



1 **RTTOV-gb v1.0 - Updates on sensors, absorption models,** 2 **uncertainty, and availability**

3 Domenico Cimini^{1,2}, James Hocking³, Francesco De Angelis², Angela Cersosimo¹, Francesco
4 Di Paola¹, Donatello Gallucci¹, Sabrina Gentile¹, Edoardo Gerdali¹, Salvatore Larosa¹,
5 Saverio Nilo¹, Filomena Romano¹, Elisabetta Ricciardelli¹, Ermann Ripepi¹, Mariassunta
6 Viggiano¹, Lorenzo Luini⁴, Carlo Riva⁴, Frank S. Marzano^{5,2}, Pauline Martinet⁶, YunYoung
7 Song⁷, Myoung Hwan Ahn⁷, and Philip W. Rosenkranz⁸.

8 ¹National Research Council of Italy, Institute of Methodologies for Environmental Analysis, Potenza, 85050,
9 Italy

10 ²Center of Excellence CETEMPS, University of L'Aquila, L'Aquila, 67100, Italy

11 ³MET OFFICE, Exeter, United Kingdom

12 ⁴DEIB - Politecnico di Milano, IEIT – CNR, Milano, Italy

13 ⁵University of Rome La Sapienza, Rome, Italy

14 ⁶CNRM, Université de Toulouse, Météo-France, CNRS, Toulouse, France

15 ⁷School of Engineering, Ewha Womans University, Seoul, South Korea

16 ⁸Massachusetts Institute of Technology, Cambridge, MA, 02139, USA

17 *Correspondence to:* D. Cimini (domenico.cimini@imaa.cnr.it)

18 **Abstract.** This paper describes the first official release (v1.0) of RTTOV-gb. RTTOV-gb is a FORTRAN 90
19 code developed by adapting the atmospheric radiative transfer code RTTOV, focused on satellite observing
20 geometry, to the ground-based observing geometry. RTTOV-gb is designed to simulate ground-based upward-
21 looking microwave radiometer (MWR) observations of atmospheric downwelling natural radiation in the
22 frequency range from 22 to 150 GHz. Given an atmospheric profile of temperature, water vapour and,
23 optionally, cloud liquid water content, and together with a viewing geometry, RTTOV-gb computes the bottom
24 of atmosphere radiances and brightness temperatures in each of the channels of the sensor being simulated. In
25 addition, it provides the sensitivity of observations to the atmospheric thermodynamical state, i.e. the Jacobians.
26 Therefore, RTTOV-gb represents the forward model needed to assimilate ground-based MWR data into
27 numerical weather prediction models, which is currently pursued internationally by several weather services.
28 RTTOV-gb is fully described in a previous paper (De Angelis et al., 2016), while several updates are described
29 here. In particular, two new MWR types and a new parameterization for atmospheric absorption model have
30 been introduced since the first paper. In addition, estimates of the uncertainty associated with the absorption
31 model and with the fast parameterization are given here. Brightness temperatures (T_B) computed with RTTOV-
32 gb v1.0 from radiosonde profiles have been compared with ground-based MWR observations at six channels
33 (23.8, 31.4, 72.5, 83.5, 90.0, and 150.0 GHz). The comparison shows statistics within the expected accuracy.
34 RTTOV-gb is now available to licensed users free of charge from the Numerical Weather Prediction Satellite
35 Application Facility (NWP SAF) website, after registration. Coefficients for four MWR instrument types and
36 two absorption model flavors are also freely available from the RTTOV-gb support website.

37



1

2

3 1 Introduction

4 RTTOV-gb is the FORTRAN-90 code described by De Angelis et al. (2016). RTTOV-gb is a fast radiative
5 transfer code, designed to simulate ground-based upward-looking microwave radiometer (MWR) observations of
6 atmospheric downwelling natural radiation (i.e. radiances). RTTOV-gb was developed by adapting version 11.2
7 of RTTOV, the Radiative Transfer for the TIROS Operational Vertical Sounder (TOVS), which is designed to
8 simulate the satellite observation perspective only. From its first implementation (Eyre, 1991) through to its
9 current version (Saunders et al., 2018), RTTOV simulates radiances from space-borne passive sensors, and also
10 computes the Jacobians, i.e. the gradient of the radiances with respect to the atmospheric state vector. RTTOV is
11 widely used by many national meteorological services for assimilating down-looking observations from visible,
12 infrared, and microwave radiometers, spectrometers and interferometers aboard satellite platforms. For this
13 reason, RTTOV is maintained and continuously developed by the Numerical Weather Prediction (NWP)
14 Satellite Application Facility (SAF) of the European Organization for the Exploitation of Meteorological
15 Satellites (EUMETSAT). However, satellite passive observations are known to lack accuracy and resolution in
16 the planetary boundary layer (PBL), leaving a so-called observational gap between surface and upper
17 troposphere (National Research Council, 2008). Therefore, in the last decade there has been increasing interest
18 for ground-based sensors that could help bridging the PBL observational gap (Illingworth et al., 2015;
19 Illingworth et al., 2019), including ground-based microwave radiometers (MWR). Ground-based MWR
20 observations are also widely used for radiopropagation studies and the characterization of atmospheric
21 attenuation for telecommunication channels (Riva et al., 2014).

22 Data assimilation (DA) of MWR observations into NWP models may be particularly important in forecasting
23 weather and atmospheric attenuation. In order to assimilate ground-based radiometric observations, namely
24 brightness temperatures (T_B), a fast radiative transfer forward model is needed. This model allows rapid
25 simulations of T_B at selected radiometer channels based on the NWP model state vector, i.e. atmospheric
26 temperature and humidity profiles, similar to what RTTOV does for satellite sensors. Therefore, in the
27 framework of the COST Actions EG-CLIMET and TOPROF, there have been continuous activities to develop a
28 ground-based version of RTTOV: RTTOV-gb (De Angelis et al., 2016). RTTOV-gb is a one-dimensional
29 radiative transfer model: it takes vertical profiles of atmospheric temperature, water vapour, and cloud liquid
30 water specified on an arbitrary set of pressure levels and from them it simulates T_B as well as the Jacobians
31 corresponding to ground-based upward-looking microwave radiometers. As hoped, the availability of RTTOV-
32 gb is fostering wider use of MWR observations in NWP models, as demonstrated by the current use at some of
33 the most relevant meteorological services in Europe as well as outside, such as Météo-France, the German
34 Meteorological Service (Deutscher Wetterdienst, DWD), Korean Meteorological Administration (KMA).

35 This paper introduces several updates of RTTOV-gb since its first development (De Angelis et al., 2016). In
36 section 2 we introduce a new absorption model parameterization and two new sensors that have been added as
37 options. Section 3 presents the evaluation against the reference line-by-line radiative transfer model and real
38 radiometric ground-based observations. Section 4 summarizes the findings while Section 5 provide instructions
39 for code and data access and use.



1

2 **2 RTTOV-gb updates**3 **2.1 New sensors**

4 Similar to RTTOV, RTTOV-gb was designed to simulate observations and Jacobians for a suite of instruments,
5 in this case ground-based instead of satellite-borne sensors. The RTTOV-gb optical depth calculation is a
6 parameterisation which requires pre-computed coefficients. These coefficients are specific to each instrument
7 and are stored in coefficient files. Every time a new sensor is added to the sensor suite, a dedicated coefficient
8 file must be generated. The coefficient file contains the regression coefficients to estimate the optical depth for
9 each atmospheric layer and each sensor channel from the thermodynamical properties of the layer through a set
10 of predictors. The predictors are derived from the input state vector profile and depend on the elevation angle θ
11 and pressure P, temperature T, and specific humidity Q at the considered and surrounding levels. Pressure levels
12 and regression limits for T and Q are reported in Table 1. The coefficients are based on a set of 101 pressure
13 levels specifically created for RTTOV-gb which are more dense in the lower atmosphere than the RTTOV
14 coefficient levels usually used for space-borne sensors.

