microwave spectral range with the exception of strongly scattering precipitation schemes situations where the radiation does not originate within the instruments' field of view (Battaglia and Tanelli, 2011). The RT equation is described by the formulation in Eq. 2.22 by Evans and Stephens (1993) or Eq. 1 in Evans and Stephens (1995). It is solved numerically by the doubling and adding method (Liou, 2002, p. 290), which follows the interaction principle. It relates the interaction of radiation with a medium by relating the radiation emerging from an atmospheric layer to the radiation incident upon and generated within this layer. Thereby, homogeneous atmospheric layers are subdivided into thinner sub-layers such that finite differences can be applied as a good approximation. The transformation from finite differences to the interaction principle is called initialization. By this initialization, the scattering matrices are related to reflection and transmission matrices. The integration over these thin sub-layers is performed by the doubling algorithm. Afterwards, for each output level, the adding algorithm is applied, i.e., transmission, reflection, and emissions of the layers above and below the output layer are added which is formulated and described in detail by several textbooks (i.e. Liou, 2002, p. 290). RT4 requires as input the vertical profiles of temperature and gaseous absorption coefficients and a lower and upper boundary condition. If hydrometeors are present, the profiles of the single scattering properties are required as well. Since a plane-parallel geometry with isotropic thermal emission is considered and all the particles are assumed to be azimuthally randomly oriented and mirror-symmetric, the radiation fluxes are also isotropic in azimuth. This symmetry in azimuth implies that the third and fourth Stokes components are zero and the RT problem simplifies to the first two components. RT4 does not make use of the Rayleigh-Jeans approximation which relates the Planck function linearly to the brightness temperature and is widely used in the microwave regions.

As mentioned above, the doubling and adding is are done in PAMTRA by RT4. In comparison to the formerly introduced RT3 (Evans and Stephens, 1991), RT4 enables the user to perform polarized RT calculations for particles that are non-spherical and oriented particlesholding a preferential horizontal alignment. Since one of the major goals in developing PAMTRA was to handle hydrometeor interactions as flexible as possible, the possibility to simulate the RT for oriented particles with any shape is mandatory. RT4 calculates polarized T_B (vertical and horizontal) for each discrete quadrature angle and frequency and upand downward looking geometries at any height within the atmosphere.

2.2 Radar simulator

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The PAMTRA radar simulator estimates the full radar Doppler spectrum based on the single scattering properties of each hydrometeor species (see Sect. 2.5.2); it is mainly based on the concepts developed by Oue et al. (2019). First, the back-scattering cross section $\sigma_B(D)$ in m^2 of the individual hydrometeor particles with maximum dimension D is converted to the volumetric back-scattering $\eta_D(D)$ in the unit of spectral radar reflectivity $m^6 m^{-3} m^{-1}$

$$\eta_D(D) = 10^{18} \sigma_{\rm B}(D) \, n(D) \frac{\lambda^4}{\pi^5 |K_w|^2},\tag{1}$$

where λ is the wavelength in m, n(D) is the normalized PSD in m⁻⁴, and $|K_w^2|$ is the dielectric factor of water related to the refractive index. It is a common convention to use the value for liquid water at centimeter wavelengths ($|K_w^2| = 0.93$, Ulaby et al., 1981) regardless of whether ice or liquid clouds are observed. Nevertheless, as $|K_w^2|$ also depends on frequency, it is possible to change it for optimal adaptation to a specific problem. However, multiple scattering can affect radar measurements in cases of strong precipitation, short wavelength and large radar footprint (Battaglia et al., 2010). These effects can be particularly relevant for satellite radar observations of strong precipitation at W-band such as those of CloudSat (Matrosov and Battaglia, 2009), but can Currently, the simulation of multiple-scattering effects is not implemented in PAMTRA. Multiple-scattering generally increases with the amount of scatterers, with larger measurement volume, and with increasing radar frequency (Battaglia et al., 2010). For satellite radars, such as CloudSat, multiple scattering effects have to be accounted for in case of heavy precipitation events (Matrosov and Battaglia, 2009). Due to the smaller measurement volume of common ground-based cloud radars, multiple scattering can usually be neglected for most other cloud radar applications is not considered by PAMTRAthis applications.

The radar reflectivity factor Z_e can be simply simply be obtained from Eq. 1 by integrating over the PSD. Unlike other radar simulators (e.g., Haynes et al., 2007; Buehler et al., 2005) (e.g. Haynes et al., 2007; Buehler et al., 2005), PAMTRA provides the option to simulate the full radar Doppler spectrum which is necessary to derive higher radar moments (mean Doppler velocity MDV, skewness, kurtosis). Deriving the radar moments from the simulated spectra also allows to account for instrument-specific characteristics such as the minimal sensitivity intensity and variance of the spectral noise. Those instrument characteristics can have an impact on the derived moments and hence are important to take into account when aiming to simulate observations of a specific radar system (see also examples in Sect. 3). The radar Doppler spectra simulator included in PAMTRA is partly based on the concept of Kollias et al. (2011, 2014), and the basic working principle is explained in the following.

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As the measured Doppler spectrum is a function of fall velocity rather than particle size, the size descriptor of the spectral radar reflectivity η_D is changed from D to fall velocity v with

$$\eta_v(v) = \eta_D(D) \frac{\partial D}{\partial v}.$$
 (2)

where, after the transformation by the differential $\frac{\partial D}{\partial v}$ (measured in s), the spectral radar reflectivity $\eta_v(v)$ assumes the measuring units of $\mathrm{mm^6m^{-3}sm^{-1}}$. The user can choose between various hydro-dynamical models to estimate hydrometeor terminal velocity v and $\frac{\partial D}{\partial v}$ based on their physical properties and environmental conditions such as air density, temperature, and pressure. For liquid drops, PAMTRA uses the relation provided by Khvorostyanov and Curry (2002) as a default. For ice and snow particles, the modified relation by Heymsfield and Westbrook (2010) is recommended because it is in better agreement with recent experiments using ice analogues (Westbrook and Sephton, 2017). For PAMTRA, v is defined such that positive values refer to particles moving towards the radar.

The velocity resolution of $\eta_v(v)$ is related to the bin spacing of the particle size distribution. However, real radar Doppler spectra have boundaries of The maximum/minimum Doppler velocity $v_{nyq} = PRF * \lambda/4$ of a real radar Doppler spectra is determined by the Nyquist velocity $v_{nyq} = PRF * \lambda/4$, which depends on the pulse repetition frequency PRF and the radar

wavelength λ used. The velocity resolution is determined by the number of Fast Fourier Transform (FFT) points $n_{\rm fft}$ used to derive the radar Doppler spectrum. These parameters are adjustable in the radar Doppler spectra simulator. $\eta_v(v)$ of all hydrometeors is then linearly interpolated onto the spectral resolution of the simulated radar. Furthermore, if the fall velocity exceeds $v_{\rm nvq}$, the simulator adds velocity folding effects (aliasing) to the spectrum.

