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Homogenization of Geostationary Infrared Imager Channels for Cold Cloud Studies Using Megha-Tropiques/ScaRaB

Thomas Fiolleau^{ID}, Rémy Roca, Sophie Cloché, Dominique Bouniol, and Patrick Raberanto

Abstract—Infrared (IR) observations from the fleet of multi-agencies meteorological geostationary satellites have a great potential to support scientific and operational investigations at a quasi-global scale. In particular, such a data record, defined as the GEOring data set, is well suited to document the tropical convective systems life cycles by applying cloud tracking algorithms. Yet, this GEOring data set is far from being homogeneous, preventing the realization of its potential. A number of sources of inhomogeneities are identified ranging from spatiotemporal resolutions to spectral characteristics of the IR channels and calibration methodologies. While previous efforts have attempted to correct such issues, the adjustment of the cold part of the IR spectrum remains unfit for cold cloud studies. Here, a processing method is introduced to minimize the inhomogeneities against a reference observational data set from the Scanner for Radiation Budget (ScaRaB) instrument onboard the Megha-Tropiques satellite. The method relies on the collocations between the geostationary observations and the reference. The techniques exhibit significant sensitivity to the selection of the relevant pairs of observations requiring a dedicated filtering of the data. A second effort is then proposed to account for the limb-darkening effect and a method is developed to correct the brightness temperature (BT) dependence on the geostationary viewing zenith angle (VZA). Overall, results show a residual after the processing of 0 K between any of the geostationary data and the ScaRaB reference. The final calibrated and limb-adjusted IR observations are then homogeneous for cold BT lower than 240 K with a standard deviation lower than 1.5 K throughout the GEOring.

Index Terms— Calibration and spectral corrections, cold cloud studies, geostationary satellites, infrared (IR) image sensors, limb-darkening corrections.

I. INTRODUCTION

METEOROLOGY agencies monitor individually operational geostationary platforms, which when used all together permit an observation for all longitudes of the Earth

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on a large latitudinal band ($\sim 60^\circ$ S– 60° N) with similar instruments. The potential benefit of such an observing capability has triggered numerous efforts in the past 40 years to make use of it for various applications, ranging from the elaboration of a unique perspective on clouds studies [1], [2] for precipitation estimation [3], climate studies [4], and global visualization applications [5]. Yet, when carefully examined, this GEOring appears to be far from a homogenous suite of instruments operated in a similar fashion. The space/time resolution and sampling differ across the platforms. The acquisition procedure also differs among the satellites (from South to North, North to South, or by sector). The calibration procedure of each instrument is also performed at the individual level with the instruments' specific modes of operation. Since 2011, for the infrared (IR) channels, the agencies provide alternative calibration coefficients obtained from a statistical comparison to hyperspectral IR measurements from low earth-orbiting satellites using the common methodology of the Global Space-Based InterCalibration System (GSICS) [6]. Nevertheless, the GSICS currently only provides calibration coefficients referenced to the NASA Earth Observing System (EOS), Aqua Atmospheric IR Sounder (AIRS), and the Exploitation of Meteorological Satellites (EUMETSAT)–Centre National d'Etudes Spatiales (CNES), Meteorological Operation (MetOp), and IR Atmospheric Sounding Interferometer (IASI) calibration standard. This harmonization effort does not take into account spectral band responses and viewing angle effects, which are needed for sensor homogenization. Indeed, while often referred to as a unique IR channel, the multiple instruments do show some spectral differences as well.

Most of the spectral imagers onboard geostationary satellites monitor with a relatively high temporal frequency the calibration coefficients for thermal IR channels by using onboard blackbody references. Some of the differences in the temperature measurements from one geostationary satellite to another can be explained by the differences in the calibration references, methods, as well as in the spectral filter functions. Moreover, it has been observed that the spectral imagers may degrade over time with different rates [7], [8]. The geostationary imagers can also be affected by short-term variations. The accumulation of ice on the surface of the imager optics modifies their spectral response function (SRF) and consequently affects the measurement of the temperatures. To face these issues, decontamination events are regularly applied to the radiometers inducing

TABLE I
SUMMARY OF THE TECHNICAL CHARACTERISTICS OF THE VARIOUS EXISTING IR GEORING DATA SETS

Products	Spatial resolution	Temporal resolution	Spatial coverage	Temporal coverage	Calibration and temporal normalization	Limb darkening correction	Data source
Gridsat-B1	0.07°	180 min	70°S-70°N	1980-present	HIRS channel 8 [14]	Yes	ISCCP-B1
CPC	~ 0.04°	30 min	60°S-60°N	2000-present	No absolute calibration and no temporal normalization	Yes	Native
CLAUS	0.5°	180 min	Global*	1983-2006	ISCCP- B3	Yes	ISCCP-B3
CLAUS	0.3°	180 min	Global*	1985-2008	ISCCP- B3	Yes	ISCCP-B3
ISCCP-B1	Native FoV subsampled to ~ 0.1°	180 min	70°S-70°N	1980-present	AVHRR channel 5 [13]	No	Native
ISCCP-B3	Native FoV subsampled to ~ 0.3°	180 min	70°S-70°N	1983-2009	AVHRR channel 5 [13]	No	Native
This study	0.04°	30 min	35°S-35°N	2012-2016	SCARAB-IR channel 4	yes	Native

instability in the temperature measurements [9]. This effect has been particularly shown for the Meteosat Visible IR Imager (MVISIR) onboard the Meteosat-7 geostationary platform.

A diurnal cycle effect can also affect the imager calibration, especially for the instruments on the three-axis stabilized geostationary platforms [7], [9]. It has been shown that the sun-synchronous orbits of MetOp/IASI and EOS/AIRS used in the GSICS calibration are not sufficient to completely correct the diurnal IR calibration issues [10]. The precessing orbit of the tropical rainfall measuring mission (TRMM) platform and the observations of its visible and IR scanner (VIRS) have then been used to quantify and correct such a midnight IR calibration anomaly [10]. IR observations are also dependent on the geostationary viewing zenith angle (VZA). The issue is called limb darkening, corresponding to a decrease in the temperature as the VZA increases.

These multiple sources of inhomogeneity have prevented the full use of this unique observational capability, although a number of successful projects have attempted to overcome the above-listed limitations. Indeed, a number of significant efforts have been proposed in the last two decades to produce more or less homogeneous “global” archive of geostationary data sets for various applications. It ranges from cloud climatology computations (International Satellite Cloud Climatology Project (ISCCP)-B1, ISCCP-B3, and Cloud Archive User Service (CLAUS) [11]), hurricane trend analysis, support of precipitation estimation (Gridded Satellite (GridSat) [12]), real-time monitoring of precipitation (Climate Prediction Center (CPC) [13]), for the enhancement of radiation budget averages [8]. Additional information on these GEORing data sets is listed in [4]. Table I summarizes the available IR GEORing data sets altogether with their respective properties

and homogeneity levels. The ISCCP-B1 and B3 have been subsampled to a 180-min temporal resolution and to, respectively, ~0.1° and ~0.3° spatial resolution. They have been calibrated and spectrally normalized using advanced very high-resolution radiometer (AVHRR) observations [14]. The CLAUS data sets rely on the ISCCP-B3 and have been reprojected to an equal angle map projection with a spatial resolution of 0.3° and 0.5°. The GridSat-B1 is based on the ISCCP-B1 data set remapped to a 0.07° regular grid. In addition to the intercalibration used for the ISCCP data set, a second calibration using the high-resolution IR radiation sounder (HIRS) observations has been applied to correct a bias at cold temperatures observed after 2001. The CPC product is built in real time from native geostationary observations. The data are remapped to a regular grid of ~0.04° spatial resolution and are available every 30 min. To reduce the spectral differences and calibration issues, a complex, multistep, multiregional procedure is applied by comparing the brightness temperatures (BTs) from pairs of geostationary platforms. However, this procedure does not provide any absolute calibration. Also note that a limb correction has been performed on the GridSat-B1, CPC, and CLAUS data. In short, while a number of efforts have paved the way for a quantitative use of the GEORing data sets, no dedicated effort toward the cold BT regimes has been promoted so far. Yet, cold cloud studies benefit a lot from geostationary IR observations, for cloud microphysical or macrophysical parameters retrievals. The spatiotemporal resolution of the measurements further allows the life cycle analysis of the cold cloudiness and the estimation of related parameters. The present GEORing attempt is directed toward these scientific applications. In particular, it will serve as the input of the realization of a tropical mesoscale convective system data set using the TOOCAN algorithm [15].

TABLE II
 TECHNICAL CHARACTERISTICS OF THE OPERATIONAL GEOSTATIONARY SATELLITES FLEET AND THE ASSOCIATED
 IMAGERS USED OVER THE 2012–2016 PERIOD

Platform	Nadir location	Instrument	Central wavelength	Spectral interval	Spatial resolution at nadir	Temporal resolution	Region of interest	Source	Period
GOES-15	135° W	IMAGER	10.7 μm	10.2-11.2 μm	4 km	30 min	180° W-105° W 35° S-35° N	NOAA / DWD	Jan 2012- Dec 2016
GOES-13	75° W	IMAGER	10.7 μm	10.2-11.2 μm	4 km	30 min	111° W-30° W 35° S-35° N	NOAA / DWD	Jan 2012- Dec 2016
METEOSAT-8/9/10	0°	SEVIRI	10.8 μm	9.8-11.8 μm	3 km	15 min	45° W-45° E 35° S-35° N	EUMETSAT/ AERIS	Jan 2012- Dec 2016
METEOSAT-7 (IODC)	57.5° E	MVIRI	11.5 μm	10.5-12.5 μm	5 km	30 min	12° E-107° E 35° S-35° N	EUMETSAT/ AERIS	Jan 2012- Dec 2016
MTSAT-2	145° E	IMAGER	10.8 μm	10.3-11.3 μm	4 km	30 min	94° E-170° W 35° S-35° N	AERIS/ CIMSS	Jan 2012- May 2015
HIMAWARI-8	140.7° E	AHI	11.2 μm	11.0-11.4 μm	2 km	10 min	94° E-170° W 35° S-35° N	AERIS/ JMA	Jun 2015- Dec 2016

148 Such analyses require observations with a minimal temporal
 149 resolution of 30 min [15], [16] to ensure tracked objects
 150 overlap between two images. Such kind of studies requires the
 151 use of thresholds applied to BT, ranging from 200 to 240 K,
 152 depending upon the analysis [17], [15]. A homogeneous data
 153 set is then mandatory, if one wants to proceed with the whole
 154 tropical belt. The homogeneity and basic requirements for a
 155 GEOring IR data set geared toward tropical cold cloud tracking
 156 applications can be summarized as follows:

- 157 1) High-resolution spatial footprint and high spatial reso-
 158 lution (~ 5 km).
- 159 2) A minimum of 30 min time sampling.
- 160 3) A spatial coverage dedicated to the tropics.
- 161 4) Homogenized BTs through spectral and calibration
 162 adjustment.
- 163 5) Limb adjustment of IR observations.

164 Matching these requirements with the specifications of the
 165 IR data sets summarized in Table I confirms the need for
 166 a dedicated level 1c GEOring IR tailored for cold cloud
 167 tracking applications. Moreover, the current implementation
 168 of the GSICS coefficients computation is directed toward the
 169 warm part of the spectrum [8], ruling out its use for cold
 170 cloud studies calling for an alternative, cold cloud compliant,
 171 absolute calibration reference for the GEOring. The limb
 172 effect is less strong over the cold part of the temperature
 173 spectrum than at the warm and clear sky end of the spectrum,
 174 and the limb correction is easier to set up for cold cloudy
 175 scenes (BTs < 240 K) than for the clear atmosphere. This

effect is strongly nonlinear and is thought to be impactful
 only for the viewing angle above 26.5° [11]; it may still
 require a correction for a homogeneous interpretation of the
 BT field all through the image. It is difficult to establish
 requirements once and for all in terms of residuals difference
 in the homogeneous data, based on the radiometric noise of the
 first generation sensors and the various sources of uncertainties
 in the adjustment procedure and basic cloud-oriented retrievals
 sensitivity. We propose to target a less than 1.5 K standard
 deviation among any of the two geostationary satellites in the
 final calibrated and limb-adjusted product for BTs < 240 K.

II. DATA

A. Five-Year Database of IR Geostationary Observations Over the Entire Tropics

Thermal channel BT images obtained by the operational
 meteorological geostationary satellite fleet are used over
 the 35°S–35°N latitude belt and for the whole 2012–2016
 period. Table II shows the geostationary IR data used to
 cover the entire tropics for this period. The geostationary
 data set has been collected from different sources: the U.S.
 National Oceanic and Atmospheric Administration (NOAA),
 the European Organization for the Exploitation of Meteorological Satellites (EUMETSAT), the Japan Meteorological Agency (JMA), and the French Atmosphere and Service Data Pole (AERIS). Note that the Multifunctional Transport Satellite (MTSAT) data set has also been completed from the Space Science and Engineering Center (SSEC) of the University of

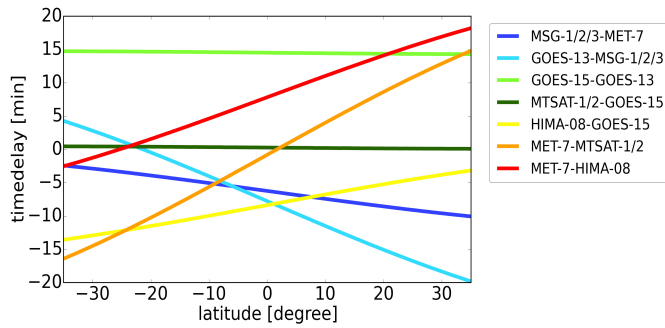


Fig. 1. Time delay of acquisition between pairs of geostationary images according to the latitude.