15 While introducing RTTOV-gb, De Angelis et al. (2016) presented results for two sensors, among the most
16 common ground-based MWR worldwide: the Humidity And Temperature Profiler (HATPRO) manufactured
17 by RPG and the MP3000A manufactured by Radiometrics. Since then, two more sensors have been added to the
18 suite: the microwave temperature radiometer TEMPERA (Navas-Guzmán et al., 2017) and the Liquid Water
19 Path K-to-W-band radiometer (LWP_K2W). Note that LWP_K2W is a virtual instrument which includes all the
20 channels offered by the LWP family of ground-based radiometers manufactured by RPG
21 ([https://www.radiometer-physics.de/products/microwave-remote-sensing-instruments/radiometers/lwp-](https://www.radiometer-physics.de/products/microwave-remote-sensing-instruments/radiometers/lwp-radiometers/)
22 [radiometers/](https://www.radiometer-physics.de/products/microwave-remote-sensing-instruments/radiometers/lwp-radiometers/)). The list of currently supported sensors is given in Table 2.

23

24 **2.2 Absorption model**

25 Similar to RTTOV, RTTOV-gb is a parametrized atmospheric radiative transfer code. In the microwave region
26 and for clear sky conditions, the parameterization only affects the atmospheric gas absorption. This means that
27 the optical depth of each layer is only due to absorption by atmospheric gases (mainly oxygen, water vapor, and
28 nitrogen). The parameterization consists in the fact that the layer optical depth is not computed from a complex
29 line-by-line (LBL) absorption model (Clough et al. 2005), but rather from a simplified parametrized model. The
30 simplified model consists in a linear regression, which relates the layer optical depth to predictors derived from
31 the layer atmospheric thermodynamical properties (i.e. pressure, temperature, and humidity). The regression
32 coefficients are computed off-line from a diverse training dataset of atmospheric thermodynamical profiles and
33 corresponding optical depths computed with a LBL model. Thus, RTTOV-gb provides a fast parameterization of
34 the LBL model adopted for the training of the regression coefficients. For the microwave frequency range (10–
35 200 GHz), the regression coefficients of RTTOV are trained using the AMSUTRAN LBL model developed at
36 the Met Office (Turner et al., 2018) which is based on the millimeter-wave propagation model (MPM)



1 introduced by Liebe (1989), with some modifications following Treyakov et al. (2005), Lilijegren et al. (2005),
2 and Payne et al. (2008) (Saunders et al. 2017). Conversely, RTTOV-gb was trained using a later version of
3 MPM, described by Rosenkranz (1998, hereafter R98), which is probably the most used among the ground-based
4 microwave radiometry community. This model is continuously revised and freely available (Rosenkranz, 2017
5 hereafter R17), and its uncertainty has been carefully investigated (Cimini et al., 2018). Therefore, RTTOV-gb
6 has been trained using the R17 model also (version of 17/05/2017 available at
7 http://cetemps.aquila.infn.it/mwrnet/lblmrt_ns.html). Coefficients for both R98 and R17 models are now
8 available. Extending the results in Cimini et al. (2018) from 60 to 150 GHz, Figure 1 shows clear-sky zenith
9 downwelling T_B computed with R17 model and the difference between T_B computed with the two model
10 versions, for six reference atmosphere climatology conditions. The difference spans from -2 to +3 K in the
11 considered frequency range and thus it is not negligible for the sensors currently available for RTTOV-gb.

12 As mentioned, Cimini et al. (2018) investigated the uncertainty of T_B computed with R17 model due to the
13 laboratory uncertainty of the adopted spectroscopic parameters. Through a sensitivity test, they identified 111
14 parameters (6 for water vapor and 105 for oxygen), whose contribution to the total uncertainty was dominant
15 with respect to others. For these 111 parameters, Cimini et al. (2018) estimated the full uncertainty covariance
16 matrix ($\mathbf{Cov}(\mathbf{p})$), from which the T_B uncertainty covariance matrix ($\mathbf{Cov}(\mathbf{T}_B)$) and the square root of its diagonal
17 terms ($\sigma(\mathbf{T}_B)$) were computed. $\sigma(\mathbf{T}_B)$ represents the standard deviation of typical spectroscopic uncertainties to
18 be expected from T_B computed with R17 model. Figure 2 shows $\sigma(\mathbf{T}_B)$ for zenith observations in six
19 climatological atmospheric conditions. Note that uncertainties used here are at 1-sigma level, i.e. applying an
20 unitary coverage factor ($k=1$, as defined by JCGM, 2008).

21 Note that the analysis of Cimini et al. (2018) was limited to the 20-60 GHz range. Here, a new sensitivity
22 analysis has been performed to cover the frequency range of sensors available for RTTOV-gb (20 to 150 GHz).
23 One additional parameter was found to contribute dominantly, namely the water vapor self-broadened continuum
24 temperature dependence exponent n_{cs} , contributing with its uncertainty by 0.2-0.6 K to the total uncertainty of
25 downwelling T_B between 70-150 GHz. By applying the same approach described in Cimini et al. (2018) for
26 other water vapor continuum parameters, the covariance and correlation between n_{cs} and the self-broadened
27 continuum parameter C_s were estimated to be $\mathbf{Cov}(C_s, n_{cs}) = -3.6208 \times 10^{-10} \text{ (km}^{-1} \text{ mb}^{-2} \text{ GHz}^{-2}\text{)}$ and $\text{Cor}(C_s, n_{cs}) = -0.183$,
28 respectively. The covariance of n_{cs} with respect to the other 111 parameters is estimated to be negligible.

29 For more details on RTTOV and the differences between RTTOV-gb and RTTOV, see Hocking et al. (2015),
30 Saunders et al. (2018), and De Angelis et al. (2016).

31

32 3 Validation with reference model and real observations

33 The accuracy of RTTOV-gb T_B simulations has been tested against both the reference LBL model and real
34 ground-based observations. As described by De Angelis et al. (2016), the approach for testing RTTOV-gb
35 against the reference LBL model used for training (i.e. R98 or R17) consists in computing T_B simulations with
36 both models from a set of independent profiles (i.e., not used for training) and to evaluate the statistics of their
37 difference, namely the mean (bias) and root-mean-square (rms) difference. For the original two sensors
38 (HATPRO and MP3000A), De Angelis et al. (2016) reported in their Tables 2 and 3 the statistics (bias and rms)
39 for the comparison between RTTOV-gb and the LBL model used for training (R98 in their case) against an



1 independent profile set at four elevation angles (90, 30, 19, and 10°). Similarly, here we report the statistics for
2 the two new sensors (i.e. TEMPERA and LWP_K2W) and the same R98 LBL model, respectively in Table 3
3 and 4. These two tables show that the discrepancies between RTTOV-gb and LBL optical depths lead to
4 negligible T_B main differences. The rms differences at zenith are lower than 0.18 K for all channels. When
5 decreasing the elevation angle, the rms differences decrease for 50-57 GHz channels, while they increase for
6 23/31 and 70-150 GHz channels, in accordance with the different atmospheric opacity. The highest rms
7 differences (0.3 K) are found for window channels 31 and 150 GHz at 10° elevation. Similarly to De Angelis et
8 al. (2016), the main conclusion is that the uncertainty introduced by the fast model approximation is within the
9 typical instrument uncertainty and thus does not dominate the uncertainty budget of observations vs. simulations.
10 Let us underline that Tables 2 and 3 of De Angelis et al. (2016) and Tables 3 and 4 of this paper report statistics
11 when using R98 LBL model for training. The analogous rms obtained using the LBL model R17 are reported in
12 Table 5 as “fast parameterization uncertainty”. As expected, rms values do not differ significantly from those
13 obtained against R98. In fact, this test only tells about the accuracy of the parametrized regression in reproducing
14 the LBL model radiances, which is largely independent of the choice of the LBL model. Table 5 also reports the
15 T_B uncertainty contribution due to the uncertainty of spectroscopic parameters (from Figure 2). The estimated
16 total uncertainty is computed as the sum in quadrature of two contributions: the uncertainty due to fast
17 parameterization and absorption model spectroscopic parameters. The latter dominates the uncertainty budget.
18 The total uncertainty so estimated is reported in Table 5 for each sensor and channel available in RTTOV-gb.
19 RTTOV-gb T_B simulations have been previously compared with real ground-based observations from six
20 HATPRO and one MP3000-A (De Angelis et al., 2016; 2017). The frequency range covered by HATPRO and
21 MP3000-A channels overlaps the frequency range of TEMPERA, so we assume RTTOV-gb has been tested for
22 this sensor as well. Conversely, the frequency range of LWP_K2W extends to higher frequencies (up to 150
23 GHz) to include all the channels offered by the RPG LWP ground-based radiometer family (LWP, LWP-U90,
24 LWP-U72-82, LWP-U150, LWP-90-150). Thus, in the following we present a comparison with observations
25 from a LWP-U72-82 radiometer located at the Polytechnic University campus in Milan (Italy, 45.450 N, 9.183
26 E), and from a LWP-90-150 radiometer located at the Atmospheric Radiation Measurement (ARM) program
27 Southern Great Plains (SGP) central facility in Lamont (OK, USA, 36.605 N, 97.485 W).
28 The LWP-U72-82 instrument has four channels (23.84, 31.4 GHz, 72.5, and 82.5 GHz) and it is mainly used for
29 radiopropagation studies. The available dataset extends for one month (from 16 June to 15 July 2018), including
30 radiometric observations and pressure, temperature, and humidity profiles measured by radiosonde ascents
31 launched twice-daily from the Milan Linate airport (~20 km from the Politechnic University campus).
32 Radiometric observations are collected at a fixed elevation angle (35.3°), matching the direction of the Alphasat
33 telecommunication link. An example of data is shown in Figure 3 for three consecutive days. Here, T_B observed
34 at the four channels are plotted together with RTTOV-gb simulations and their estimated uncertainty. It appears
35 that simulations usually fit the observations within uncertainty, except for periods with clouds (at ~167.0, i.e.
36 00:00 of June 16) and rain (~169.0, i.e. 00:00 of June 18). This is expected as RTTOV-gb simulations are
37 computed from radiosonde measurements, which do not include hydrometeor content, and thus do not take into
38 account the radiative contribution of clouds and rain. Thus, for a fair clear-sky comparison, data affected by
39 either rain or clouds must be screened out. As illustrated in Figure 3, the LWP-U72-82 is equipped with a rain
40 sensor, indicating either rain or no-rain on the antenna radome. Observations during rain, as flagged by the rain