In reality, the idealized $\eta_v(v)$ spectrum is affected by dynamical and instrument effects such as attenuation, kinematic broadening, vertical air motion, and radar noise (Doviak and Zrnic, 1993). In PAMTRA, the attenuated $\eta_v(v)$ is obtained by subtracting the cumulative path integrated attenuation, which is estimated from the extinction of gases and hydrometeors depending on measurement geometry (ground-based, airborne, or space-based). Kinematic broadening is assumed to have a Gaussian distribution that is convoluted with $\eta_v(v)$ to simulate the broadening of the Doppler spectrum due to air motions (Gossard and Strauch, 1989) as described in detail in Maahn et al. (2015). In addition to kinematic broadening, also a constant vertical air motion $V_{\rm air}$ can be added that shifts the Doppler peak in the velocity spectrum without broadening the peak. To account for the radar receiver noise, the radar receiver noise power N_P (in units of mm⁶m⁻³ in accordance with Z_e) is added to the spectrum $\eta_v(v)$. To account for the loss of radar sensitivity due to range, N_P is scaled with range squared. Because the noise is assumed to be white, random perturbations are added to every bin i of the spectrum in order to account for random noise effects following (Zrnić, 1975). To make the simulations by the radar simulator reproducible, the random seed used to obtain r(i) can be defined in PAMTRA. Finally, the spectrum is successively averaged $n_{\rm ave}$ times to account for smoothing and turbulence broadening.

Once the simulation of the non-idealized radar Doppler spectrum is complete, the corresponding moments are estimated similar to a real radar data processing scheme (e.g., Maahn and Kollias, 2012)(e.g. Maahn and Kollias, 2012): first, the noise is removed from the spectrum, and second the moments of the hydrometeor peak are determined. In case of several hydrometeor peaks in the same spectrum that are fully separated by the noise floor (multi-modal spectra, e.g., Williams et al., 2018) (multi-modal spectra, e.g. Williams et al., 2018), PAMTRA can estimate the moments of individual peaks independently ordered by maximum spectral reflectivity. The main difference to a real radar data processing scheme is that the noise N_i is known already in advance. Therefore, the user can choose between using the known N_i or applying the method by Hildebrand et al. (2002) for estimating N_i . Based on the noise-corrected radar Doppler spectrum $\eta_v(i)'$, the moments (Z_e , MDV, Doppler spectrum width σ , skewness γ , and kurtosis κ) and slopes (left, right) of the radar Doppler spectrum are estimated as discussed in (Maahn and Löhnert, 2017). The higher moments and the slopes depend on the instrument noise, therefore it is crucial to configure PAMTRA in accordance with the radar specifications. All radar moments and the Doppler spectrum are available non-polarized (NN), but if required also for HV (horizontal receive, vertical transmit), VH, VV, and HH polarization. This allows estimation of differential reflectivity ZDR = Z_e^{HH}/Z_e^{VV} and linear depolarization ratio LDR = Z_e^{HV}/Z_e^{HH} , among others.

2.3 Gaseous absorption

Absorption by atmospheric gases in the microwave range can be separated into contributions by resonant line absorption (i.e. $\frac{1}{2}$ H₂O, O₂, $\frac{1}{2}$ and O₃) and the water vapor and dry continuum. PAMTRA implements various models to calculate the absorption

coefficients of atmospheric gases. The model by Rosenkranz (2015) including modifications of the water vapor continuum absorption (Turner et al., 2009) and the line width modification of the 22.235 GHz H₂O line (Liljegren et al., 2005) is selected as default. Alternatively, the Millimeter-wave Propagation Model (MPM93) developed by Liebe et al. (1993) can be used to simulate the absorption by the gaseous atmosphere. The clear interface structure of PAMTRA gives the possibility to easily include future improvements in gas absorption models, i.e., developments of models in the sub-millimeter wavelength range (Mattioli et al., 2019), as well as the implementation of absorption catalogs (Feist, 2004). This provides the possibility to also account for trace gases, which show abundant but weak absorption features in the microwave frequency range above 200 GHz.

2.4 Boundary conditions

The atmosphere is bounded at its upper end by the free space. The radiation emitted by this upper boundary can be described by the cosmic background with its mean radiative temperature of 2.73 K (Fixsen, 2009). The lower boundary of the atmosphere interacting with radiation is the Earth's surface. Thereby, the amount of radiation emitted in each upward direction is defined by the surface temperature and its type, which is determined by setting the emissivity and the model of scattering or reflection. This is not only important for up-welling geometries but also for down-welling in case of a strongly scattering atmosphere (Kneifel et al., 2010). In PAMTRA, scattering or reflection properties of the surface are estimated assuming either a specular, Lambertian, or Fresnel reflection types (Mätzler, 2006, p. 225). Reflection on natural surfaces can be described by Fresnel equations. For idealized simulations, the emissivity can be fixed. Over land surfaces, PAMTRA makes use of the Tool to Estimate Land Surface Emissivity from Microwave to sub-Millimeter waves (TELSEM²; Wang et al., 2017; Aires et al., 2011) which provides emissivities based on geographic location and time information as angular and frequency dependent monthly mean values based on satellite observations.

The reflection of flat ocean surfaces can also be calculated with the Fresnel reflection formulae. The intensity of the reflection is strongly polarization and angle dependent and characterized by the dielectric properties of the ocean surface as a function of the sea surface temperature and salinity. With the Fresnel reflection formulae, the reflection coefficients and the Stokes reflection matrix can be calculated, as well as the angle and polarization dependent emissivity. Since the reflection and emissivity calculated with the Fresnel relations are valid for calm surfaces and deviate significantly for high wind speeds, corrections for wind speed and therefore sea surface roughness and foam coverage have to be applied. PAMTRA utilizes the Tool to Estimate Sea-Surface Emissivity from Microwaves to sub-Millimeter waves (TESSEM²; Prigent et al., 2017). It is based on the community model FAST microwave Emissivity Model (FASTEM; Liu et al., 2011) and is designed for frequencies up to 700 GHz.

2.5 Hydrometeor description

PAMTRA has been designed to be flexible considering the treatment of hydrometeors enabling the use of a wide variety of input data. Hydrometeor classes can be defined in a flexible way that allows to exactly match the properties of particles measured, e.g., by in situ microphysical probes or to be consistent with assumptions on PSD, density, shape, etc., made in CRMs. In addition to the assumptions on hydrometeors, the calculation of their interaction parameters (mainly absorption, scattering,

and back-scattering) and the integration over the PSD of the specific hydrometeor class is a central part of the RT framework (Johnson et al., 2012).

PAMTRA can handle a flexible number of hydrometeor classes. As an example, for the simulations based on the output of a CRM (see example Sect. 3.2), which provides hydrometeor content and total number concentration for cloud liquid, cloud ice, graupel, snow, rain, and hail the number of hydrometeor classes would be six. For each hydrometeor class, the user can specify their microphysical and scattering properties.

The single particle properties are defined in PAMTRA with respect to the particle maximum extend 3D maximum dimension (D) and particle, which is uniquely defined also for complex shapes such as snowflakes. Microphysical properties, such as PSD, mass—size and velocity—size relation can be easily defined by the user with the help of built-in functions. The user can select either size-resolved distributions of particles directly or the functional form of the PSD.

5 2.5.1 Particle size distribution

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Most atmospheric models assume a moment-based <u>bulk</u> microphysical scheme for the treatment of cloud processes. In these schemes, the PSD for each hydrometeor category is assumed to follow a predefined functional form and one or multiple moments of the PSD are simulated as prognostic variables. Using PAMTRA, it is straightforward to ingest the moments of the hydrometeor distributions and reconstruct the full PSD from them.