Wisconsin–Madison. The technical characteristics of each geostationary platform and their imagers are described in Table II and can differ from one platform to another. First, considering the temporal resolution, Meteosat Second Generation (MSG) and HIMAWARI-8 observe the Earth, respectively, with a 15- and 10-min time resolution, while the other platforms display a 30-min temporal resolution. Similarly, the scan-times vary from 7 min for HIMAWARI-8 to 25 min for Meteosat-7. The Geostationary Operational Environmental Satellite (GOES)-13 and GOES-15 sensors follow a complex scanning sector schedule. The full disk images are produced every 3 h, while the northern hemisphere and the southern hemisphere images are produced every 30 min with a time lag of few minutes between each scan. Also, note that the scanning schedule of MTSAT-2 does not provide half-hourly sampling of the southern hemisphere region. Contrary to the other imagers, the GOES-13 imager starts its earth scan at 15 and 45 min of the hour. Also note that Meteosat first- and second-generation imagers start their earth scan from the southeast corner, whereas NOAA and JMA platforms begin their earth scan from the northwest corner. All these temporal scanning differences imply some time delays between neighboring platforms observing the same region (Fig. 1). While the 15-min time delay is constant for the pairs of the GOES-15/GOES-13 and MTSAT-2/GOES-15 platforms, the time delay varies according to the latitude for the other pairs of geostationary platforms. At 35° S, a maximum time delay is found for the pair of MSG/GOES-13 platform of around 20 min. The maximum time delay amplitude accounted for the pair of the Meteosat-7/MTSAT-2 platforms and varies between -16 and 15 min.

Depending on the considered satellite, spatial resolution at nadir differs from one geostationary satellite to another and ranges from 2 km for HIMAWARI-8 to 5 km for Meteosat-7. In addition to the resolutions disparities, the GEOring is also characterized by differences in the central wavelength and the associated SRFs of each of the instruments (Fig. 2). While the central wavelength ranges between 10.7 and 11.5 μm for the GOES imagers and the Meteosat-7 imager, respectively, two types of SRFs can define the IR channels. The GOES-13, GOES-15, MTSAT-2, and HIMAWARI-8 imagers display narrowband channels with a bandwidth lower than 1 μm , whereas the Meteosat-7/MVIRI and MSG/SEVIRI imagers exhibit

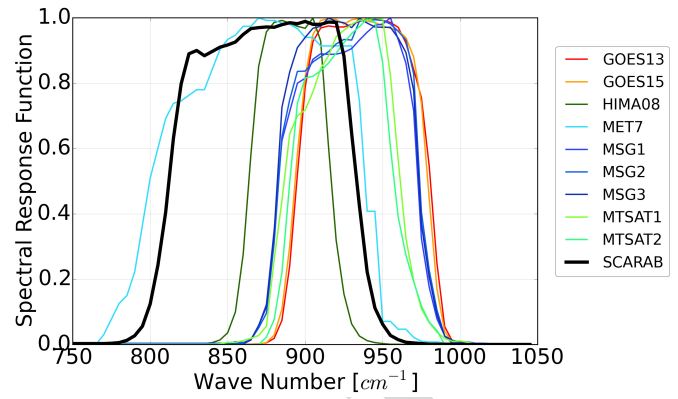


Fig. 2. SRF of the IR channel of the GEOring. The black line is the spectral response of the channel 4 of the ScaRaB instrument.

broader channels characterized by a 2 μm bandwidth. A broadband channel is highly sensitive to ice contamination on the optics of the imagers, modifying their SRFs, and consequently, introducing a calibration error. Operational decontamination procedures are applied regularly to remove the ice build-up on the optics. Thus, spectral differences, as well as the individual satellite calibration procedures, contribute to the GEOring radiometric inhomogeneities.

B. Megha-Tropiques/ScaRaB-3 Observations

The Scanner for Radiation Budget (ScaRaB) instrument on the Megha-Tropiques platform [18] is the third of its kind [19], [20] and has been designed to measure the Earth radiation components at the top of the atmosphere with high accuracy (<1%). The instrument acquires data across the satellite track with a swath of ~ 2200 km from 30° S to 30° N. It is a broadband radiometer with four channels. The total channel measures the total energy between 0.2 and 100 μm . The shortwave channel (0.2–4 μm) is subtracted from the total channel to obtain the longwave part of the spectrum [21]. The fourth channel is an IR thermal channel (10.5–12.5 μm) which is used in this article (Fig. 2). The instrument operates nominally since the beginning of the mission and its operational performances are well in line with the specifications [22]–[24]. The nominal resolution of the ScaRaB footprint at nadir is 40 km and is detailed in [18].

The stability of the instrument is monitored by the CNES on a daily basis and so far, the Megha-Tropiques instruments have shown remarkable stability [25]. The use of geophysical targets like deep convective clouds (DCC) has been shown to be useful for geophysical calibration and cross-calibration of either broad radiometers [26], [27]. It is used here to showcase the relative stability of the channel 4 to that of the longwave channel. Fig. 3 shows the ratio between the radiance in channel 4 and channel 1 over the studied period for various deep convective regimes. The two channels of the instrument show no sign of relative degradation and a very steady behavior. Comparisons with the Clouds and the Earth's Radiant Energy System (CERES) instrument using dedicated collocation campaigns [28] further indicate a very good agreement between the two instruments' longwave channels [29].

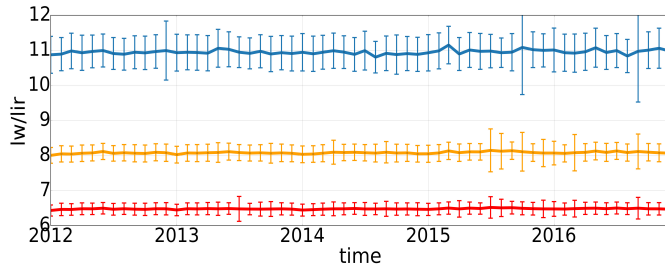


Fig. 3. Evolution of the ratio between measured radiance in the longwave band and in the IR channel of ScaRaB over the 2012–2016 period for various deep convective regimes. The blue curve corresponds to an IR radiance lower than $2.25 \text{ W} \cdot \text{m}^{-2} \cdot \text{sr}^{-1}$ corresponding to a 187 K BT, the yellow one lower than $4.25 \text{ W} \cdot \text{m}^{-2} \cdot \text{sr}^{-1}$ (207.4 K), and the red curve corresponds to an IR radiance lower than $7.75 \text{ W} \cdot \text{m}^{-2} \cdot \text{sr}^{-1}$ (224.5 K).

286 The ScaRaB instrument is also thermally controlled and is not
 287 impacted by icing events as shown in the monitoring of the
 288 instrument performances [22], [23]. The combination of an
 289 accurate and stable instrument makes it a well-suited reference
 290 to adjust the geostationary data.

291 Here we use the so-called level 2B products that consist
 292 of a 0.5° regularly gridded instantaneous directional radiances
 293 ($\text{W} \cdot \text{m}^{-2} \cdot \text{sr}^{-1}$). The original measurements are averaged on the
 294 regular grid using a statistical technique and the point spread
 295 function of the instrument [30]. Since 2011, the availability
 296 of this product is about 99.9%, making it a useful resource as
 297 a reference instrument for the spectral homogenization of the
 298 GEORing data. Note that unlike sun-synchronous platforms,
 299 the precessing orbit of Megha-Tropiques samples all the local
 300 times every 51 days allowing the collocation with the geo-
 301 stationary data all through the day and then the correction of
 302 the midnight IR anomaly [10], [18], [31]. The low inclination
 303 at the equator (20°) and high altitude of flight (865 km) also
 304 allow high repetitive measurements in the tropics [18].

305 III. HOMOGENOUS LEVEL 1C IR DATA SET 306 FOR THE TROPICS FOR 2012–2016

307 A. Homogenization of the Temporal Resolution

308 Given the requirements for cold cloud studies (see
 309 Section I), the lack of temporal resolution homogeneity
 310 between the geostationary imagers is accounted for by using
 311 a 30-min temporal frequency for all the platforms, preventing
 312 the use of MTSAT data in the southern hemisphere.

313 B. Homogenization of the Spatial Resolution

314 A common equal angle grid of $0.04^\circ \times 0.04^\circ$ has been
 315 selected for all the platforms to account for this source of
 316 inhomogeneity. This resolution is very close to the native
 317 resolution of GOES-13, GOES-15, MTSAT-2, and Meteosat-7
 318 data. The regridding process is performed by applying the
 319 inverse distance weighting method in the radiance space with
 320 a maximum search radius corresponding to the sum of half
 321 of the geostationary spatial resolution for a given pixel plus
 322 a half of the equal angle grid resolution. Then, the average
 323 radiance is transformed in BT using the Planck function in
 324 order to account for its nonlinearity.

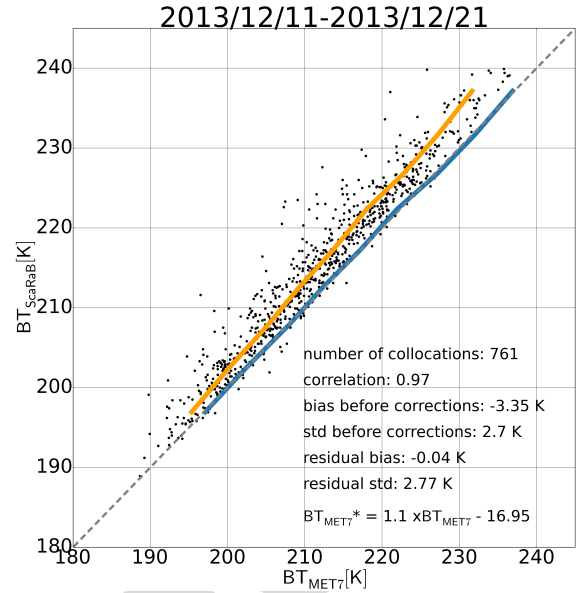


Fig. 4. Scatter plot between BT_{MET7} and $\text{BT}_{\text{ScaRaB}}$ obtained after the filtering procedure over a ten-day period starting in 2013/12/11 for BT in the range [180–240 K]. The orange line corresponds to the average BT for each 5 K bin. The blue line corresponds to the residual bias.

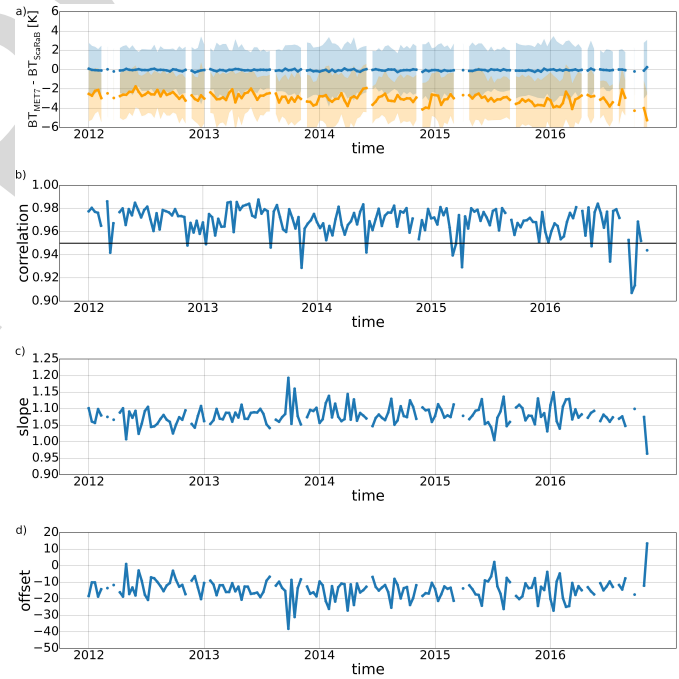


Fig. 5. (a) Time series of decadal initial BT bias (orange) and corrected BT bias (blue) for Meteosat-7 IR observations with respect to ScaRaB in the range [180–240 K]. The filled and transparent areas in orange and blue represent the standard deviations: (b) Time series of decadal linear regression correlations, (c) slope, and (d) offset between the ScaRaB and the Meteosat-7 IR observations in the range [180–240 K].

325 C. Intercalibration and Spectral Normalization

326 1) *Regression Technique*: The use of a common low
 327 earth-orbiting satellite, carrying an IR radiometer to anchor
 328 each of the platforms to a common reference data set, forms
 329 the basis of the different intercalibration and normalization
 330



Fig. 6. (a) Time series of initial BT bias of geostationary IR observations with respect to ScaRaB in the range [180–240 K]. (b) Time series of BT bias between the geostationary IR observations and the ScaRaB observations in the range [180–240 K] after spectral and calibration corrections. (c) Time series of correlations between the ScaRaB and the geostationary IR observations in the range [180–240 K]. Each dot corresponds to a ten-day period.

efforts listed in Table I [14], [32] or under the GSICS calibration procedure [7]. Regression-based techniques are the most commonly used approaches. The difficulty arises in building the pairs (space/time/angular collocation) as well as in selecting the pairs of observations that will be regressed. This selection can have a strong impact on the result. Hence, it has been noted that the original ISCCP intercalibration was skewed toward a warm scene due to the implementation of the pairs selection filter prior to the regression computation, yielding a ~ 4 K cold bias on the cold part of the spectrum [33]. Similarly, most of the tropical observations are located over large BT values and only a few of them are explained by DCCs. As the larger population of collocations is explained by high BT, the GSICS corrections are optimized for clear sky conditions and are not well suited for the cold cloud scenes [7]. A specific homogenization of the IR geostationary database is then required to fit with the objective of cold cloud tracking applications. This homogenization consists of a spectral normalization as well as an intercalibration of the various geostationary imagers focused on high cold clouds and on cold BTs. For this purpose, the BTs from the geostationary imagers are calibrated and spectrally normalized against the ScaRaB channel 4 measurements. The final data set then consists of a 10.5–12.5 μm IR equivalent BT. Calibration uncertainty is expected to be within 1 K, which is the typical error for operational satellite calibration [34]. Contrary to other geostationary imagers, the correction of the Meteosat-7/MVIRI imager is restricted to an intercalibration procedure,

since its broad SRF is similar to the ScaRaB/channel 4 and both of them are centered on 11.5 μm .