1 sensor, have been discarded. In addition, cloudy conditions have been identified by setting a threshold on the
2 standard deviation of $T_B(31.4 \text{ GHz})$ over a time period. This approach has been previously proposed (Turner et
3 al. 2007; De Angelis et al. 2017), using a threshold of 0.5 K over 1-hour period. Here, we prefer to use a shorter
4 period (10 minutes) and thus we reduced the threshold to 0.2 K accordingly. Thus, data identified as cloudy by
5 the standard deviation of $T_B(31.4 \text{ GHz})$ over a 10-minute period have been discarded. The cloud and rain
6 screening reduced the dataset by $\sim 33\%$, leaving 40 match-ups between clear sky radiosonde and radiometric
7 observations (averaged within ± 5 minutes from the radiosonde launch). Scatter plots of simulated vs. observed
8 T_B at 35.3° elevation for the four channels of LWP-U72-82 are shown in Figure 4. Note that the correlation
9 coefficient is 0.98 for all four channels. The slope is within 5% for all channels but 72.5 GHz ($\sim 8\%$), for which
10 the difference between observations and simulations tend to increase as T_B decrease. This may be due to
11 conditions-dependent uncertainty for this channel, as well as an issue with the instrument gain calibration.
12 Statistics at 23.84 and 31.4 GHz are of the same magnitude of those reported by De Angelis et al. (2017) at 30°
13 elevation (their Figure 5, panel C).

14 The LWP-90-150 instrument has two channels (90.0 and 150.0 GHz) and it is mainly used for the retrieval of
15 total column cloud liquid water content. The instrument considered here has been running at the ARM SGP
16 central facility between November 2006 and November 2013 (not continuously). Here we exploit a 2-month
17 dataset of radiometric and radiosonde observations (ARM, 2018a & 2018b) collected in January-February 2012.
18 This dataset corresponds to relatively dry midlatitude winter conditions. An example of data is shown in Figure 5
19 for three consecutive days, corresponding to a dry clear-sky period with intermittent thick clouds and rain.
20 Observations flagged by the rain sensor have been discarded. In addition, cloudy conditions have been identified
21 with the same approach as described above, i.e. setting a threshold on the 10-min standard deviation of T_B at a
22 window channel, here replacing the 31.4 GHz with the 90.0 GHz channel. However, since $T_B(90\text{GHz})$ has ~ 6
23 times larger sensitivity to water vapor (Cimini et al., 2007), the clear-sky threshold is increased by the same
24 factor, i.e. 1.2 K. Thus, data with 10-minute standard deviation of $T_B(90 \text{ GHz})$ larger than 1.2 K have been
25 discarded. The cloud and rain screening reduced the dataset by $\sim 26\%$, leaving 173 match-ups between clear sky
26 radiosonde and radiometric observations (averaged within ± 5 minutes from the radiosonde launch). Scatter plots
27 of simulated vs. observed T_B at 90° elevation for the two channels of LWP-90-150 are shown in Figure 6. The
28 correlation coefficient is 0.95 and 0.99 for 90 and 150 GHz, respectively, while the slope is within 4% for both
29 channels.

30 Overall, the average differences at all the six LWP_K2W channels are close to the accuracy estimated in Table
31 5D. A direct comparison is given in Figure 7. Here, the estimated uncertainty for the six LWP_K2W channels is
32 compared with the experimental mean difference between simulations and observations. Note that radiometric
33 observations at the four lower frequency channels were collected in June-July in Milan (45°N), while in January-
34 February in Lamont (36°N) at the two higher frequency channels. Thus, the simulation uncertainty is estimated
35 using midlatitude summer conditions for the four lower frequency channels, while midlatitude winter conditions
36 for the two higher frequency channels. The experimental bias is generally larger than the simulation estimated
37 uncertainty, as one would expect since the observations are also affected by uncertainty. Except for the 72.5 GHz
38 channel, the estimated uncertainty and experimental bias are within 0.5 K, which corresponds to the absolute T_B
39 accuracy claimed by the manufacturer for the LWP radiometer series. At 72.5 GHz, as anticipated, observations-



1 simulations differences tend to increase as T_B decrease, possibly due to either conditions-dependent uncertainty
2 or an issue with the instrument gain calibration. This will be subject of future investigation.

3

4 **4 Summary**

5 RTTOV-gb v1.0 is now freely available, after website registration (see Section 5). The updates with respect to
6 the original development (described in De Angelis et al., 2016) are presented here, including two additional
7 sensors, an additional parameterization for the training atmospheric absorption model, and an estimate of the T_B
8 uncertainty.

9 RTTOV-gb v1.0 has been trained and validated against two versions of a reference line-by-line absorption
10 model, i.e. R98 (Rosenkranz, 1998) and R17 (Rosenkranz, 2017). In the frequency range commonly covered by
11 RTTOV-gb v1.0 sensors, T_B rms differences are smaller than typical sensor uncertainties at all considered
12 channels and for both the reference absorption models. T_B computed with RTTOV-gb v1.0 from radiosonde
13 profiles have been compared with simultaneous ground-based radiometric observations at six channels (23.84,
14 31.4, 72.5, 82.5, 90.0, and 150.0 GHz) and two observing elevation angle (35.3° and 90°). Differences between
15 simulated and measured T_B are within uncertainty as expected from instrumental and simulation contributions.

16 We hope this paper will provide a reference for the exploitation of RTTOV-gb for MWR data assimilation into
17 NWP models, as already started at some meteorological services in Europe as well as in other continents.

18

19 **5 Code and data availability**

20 RTTOV-gb v1.0 is available to licensed users free of charge. RTTOV-gb may be obtained by registering
21 (<https://www.nwpsaf.eu/site/register/>) with the NWP SAF website (<https://www.nwpsaf.eu/>) and then selecting
22 RTTOV-gb in your software preferences. Instructions for compiling and running RTTOV-gb are provided in the
23 RTTOV-gb User Guide within the software package. The software package also includes scripts to verify the
24 installation and FORTRAN code examples for running the RTTOV-gb forward and K (Jacobian) modules.
25 RTTOV-gb is designed for UNIX/Linux systems. The software is now successfully tested on the following
26 architectures and Fortran 90 compilers: Intel systems with gfortran, ifort, NAG, and pgf90, and Apple Mac
27 systems with gfortran.