The PSD forms which are built-in in PAMTRA include the mono-disperse, the inverse exponential, the modified gamma, and log-normal distributions. Some variations of these four main distributions have been implemented to facilitate the interface with some specific weather models. As an example, the formulation used by the two-category ice scheme in the COSMO model (see Doms et al., 2005, p. 69) assumes mono-disperse distribution where the number concentration is dependent on the ambient temperature. Also the relations reported by Field et al. (2005) and Ryan (2002) relating the inverse exponential distribution parameters to the atmospheric temperature and hydrometeor content are already implemented.

In the PSD construction, one or two PSD parameters are free parameters depending on the settings. PAMTRA derives the values of the unknown parameters by resolving the system of equations for the moments M_k that are given by the model output. The k-th moment of a generic size distribution f(D) is defined as $M_k = \int D^k f(D) dD$. At the current development stage, PAMTRA can use three different quantities related to the PSD moments as input, namely: the total number concentration $N_T = M_0$, the effective radius $r_e = M_3/2M_2$, and the mass mixing ratio $q = aM_b$, where a and b are the parameters of the power-law defining the mass—size relation $m(D) = aD^b$.

PAMTRA can also handle size-resolved distributions of particles giving the largest flexibility in the definition of hydrometeor content and properties. With this tool, it is possible to set the properties (i.e., mass, area, hydrometeor terminal velocity) of particles for each size range, allowing PAMTRA to ingest in situ observations (see example Sect. 3.4) or the output of numerical models employing size-resolved (binned) microphysical schemes. This flexible interface can also be used to connect PAMTRA with atmospheric models that do not require predefined hydrometeor properties such as those involving the Particle Prediction Properties (P3; Morrison and Milbrandt, 2015) microphysical scheme, or even the semi-Lagrangian super-particle models used for snow (McSnow; Brdar and Seifert, 2018) or drizzle formation (Hoffmann et al., 2017; Maahn et al., 2019).

2.5.2 Single scattering and absorption properties

For liquid hydrometeors, such as cloud droplets, drizzle, or rain dropsraindrops, the single scattering properties are calculated using Mie theory (Mie, 1908). A large number of refractive index models for liquid water have been published over the last decades (Liebe et al., 1991, 1993; Ellison, 2006, 2007; Stogryn, 1995; Rosenkranz, 2015; Turner et al., 2016). Some of them, such as Liebe et al. (1991, 1993), are well accepted and very commonly used for liquid water in microwave RT. For liquid water at temperatures higher than 0 °C and the lower frequency range (<150 GHz), the refractive indices of various models are relatively similar. However, for super-cooled liquid water (i.e., liquid water at temperatures below freezing) and higher frequencies, the models increasingly deviate from each other because laboratory measurements of the refractive index in this region are lacking (Kneifel et al., 2014; Cadeddu and Turner, 2011). Recent observations of super-cooled clouds at various sites (Kneifel et al., 2014) triggered the development of new refractive index models which combine the existing laboratory data set with the new cloud observations (Rosenkranz, 2015; Turner et al., 2016). The model of Turner et al. (2016) is used as the default liquid water refractive index model in PAMTRA. Other models, such as Liebe et al. (1993), Ray (1972), Stogryn (1995), and Ellison (2006), can be chosen by the user in order to allow comparison studies with other RT models or with previous RT simulations.

Frozen hydrometeors, such as ice crystals, snowflakes, or rimed particles, comprise a large natural variability of habits, densities, and orientations. This variability also affects their interaction with electromagnetic radiation, which explains the still large uncertainties in simulating their radiative properties. As a result, the number of scattering databases with various amount levels of complexity is rapidly increasing (Kneifel et al., 2018). In PAMTRA, there are several options regarding the definition of particle properties as well as the selection of scattering models. If the ice refractive index is not implicitly included in the selected scattering database, it is calculated using the model by Mätzler (2006).

One of the most widely used approximations for ice and snow particles are spheres or spheroids (Bennartz and Petty, 2002; Petty, 2001; Honeyager et al., 2016; Hogan et al., 2012; Tyynela et al., 2011; Matrosov, 2015). Frozen hydrometeors are usually not composed of a homogeneous medium but rather a mixture of ice, air, or liquid water. Hence, spheroidal approximations always require the calculation of an effective refractive index. In PAMTRA, the generic mixing rule by Sihvola and Shivola (1989) is used. It should be noted that differences between various mixing formulas might be significant and we have adopted the mixing rule that Petty and Huang (2009) found to cause the smallest deviation of the scattering properties of spheres when compared with more realistic snowflake shapes. PAMTRA allows to define defining either a constant density or a size dependent mass—size relation. The scattering properties are then calculated using the cost effective Dependent on the requirements on computational speed and particle properties, the scattering properties can be calculated using Mie (Mie, 1908) or the more time consuming T-matrix theory (Mishchenko and Travis, 1994); for the latter, also the latter also requires the definition of orientation and aspect ratio of the particles have to be defined.

The spheroidal approximations and in particular the effective refractive index calculations become increasingly unrealistic as soon as the wavelength becomes similar to the particle size. However, the size at which more complex particle models should be used also depends on the scattering variable (Schrom and Kumjian, 2017). The Discrete Dipole Approximation (DDA; Purcell

and Pennypacker, 1973) is considered as a reference method to compute scattering properties of complex shaped particles. An increasing number of databases with various particles and scattering variables have been developed during recent years (Kneifel et al., 2018). At the moment, the user can select particles of the DDA databases from Liu (2008) and Hong et al. (2009) which provide a number of single ice crystal types as well as a small number of aggregates. In particular for aggregates, PAMTRA also includes the very recent SSRGA for active and passive simulations (Hogan et al., 2017). The SSRGA is a cost effective method to calculate the full phase function representative for an ensemble of aggregates. Unlike the soft spheroidal approximations, no effective refractive index is needed but the fluctuations of mass, which mainly characterize the non-Rayleigh scattering, are described with a number of coefficients. Those have been derived from a large ensemble of aggregates as described in detail in Hogan et al. (2017). The method is also applicable with reasonable accuracy to light and moderately rimed aggregates (Leinonen et al., 2018). A limitation of the SSRGA is that polarimetric variables cannot be estimated because the interaction of the scattering elements inside the particle are is neglected. To our knowledge, PAMTRA is the first RT model which allows to use the use of SSRGA for passive microwave simulations.

3 Application examples

When developing PAMTRA specific emphasis has been on its ability to interface a broad spectrum of microwave instruments and observing geometries with common atmospheric models and their different output variables and hydrometeor schemes. Here, we demonstrate the high versatility of PAMTRA with a number of application examples based on data from recent field campaigns and state-of-the-art atmospheric models.

All following simulations assume spheres (Mie) for the hydrometeor categories liquid water, rain, graupel, and hail. The SSRGA with the coefficients as in Mason et al. (2019) are used for cloud ice and snow. Herein, using the SSRGA allows us to ensure maximum consistency regarding particle properties such as mass—size relation assumed in the microphysical schemes. The benefit of the SSRGA is illustrated for satellite measurements within a eases—case study using the Integrated Forecasting System (IFS) from the European Centre for Medium-Range Weather Forecasts (ECMWF) model output together with passive microwave satellite observations (Sect. 3.1). How PAMTRA can be used for understanding cloud and precipitation processes as well as their representation in the novel ICOsahedral Non-hydrostatic atmosphere model (ICON; Zängl et al., 2015) is shown for both ground-based (Sect. 3.2) and airborne measurements (Sect. 3.3). While PAMTRA is interfaced with the two-moment microphysical scheme by Seifert and Beheng (2006) for the ICON application, Sect. 3.4 demonstrates PAMTRA's ability to ingest spectrally resolved information - in this case provided by airborne in situ measurements. The eode scripts to reproduce the simulations and all following figures is are available in the supplement.