2) *Selection of the Data Match-Ups for the Regression:* The IR geostationary data from all the platforms are first remapped from their native formats to a regular lon/lat 0.5° grid every 30 min to allow direct comparison with the ScaRaB/L2B data. Prior to computing the averaged radiance and the associated spatial standard deviation for each 0.5° grid point, the BTs of each geostationary platform are converted in radiances ($\text{mW} \cdot \text{m}^{-2} \cdot \text{sr}^{-1} \cdot \text{cm}^{-1}$) by applying the Planck function. The information on geostationary scan-time is also indicated for every grid point. The collocation between the ScaRaB-L2B observation and the IR $0.5^\circ \times 0.5^\circ$ gridded geostationary data is reached when both view the same scene with a predefined time delay and similar viewing geometry. The predefined criteria are.

- 1) A maximum VZA of 20° and 26° , respectively, for the ScaRaB and the geostationary observation, to ensure alignment of the collocated pixels in viewing geometry and to avoid limb-darkening issues [35].
- 2) A maximum time delay of 10 min between two collocated pixels, which is a tradeoff between a relevant number of match-ups to populate the decadal regressions and the best possible temporal precision.
- 3) The IR-gridded radiances are regressed in units of temperature.

As we focus on high cold clouds scenes, we have developed a filtering procedure that keeps more collocation scenes at a colder temperature. Indeed, a bulk of collocations occurs at

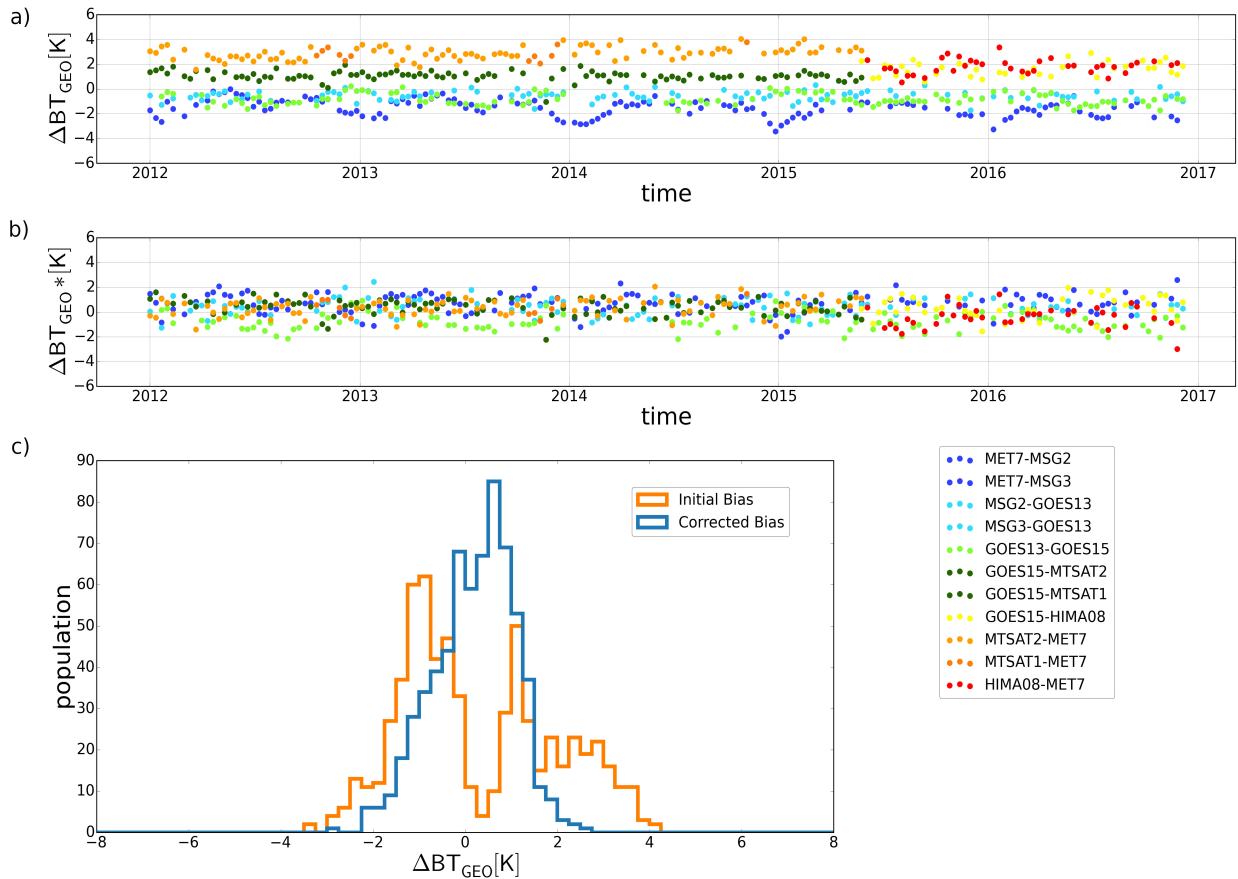


Fig. 7. (a) Time series of initial BT bias between pairs of geostationary platforms observing common areas with an equivalent VZA in the range [180–240 K]. (b) After spectral and calibration corrections; each dot corresponds to a ten-day period. (c) Distribution of the bias before and after corrections over the 2012–2016 period for all the pairs of geostationary platforms for all the ten-day periods.

386 a warmer temperature and displays a lower standard deviation [33]. To separate the high cold cloud population from
 387 the warmer targets in the scatterplot, thresholds are applied
 388 to the GEO standard deviation that depends on the BT.
 389 Pixels presenting a $BT_{GEO} > 240$ K are rejected if their
 390 standard deviations are larger than 0.5 K and pixels with
 391 $BT_{GEO} < 240$ K are kept if their standard deviations are
 392 lower than 2 K. Besides separating the high cold cloud popu-
 393 lation from clear sky pixels, this filtering procedure ensures
 394 that heterogeneous moving objects do not contaminate the
 395 collocation scenes and that the two instruments observe the
 396 same high cloud target. Assuming also that the IR radiometers
 397 onboard geostationary platforms have a linear response when
 398 observing high cold cloud homogeneous scenes [8], [36], the
 399 calibration and spectral normalization corrections are then
 400 based on linear regressions computed over a 180–240 K
 401 range. For this, the collocated BT_{GEO} and BT_{ScaRaB} are binned
 402 and averaged for every BT_{ScaRaB} interval of 5 K from 180 to
 403 240 K. The regression coefficients are then computed over
 404 these binned data every ten days and over a ten-day period.
 405 This is required to ensure statistical robustness but prevent
 406 higher frequency variation (<10 days) of the calibration
 407 issues (due to decontamination for instance) to be accounted
 408 for. As shown below, such higher frequency effects do not
 409 impact the residuals of the regression. Correlations are used to

determine the ten-day period that might have match-ups errors.
 If the correlations are lower than 0.95 or if the population
 of collocations does not exceed ten matchups, we consider
 that the linear regression cannot be computed and that every
 ten-day period is removed from our analysis. In this case,
 the previous regression coefficients are replicated.

3) *Results:* Fig. 4 shows an example of such a regression
 computed from 761 collocated pixels between the Meteosat-7/
 MVIRI imager and the Megha-Tropiques/ScaRaB observations
 acquired over a ten-day period from 2013/2012/2011 and
 for a range between 180 and 240 K. The $11.5 \mu m$ channel
 of Meteosat-7 exhibits a large negative bias (-3.35 K) for
 this considered BT and for this specific period. The blue
 line shows the regression fit given by a slope of 1.11 and
 an offset of -16.95 K. The correction has produced a very
 low residual bias, which averages at -0.04 K over this
 decadal period, and the correlation higher than 0.97 between
 the two data sets shows the high quality of the regression
 computation. Fig. 5(a) shows the time series of difference
 between Meteosat-7/MVIRI and ScaRaB before and after
 calibration over the 2012–2016 period. The Meteosat-7 BT
 is corrected every ten days using the slopes and offset com-
 puted from the linear regression. Over the period and for
 the specific Meteosat-7 geostationary platform, 12% of the
 decadal scatterplots are excluded from our analysis following

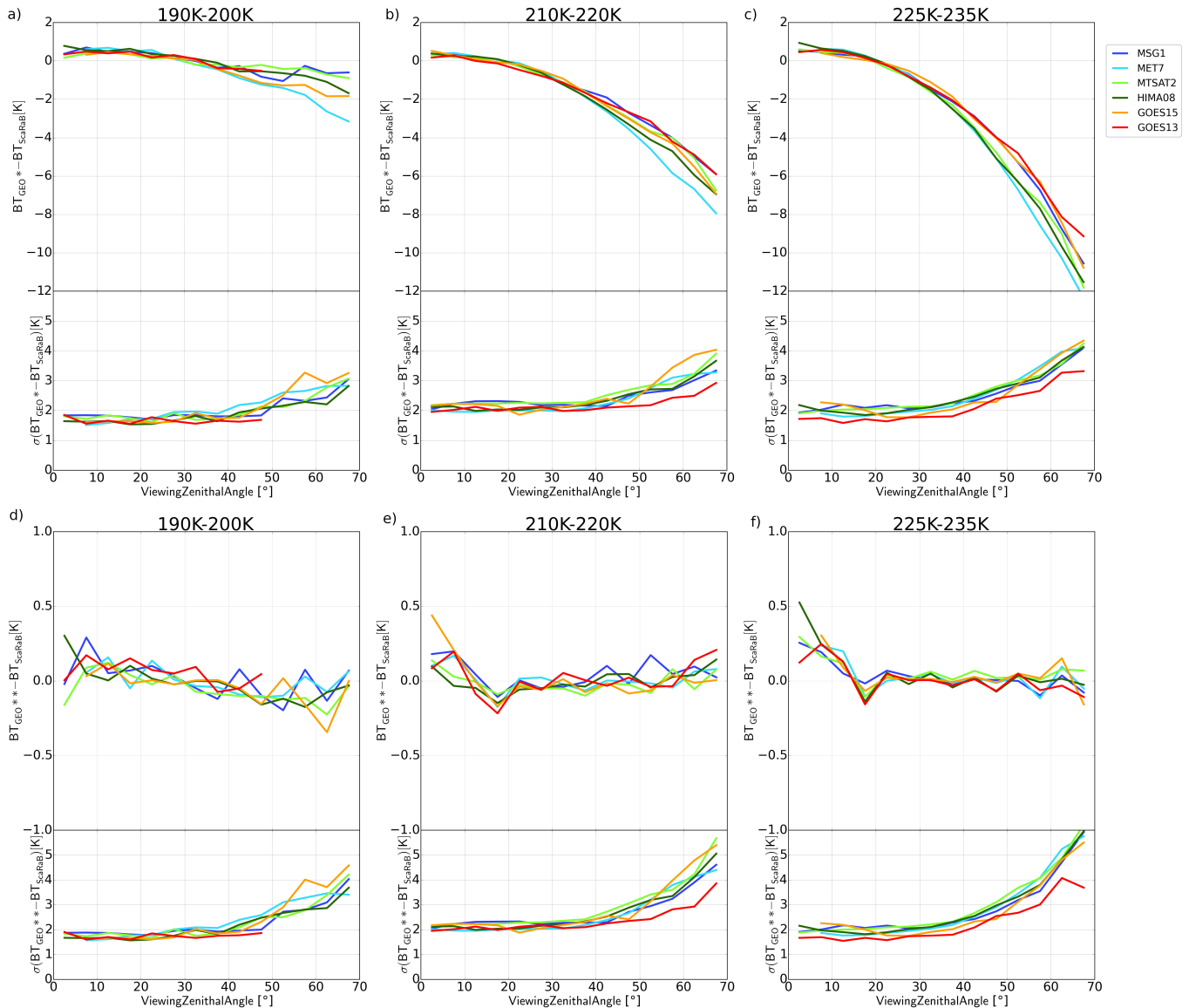


Fig. 8. Variation of the BT bias and the corresponding standard deviation between the GEO and the SCARAB observations according to the geostationary VZA, for a $VZA_{ScaRaB} < 20^\circ$ and for three different ranges of BT_{ScaRaB} . (a)–(c) Before VZA corrections. (d)–(f) After VZA corrections.