28 The RTTOV-gb v1.0 code is based on RTTOV v11.2 and the programming interface is identical to that version
29 of RTTOV, though some inputs and outputs are not used by RTTOV-gb. The original RTTOV v11.2 can be
30 obtained from the NWP SAF web site (<http://nwpsaf.eu/site/software/rttov/rttov-v11/>). Thus, the computational
31 performances of RTTOV-gb is similar to that of RTTOV v11.2, which have been reported
32 (https://www.nwpsaf.eu/site/download/documentation/rtm/docs_rttov11/Performance_Tests_RTTOV_v11.2.pdf
33). For clear-sky microwave simulations, the main factor in simulation speed is the number of coefficient levels,
34 which is 101 for RTTOV-gb. Typical clear-sky run-times for RTTOV-gb are ~0.25 ms per profile for the direct
35 model and ~1.0 ms per profile for the Jacobian model, though timings are dependent on the hardware, compiler,
36 and compiler flags being used, as well as, for example, the number of levels in the input profile, the number of
37 channels simulated per profile, and the inclusion or not of cloud liquid water.



1 Note that RTTOV-gb is not supported by NWP SAF. All questions, bug reports or requests for new coefficients
2 should be sent to rttovgb@aquila.infn.it. Always refer to the RTTOV-gb web page for bug fixes, new
3 coefficients, and code updates: <http://cetemps.aquila.infn.it/rttovgb/rttovgb.html>.

4 The RTTOV-gb package contains optical depth coefficient files for sensors supported by RTTOV-gb at the time
5 of release. Coefficients for sensors not currently considered can be requested to rttovgb@aquila.infn.it. Note that
6 RTTOV-gb only supports microwave sensors currently. Other resources include:

7 • Default pressure levels: http://cetemps.aquila.infn.it/mwrnet/main_files/DAT/RTTOVgb_101_levels_p.dat

8 • Regression coefficients: http://cetemps.aquila.infn.it/mwrnet/rttovgb_coefficients.html

9 • Regression limits:
10 [http://cetemps.aquila.infn.it/mwrnet/main_files/DAT/RTTOVgb_101_pressure_levels_and_regression_limits_x](http://cetemps.aquila.infn.it/mwrnet/main_files/DAT/RTTOVgb_101_pressure_levels_and_regression_limits_xlsx)
11 [lsx](#)

12 • NWP SAF profile sets used for the RTTOV-gb training and independent test:
13 https://nwpsaf.eu/deliverables/rtm/profile_datasets.html.

14 For more information on reference profiles and regression limits see the related link on the official RTTOV
15 website ([https://www.nwpsaf.eu/site/software/rttov/download/coefficients/coefficient-](https://www.nwpsaf.eu/site/software/rttov/download/coefficients/coefficient-download/#Reference_profiles_and_regression_limits)
16 [download/#Reference_profiles_and_regression_limits](#)).

17 Finally, the absorption model by Rosenkranz (2017) is available as a FORTRAN 77 code at
18 <http://doi.org/10.21982/M81013>. Older versions, including the one used here (2017/05/15), are available at
19 http://cetemps.aquila.infn.it/mwrnet/lblmrt_ns.html.

20

21 6 Acknowledgements

22 This work has been stimulated through COST Actions supported by COST (European Cooperation in Science
23 and Technology). Support from the European Space Agency through the WRad project (ESA Contract No.
24 4000125141/18/NL/AF) is acknowledged. Part of this research was funded by the Korea Meteorological
25 Administration Research and Development Program under Grant KMI2018-07410. Microwave radiometer and
26 radiosonde data in Lamont (OK, USA) were obtained from the Atmospheric Radiation Measurement (ARM)
27 User Facility, a U.S. Department of Energy (DOE) Office of Science user facility managed by the Office of
28 Biological and Environmental Research.

29

30 *Author contributions.* DC, JH, and FDA designed the research, contributed to data processing and analysis, and
31 wrote the original manuscript. PWR, PM, YS, and MHA contributed to the investigation in Section 2. LL, CR,
32 FSM contributed with curation of observed data. FDP, DG, SG, FR, ER, and ER contributed with software
33 development. AC, EG, SL, SN, MV contributed to validation data analysis. All the co-authors helped to revise
34 the manuscript.

35

36 References

37 Atmospheric Radiation Measurement (ARM) user facility. 2006, updated daily. Microwave Radiometer - High
38 Frequency (MWRHFCAL150). 2012-01-01 to 2012-02-29, Southern Great Plains (SGP) Central Facility,



- 1 Lamont, OK (C1). Compiled by M. Cadeddu and V. Ghate. ARM Data Center. Data set accessed 2018-10-17
2 at <http://dx.doi.org/10.5439/1150245>, 2018a.
- 3 Atmospheric Radiation Measurement (ARM) user facility. 1994, updated daily. Balloon-borne sounding system
4 (SONDEWNP). 2012-01-01 to 2012-02-29, Southern Great Plains (SGP) Central Facility, Lamont, OK
5 (C1). Compiled by R. Coulter, J. Prell, M. Ritsche, and D. Holdridge. ARM Data Center. Data set accessed
6 2018-10-17 at <http://dx.doi.org/10.5439/1150245>, 2018b.
- 7 Cimini, D., Rosenkranz, P. W., Tretyakov, M. Y., Koshelev, M. A., and Romano, F.: Uncertainty of atmospheric
8 microwave absorption model: impact on ground-based radiometer simulations and retrievals, *Atmos. Chem.*
9 *Phys.*, 18, 15231-15259, <https://doi.org/10.5194/acp-18-15231-2018>, 2018.
- 10 Cimini, D., E. R. Westwater, A. J. Gasiewski, M. Klein, V. Leusky, and J. Liljegren, Ground-based millimeter-
11 and submillimeter-wave observations of low vapor and liquid water contents, *IEEE Transactions on*
12 *Geoscience and Remote Sensing*, 45(7), 2169-2180, doi:10.1109/TGRS.2007.897450, July 2007.
- 13 Clough S. A., M.W. Shephard, E.J. Mlawer, J.S. Delamere, M.J. Iacono, K. Cady-Pereira, S. Boukabara, and
14 P.D. Brown: Atmospheric radiative transfer modeling: a summary of the AER codes. *J. Quant. Spectr. Rad.*
15 *Trans.*, 9, 233-244, 2005.
- 16 De Angelis, F., Cimini, D., Löhnert, U., Caumont, O., Haeefe, A., Pospichal, B., Martinet, P., Navas-Guzmán,
17 F., Klein-Baltink, H., Dupont, J.-C., and Hocking, J.: Long-term observations minus background monitoring
18 of ground-based brightness temperatures from a microwave radiometer network, *Atmos. Meas. Tech.*, 10,
19 3947-3961, doi:10.5194/amt-10-3947-2017, 2017.
- 20 De Angelis, F., Cimini, D., Hocking, J., Martinet, P., and Kneifel, S.: RTTOV-gb – adapting the fast radiative
21 transfer model RTTOV for the assimilation of ground-based microwave radiometer observations, *Geosci.*
22 *Model Dev.*, 9, 2721-2739, doi:10.5194/gmd-9-2721-2016, 2016.
- 23 Hocking, J., Rayer, P., Saunders, R., Madricardi, M., Geer, A., Brunel, P., Vidot, J., RTTOV v11 Users Guide,
24 Doc ID: NWPSAF-MO-UD-028 (online:
25 https://www.nwpsaf.eu/site/download/documentation/rtm/docs_rttov11/users_guide_11_v1.4.pdf), 2015.
- 26 Illingworth, A. J., Cimini, D., Haeefe, A., Haeffelin, M., Hervo, M., Kotthaus, S., Löhnert, U., Martinet, P.,
27 Mattis, I., O'Connor, E. J., and Potthast, R.: How can Existing Ground-Based Profiling Instruments Improve
28 European Weather Forecasts? *Bull. Amer. Meteor. Soc.* doi:10.1175/BAMS-D-17-0231.1, in press, 2018.
- 29 Illingworth, A. J., Cimini, D., Gaffard, C., Haeffelin, M., Lehmann, V., Löhnert, U., O'Connor, E., Ruffieux, D.,
30 Exploiting Existing Ground-Based Remote Sensing Networks To Improve High Resolution Weather
31 Forecasts, *Bull. Amer. Meteor. Soc.* doi: 10.1175/BAMS-D-13-00283.1, February, 2015.
- 32 Joint Committee for Guides in Metrology (JCGM): Evaluation of Measurement Data - Guide to the Expression
33 of Uncertainty in Measurement, 2008. Online (last access: 25 May 2018):
34 https://www.bipm.org/utis/common/documents/jcgm/JCGM_100_2008_E.pdf
- 35 Liebe, H. J.: MPM—An atmospheric millimeter wave propagation model, *Int. J. Infrared Millimeter Waves*,
36 10(6), 631–650, 1989.
- 37 Liljegren, J. C., Boukabara, S. A., Cady-Pereira, K., and Clough, S. A.: The effect of the half-width of the 22-
38 GHz water vapor line on retrievals of temperature and water vapor profiles with a twelve-channel microwave
39 radiometer, *IEEE Trans. Geosci. Remote Sens.*, 43, 1102–1108, doi:10.1109/TGRS.2004.839593, 2005.