3.1 Satellite perspective

Microwave satellite observations from polar orbiters, e.g.,—Advanced Microwave Sounding Unit-A/B (AMSU-A/B) or Microwave Humidity Sounder (MHS), have provided fundamental insights into tropical storms due to their unique ability to penetrate even opaque cloud systems (Kidder et al., 2000). Furthermore, their assimilation into NWP strongly contributes to

forecast skills (Geer et al., 2017), though the assimilation at higher frequencies becomes difficult due to the complex interaction of microwave radiation with frozen hydrometeors in the forward simulation.

To illustrate the benefit of the SSRGA compared to the conventional Mie approach for frozen hydrometeors, we selected a scene from ex-tropical storm Karl, which has been investigated during the North Atlantic Waveguide and Downstream impact EXperiment (NAWDEX; Schäfler et al., 2018). The ECMWF IFS cycle 41r2 with a 0.1° grid (6 to 7 km) resolution provides the atmospheric input fields for PAMTRA. The IFS applies a one-moment microphysical scheme having four hydrometeor categories with mono-disperse cloud categories liquid and ice and exponentially distributed rain and snow (Forbes et al., 2011) as prognostic variables. Because the convective rain and snow flux profiles are not available in the standard output, we - in contrast to Geer and Baordo (2014) - can not consider their contribution which may modify the results. More details on the treatment of the IFS microphysics in PAMTRA are given in appendix A1.

Microwave T_B contain frequency dependent contributions from atmospheric gases and hydrometeors which are difficult to disentangle as can be seen for AMSU-A/MHS measurements (Fig. 2a–c). Liquid clouds and precipitation usually appear as enhanced T_B over radiatively cold surfaces like the ocean (emissivity of 0.5–0.7). Scattering at frozen hydrometeors, i.e. ; ice, snow, graupel, or hail, lead leads to a depression in T_B (observed from space), which becomes stronger with increasing frequency. Due to the also increasing absorption by water vapor with higher frequencies, the surface influence is reduced and the scattering effects are better distinguishable from the surface effects (Skofronick-Jackson and Johnson, 2011). In order to illustrate the scattering effect, three window channels, i.e. ,50.3, 89, and 157 GHz were selected. The observed scenes clearly reveal the cyclonic nature of the storm. The occurrence of snow precipitation in its north-easterly sector can be clearly identified by its scattering effect which leads to stronger decreases in T_B with increasing frequency (Fig. 2a–c).

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PAMTRA was run twice, once using Mie theory for the calculation of the single scattering properties of cloud ice and snow particles and once using the SSRGA. To match the output with the satellite observation, simulated T_B were convoluted according to the satellite geometry. Looking at the differences in T_B between observation and both simulations (j–o), it can be seen that especially for the lower two frequencies the simulations show slightly lower values especially in the southern part of the area. At this these frequencies, the signal is mainly driven by emission from the surface, the water vapor, and liquid hydrometeors, and not so much by scattering at frozen hydrometeors (Skofronick-Jackson and Johnson, 2011). Therefore Since the ocean surface signal in the microwave region can be model quite well by TESSEM² and the sea surface temperature in the model and reality does not differ that much, the differences in the T_B can be most likely attributed to an underestimation of the liquid water contents (Fig. 2s2s) or to the water vapor field (not shown). Differences in the surface signal can be excluded, since the differences are still present at the highest simulated. At the higher frequency of 157 GHz (Fig. 2l,o), where the surface influence can be neglected. At this higher frequency the scattering at larger frozen hydrometeors becomes more important. In the Mie simulation, a clear underestimation of the scattering effect can be noted as no T_B depression is present in the simulated field (Fig. 2f), although the IFS produces considerable amounts of snow as shown in the integrated hydrometeor contents with a snow water path up to 6 kg m^{-2} (Fig. 2u). The underestimation of the scattering by Mie theory is in accordance with previous studies which find spherical particles introducing a significant positive biases in simulated T_B (Geer and Baordo, 2014). In contrast, simulations with SSRGA (Hogan and Westbrook, 2014; Hogan et al., 2017) are in general able to produce T_B

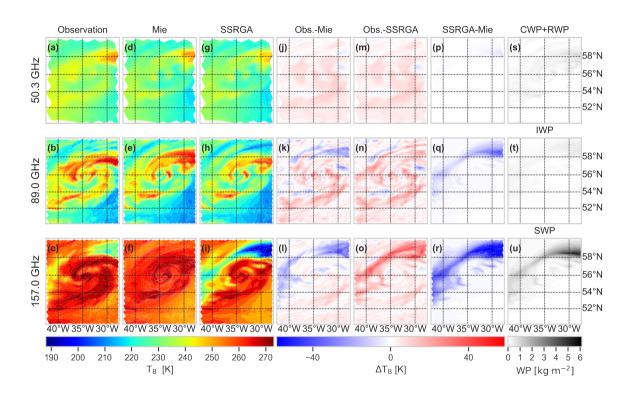


Figure 2. Observations with AMSU-A at 50.3 GHz (a) and MHS for at 89.0 (b) and 157.0 GHz (c) for ex-tropical cyclone Karl on 26 September 2016 during the NAWDEX campaign over the North Atlantic; simulations with IFS and PAMTRA with single scattering properties calculated with Mie theory (d–f) and simulations with SSRGA for ice and snow (g–i). Difference in T_B for Obs-Mie (j–l), Obs-SSRGA (m–o), and SSRGA-Mie (p–r). Integrated contents as the sum of cloud water and of rain water path (CWP+RWP) (s), ice water path (IWP) (t), and snow water path (SWP) (u).

depressions in agreement with the observations. For the simulations shown here based on IFS and PAMTRA using SSRGA for the frozen hydrometeors, the depression is much stronger than for Mie (Fig. 2r) and comparing it to the observation (Fig. 2o), it can be seen that it is even stronger than in the observation of MHS for the north-eastern area, although the contributions to the total precipitating hydrometeor amount through convection are not included in the simulations. With the aforementioned capability of SSRGA to reproduce T_B depressions in agreement with observations, this overestimation can be linked might be either connected to an overestimation of the snow water content of ECMWF IFS especially in the middle and upper troposphere or to an overestimation of the scattering by the SSRGA.

3.2 Ground-based perspective

Novel remote sensing instrumentation combined with high resolution modeling is seen as a way forward to better understand cloud and precipitation processes. In this case study, we demonstrate how PAMTRA can be used to simulate a wealth of state-of-the-art ground-based active and passive microwave of observations including radar Doppler spectra at multiple frequencies. The observations shown in Fig. 3 have been recorded on 19 November 2015 as part of the TRIple-frequency and Polarimetric radar Experiment for improving process observation of winter precipitation campaign (TRIPEx; Dias Neto et al., 2019) at the Jülich Observatory for Cloud Evolution Core Facility (JOYCE-CF; Löhnert et al., 2015). The data have been carefully quality controlled and corrected for radar calibration biases and attenuation by gases and hydrometeors as described in detail in Dias Neto et al. (2019).