the filtering procedure [Fig. 5(b)] and 87 405 collocations have then been used to compute the decadal regressions. One can observe that over the 2012–2016 period, and before the corrections, the Meteosat-7/MVIRI exhibits a negative bias, averaging at -2.98 K and negatively increasing slightly with time due to contamination on the optics. This radiometric issue has been fully documented in [7]. The variations of the slopes and offsets are relatively stable over the period [Fig. 5(c) and (d)], showing the robustness of the methodology. After applying the calibration correction, the residual bias is very smooth, stable, and close to zero over the whole period. The calibration and spectral normalization corrections are applied on all the geostationary platforms available over the 2012–2016 period. The results have been obtained over the entire period, and for all the geostationary platforms by filtering 8% of the decadal scatterplots which did not pass our

quality control [Fig. 6(c)], leading to a comparison of around 655 000 collocations. Fig. 6(a) shows the time series of the decadal initial bias in BT for all the geostationary imagers with respect to the ScaRaB observations in the range 180–240 K. Over the 2012–2016 period, the Meteosat-7/MVIRI exhibits the highest error as discussed previously. It is also to be noted that the MSG/SEVIRI imager shows a relatively high negative bias (-1.39 K) to be compared to the other instruments whose biases are mainly in the range -0.5 – 0.5 K over the entire period. While the calibration issues only explain the bias between Meteosat-7 and ScaRaB, the differences of temperature between ScaRaB and the other geostationary platforms can be explained by both a poor calibration and some differences in the SRF. The results of the intercalibration and spectral normalization are shown in Fig. 6(b). Over the entire period and for all the geostationary platforms, the decadal

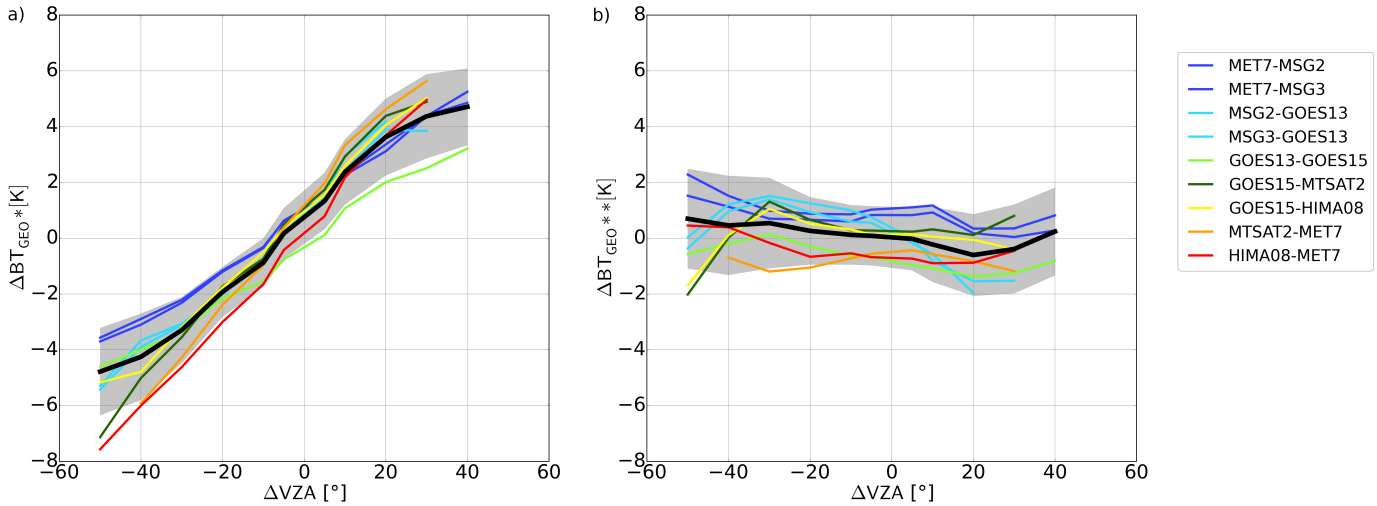


Fig. 9. Variation of the BT bias according to the VZA differences between pairs of geostationary platforms observing common areas and for BT_{ScaRaB} in the range [180–235 K]. (a) Before VZA corrections. (b) After VZA corrections. The BT bias and its standard deviation for all the pairs of geostationary platforms are represented respectively by the black line and the filled area in gray over the 2012–2016 period for all the ten days.

468 residual bias indicates a very low and stable bias, around 0 K,
 469 with a standard deviation lower than 0.08 K between BT_{ScaRaB}
 470 and BT_{GEO} . The corrections are then directly applied to the
 471 original geostationary data.

472 To fully describe the impact of the ScaRaB calibration
 473 and spectral corrections, an independent validation has been
 474 developed. The bias of uncorrected and corrected BT_{GEO} is
 475 then computed in the range of 180–240 K every ten days
 476 between pairs of neighbored geostationary platforms observing
 477 common areas and sharing an equivalent VZA (Fig. 7). Indeed,
 478 for two adjacent geostationary satellites, the BT bias at the
 479 middle of the overlap region should average to zero, due to
 480 the similar geometry for both platforms. Fig. 7(a) shows the
 481 time series of decadal mean differences between uncorrected
 482 BTs of each pair of geostationary satellites. Each color
 483 represents a pair of geostationary platforms. Results indicate
 484 a scatter of the decadal bias between -4 and 4 K. The
 485 maximum bias occurs for the BT differences between
 486 MTSAT-1/2 and Meteosat-7, which averages at 2.9 K from
 487 the beginning of 2012 to June 2015. The bias between
 488 uncorrected BT_{MET-7} and $BT_{MSG-1/2/3}$ exhibits a relatively
 489 high bias (-1.62 K on average) and is distinguishable from
 490 the other pairs of geostationary satellites, in contrast to
 491 its relatively high seasonal variations. The distribution of
 492 the decadal initial bias for all the geostationary platforms
 493 over the 2012–2016 period [Fig. 7(c)] reveals a multimodal
 494 distribution, which averages at 0.25 K and displays a standard
 495 deviation of 1.65 K. Fig. 7(b) shows that the corrections
 496 applied to the geostationary IR observations improve the error
 497 as well as the disparity of the scatter plot over the whole
 498 period and for all the geostationary platforms. These results
 499 are confirmed by the Gaussian distribution of the decadal
 500 bias for corrected BT_{GEO} shown in Fig. 7(c), which averages
 501 at 0.19 K with a standard deviation of 0.87 K. The resulting
 502 geostationary satellite calibration residuals specifications are
 503 hence well within the 1 K limit previously mentioned and

demonstrate the importance to develop a correction procedure
 for the geostationary IR observations at cold temperatures.

D. Limb-Darkening Adjustment

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 507 1) *Methodological Considerations:* To complete the
 508 homogenization procedure of the geostationary database, we
 509 focus now on the dependence of the BT_{GEO} on the VZA.
 510 This issue, also called limb darkening, corresponds to a
 511 decrease in the temperature as the VZA increases. The greater
 512 optical path length of the absorbing atmosphere, as the
 513 VZA increases, results in a larger atmospheric absorption.
 514 Indeed, a longer optical path length contains much more
 515 water vapor and ozone explaining the observation of colder
 516 temperatures [35], [37], [38]. Cloudy scenes imply a second
 517 mechanism in the VZA issues [35]. A geometric effect may
 518 also be involved when the sides of the clouds obstruct the
 519 Earth’s emitted radiation at a large VZA. Some studies on
 520 these geometric effects discussed the different configurations
 521 of cloud fields [39]. To prevent an erroneous analysis between
 522 meteorological situations, which occurred at nadir and at a
 523 large VZA, the BTs have to be limb-adjusted. Some studies
 524 have been carried out to limb adjust the IR observations
 525 for cloudy regions by establishing empirical limb correction
 526 functions, depending on the cosine of zenith angle from the
 527 radiative transfer model for low earth orbit platforms [39]
 528 and IR geostationary observations [40]. However, it has
 529 been shown [41] that the corrections developed by [39]
 530 underestimate the observed BTs. The variation of the radiance,
 531 according to the zenith angle, is approximated in the CLAUS
 532 data set, by applying a function of the cosine of the zenith
 533 angle. Another way to face the limb-darkening issues is to use
 534 observations from low earth orbit platforms. Limb-correction
 535 algorithms have been developed for microwave observations
 536 from the AMSU-A and are based on a physical-statistical
 537 methodology [37]. The limb-darkening problem has also been

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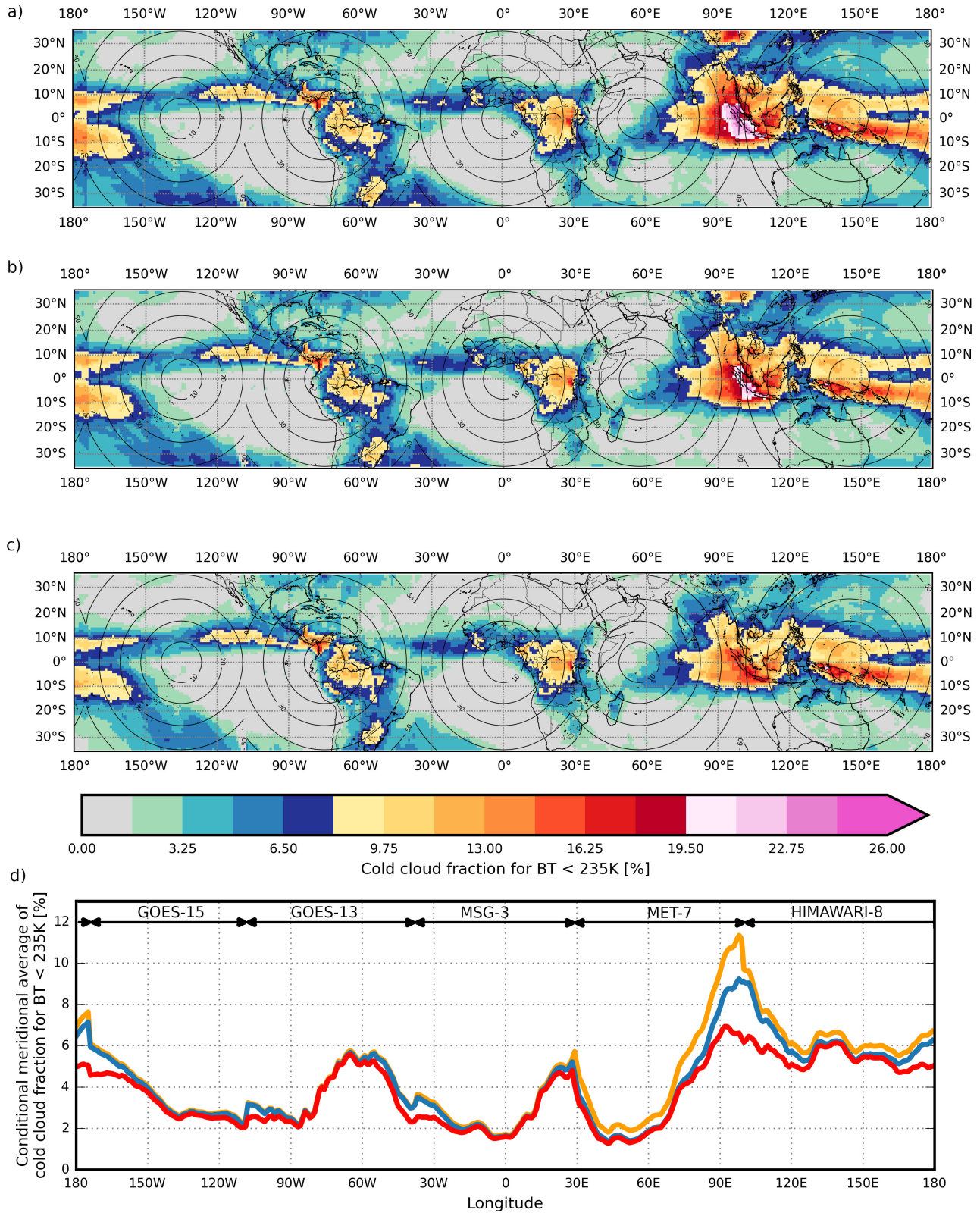


Fig. 10. Map of the mean cold cloud fraction (%) for $BT_{GEO} < 235\text{ K}$ over 2016 (a) by using the initial BT_{GEO} before any corrections, (b) by using the calibrated and spectrally corrected BT_{GEO} , and (c) after applying the VZA corrections on BT_{GEO} . The black circles represent the different levels of VZA. (d) Conditional meridional average of the cold cloud fraction for $BT_{GEO} < 235\text{ K}$ before any corrections (orange line), after applying the calibration and spectral corrections (blue line) and after applying the VZA corrections (red line).

538 explored by comparing the radiances from the geostationary
539 platforms with observations from polar orbiter platforms [40].
540 Some limb-darkening corrections have been computed from

collocated pairs of GOES platforms (GOES-8/GOES-10) and
541 and Meteosat platforms (Meteosat-5/Meteosat-7) [35]. They
542 proposed a two-step correction. First, a correction table of
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544 the BT_{GEO} as a function of VZA is applied to BT_{GEO} only
 545 for observations in the tropics. For mid-latitude regions,
 546 an additional correction is applied to tackle the latitudinal
 547 and seasonal dependence of BT_{GEO} . This dependence is
 548 explained by a smaller contrast between the cloud top and
 549 the Earth's surface in mid-latitudes than over the tropics.
 550 This limb-darkening correction is used in GridSat and ISCCP
 551 data sets [4]. However, this methodology is likely to suffer
 552 from limitations to be used to homogeneously limb adjust a
 553 fleet of geostationary satellites. Indeed, the limb adjustment,
 554 developed in the 2000s and empirically derived from a
 555 combination of a specific and limited set of instruments,
 556 may also be biased for other and new generation of
 557 geostationary instruments [12]. Moreover, the computation of
 558 the limb-darkening correction is limited by the range of VZA,
 559 due to the combination of the fixed location of geostationary
 560 satellites inducing few observations of cold BT_{GEO} .