- 1 National Research Council, Committee on Developing Mesoscale Meteorological Observational Capabilities to
2 Meet Multiple Needs, Observing Weather and Climate from the Ground Up: A Nationwide Network of
3 Networks, ISBN: 978-0-309-12986-2, 250 pages, 2008.
- 4 Navas-Guzmán, F., Kämpfer, N., Schranz, F., Steinbrecht, W., and Haefele, A.: Intercomparison of stratospheric
5 temperature profiles from a ground-based microwave radiometer with other techniques, *Atmos. Chem. Phys.*,
6 17, 14085-14104, <https://doi.org/10.5194/acp-17-14085-2017>, 2017.
- 7 Payne, V.H., J. S. Delamere, K. E. Cady-Pereira, R. R. Gamache, J-L. Moncet, E. J. Mlawer and S. A. Clough:
8 Air-broadened half-widths of the 22 and 183 GHz water vapor lines, *IEEE Trans. Geosci. Remote Sens.*, vol.
9 46, no. 11, pp3601-3617, 2008.
- 10 Riva, C., Capsoni, C., Luini, L., Luccini, M., Nebuloni, R., and Martellucci, A., The challenge of using the W
11 band in satellite communication, *Int. J. Satell. Commun. Network.*, 32:187–200, doi:10.1002/sat.1050, 2014.
- 12 Rosenkranz, P.W.: Line-by-line microwave radiative transfer (non-scattering), *Remote Sens. Code Library*,
13 doi:10.21982/M81013, 2017.
- 14 Rosenkranz, P.W., Water vapor microwave continuum absorption: A comparison of measurements and
15 models. *Radio Science* 33: doi: 10.1029/98RS01182. issn: 0048-6604, 1998.
- 16 Saunders, R., Hocking, J., Turner, E., Rayer, P., Rundle, D., Brunel, P., Vidot, J., Roquet, P., Matricardi, M.,
17 Geer, A., Bormann, N., and Lupu, C., An update on the RTTOV fast radiative transfer model (currently at
18 version 12), *Geosci. Model Dev.*, 11, 2717-2737, doi: 10.5194/gmd-11-2717-2018, 2018.
- 19 Saunders, R., Hocking, J., Rundle, D., Rayer, P., Havemann, S., Matricardi, M., Geer, A., Lupu, C., Brunel, P.,
20 Vidot, J., RTTOV-12 Science and validation report, Doc ID: NWPSAF-MO-TV-41, Version 1.0,
21 16/02/2017, online: https://www.nwpsaf.eu/site/download/documentation/rtm/docs_rttov12/rttov12_svr.pdf,
22 Last downloaded: Jan 11 2018, 2017.
- 23 Saunders, R.W., Matricardi, M., Brunel, P., An Improved Fast Radiative Transfer Model for Assimilation of
24 Satellite Radiance Observations, *Q.J.Royal Meteorol. Soc.*, 125, 1407-1425, doi:
25 10.1002/qj.1999.49712555615, 1999.
- 26 Tretyakov, M. Yu., Koshelev, M.A., Dorovskikh, V.V., Makarov, D.S., and Rosenkranz, P.W.: 60 GHz oxygen
27 band: precise broadening and central frequencies of fine structure lines, absolute absorption profile at
28 atmospheric pressure, and revision of mixing coefficients, *J.Mol.Spectrosc.* 231, pp.1-14, doi:
29 10.1016/j.jms.2004.11.011, 2005.
- 30 Turner, E., Rayer, P., Saunders, R., AMSUTRAN: A microwave transmittance code for satellite remote sensing,
31 *submitted to Geosci. Model Dev.*, 2018
- 32 Turner, D. D., Clough, S. A., Liljegren, J. C., Clothiaux, E. E., Cady-Pereira, K. E., and Gaustad, K. L.:
33 Retrieving liquid water path and precipitable water vapor from the Atmospheric Radiation Measurement
34 (ARM) microwave radiometers, *IEEE T. Geosci. Remote*, 45, 3680–3690, doi: 10.1109/TGRS.2007.903703,
35 2007.
- 36



1

2 **Table 1:** A selection of the 101 pressure levels adopted for RTTOV-gb (De Angelis et al., 2016). The
 3 table reports also the limits for temperature (T) and specific humidity (Q) at each level representing
 4 the range of values used when training the regression coefficients. Note that Q is in ppmv over dry air.
 5 The full matrix is provided as supplement to this manuscript and freely available online¹.

Level (#)	Pressure (1e3 hPa)	Minimum T (K)	Maximum T (K)	Minimum Q (ppmv)	Maximum Q (ppmv)
1	0.0000	143,65	245,95	9,1330E-01	5,2410E+00
11	0.0379	162,77	279,05	1,3280E+00	6,0170E+00
21	0.1349	169,71	259,26	1,2860E-02	1,0250E+02
31	0.2700	182,27	278,60	1,2860E-02	4,5660E+03
41	0.4251	195,91	303,26	2,3870E+00	1,6690E+04
51	0.5841	196,73	315,57	4,8630E+00	2,8090E+04
61	0.7336	189,96	332,20	8,8570E+00	3,7010E+04
71	0.8624	189,96	342,43	7,5350E+00	4,4160E+04
81	0.9618	189,96	349,92	6,7550E+00	5,1280E+04
91	1.0256	189,96	350,08	6,3350E+00	4,7540E+04
101	1.0500	189,96	350,08	6,1880E+00	4,7640E+04

6

7

¹http://cetemps.aquila.infn.it/mwrnet/main_files/DAT/RTTOVgb_101_pressure_levels_and_regression_limits.xlsx



1 **Table 2:** Sensors supported by RTTOV-gb as for October 2018.

Sensor	RTTOV-gb ID	Sensor Chans (#)	Sensor Chans (GHz)
HATPRO	1	14	22.24; 23.04; 23.84; 25.44; 26.24; 27.84; 31.40; 51.26; 52.28; 53.86; 54.94; 56.66; 57.30; 58.00
MP3000A	2	22	22.234; 22.500; 23.034; 23.834; 25.000; 26.234; 28.000; 30.000; 51.248; 51.760; 52.280; 52.804; 53.336; 53.848; 54.400; 54.940; 55.500; 56.020; 56.660; 57.288; 57.964; 58.800;
TEMPERA	3	12	51.25; 51.75; 52.25; 52.85; 53.35; 53.85; 54.40; 54.90; 55.40; 56.00; 56.50; 57.00
LWP_K2W	4	6	23.84; 31.40; 72.50; 82.50; 90.0; 150.0

2

3



1 **Table 3:** Statistics for the comparison between RTTOV-gb and the line-by-line model R98
 2 (Rosenkranz, 1998) used for training against an independent profile set. The TEMPERA instrument
 3 channel number (Chan no.), the channel central frequency, bias, and rms at four elevation angles are
 4 reported.

Chan no. (#)	Central frequency (GHz)	Bias (K)				rms (K)			
		90°	30°	19°	10°	90°	30°	19°	10°
1	51.25	-0.003	-0.019	-0.018	-0.043	0.153	0.158	0.125	0.077
2	51.75	-0.003	-0.016	-0.012	-0.031	0.160	0.148	0.104	0.049
3	52.25	-0.004	-0.010	-0.006	-0.020	0.167	0.131	0.077	0.029
4	52.85	-0.003	0.001	-0.002	-0.010	0.165	0.093	0.041	0.019
5	53.35	-0.001	0.006	-0.003	-0.004	0.141	0.054	0.021	0.015
6	53.85	-0.001	0.002	-0.001	-0.002	0.095	0.026	0.015	0.012
7	54.40	0.001	-0.002	-0.001	-0.001	0.047	0.015	0.011	0.007
8	54.90	0.002	0.000	-0.000	-0.000	0.024	0.011	0.008	0.004
9	55.40	0.002	0.001	0.000	-0.000	0.017	0.008	0.005	0.002
10	56.00	0.003	0.000	0.000	0.000	0.013	0.005	0.003	0.001
11	56.50	0.002	0.001	0.000	0.000	0.011	0.004	0.002	0.001
12	57.00	0.002	0.000	0.000	0.000	0.009	0.003	0.001	0.000

5

6



1 **Table 4:** Same as Table 3 but for the LWP_K2W instrument.