The novel ICON model in its Large Eddy version (ICON-LEM; Heinze et al., 2017) with a horizontal resolution of 600 m and 150 vertical layers is used as input to PAMTRA. ICON-LEM is forced by initial and lateral boundary conditions from the ECMWF IFS. The ICON-LEM model used here implements the 2-moments microphysical scheme from Seifert and Beheng (2006). The cloud scheme has six hydrometeor classes (cloud drops, rain, cloud ice, snow, graupel, and hail) which are assumed to be distributed according to a modified gamma function (Petty and Huang, 2011). The model simulates the evolution of two moments of the hydrometeor distributions, namely the mass mixing ratio *q* and the total number concentration *N*. Details on the treatment of the ICON microphysical scheme are given in appendix A2. The forward simulations take the different radar specifications (e.g., sensitivity, beam widths, and averaging interval) as described in Dias Neto et al. (2019) into account. T_B are simulated for the 14 channels of a Humidity and Temperature PROfiler (HATPRO; Rose et al., 2005). The passage of a cold front on 19 December November 2015 (Fig. 3) is nicely captured by the ICON simulations both regarding vertical and temporal evolution. The first 6 h of model simulations are likely affected by the spin-up of the model (started at 00 UTC) and therefore have been excluded from the figure.

The cloud and precipitation field associated to with the cold front causes similar reflectivity structures in the forward simulations as observed. Although the ICON/PAMTRA setup currently does not include a melting layer model, the transition from ice to rain at 1.5-2 km can be clearly seen in the reflectivity and particularly in the mean Doppler velocity (note the well-matched increase in melting layer height). During periods of most intense rainfall (up to 5.6 mm h⁻¹ between 14 and 15 UTC), attenuation effects are somewhat overestimated in the model but overall the observed signatures are well captured.

Passive observations are unreliable during rainy periods due to potential liquid water on the radome (Cadeddu et al., 2017). During non-precipitating conditions, the overall spectral response of the different channels matches the observations very well albeit the high-frequent fluctuations associated with liquid cloud in the observed T_B are missed. This might be due to lower spatio-temporal resolution of the ICON simulation which reproduces the basic temporal evolution but small scale fluctuations cannot be captured with the resolution used.

A more detailed comparison of modeled and observed microphysical processes is possible due to the ability of PAMTRA to simulate the entire radar Doppler spectrum (Fig. 4). The vertical distribution of the Doppler spectrum during the core precipitation period nicely shows the transition from the slow and narrow ice and snow spectra to wider and faster rainfall at around 2 km height in both simulations and observations. In the ice part, the simulated spectra sometimes reveal bi-modalities and too large fall velocities (up to 2 m s⁻¹). This might indicate some discrepancies in the transition from ice to the snow hydrometeor category in the ICON model or rimed particles in the model with larger fall velocities which are not observed. The observations show at certain heights dynamical effects such as shifting due to vertical air motions or broadening due to turbulence. PAMTRA is principally able to account for these effects if vertical air motion or Eddy dissipation rate is provided. Looking at individual spectra in the ice part (Fig. 4c,d), one can see that the noise levels, shape and velocity region of the ice Doppler spectra are very well matched. In the rain part (Fig. 4e,f), the Doppler spectra reveal typical resonances at larger drop diameters (first minimum at 6 m s⁻¹ corresponding to 1.7 mm size drops (Kollias et al., 2002)) which are also well captured by the PAMTRA forward simulations. The differences in the noise levels (especially Ka-band) are due to known saturation effects in the Ka-band receiver which enhances the spectral noise. The slight mismatch of the W-band noise level is due to height dependent chirp table configuration and associated variable sensitivity (Küchler et al., 2018). In this simulation, PAMTRA was configured to match the highest chirp sequence and hence the noise level at lower ranges is underestimated.

The ability of PAMTRA to consistently simulate a multitude of radar observables in combination with passive observations provides new opportunities to evaluate microphysics schemes on a process level. For example, the multi-frequency radar observations can be used to distinguish aggregation and riming dominated regions (Kneifel et al., 2015). Additional constraints on e.g. the assumed PSD or terminal velocity—size relation used in a microphysics scheme can be provided by multi-frequency Doppler spectra (Li and Moisseev, 2019; Kneifel et al., 2016). Finally, the passive observations add information on temperature and humidity profile as well as on vertically integrated liquid water and ice content (Kneifel et al., 2010). PAMTRA is thus not only an important tool to derive new retrievals (Maahn and Löhnert, 2017) but can also be used to develop new microphysical parameterizations as new schemes can be directly confronted with observational characteristics (e.g., typical properties of the radar Doppler spectra).

3.3 Airborne remote sensing perspective

Widespread arctic Arctic mixed-phase clouds present one of the largest challenges to atmospheric models for weather and climate applications (Morrison and Milbrandt, 2015). Airborne campaigns can provide unique information in this measurement void area area where ground based observations are made at very few stations and where polar orbiting satellites have a rather coarse spatial resolution. Here we want to demonstrate how airborne active and passive microwave observations can be ex-

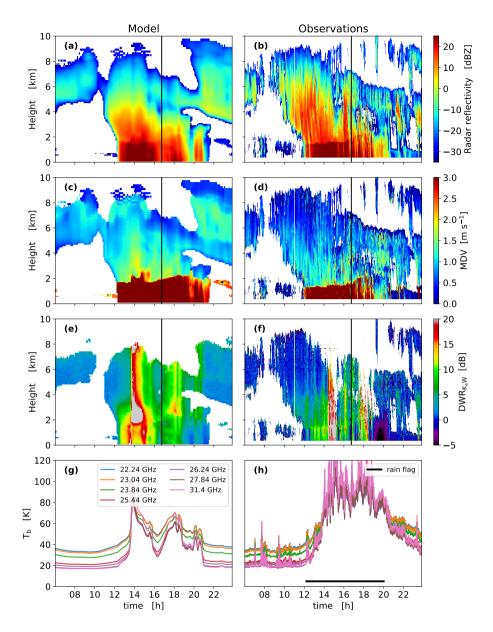


Figure 3. Case study of a cold front observed at JOYCE site on 19 November 2015. From top to bottom: time versus height above ground showing equivalent radar reflectivity factor Z_e [dBZ] at Ka-band (a) and (b), mean Doppler velocity MDV [m s⁻¹] (c) and (d), dual wavelength ratio DWR between Ka- and W-band [dB] (e) and (f), T_B [K] of the HATPRO microwave radiometer for the seven water vapor (g) and (h). The right side shows the observations, PAMTRA simulations based on the ICON-LEM output are shown in the left column. The black horizontal bar in the HATPRO observation plots indicates periods of active rain flag of the microwave radiometer; data during this period are likely to be disturbed by rain on the radome. The vertical black line in the radar time-height plots indicate the time that is used for the spectra comparison in Fig. 4.