561 It appears important, for this article, to tackle limb-
 562 darkening issues on geostationary satellite per geostationary
 563 satellite over the full time period by using a common reference,
 564 such as the IR observation from the ScaRaB instrument.
 565 This correction focuses on cold clouds monitored by the IR
 566 imagery of geostationary satellites previously intercalibrated
 567 and normalized into a $10.5\text{--}12.5\ \mu\text{m}$ IR equivalent BT, called
 568 BT_{GEO}^* . For this purpose, we perform for each GEO platform
 569 a collocation procedure between the regular lon/lat 0.5° and
 570 corrected IR geostationary data, presented previously, and the
 571 ScaRaB/L2B IR observation over the entire 2012–2016 period.
 572 The collocation procedure is applied under some criteria.
 573 Where the zenithal angles of ScaRaB do not exceed 20° ,
 574 the geostationary zenith angles can range from 0° to 70° . The
 575 collocation is also performed with a time delay of less than
 576 10 min between the GEO and the ScaRaB observations and for
 577 BT_{ScaRaB} , which does not exceed 235 K for standard deviations
 578 lower than 2 K.

579 GEO-ScaRaB match-ups are binned into geostationary
 580 zenith angle intervals of 2° and for zenith angles ranging
 581 from 20° to 70° . For GEO VZAs lower than 20° , a unique
 582 limb correction is performed. Moreover, to ensure a relevant
 583 sample size on each 2° zenith angle bin, the regressions are
 584 applied annually. For each zenith angle interval, the collocated
 585 BT_{GEO}^* and BT_{ScaRaB} are binned and averaged for every
 586 BT_{ScaRaB} interval of 5 K, from 180 to 235 K. To filter corrupted
 587 collocated points, the binning procedure is performed when
 588 at least ten samples are present. The limb correction then
 589 consists of applying annual regressions between the binned
 590 GEO and ScaRaB data for every 2° zenith angle bin and
 591 for each geostationary platform. A second-order polynomial
 592 regression has been preferred over the linear regression, due
 593 to slight improvements in the minimization of the residuals.

594 2) *Results*: Fig. 8(a)–(c) shows the variation of the initial
 595 bias between each GEO platform and ScaRaB, according
 596 to the VZA for different bins of ScaRaB temperatures and
 597 over the 2012–2016 period. First, one can observe a negative
 598 increase in the bias as the VZA increases, whatever the range
 599 of BT. For VZA lower than 30° , the absolute bias seems to be
 600 lower than 1 K, whatever the BT_{ScaRaB} is. However, we can
 601 observe drastic differences in the bias evolution depending on

the BT_{ScaRaB} bin. Indeed, at warmer BT_{ScaRaB} [225–235 K],
 the negative increase in the bias according to the VZA is more
 pronounced than for colder values of BT_{ScaRaB} [190–200 K].
 For a VZA at 40° , the bias averages at -0.63 K for BT_{ScaRaB} ,
 ranging from 190 to 200 K and for all the GEO platforms,
 while the bias averages at -2.9 K for BT_{ScaRaB} greater than
 225 K and for a similar VZA. It is also to be noticed that
 the disparity of the bias among GEO platforms increases with
 the VZA. Thus, at a 60° zenith angle and for the 225–235 K
 range, the bias varies from -7.4 to -9.5 K for GOES-15 and
 Meteosat-7, respectively. Such large errors have been previ-
 ously reported for an older Meteosat first-generation satellite
 (Meteosat-4) compared to NOAA/AVHRR reference [27].
 Note that the uncertainty does not increase much with the BT
 but increases with the VZA. Fig. 8(d)–(f) shows the results
 of the limb-darkening corrections according to the variation
 of the zenith angle for all the geostationary platforms and for
 different ranges of BT_{ScaRaB} . Over the whole range of ScaRaB
 BTs, the residual biases average at 0 K, regardless of the
 geostationary platform and the zenith angles. These results
 demonstrate the capability of the limb-darkening correction
 methodology we have developed. An independent validation
 is provided in Fig. 9 by comparing pairs of geostationary
 platform observations. It shows the variation of the decadal
 biases of BT_{GEO}^* between pairs of geostationary satellites,
 according to the difference of their VZAs (ΔVZA), before
 and after applying limb-darkening corrections. A ΔVZA of
 0° means that two adjacent geostationary platforms observe
 the same region with an equivalent VZA. On the contrary,
 a large ΔVZA indicates that a given geostationary platform
 observes a region with a nadir zenith angle, while the adjacent
 geostationary platform monitors the same region with a limb
 zenith angle. Before applying the zenith angle corrections,
 we can observe an increase of BT_{GEO}^* bias for all the
 platforms from ~ -5 to ~ 5 K as the ΔVZA moves from
 -50° to 40° [Fig. 9(a)]. The standard deviation, on its side,
 varies between 0.86 K for a low ΔVZA and 1.57 K for
 large ΔVZA . Results indeed indicate a larger disparity of
 the BT_{GEO}^* biases between pairs of geostationary satellites
 for large ΔVZA . After applying the zenith angle correction
 [Fig. 9(b)], the BT_{GEO}^{**} bias is relatively stable and averages
 at 0.09 K with a standard variation of 1.43 K, regardless of
 the variation of ΔVZA . The standard deviation varies from
 1.05 to 1.78 K for low and large ΔVZA , respectively. Note
 that when the limb-darkening issue is clearly improved and
 corrected, the standard deviation for a low VZA is a little
 larger than before applying the VZA correction.

IV. DISCUSSION AND CONCLUSION

In summary, a level 1c IR GEOing data set is introduced.
 This data set is a consistent $10.5\text{--}12.5\ \mu\text{m}$ IR equivalent BT
 data set, with a homogeneous recalibration, a $0.04^\circ \times 0.04^\circ$
 spatial resolution, a 30 min common time resolution, and a
 correction for limb effect. The data set covers the 2012–2016
 period. The global homogeneity of the IR GEOing data set,
 regardless of the variation of VZA, is then characterized by a
 standard deviation of 1.43 K within any of two geostationary
 satellites.

As discussed in the Introduction section, global geostationary IR observations provide useful resources to carry out studies on convective systems. Cold cloud tracking algorithms usually delineate cold clusters by applying a 235 K threshold on IR geostationary data. To evaluate the impact of the geostationary IR data homogenization on high cold clouds, we compute the cold cloud fraction. The fraction is determined by applying a 235 K threshold on the initial BT_{GEO} before any corrections, on the spectrally adjusted and calibrated BT_{GEO} and finally on the VZA corrected IR data for the 2016 period and over the entire tropics. The computations are performed by using the selected configuration of the geostationary fleet and the map of cold cloud fraction provided on a $1^\circ \times 1^\circ$ grid [Fig. 10(a)–(c)]. Before corrections, a local maximum of cold cloud fraction reaching 26% is seen in the Meteosat-7 area over the west coast of the Indo–China peninsula. However, the calibration correction allows to attenuate this fraction which falls to 22%. Also, note that this region is observed by Meteosat-7 with a VZA of 47° . After the limb-adjustment procedure, the value of cold cloud fraction for this specific region decreases to 17%. The maximum of cold cloud fraction is now located over the warm pool and is observed by the HIMAWARI-8 platform. One can also see that the local maximum of cold cloud fraction over the Tibetan plateau is strongly attenuated with all the corrections. Before applying the homogenization procedure, one can observe some steps of the conditional meridional average of cold cloud cover at the boundary of Meteosat-7 and HIMAWARI-8 and between HIMAWARI-8 and GOES-15, reaching 1.65% and 1.47%, respectively [Fig. 10(d)]. The combined corrections improve these issues and show a cold cloud cover exhibiting a smoother transition from one platform to another. Fig. 10(d) shows similar cold cloud covers for low VZA between a unique calibration/spectral correction and the combined calibration/spectral and VZA corrections, while one can observe a decrease in cold cloud cover for large VZA between the two corrections.

The extension of the current database beyond 2016 is under consideration. The configuration of the fleet, nevertheless, drastically changes in 2017 with the end of operation of Meteosat-7 and the arrival of MSG-1 in February on a shifted position, although not fully covering the Indian Ocean. Tests are needed to explore the sensitivity of the database to this new configuration. The use of INSAT-3D is also contemplated as a better way to bridge MSG-1 and HIMAWARI-8 data. GOES-R has become operational and the stream of GOES-16 replaces GOES-13 in December 2017. Similarly, GOES-17 is now operational as the new GOES west coverage replacing the GOES-15 platform from February 2019.

Up to the end of 2018, the Megha-Tropiques mission has been operated nominally granting the possibility of extending the present effort up to that time. While the present work relies on the ScaRaB instrument onboard Megha-Tropiques, it can easily be applied to alternative IR reference observations, from hyperspectral sounders, for instance. This article indicates that the final calibrated and limb-adjusted IR observations for $BT < 240$ K can be homogeneous throughout the GEOring

with less than 1.5 K standard deviation, and that further future efforts should strive for such, or better, accuracy.

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Homogenization of Geostationary Infrared Imager Channels for Cold Cloud Studies Using Megha-Tropiques/ScaRaB

Thomas Fiolleau¹, Rémy Roca, Sophie Cloché, Dominique Bouniol, and Patrick Raberanto

Abstract—Infrared (IR) observations from the fleet of multi-agencies meteorological geostationary satellites have a great potential to support scientific and operational investigations at a quasi-global scale. In particular, such a data record, defined as the GEOring data set, is well suited to document the tropical convective systems life cycles by applying cloud tracking algorithms. Yet, this GEOring data set is far from being homogeneous, preventing the realization of its potential. A number of sources of inhomogeneities are identified ranging from spatiotemporal resolutions to spectral characteristics of the IR channels and calibration methodologies. While previous efforts have attempted to correct such issues, the adjustment of the cold part of the IR spectrum remains unfit for cold cloud studies. Here, a processing method is introduced to minimize the inhomogeneities against a reference observational data set from the Scanner for Radiation Budget (ScaRaB) instrument onboard the Megha-Tropiques satellite. The method relies on the collocations between the geostationary observations and the reference. The techniques exhibit significant sensitivity to the selection of the relevant pairs of observations requiring a dedicated filtering of the data. A second effort is then proposed to account for the limb-darkening effect and a method is developed to correct the brightness temperature (BT) dependence on the geostationary viewing zenith angle (VZA). Overall, results show a residual after the processing of 0 K between any of the geostationary data and the ScaRaB reference. The final calibrated and limb-adjusted IR observations are then homogeneous for cold BT lower than 240 K with a standard deviation lower than 1.5 K throughout the GEOring.

Index Terms— Calibration and spectral corrections, cold cloud studies, geostationary satellites, infrared (IR) image sensors, limb-darkening corrections.

I. INTRODUCTION

METEOROLOGY agencies monitor individually operational geostationary platforms, which when used all together permit an observation for all longitudes of the Earth

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on a large latitudinal band ($\sim 60^\circ$ S– 60° N) with similar instruments. The potential benefit of such an observing capability has triggered numerous efforts in the past 40 years to make use of it for various applications, ranging from the elaboration of a unique perspective on clouds studies [1], [2] for precipitation estimation [3], climate studies [4], and global visualization applications [5]. Yet, when carefully examined, this GEOring appears to be far from a homogenous suite of instruments operated in a similar fashion. The space/time resolution and sampling differ across the platforms. The acquisition procedure also differs among the satellites (from South to North, North to South, or by sector). The calibration procedure of each instrument is also performed at the individual level with the instruments' specific modes of operation. Since 2011, for the infrared (IR) channels, the agencies provide alternative calibration coefficients obtained from a statistical comparison to hyperspectral IR measurements from low earth-orbiting satellites using the common methodology of the Global Space-Based InterCalibration System (GSICS) [6]. Nevertheless, the GSICS currently only provides calibration coefficients referenced to the NASA Earth Observing System (EOS), Aqua Atmospheric IR Sounder (AIRS), and the Exploitation of Meteorological Satellites (EUMETSAT)–Centre National d'Etudes Spatiales (CNES), Meteorological Operation (MetOp), and IR Atmospheric Sounding Interferometer (IASI) calibration standard. This harmonization effort does not take into account spectral band responses and viewing angle effects, which are needed for sensor homogenization. Indeed, while often referred to as a unique IR channel, the multiple instruments do show some spectral differences as well.

Most of the spectral imagers onboard geostationary satellites monitor with a relatively high temporal frequency the calibration coefficients for thermal IR channels by using onboard blackbody references. Some of the differences in the temperature measurements from one geostationary satellite to another can be explained by the differences in the calibration references, methods, as well as in the spectral filter functions. Moreover, it has been observed that the spectral imagers may degrade over time with different rates [7], [8]. The geostationary imagers can also be affected by short-term variations. The accumulation of ice on the surface of the imager optics modifies their spectral response function (SRF) and consequently affects the measurement of the temperatures. To face these issues, decontamination events are regularly applied to the radiometers inducing

TABLE I
SUMMARY OF THE TECHNICAL CHARACTERISTICS OF THE VARIOUS EXISTING IR GEORING DATA SETS

Products	Spatial resolution	Temporal resolution	Spatial coverage	Temporal coverage	Calibration and temporal normalization	Limb darkening correction	Data source
Gridsat-B1	0.07°	180 min	70°S-70°N	1980-present	HIRS channel 8 [14]	Yes	ISCCP-B1
CPC	~ 0.04°	30 min	60°S-60°N	2000-present	No absolute calibration and no temporal normalization	Yes	Native
CLAUS	0.5°	180 min	Global*	1983-2006	ISCCP- B3	Yes	ISCCP-B3
CLAUS	0.3°	180 min	Global*	1985-2008	ISCCP- B3	Yes	ISCCP-B3
ISCCP-B1	Native FoV subsampled to ~ 0.1°	180 min	70°S-70°N	1980-present	AVHRR channel 5 [13]	No	Native
ISCCP-B3	Native FoV subsampled to ~ 0.3°	180 min	70°S-70°N	1983-2009	AVHRR channel 5 [13]	No	Native
This study	0.04°	30 min	35°S-35°N	2012-2016	SCARAB-IR channel 4	yes	Native

instability in the temperature measurements [9]. This effect has been particularly shown for the Meteosat Visible IR Imager (MVISIR) onboard the Meteosat-7 geostationary platform.