Chan no. (#)	Central frequency (GHz)	Bias (K)				rms (K)			
		90°	30°	19°	10°	90°	30°	19°	10°
1	23.84	0.008	0.004	-0.009	-0.086	0.027	0.032	0.040	0.141
2	31.40	0.008	-0.004	-0.011	-0.107	0.035	0.044	0.059	0.302
3	72.50	0.007	-0.027	-0.038	-0.094	0.146	0.155	0.170	0.185
4	82.50	0.027	-0.024	-0.043	-0.078	0.138	0.138	0.174	0.238
5	90.00	0.030	-0.025	-0.045	-0.067	0.148	0.140	0.180	0.251
6	150.00	-0.006	-0.061	-0.044	0.077	0.172	0.133	0.157	0.301

2

3



1 **Table 5:** RTTOV-gb T_B uncertainty due to forward model and fast parameterization, and their total
 2 squared sum for two extreme climatology conditions. Channels for the four sensors considered in the
 3 current version of RTTOV-gb are given in Tables 5A (HATPRO), 5B (MP3000A), 5C (TEMPERA),
 4 and 5D (LWP_K2W). Values are given for zenith observations.

5A - HATPRO						
Chan no. (#)	Central frequency (GHz)	Fast parameterization uncertainty	Absorption model uncertainty		Total uncertainty	
			Tropical	Subarctic winter	Tropical	Subarctic winter
1	22.240	0.037	0.665	0.290	0.666	0.292
2	23.040	0.030	0.621	0.296	0.621	0.297
3	23.840	0.026	0.542	0.303	0.543	0.304
4	25.440	0.028	0.480	0.322	0.481	0.323
5	26.240	0.027	0.480	0.332	0.481	0.333
6	27.840	0.026	0.506	0.356	0.506	0.357
7	31.400	0.030	0.609	0.420	0.610	0.421
8	51.260	0.148	2.623	3.119	2.628	3.123
9	52.280	0.167	2.727	3.301	2.732	3.305
10	53.860	0.094	1.003	1.132	1.007	1.136
11	54.940	0.024	0.126	0.089	0.128	0.093
12	56.660	0.011	0.023	0.001	0.026	0.011
13	57.300	0.009	0.019	0.003	0.021	0.009
14	58.000	0.008	0.018	0.003	0.020	0.009

5

5B – MP3000						
Chan no. (#)	Central frequency (GHz)	Fast parameterization uncertainty	Absorption model uncertainty		Total uncertainty	
			Tropical	Subarctic winter	Tropical	Subarctic winter
1	22.234	0.037	0.665	0.290	0.666	0.292
2	22.500	0.036	0.663	0.292	0.664	0.294
3	23.034	0.030	0.621	0.296	0.622	0.297
4	23.834	0.026	0.543	0.303	0.543	0.304
5	25.000	0.028	0.487	0.316	0.487	0.317
6	26.234	0.027	0.480	0.332	0.481	0.333
7	28.000	0.026	0.509	0.358	0.510	0.359
8	30.000	0.028	0.564	0.393	0.565	0.394
9	51.248	0.148	2.619	3.114	2.624	3.117
10	51.760	0.157	2.744	3.299	2.749	3.302
11	52.280	0.166	2.727	3.301	2.732	3.305
12	52.804	0.165	2.434	2.943	2.440	2.948
13	53.336	0.141	1.793	2.129	1.798	2.134
14	53.848	0.094	1.020	1.153	1.024	1.156
15	54.400	0.046	0.390	0.388	0.393	0.391
16	54.940	0.024	0.126	0.089	0.128	0.093
17	55.500	0.016	0.052	0.018	0.054	0.024
18	56.020	0.013	0.033	0.004	0.035	0.014
19	56.660	0.011	0.023	0.001	0.026	0.011



1

20	57.288	0.009	0.019	0.003	0.021	0.009
21	57.964	0.008	0.018	0.003	0.020	0.009
22	58.800	0.007	0.018	0.004	0.019	0.008

2

5C - TEMPERA						
Chan no. (#)	Central frequency (GHz)	Fast parameterization uncertainty	Absorption model uncertainty		Total uncertainty	
			Tropical	Subarctic winter	Tropical	Subarctic winter
1	51.250	0.148	2.620	3.115	2.624	3.118
2	51.750	0.157	2.743	3.296	2.747	3.300
3	52.250	0.166	2.733	3.307	2.738	3.311
4	52.850	0.164	2.393	2.892	2.398	2.896
5	53.350	0.141	1.773	2.104	1.778	2.109
6	53.850	0.094	1.017	1.149	1.021	1.153
7	54.400	0.046	0.390	0.388	0.393	0.391
8	54.900	0.024	0.136	0.100	0.138	0.103
9	55.400	0.017	0.059	0.023	0.061	0.029
10	56.000	0.013	0.033	0.004	0.036	0.014
11	56.500	0.011	0.025	0.000	0.027	0.011
12	57.000	0.010	0.021	0.002	0.023	0.010