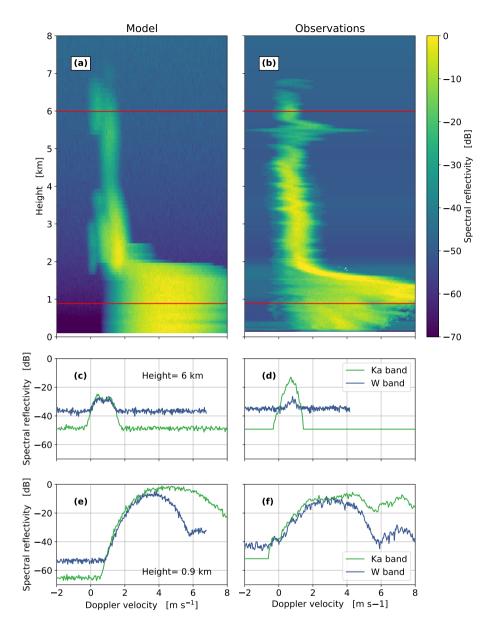


Figure 4. Comparison of simulated (left) and observed (right) Doppler spectra at 16:20 UTC of the same frontal case from 19 November 2015 shown in Fig. 3. From top to bottom: Ka-band spectrograms (a) and (b), example Ka- and W-band Doppler spectra in the ice (c) and (d) and the rain (e) and (f) parts of the cloud. The horizontal red lines in the spectrograms indicate the heights at which the example spectra of the successive panels have been extracted.

ploited with the help of PAMTRA simulations to constrain microphysical schemes in CRMs. For this purpose, PAMTRA settings are adapted to mimic the measurements of upward directed passive and active radiation made by the airborne Microwave Radar/radiometer for Arctic Clouds (MiRAC; Mech et al., 2019a)flown during the Arctic CLoud Observations Using airborne measurements during polar Day (ACLOUD; Wendisch et al., 2019) campaign. MiRAC combines a 94 GHz frequency modulated continuous wave (FMCW) radar with its integrated passive 89 GHz channel and novel 180–340 GHz radiometer. It was operated aboard the *Polar 5* research aircraft of the Alfred Wegener Institute, Helmholtz Centre for Polar and Marine Research (AWI) over the Arctic ocean Ocean and the sea ice north-west of Svalbard in Mai/June 2017.

The measurements were taken on a flight section of research flight 5 in a cold air outbreak over the Fram street West of Svalbard on 25 May 2017 between 11:30 and 12:00 UTC. The aircraft was flying from West to East over open ocean perpendicular to the atmospheric flow. The reflectivity measurements shown in Fig.5 nicely depict the typical roll cloud structure that develops when an Arctic air mass transitions from the central arctic to the open ocean during a cold air outbreak (Liu et al., 2006). Their vertical extent extent is around 750 m and has horizontal length scales of up to 3 km in the observations. By the strong reflectivities in the lowest atmospheric layers, it can be seen that some of the rolls are connected to precipitation, most likely as snow. The enhanced T_B of the 89 GHz passive channel indicate the presence of liquid water over the radiatively cold ocean. A simple regression algorithm for liquid water path (LWP) has been derived from PAMTRA simulated T_B using nearby dropsondes and artificial clouds, giving an estimate of a maximum LWP of 80 g m⁻².

Similar to the example shown for ground-based perspectives (Sect. 3.2), the ICON-LEM model was used to simulate the atmospheric conditions this time in a nested approach with the final horizontal resolution of 150 m. Two different simulations have been performed, the first with a standard, fixed vertical profile for cloud condensation nuclei (CCN) and ice nuclei (IN) and second with a parameterization for CCN/IN activation based on (Hande et al., 2016). For the second realization, the fixed profile was replaced by prognostic CCN and IN and the major part of the change was caused by the activation scheme by (Phillips et al., 2008). The sea surface emissivity is calculated by TESSEM² based on the ICON-LEM input, i.e. , wind speed and sea surface temperature. Gaseous absorption has been calculated according to the Rosenkranz 98 m-odelmodel.

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As can be seen in Fig. 5, the general structure of the roll clouds with approx. $800 \,\mathrm{m}$ top height is well captured with the ICON model resolution of $150 \,\mathrm{m}$ in both simulations. The vertical, as well as the horizontal scales of the roll clouds, are similar to the observations. The simulated radar reflectivities with the ICON standard setup (Fig. 5c) are much lower than the observed ones and basically confined to a few $100 \,\mathrm{m}$ thick cloud layer. Hardly any precipitation reaching the ground is visible in stark contrast to observations. Since the reflectivity at this frequency is mainly driven by large frozen hydrometeors, this indicates too few snow hydrometeors in the simulations. The simulated brightness temperatures agree better with their observational counterpart though a slightly enlarged amplitude - indicating higher LWP - can be seen in the simulations. Looking at the ICON simulation with modified CCN/IN activation radar reflectivities are generally enhanced compared to the original simulation with maxima of $+10 \,\mathrm{dBZ}$ compared to $-20 \,\mathrm{dBZ}$ in the original run and are now much closer to the observations. As the amplitude of the T_{B} signal is slightly reduced one can conclude that the modified scheme is able to convert liquid water more efficiently into ice precipitation.

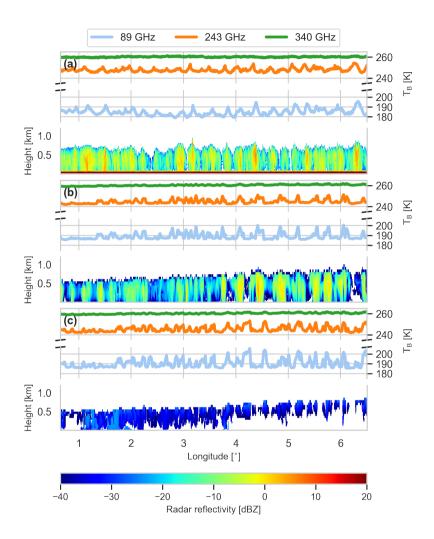


Figure 5. Radar reflectivity at 94 GHz and T_B at 89 GHz (blue) with horizontal polarization and 243 (orange) and 340 GHz (green) with mixed polarization as measured by the MiRAC instrument (a) during a 30 min flight section in west-east direction over the Fram street on the 25 May 2017 and simulated radar reflectivity and T_B with ICON-LEM and PAMTRA after improvements made to the microphysical scheme in ICON-LEM (b) and before (c). Note, that radar and 89 GHz measurements were performed under an angle of 25 $^{\circ}$ backwards.

For the same scene also coarser resolution ICON standard runs were evaluated which revealed an even larger underestimation of radar reflectivities. While much larger samples of observations and simulations are needed to draw solid conclusions this case study demonstrates the ability of PAMTRA in testing different microphysical schemes and suitability.

10 3.4 Airborne in situ perspective

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In the previous examples, typical bulk microphysical schemes were employed in the atmospheric models which implicitly assume a functional relation of hydrometeor properties (e.g., PSD) to which the prognostic model variables, e.g. mixing ratios, number concentrations, can be directly related. New model developments, such as the P3 microphysical scheme (Morrison and Milbrandt, 2015) pose a challenge for RT as their hydrometeor properties, e.g. density, are variable. Similarly, Lagrangian super-particle models (Brdar and Seifert, 2018), models with full-bin microphysics or box models (Hoffmann et al., 2017) require similar flexibility in the assumptions of hydrometeor properties from the RT. PAMTRA addresses those needs with a full-bin interface (Maahn et al., 2019). In order to demonstrate this feature, we simulate radar Doppler spectra based on airborne in situ observations of liquid clouds. The direct use, i.e., without the need to fit any functional form to the particle properties, of in situ observations in the RT provides numerous possibilities for closure studies between in situ and remotely sensed observations.