A diurnal cycle effect can also affect the imager calibration, especially for the instruments on the three-axis stabilized geostationary platforms [7], [9]. It has been shown that the sun-synchronous orbits of MetOp/IASI and EOS/AIRS used in the GSICS calibration are not sufficient to completely correct the diurnal IR calibration issues [10]. The precessing orbit of the tropical rainfall measuring mission (TRMM) platform and the observations of its visible and IR scanner (VIRS) have then been used to quantify and correct such a midnight IR calibration anomaly [10]. IR observations are also dependent on the geostationary viewing zenith angle (VZA). The issue is called limb darkening, corresponding to a decrease in the temperature as the VZA increases.

These multiple sources of inhomogeneity have prevented the full use of this unique observational capability, although a number of successful projects have attempted to overcome the above-listed limitations. Indeed, a number of significant efforts have been proposed in the last two decades to produce more or less homogeneous “global” archive of geostationary data sets for various applications. It ranges from cloud climatology computations (International Satellite Cloud Climatology Project (ISCCP)-B1, ISCCP-B3, and Cloud Archive User Service (CLAUS) [11]), hurricane trend analysis, support of precipitation estimation (Gridded Satellite (GridSat) [12]), real-time monitoring of precipitation (Climate Prediction Center (CPC) [13]), for the enhancement of radiation budget averages [8]. Additional information on these GEORing data sets is listed in [4]. Table I summarizes the available IR GEORing data sets altogether with their respective properties

and homogeneity levels. The ISCCP-B1 and B3 have been subsampled to a 180-min temporal resolution and to, respectively, ~0.1° and ~0.3° spatial resolution. They have been calibrated and spectrally normalized using advanced very high-resolution radiometer (AVHRR) observations [14]. The CLAUS data sets rely on the ISCCP-B3 and have been reprojected to an equal angle map projection with a spatial resolution of 0.3° and 0.5°. The GridSat-B1 is based on the ISCCP-B1 data set remapped to a 0.07° regular grid. In addition to the intercalibration used for the ISCCP data set, a second calibration using the high-resolution IR radiation sounder (HIRS) observations has been applied to correct a bias at cold temperatures observed after 2001. The CPC product is built in real time from native geostationary observations. The data are remapped to a regular grid of ~0.04° spatial resolution and are available every 30 min. To reduce the spectral differences and calibration issues, a complex, multistep, multiregional procedure is applied by comparing the brightness temperatures (BTs) from pairs of geostationary platforms. However, this procedure does not provide any absolute calibration. Also note that a limb correction has been performed on the GridSat-B1, CPC, and CLAUS data. In short, while a number of efforts have paved the way for a quantitative use of the GEORing data sets, no dedicated effort toward the cold BT regimes has been promoted so far. Yet, cold cloud studies benefit a lot from geostationary IR observations, for cloud microphysical or macrophysical parameters retrievals. The spatiotemporal resolution of the measurements further allows the life cycle analysis of the cold cloudiness and the estimation of related parameters. The present GEORing attempt is directed toward these scientific applications. In particular, it will serve as the input of the realization of a tropical mesoscale convective system data set using the TOOCAN algorithm [15].

TABLE II
 TECHNICAL CHARACTERISTICS OF THE OPERATIONAL GEOSTATIONARY SATELLITES FLEET AND THE ASSOCIATED
 IMAGERS USED OVER THE 2012–2016 PERIOD

Platform	Nadir location	Instrument	Central wavelength	Spectral interval	Spatial resolution at nadir	Temporal resolution	Region of interest	Source	Period
GOES-15	135° W	IMAGER	10.7 μm	10.2-11.2 μm	4 km	30 min	180° W-105° W 35° S-35° N	NOAA / DWD	Jan 2012- Dec 2016
GOES-13	75° W	IMAGER	10.7 μm	10.2-11.2 μm	4 km	30 min	111° W-30° W 35° S-35° N	NOAA / DWD	Jan 2012- Dec 2016
METEOSAT-8/9/10	0°	SEVIRI	10.8 μm	9.8-11.8 μm	3 km	15 min	45° W-45° E 35° S-35° N	EUMETSAT/ AERIS	Jan 2012- Dec 2016
METEOSAT-7 (IODC)	57.5° E	MVIRI	11.5 μm	10.5-12.5 μm	5 km	30 min	12° E-107° E 35° S-35° N	EUMETSAT/ AERIS	Jan 2012- Dec 2016
MTSAT-2	145° E	IMAGER	10.8 μm	10.3-11.3 μm	4 km	30 min	94° E-170° W 35° S-35° N	AERIS/ CIMSS	Jan 2012- May 2015
HIMAWARI-8	140.7° E	AHI	11.2 μm	11.0-11.4 μm	2 km	10 min	94° E-170° W 35° S-35° N	AERIS/ JMA	Jun 2015- Dec 2016

148 Such analyses require observations with a minimal temporal
 149 resolution of 30 min [15], [16] to ensure tracked objects
 150 overlap between two images. Such kind of studies requires the
 151 use of thresholds applied to BT, ranging from 200 to 240 K,
 152 depending upon the analysis [17], [15]. A homogeneous data
 153 set is then mandatory, if one wants to proceed with the whole
 154 tropical belt. The homogeneity and basic requirements for a
 155 GEOring IR data set geared toward tropical cold cloud tracking
 156 applications can be summarized as follows:

- 157 1) High-resolution spatial footprint and high spatial reso-
 158 lution (~5 km).
- 159 2) A minimum of 30 min time sampling.
- 160 3) A spatial coverage dedicated to the tropics.
- 161 4) Homogenized BTs through spectral and calibration
 162 adjustment.
- 163 5) Limb adjustment of IR observations.

164 Matching these requirements with the specifications of the
 165 IR data sets summarized in Table I confirms the need for
 166 a dedicated level 1c GEOring IR tailored for cold cloud
 167 tracking applications. Moreover, the current implementation
 168 of the GSICS coefficients computation is directed toward the
 169 warm part of the spectrum [8], ruling out its use for cold
 170 cloud studies calling for an alternative, cold cloud compliant,
 171 absolute calibration reference for the GEOring. The limb
 172 effect is less strong over the cold part of the temperature
 173 spectrum than at the warm and clear sky end of the spectrum,
 174 and the limb correction is easier to set up for cold cloudy
 175 scenes (BTs < 240 K) than for the clear atmosphere. This

effect is strongly nonlinear and is thought to be impactful
 only for the viewing angle above 26.5° [11]; it may still
 require a correction for a homogeneous interpretation of the
 BT field all through the image. It is difficult to establish
 requirements once and for all in terms of residuals difference
 in the homogeneous data, based on the radiometric noise of the
 first generation sensors and the various sources of uncertainties
 in the adjustment procedure and basic cloud-oriented retrievals
 sensitivity. We propose to target a less than 1.5 K standard
 deviation among any of the two geostationary satellites in the
 final calibrated and limb-adjusted product for BTs < 240 K.

II. DATA

A. Five-Year Database of IR Geostationary Observations Over the Entire Tropics

Thermal channel BT images obtained by the operational
 meteorological geostationary satellite fleet are used over
 the 35°S–35°N latitude belt and for the whole 2012–2016
 period. Table II shows the geostationary IR data used to
 cover the entire tropics for this period. The geostationary
 data set has been collected from different sources: the U.S.
 National Oceanic and Atmospheric Administration (NOAA),
 the European Organization for the Exploitation of Meteorological Satellites (EUMETSAT), the Japan Meteorological Agency (JMA), and the French Atmosphere and Service Data Pole (AERIS). Note that the Multifunctional Transport Satellite (MTSAT) data set has also been completed from the Space Science and Engineering Center (SSEC) of the University of

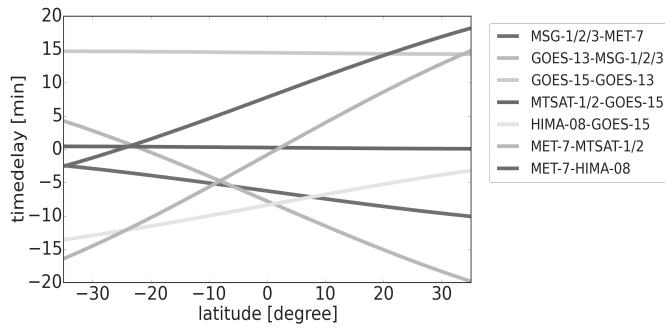


Fig. 1. Time delay of acquisition between pairs of geostationary images according to the latitude.

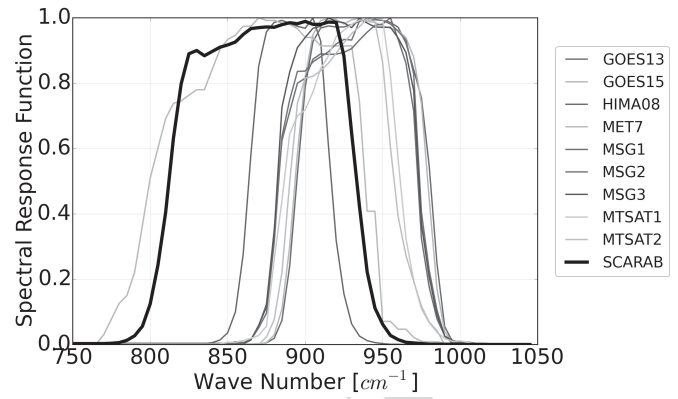


Fig. 2. SRF of the IR channel of the GEOring. The black line is the spectral response of the channel 4 of the ScaRaB instrument.

Wisconsin–Madison. The technical characteristics of each geostationary platform and their imagers are described in Table II and can differ from one platform to another. First, considering the temporal resolution, Meteorat Second Generation (MSG) and HIMAWARI-8 observe the Earth, respectively, with a 15- and 10-min time resolution, while the other platforms display a 30-min temporal resolution. Similarly, the scan-times vary from 7 min for HIMAWARI-8 to 25 min for Meteorat-7. The Geostationary Operational Environmental Satellite (GOES)-13 and GOES-15 sensors follow a complex scanning sector schedule. The full disk images are produced every 3 h, while the northern hemisphere and the southern hemisphere images are produced every 30 min with a time lag of few minutes between each scan. Also, note that the scanning schedule of MTSAT-2 does not provide half-hourly sampling of the southern hemisphere region. Contrary to the other imagers, the GOES-13 imager starts its earth scan at 15 and 45 min of the hour. Also note that Meteorat first- and second-generation imagers start their earth scan from the southeast corner, whereas NOAA and JMA platforms begin their earth scan from the northwest corner. All these temporal scanning differences imply some time delays between neighboring platforms observing the same region (Fig. 1). While the 15-min time delay is constant for the pairs of the GOES-15/GOES-13 and MTSAT-2/GOES-15 platforms, the time delay varies according to the latitude for the other pairs of geostationary platforms. At 35° S, a maximum time delay is found for the pair of MSG/GOES-13 platform of around 20 min. The maximum time delay amplitude accounted for the pair of the Meteorat-7/MTSAT-2 platforms and varies between −16 and 15 min.

Depending on the considered satellite, spatial resolution at nadir differs from one geostationary satellite to another and ranges from 2 km for HIMAWARI-8 to 5 km for Meteorat-7. In addition to the resolutions disparities, the GEOring is also characterized by differences in the central wavelength and the associated SRFs of each of the instruments (Fig. 2). While the central wavelength ranges between 10.7 and 11.5 μm for the GOES imagers and the Meteorat-7 imager, respectively, two types of SRFs can define the IR channels. The GOES-13, GOES-15, MTSAT-2, and HIMAWARI-8 imagers display narrowband channels with a bandwidth lower than 1 μm , whereas the Meteorat-7/MVIRI and MSG/SEVIRI imagers exhibit

broader channels characterized by a 2 μm bandwidth. A broadband channel is highly sensitive to ice contamination on the optics of the imagers, modifying their SRFs, and consequently, introducing a calibration error. Operational decontamination procedures are applied regularly to remove the ice build-up on the optics. Thus, spectral differences, as well as the individual satellite calibration procedures, contribute to the GEOring radiometric inhomogeneities.

B. Megha-Tropiques/ScaRaB-3 Observations

The Scanner for Radiation Budget (ScaRaB) instrument on the Megha-Tropiques platform [18] is the third of its kind [19], [20] and has been designed to measure the Earth radiation components at the top of the atmosphere with high accuracy (<1%). The instrument acquires data across the satellite track with a swath of ~ 2200 km from 30° S to 30° N. It is a broadband radiometer with four channels. The total channel measures the total energy between 0.2 and 100 μm . The shortwave channel (0.2–4 μm) is subtracted from the total channel to obtain the longwave part of the spectrum [21]. The fourth channel is an IR thermal channel (10.5–12.5 μm) which is used in this article (Fig. 2). The instrument operates nominally since the beginning of the mission and its operational performances are well in line with the specifications [22]–[24]. The nominal resolution of the ScaRaB footprint at nadir is 40 km and is detailed in [18].