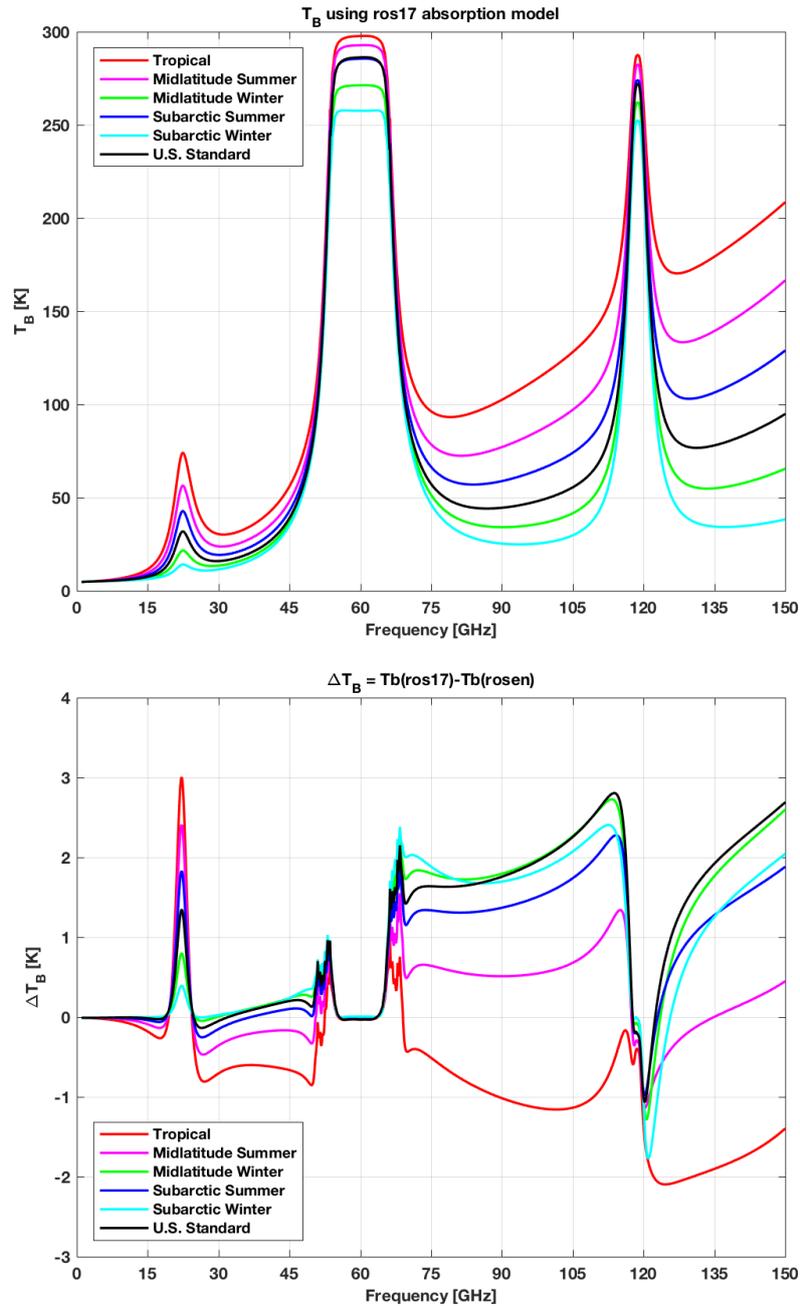
3

4

5

6

5D - LWP_K2W						
Chan no. (#)	Central frequency (GHz)	Fast parameterization uncertainty	Absorption model uncertainty		Total uncertainty	
			Tropical	Subarctic winter	Tropical	Subarctic winter
1	23.840	0.026	0.542	0.303	0.543	0.304
2	31.400	0.030	0.609	0.420	0.610	0.421
3	72.500	0.139	2.775	3.690	2.778	3.692
4	82.500	0.119	2.706	2.042	2.708	2.045
5	90.000	0.126	2.963	1.665	2.966	1.669
6	150.000	0.161	3.547	2.118	3.550	2.124

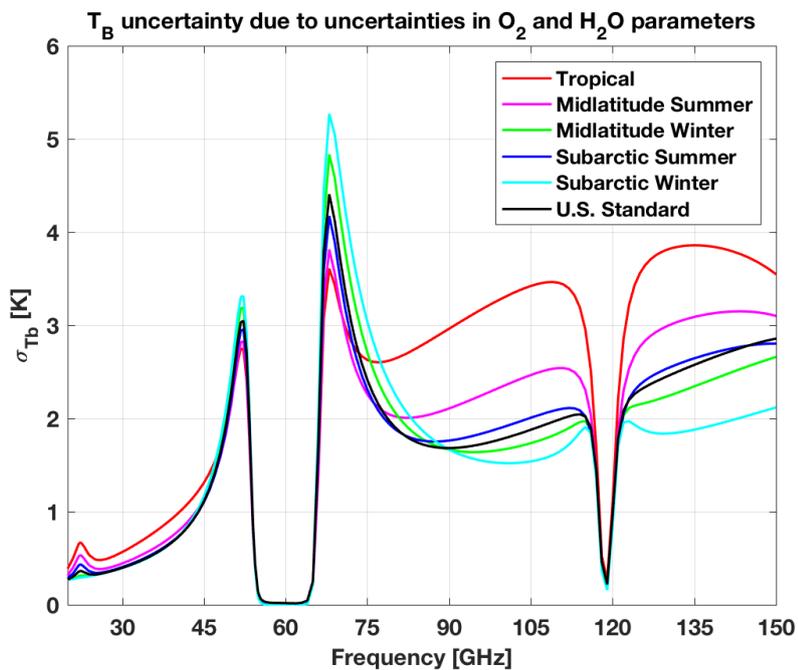


1

2

3 Figure 1: (Top) Zenith downwelling T_B computed using six reference atmosphere climatology conditions with the R17
4 model. (Bottom) Difference between T_B computed with the current and reference versions (R17 minus R98) for the six
5 atmosphere climatology conditions. This figure is similar to Figure 1 in Cimini et al. (2018), although T_B were
6 recomputed to cover a wider frequency range.

7



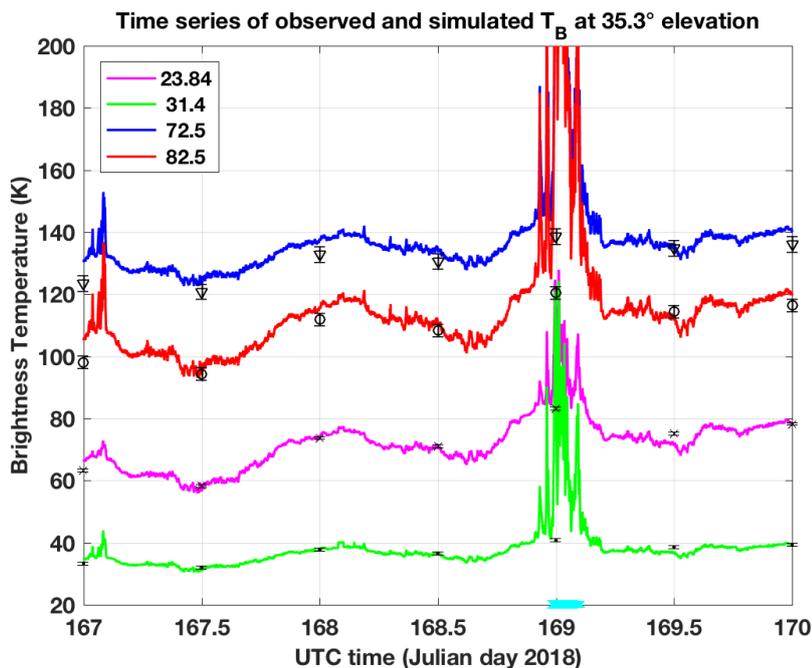
1

2 Figure 2: Zenith downwelling T_B uncertainty ($\sigma(T_B)$) due to the uncertainty in O_2 and H_2O absorption model
 3 parameters. Six climatological atmospheric conditions (color-coded) have been used to compute K_p . $\sigma(T_B)$ is
 4 computed as the square root of the diagonal terms of $Cov(T_B)$. This figure is similar to Figure 6 in Cimini et al. (2018),
 5 although $\sigma(T_B)$ was recomputed to cover a wider frequency range.