The in situ observations of liquid cloud properties have been obtained from the 5th Department of Energy Atmospheric Radiation Measurement (DOE ARM) Program's Airborne Carbon Measurements (ACME-V) campaign obtained at the North Slope of Alaska in Summer 2015. The ARM Gulfstream G-159 (G-1) aircraft of the ARM aerial facility (Schmid et al., 2014, 2016) measured the cloud droplet number concentration for droplets larger than 1.5 µm using a combination of optical cloud probes. The probes and the processing of the cloud probe data set following Wu and McFarquhar (2016) are detailed in Maahn et al. (2017). In contrast to the other examples, no particle size distribution is assumed but the measured PSD is directly used in PAMTRA through the full-bin interface. Besides the PSD, also the particle mass, density, cross section area, and aspect ratio need to be defined in PAMTRA for every size bin which is trivial for liquid particles. To focus on the idealized development of the spectrum, vertical air motions are not considered in this example.

Figure 6 shows the observed PSD and the resulting simulated radar Doppler spectrum for a vertically sampled cloud at around 22:36 UTC on 27 June 2015. The observed PSD and effective diameter (the ratio of the third and the second moment analogue to the effective radius) show clearly the near-adiabatic increase of droplet size with increasing height caused by condensation. When forward modeled with PAMTRA, this leads to an increase of Z_e because droplet backscattering scales with diameter D^6 . In the height resolved Doppler spectra (Fig. 6b), this is mainly reflected in the increased spectral reflectivity within the cloud mode with height. Herein, due to the low fall velocity of cloud droplets, their Doppler velocities are basically limited to below 0.5 m s^{-1} even close to cloud top where droplets are largest. However, the radar Doppler spectra reveal Doppler velocities up to 1 m s^{-1} in certain heights sometimes showing a clear bi-modality of the spectra, e.g. 800 and 1000 m. These can clearly be attributed to drizzle droplets that the high impact of low-concentration drizzle droplets on radar observations. These drizzle droplets are not visible in the in situ measurements (Fig. 6a) despite the logarithmic color scale as they are rather rare. Their number concentration is likely even underestimated in the ACME-V data set due to the rare. The small sampling volume of the

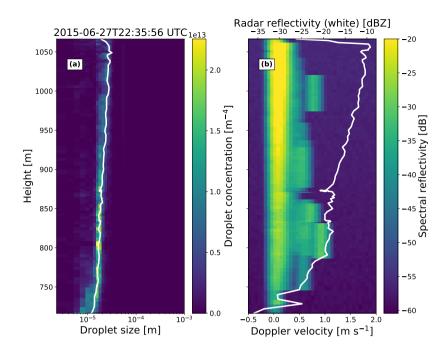


Figure 6. (a) Observed drop size distribution and (b) forward modelled radar Doppler spectrum of the sample cloud of the ACME-V campaign. The white line denote denotes (a) the effective diameter and (b) the radar reflectivity Z_e .

used optical probes and thus their influence on radar spectra is likely even underestimated optical probes used during ACME-V leads to poor statistics for drizzle drops which can explain the presence of inhomogeneities in the spectra forward modeled with PAMTRA (Fig. 6b).

This example has shown how spectrally resolved information can be exploited by PAMTRA to investigate the impact of different hydrometeors on radar Doppler spectra. Even the simple example of liquid only hydrometeors has shown the high impact of few larger particles on the shape of Doppler spectra. Therefore higher moments of the Doppler spectra, e.g. skewness, kurtosis, contain information which can be used in retrieval algorithms to disentangle the cloud and the drizzle contribution to Z_e (Küchler et al., 2018). Fingerprinting of characteristic hydrometeor signatures in the spectra becomes even more important for frozen hydrometeors allowing detailed process studies (Kalesse et al., 2016).

4 Summary and future perspectives

This study introduced the first publicly available version of the PAMTRA forward operator, whose development was motivated by the growing interest to better exploit the unique characteristics of microwave observations in providing information on clouds and precipitation. Specifically, the combination of passive and active microwave sensors on different platforms is very

attractive due to their complementary information. To fully exploit the information of the measurements for process studies, the evaluation and further development of cloud resolving models, PAMTRA has been designed as a versatile tool to be compliant with a wide variety of model output and in situ observations as input. Furthermore, PAMTRA aims to support the ongoing development and application of ground-based instrumentation in particular for multi-frequency radar Doppler spectra measurements, airborne active/passive instrument packages and satellite measurements which will be further extended into the sub-millimeter range, i.e. by the Ice Cloud Imager (ICI; Kangas et al., 2014) including channels up to 664 GHz.

PAMTRA simulates the one-dimensional radiative transfer in a plane-parallel atmosphere for polarized passive as well as the full radar Doppler spectrum for active applications in up- and downlooking mode. PAMTRA has many features already included, i.e., different gas absorption modules, models for the calculation of the surface emissivity, and different methods to calculate the single scattering properties of hydrometeors. Herein, it is unique as the SSRGA can be applied for both passive and active applications. Due to its modularity, it can be easily extended when new developments, e.g., new absorption models, single scattering databases, become available. As some applications require massive calculations calculations, e.g., databases for retrieval development or model evaluation, the implementation of parallelization features into PAMTRA supports high performance computing.

Within an example section several applications of PAMTRA as a forward simulator were introduced, which can be reproduced by the interested reader with the help of jupyter notebooks (https://github.com/igmk/pamtra/). The examples consider different geometries, i.e. ground-based, aircraft, and satellite, as well as different input sources such as airborne in situ hydrometeor spectra, two-moment cloud resolving model simulations and NWP (ECMWF-IFS) analysis. It should be noted that the modular setup of PAMTRA also allows for simpler information such as idealized atmospheric profiles or radiosonde measurements. The latter is especially common for classical retrieval or information content studies for passive microwave measurements such as in (Ebell et al., 2013).

The representation of cloud and precipitation processes is a long-standing problem for atmospheric models and the development of new parameterizations and schemes is ongoing in particular for frozen hydrometeors. Microwave scattering by frozen particles provides on the one hand insights into the dominating hydrometeors and on the other hand, it is also rather challenging due to the wide variety of particles and thus single scattering properties. Progress can only be achieved by the interplay of cloud and RT modeling and its confrontation with measurements. In this context, the first example (Sect. 3.1) demonstrates that especially for higher frequencies, i.e. millimeter and sub-millimeter range, the conventional Mie approach is not useful. PAMTRA can also use the T-matrix approach for the single scattering calculations which, however, requires knowledge on particle orientation. Therefore, the SSRGA approach already used successfully in the radar community has been implemented also for passive RT and shows promising capabilities.

That the forward approach is helpful in disentangling the different contributions of hydrometeors in the measurements is illustrated by further examples. Herein it is important that PAMTRA can be run in high consistency to the models' microphysical assumptions. From the ground where novel technologies can be deployed fast, multiple frequency radar provides exciting insights into precipitation formation via the ice phase (Sect. 3.2). Airborne measurements allow to reach reaching remote areas such as the Arctic where complex mixed-phase clouds are observed in cold air outbreaks. How active/passive microwave

measurements of such clouds can constrain microphysical schemes in the novel ICON model could be shown in Sect. 3.3 by the different response of radar reflectivities and brightness temperatures in with respect to the relative contributions by frozen and liquid hydrometeors.