The stability of the instrument is monitored by the CNES on a daily basis and so far, the Megha-Tropiques instruments have shown remarkable stability [25]. The use of geophysical targets like deep convective clouds (DCC) has been shown to be useful for geophysical calibration and cross-calibration of either broad radiometers [26], [27]. It is used here to showcase the relative stability of the channel 4 to that of the longwave channel. Fig. 3 shows the ratio between the radiance in channel 4 and channel 1 over the studied period for various deep convective regimes. The two channels of the instrument show no sign of relative degradation and a very steady behavior. Comparisons with the Clouds and the Earth's Radiant Energy System (CERES) instrument using dedicated collocation campaigns [28] further indicate a very good agreement between the two instruments' longwave channels [29].

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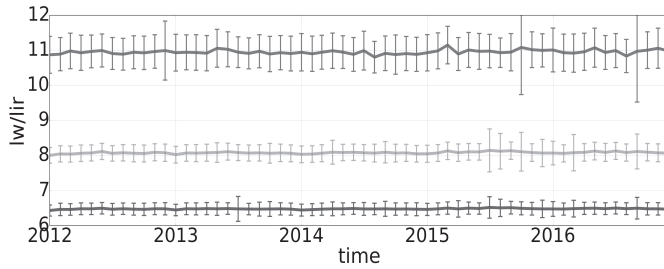


Fig. 3. Evolution of the ratio between measured radiance in the longwave band and in the IR channel of ScaRaB over the 2012–2016 period for various deep convective regimes. The blue curve corresponds to an IR radiance lower than $2.25 \text{ W} \cdot \text{m}^{-2} \cdot \text{sr}^{-1}$ corresponding to a 187 K BT, the yellow one lower than $4.25 \text{ W} \cdot \text{m}^{-2} \cdot \text{sr}^{-1}$ (207.4 K), and the red curve corresponds to an IR radiance lower than $7.75 \text{ W} \cdot \text{m}^{-2} \cdot \text{sr}^{-1}$ (224.5 K).

286 The ScaRaB instrument is also thermally controlled and is not
 287 impacted by icing events as shown in the monitoring of the
 288 instrument performances [22], [23]. The combination of an
 289 accurate and stable instrument makes it a well-suited reference
 290 to adjust the geostationary data.

291 Here we use the so-called level 2B products that consist
 292 of a 0.5° regularly gridded instantaneous directional radiances
 293 ($\text{W} \cdot \text{m}^{-2} \cdot \text{sr}^{-1}$). The original measurements are averaged on the
 294 regular grid using a statistical technique and the point spread
 295 function of the instrument [30]. Since 2011, the availability
 296 of this product is about 99.9%, making it a useful resource as
 297 a reference instrument for the spectral homogenization of the
 298 GEORing data. Note that unlike sun-synchronous platforms,
 299 the precessing orbit of Megha-Tropiques samples all the local
 300 times every 51 days allowing the collocation with the geo-
 301 stationary data all through the day and then the correction of
 302 the midnight IR anomaly [10], [18], [31]. The low inclination
 303 at the equator (20°) and high altitude of flight (865 km) also
 304 allow high repetitive measurements in the tropics [18].

305 III. HOMOGENOUS LEVEL 1C IR DATA SET 306 FOR THE TROPICS FOR 2012–2016

307 A. Homogenization of the Temporal Resolution

308 Given the requirements for cold cloud studies (see
 309 Section I), the lack of temporal resolution homogeneity
 310 between the geostationary imagers is accounted for by using
 311 a 30-min temporal frequency for all the platforms, preventing
 312 the use of MTSAT data in the southern hemisphere.

313 B. Homogenization of the Spatial Resolution

314 A common equal angle grid of $0.04^\circ \times 0.04^\circ$ has been
 315 selected for all the platforms to account for this source of
 316 inhomogeneity. This resolution is very close to the native
 317 resolution of GOES-13, GOES-15, MTSAT-2, and Meteosat-7
 318 data. The regridding process is performed by applying the
 319 inverse distance weighting method in the radiance space with
 320 a maximum search radius corresponding to the sum of half
 321 of the geostationary spatial resolution for a given pixel plus
 322 a half of the equal angle grid resolution. Then, the average
 323 radiance is transformed in BT using the Planck function in
 324 order to account for its nonlinearity.

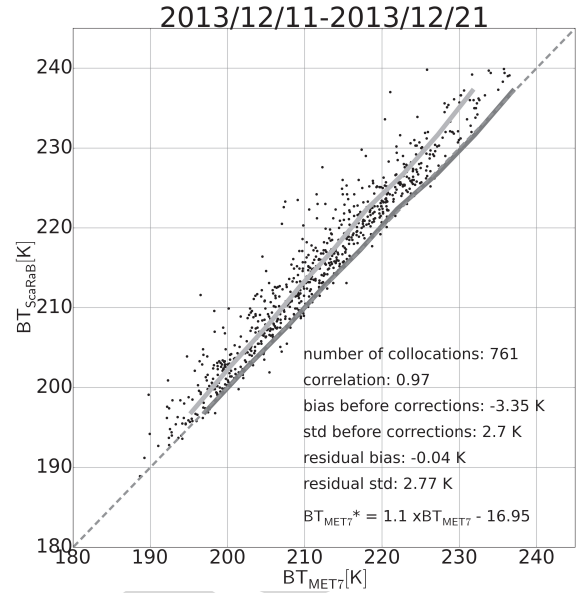


Fig. 4. Scatter plot between BT_{MET7} and $\text{BT}_{\text{ScaRaB}}$ obtained after the filtering procedure over a ten-day period starting in 2013/12/11 for BT in the range [180–240 K]. The orange line corresponds to the average BT for each 5 K bin. The blue line corresponds to the residual bias.

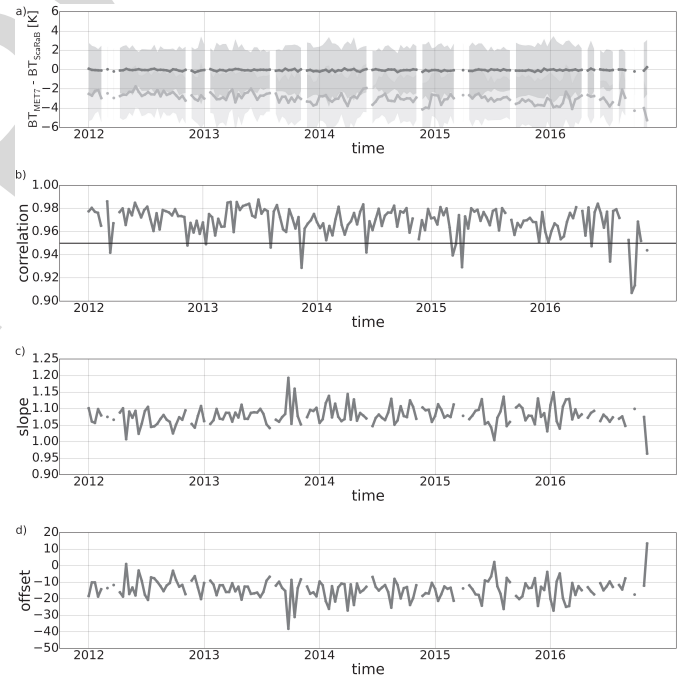


Fig. 5. (a) Time series of decadal initial BT bias (orange) and corrected BT bias (blue) for Meteosat-7 IR observations with respect to ScaRaB in the range [180–240 K]. The filled and transparent areas in orange and blue represent the standard deviations: (b) Time series of decadal linear regression correlations, (c) slope, and (d) offset between the ScaRaB and the Meteosat-7 IR observations in the range [180–240 K].

325 C. Intercalibration and Spectral Normalization

326 1) *Regression Technique*: The use of a common low
 327 earth-orbiting satellite, carrying an IR radiometer to anchor
 328 each of the platforms to a common reference data set, forms
 329 the basis of the different intercalibration and normalization
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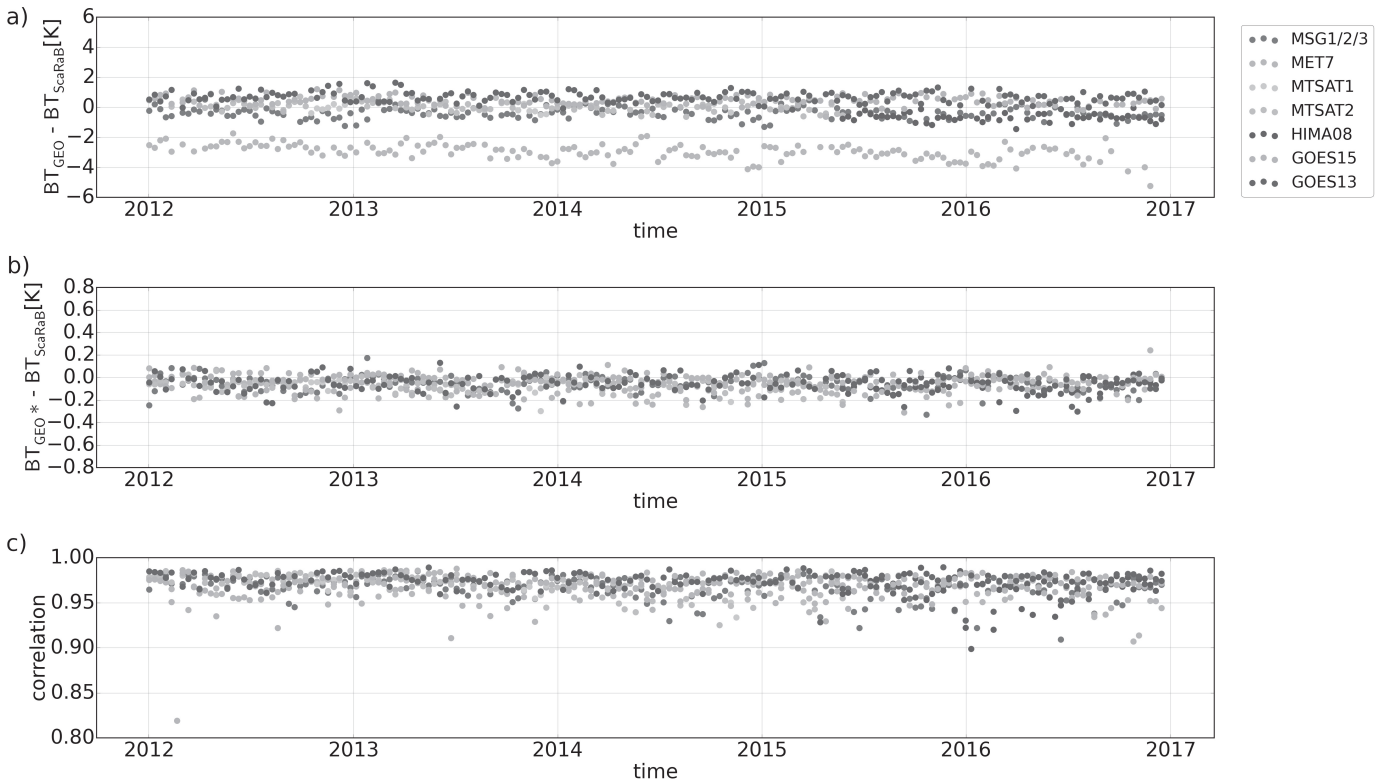


Fig. 6. (a) Time series of initial BT bias of geostationary IR observations with respect to ScaRaB in the range [180–240 K]. (b) Time series of BT bias between the geostationary IR observations and the ScaRaB observations in the range [180–240 K] after spectral and calibration corrections. (c) Time series of correlations between the ScaRaB and the geostationary IR observations in the range [180–240 K]. Each dot corresponds to a ten-day period.

330 efforts listed in Table I [14], [32] or under the GSICS
 331 calibration procedure [7]. Regression-based techniques are
 332 the most commonly used approaches. The difficulty arises in
 333 building the pairs (space/time/angular collocation) as well as
 334 in selecting the pairs of observations that will be regressed.
 335 This selection can have a strong impact on the result. Hence,
 336 it has been noted that the original ISCCP intercalibration was
 337 skewed toward a warm scene due to the implementation of
 338 the pairs selection filter prior to the regression computation,
 339 yielding a ~ 4 K cold bias on the cold part of the spectrum [33].
 340 Similarly, most of the tropical observations are located over
 341 large BT values and only a few of them are explained by
 342 DCCs. As the larger population of collocations is explained
 343 by high BT, the GSICS corrections are optimized for clear
 344 sky conditions and are not well suited for the cold cloud
 345 scenes [7]. A specific homogenization of the IR geostationary
 346 database is then required to fit with the objective of cold
 347 cloud tracking applications. This homogenization consists of
 348 a spectral normalization as well as an intercalibration of the
 349 various geostationary imagers focused on high cold clouds and
 350 on cold BTs. For this purpose, the BTs from the geostationary
 351 imagers are calibrated and spectrally normalized against the
 352 ScaRaB channel 4 measurements. The final data set then
 353 consists of a 10.5–12.5 μm IR equivalent BT. Calibration
 354 uncertainty is expected to be within 1 K, which is the typical
 355 error for operational satellite calibration [34]. Contrary to
 356 other geostationary imagers, the correction of the Meteosat-7/
 357 MVIRI imager is restricted to an intercalibration procedure,

358 since its broad SRF is similar to the ScaRaB/channel 4 and
 359 both of them are centered on 11.5 μm .