6

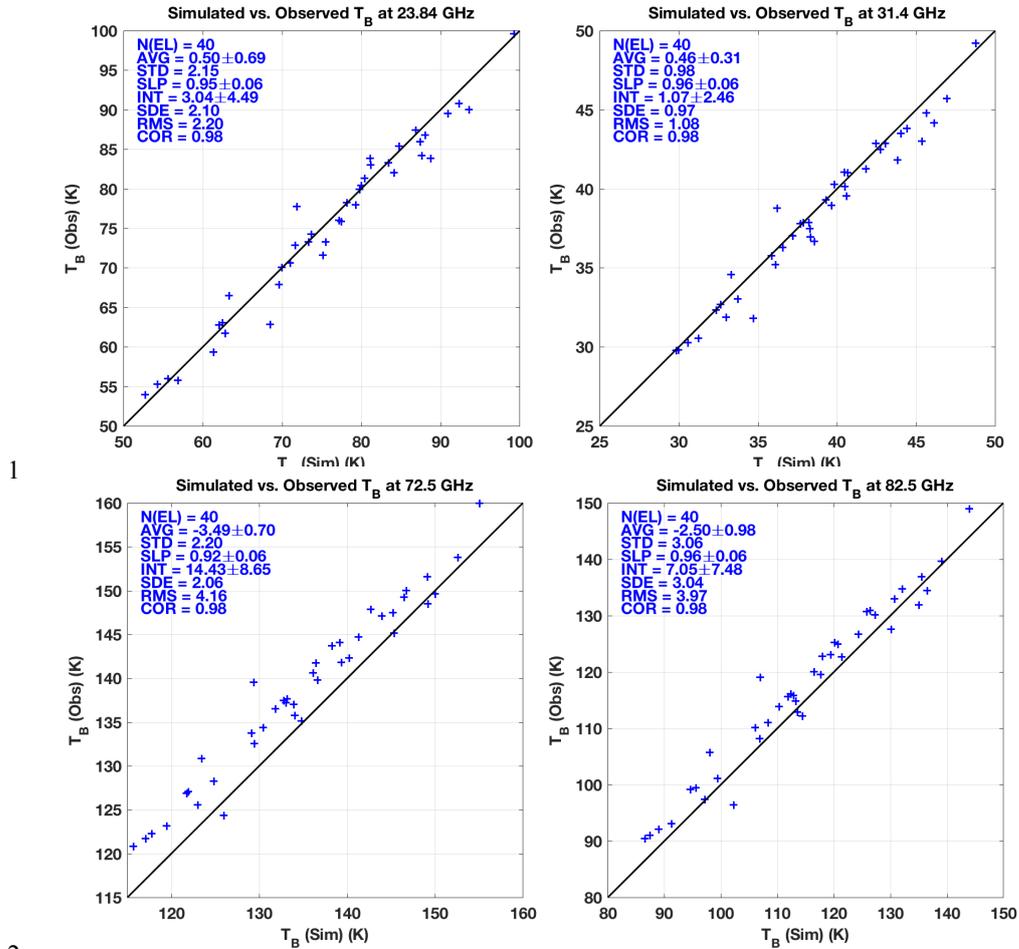
7

8

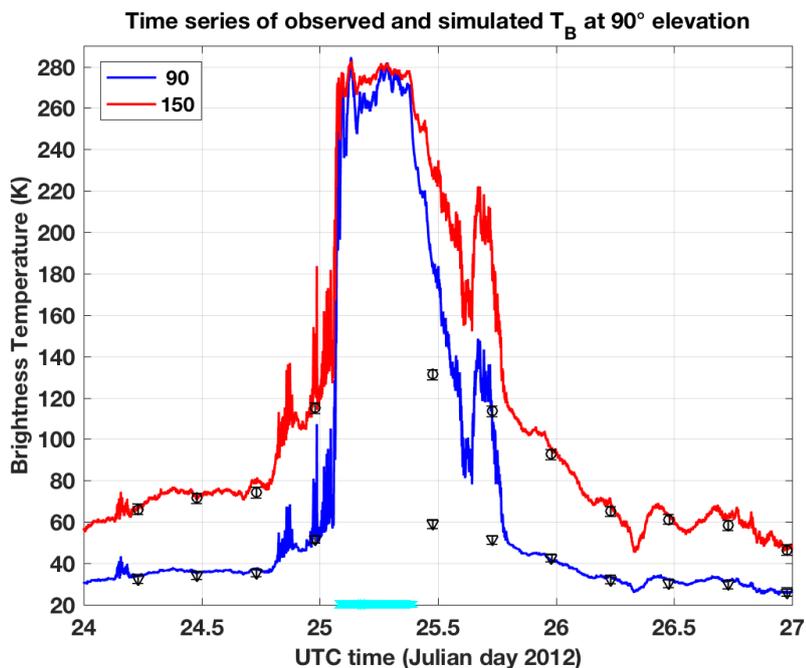


1
2
3
4
5
6
7
8
9

Figure 3: Time series of observed (lines) and simulated (markers) T_B at 35.3° elevation for four channels of LWP-U72-82. The radiometer is located at the Polytechnic University campus in Milan (Italy), while radiosondes are launched from the Milan Linate airport (~20 km from the Politechnic University campus). Channel frequencies are color-coded as reported in the legend. Simulations are reported with dots (23.84 GHz), crosses (31.4 GHz), triangles (72.5 GHz), and circles (82.5 GHz), including an indicative estimate of the total uncertainty. Presence of rain on the instrument antenna dome is indicated on the bottom by cyan crosses. The time series spans from 00:00 of 16 June (Julian day 167) to 00:00 of 19 June (Julian day 170) 2018.



2
 3 Figure 4: Scatter of simulated vs. observed T_B at 35.3° elevation for four channels of LWP-U72-82. Location of
 4 radiometer and radiosondes are as in Figure 3. The absorption model of Rosenkranz 2017 has been used. Each panel
 5 reports the number of elements (N(EL)), the average difference (AVG), the standard deviation (STD), the slope (SLP)
 6 and intercept (INT) of a linear fit, the standard error (SDE), the root-mean-square (RMS), and correlation coefficient
 7 (COR). 95% confidence intervals are given for AVG, SLP, and INT. Units for AVG, STD, SDE, RMS are Kelvin.
 8



1

2 **Figure 5: Time series of observed (lines) and simulated (markers) T_B at 90° elevation for two channels of LWP-90-150.**

3 **The radiometer and radiosondes are operated from the Atmospheric Radiation Measurement (ARM) program**

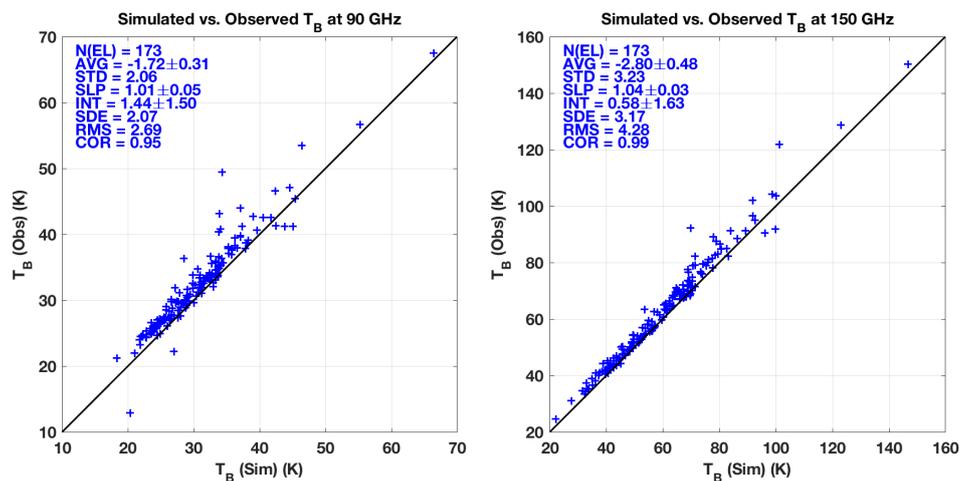
4 **Southern Great Plains (SGP) central facility (Lamont, OK, USA). Channel frequencies are color-coded as reported in**

5 **the legend. Simulations are reported with triangles (90 GHz) and circles (150 GHz), including an indicative estimate**

6 **of the total uncertainty. Presence of rain on the instrument antenna dome is indicated on the bottom by cyan crosses.**

7 **The time series spans from Jan 24 00:00 to Jan 27 00:00 UTC 2012 (Julian day 24-27).**

8



1

2

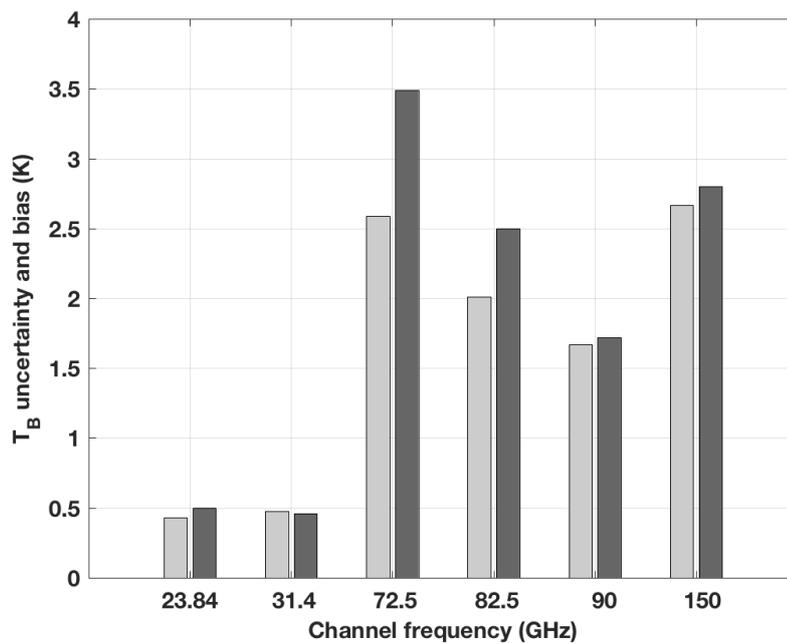
Figure 6: Same as Figure 4 but showing simulated vs. observed T_B at 90° elevation for two channels of LWP-90-150.

3

Location of radiometer and radiosondes are as in Figure 5. The absorption model of Rosenkranz 2017 has been used.

4

5



1

2 **Figure 7: Estimated uncertainty (light grey) and experimental mean difference (dark grey) for six LWP_K2W**
 3 **channels. Radiometric observations were collected in June-July in Milan (45°N) with the four lower frequency**
 4 **channels, while in January-February in Lamont (36°N) with the two higher frequency channels. Thus, uncertainty is**
 5 **estimated using midlatitude summer conditions for the four lower frequency channels, while midlatitude winter**
 6 **conditions for the two higher frequency channels.**

7

8

9