We did not show an example for the classical observation-to-model approach where data bases databases of synthetic measurements and corresponding variables of interest are generated for subsequent retrieval development, e.g., (Chaboureau et al., 2008). However, by illustrating how cloud droplet spectra measured by in situ measurements can be used as PAMTRA input for simulating radar Doppler spectra (Sect. 3.4) we could illustrate that higher moments of the spectra can be suitable as retrieval input as they show clear drizzle signatures (Acquistapace et al., 2019; Küchler et al., 2018). Along this line also the passive microwave signatures of drizzle can be simulated to support related retrieval development (Cadeddu et al., 2020). In general, PAMTRA is well suited for synergetic retrieval development as a multitude of microwave measurement quantities, i.e. multi-frequency, polarized brightness temperatures and Doppler spectra moments, can be simulated consistantly consistently for the same atmospheric scene.

For the future development of PAMTRA, namely PAMTRA2.0, it is planed planned to move on to an even more modular code based on python3 to allow an enhanced parallelization. Further features to be taken into account in the development of PAMTRA2.0 are improvements in the simulation of spectral radar polarimetry, parameterization of frozen surface emissivity as well as simpler adaptations to slant observation geometry. Interested scientists are cordially invited to contribute to the PAMTRA through our online repository.

25 Code availability. The current version of PAMTRA can be found in a publicly available GitHub repository distributed under an GPLv3.0 license found at https://github.com/igmk/pamtra. The exact version of PAMTRA as used for this manuscript is archived on Zenodo (Mech et al., 2019b) including the scripts and data to produce the plots shown in the application section. The code documentation and user manual is compiled into a Read the Docs web page available at https://pamtra.readthedocs.io, and jupyter notebooks that introduce the PAMTRA usage by presenting documented examples and links to the required data are included in the public GitHub repository.

30 Appendix A: Consistency between ICON-LEM and IFS microphysical schemes with PAMTRA

This section provides details on how the model output of IFS and ICON-LEM is converted into hydrometeor PSDs in PAMTRA for the computations showed in the example Sects. 3.1, 3.2, and 3.3.

A1 IFS microphysical scheme

The IFS model uses a 1-moment microphysical scheme with four hydrometeor classes, namely cloud drops, cloud ice, snow, and rain. The prognostic moment simulated by the scheme is the mass mixing ratio q.

The liquid cloud drops and cloud ice are assumed to be mono-dispersed with all the particles having an equal maximum diameter D and an equal mass m. As a consequence of that these particles will also have an equal density ρ (Forbes et al., 2011)

and therefore it is possible to derive the number concentration of the particles N by dividing q by the specific volume of the hydrometeors

$$N = \frac{q\rho}{m}.\tag{A1}$$

The rain and snow are assumed to be distributed according to an inverse exponential size distribution

$$N(D) = N_0 \exp(-\Lambda D). \tag{A2}$$

The relations between the mass and size of each particle and the N_0 and Λ parameters of N(D) are both assumed to follow the power-law form

$$N_0 = \alpha \Lambda^{\beta},\tag{A3}$$

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$$m = aD^b,$$
 (A4)

and

$$N(D) = \alpha \Lambda^{\beta} \exp(-\Lambda D) \tag{A5}$$

Equation A5 is solved for Λ through the analytic inversion of the equation for the mixing ratio

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$$q = \int aD^b \alpha \Lambda^\beta \exp(-\Lambda D) dD$$
 (A6)

A2 ICON-LEM microphysical scheme SB06

As already mentioned in Sects. 3.2 and 3.3, the ICON-LEM model used in the examples is implementing the 2-moments Seifert and Behens microphysical scheme (hereafter SB06). SB06 predicts the evolution of six hydrometeor classes, namely cloud drops, rain, cloud ice, snow, graupel, and hail. For each of these hydrometeor classes it is assumed that the mass m is distributed according to a modified gamma distribution. Also the mass of each hydrometeor is related to its maximum size through a power-law function

$$f(m) = N_0 m^{\mu} \exp(-\Lambda m^{\gamma}), \tag{A7}$$

and

$$m(D) = aD^b. (A8)$$

By applying the change of variable rule f(m)dm = N(D)dD, one can see that the size distribution N(D) assumed in PAMTRA is again of the modified-gamma functional form

$$N(D) = f(m(D))\frac{\mathrm{d}m}{\mathrm{d}D} = N_0 b a^{\mu+1} D^{b\mu+b-1} \exp(-\Lambda a^{\gamma} D^{b\gamma}). \tag{A9}$$

The parameters of the PAMTRA size-distribution N(D) (primed variables) can be derived from those of mass-distribution f(x) using the following substitutions

10
$$N(D) = N_0' D^{\mu'} \exp(-\Lambda' D^{\gamma'}),$$
 (A10)

with

$$N_0' = N_0 a^{\mu + 1} b, (A11)$$

$$\mu' = b\mu + b - 1,\tag{A12}$$

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$$\Lambda' = \Lambda a^{\mu},\tag{A13}$$

and

$$\gamma' = b\mu. \tag{A14}$$

For each hydrometeor class the parameters μ and γ are fixed and thus also the PAMTRA parameters μ' and γ' . PAMTRA derives the free parameters N_0' and Λ' by solving the system of equations for the two prognostic moments in the SB06 scheme

$$N = \int N(D) dD \tag{A15}$$

and

$$q = \int aD^b N(D) dD, \tag{A16}$$

which have an analytic solution through the use of the *complete gamma function* $\Gamma(x)$ for values of $\mu' > -1$.

25 Author contributions. M. Mech originally created the PAMTRA model framework consisting of the passive part and the basic methods for gaseous absorption and single scattering calculations. M. Maahn developed the active radar simulator and the pyPamtra framework and designed the documentation and example framework. M. Mech and M. Maahn are the main authors of PAMTRA. P. Kollias supported the

implementation of the radar Doppler spectra simulator in PAMTRA. D. Ori implemented the SSRGA and is strongly involved in the complete development of the single scattering section and the examples. S. Kneifel contributed significantly to the methods of the model for the single scattering calculations, the dielectric properties, and the gaseous absorption. E. Orlandi designed the hydrometeor interface and the particle size distribution methods. V. Schemann performed the cloud resolving model simulations with the ICON-LEM and is strongly involved in the interfacing of atmospheric models to PAMTRA. S. Crewell contributed to the interpretation of the simulation results and the basic model and manuscript design. M. Mech prepared the manuscript with contributions from all co-authors.

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

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Table 1. Main characteristics and features of PAMTRA. Default options are written in bold letters.

| General | Python with Fortran core |
|---------------------------------|--|
| Setup | 1D, plane-parallel, horizontally homogeneous |
| Geometry | Ground-based, airborne, and spaceborne; vertical and slanted view |
| Frequency range | $1-800~\mathrm{GHz}$ |
| Importers for various sources | GCMs, CRMs, soundings, full-bin models, in situ measurements |
| Surface emissivities | FASTEM, TESSEM ² , TELSEM ² TESSEM ² , TELSEM ² |
| Gas absorption | Rosenkranz (1998) (with improvements) Rosenkranz (1998) (with improvements), Liebe et al. (1993) |
| Dielectric properties of ice | Mätzler (2006) Mätzler (2006) |
| Dielectric properties of liquid | Turner et al. (2019) Turner et al. (2019), Ellison (2006), Liebe et al. (1993), Stogryn (1995) |
| Single scattering models | Mie, T-matrix, self-similar Rayleigh-Gans, Liu (2008), Hong (2007) |
| Passive output | Polarized brightness temperatures and radiances Polarized brightness temperatures and radiances |
| Active output | Radar reflectivity Radar reflectivity and higher moments, polarized Doppler spectrum, LDR, ZDR |