360 2) *Selection of the Data Match-Ups for the Regression:* The
 361 IR geostationary data from all the platforms are first remapped
 362 from their native formats to a regular lon/lat 0.5° grid every
 363 30 min to allow direct comparison with the ScaRaB/L2B data.
 364 Prior to computing the averaged radiance and the associated
 365 spatial standard deviation for each 0.5° grid point, the BTs of
 366 each geostationary platform are converted in radiances ($\text{mW} \cdot$
 367 $\text{m}^{-2} \cdot \text{sr}^{-1} \cdot \text{cm}^{-1}$) by applying the Planck function. The information
 368 on geostationary scan-time is also indicated for every grid
 369 point. The collocation between the ScaRaB-L2B observation
 370 and the IR 0.5° \times 0.5° gridded geostationary data is reached
 371 when both view the same scene with a predefined time delay
 372 and similar viewing geometry. The predefined criteria are.

- 373 1) A maximum VZA of 20° and 26°, respectively, for the
 374 ScaRaB and the geostationary observation, to ensure
 375 alignment of the collocated pixels in viewing geometry
 376 and to avoid limb-darkening issues [35].
- 377 2) A maximum time delay of 10 min between two col-
 378 located pixels, which is a tradeoff between a relevant
 379 number of match-ups to populate the decadal regressions
 380 and the best possible temporal precision.
- 381 3) The IR-gridded radiances are regressed in units of
 382 temperature.

383 As we focus on high cold clouds scenes, we have developed
 384 a filtering procedure that keeps more collocation scenes at a
 385 colder temperature. Indeed, a bulk of collocations occurs at

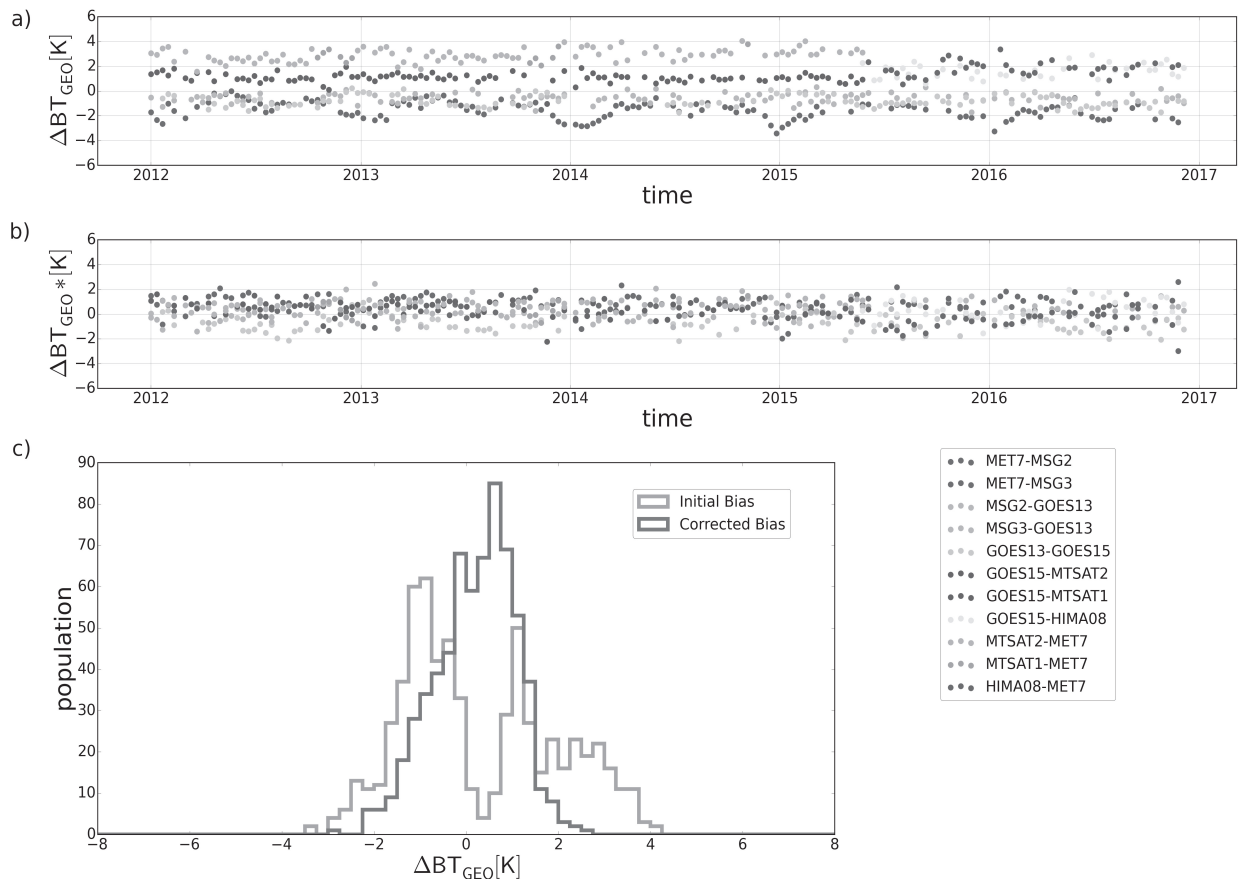


Fig. 7. (a) Time series of initial BT bias between pairs of geostationary platforms observing common areas with an equivalent VZA in the range [180–240 K]. (b) After spectral and calibration corrections; each dot corresponds to a ten-day period. (c) Distribution of the bias before and after corrections over the 2012–2016 period for all the pairs of geostationary platforms for all the ten-day periods.

386 a warmer temperature and displays a lower standard deviation [33]. To separate the high cold cloud population from
 387 the warmer targets in the scatterplot, thresholds are applied to the GEO standard deviation that depends on the BT.
 388 Pixels presenting a $BT_{GEO} > 240$ K are rejected if their standard deviations are larger than 0.5 K and pixels with
 389 $BT_{GEO} < 240$ K are kept if their standard deviations are lower than 2 K. Besides separating the high cold cloud popu-
 390 lation from clear sky pixels, this filtering procedure ensures that heterogeneous moving objects do not contaminate the
 391 collocation scenes and that the two instruments observe the same high cloud target. Assuming also that the IR radiometers
 392 onboard geostationary platforms have a linear response when observing high cold cloud homogeneous scenes [8], [36],
 393 the calibration and spectral normalization corrections are then based on linear regressions computed over a 180–240 K
 394 range. For this, the collocated BT_{GEO} and BT_{ScaRaB} are binned and averaged for every BT_{ScaRaB} interval of 5 K from 180 to
 395 240 K. The regression coefficients are then computed over these binned data every ten days and over a ten-day period.
 396 This is required to ensure statistical robustness but prevent higher frequency variation (<10 days) of the calibration
 397 issues (due to decontamination for instance) to be accounted for. As shown below, such higher frequency effects do not
 398 impact the residuals of the regression. Correlations are used to
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411 determine the ten-day period that might have match-ups errors. If the correlations are lower than 0.95 or if the population
 412 of collocations does not exceed ten matchups, we consider that the linear regression cannot be computed and that every
 413 ten-day period is removed from our analysis. In this case, the previous regression coefficients are replicated.
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417 3) *Results*: Fig. 4 shows an example of such a regression computed from 761 collocated pixels between the Meteosat-7/
 418 MVIRI imager and the Megha-Tropiques/ScaRaB observations acquired over a ten-day period from 2013/2012/2011 and
 419 for a range between 180 and 240 K. The $11.5 \mu\text{m}$ channel of Meteosat-7 exhibits a large negative bias (-3.35 K) for
 420 this considered BT and for this specific period. The blue line shows the regression fit given by a slope of 1.11 and
 421 an offset of -16.95 K. The correction has produced a very low residual bias, which averages at -0.04 K over this
 422 decadal period, and the correlation higher than 0.97 between the two data sets shows the high quality of the regression
 423 computation. Fig. 5(a) shows the time series of difference between Meteosat-7/MVIRI and ScaRaB before and after
 424 calibration over the 2012–2016 period. The Meteosat-7 BT is corrected every ten days using the slopes and offset com-
 425 puted from the linear regression. Over the period and for the specific Meteosat-7 geostationary platform, 12% of the
 426 decadal scatterplots are excluded from our analysis following
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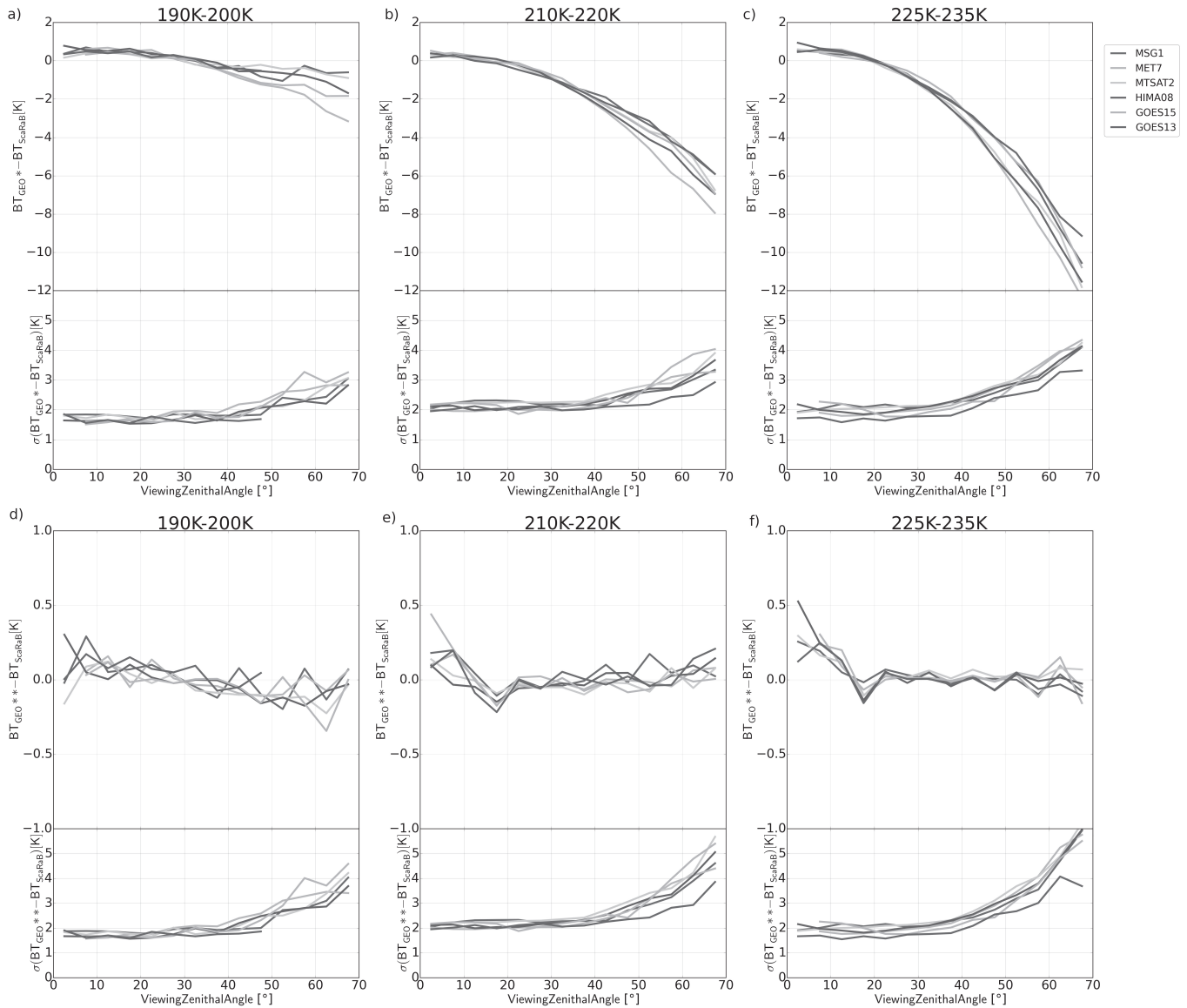


Fig. 8. Variation of the BT bias and the corresponding standard deviation between the GEO and the SCARAB observations according to the geostationary VZA, for a $VZA_{ScaRaB} < 20^\circ$ and for three different ranges of BT_{ScaRaB} . (a)–(c) Before VZA corrections. (d)–(f) After VZA corrections.

the filtering procedure [Fig. 5(b)] and 87 405 collocations have then been used to compute the decadal regressions. One can observe that over the 2012–2016 period, and before the corrections, the Meteosat-7/MVIRI exhibits a negative bias, averaging at -2.98 K and negatively increasing slightly with time due to contamination on the optics. This radiometric issue has been fully documented in [7]. The variations of the slopes and offsets are relatively stable over the period [Fig. 5(c) and (d)], showing the robustness of the methodology. After applying the calibration correction, the residual bias is very smooth, stable, and close to zero over the whole period. The calibration and spectral normalization corrections are applied on all the geostationary platforms available over the 2012–2016 period. The results have been obtained over the entire period, and for all the geostationary platforms by filtering 8% of the decadal scatterplots which did not pass our

quality control [Fig. 6(c)], leading to a comparison of around 655 000 collocations. Fig. 6(a) shows the time series of the decadal initial bias in BT for all the geostationary imagers with respect to the ScaRaB observations in the range 180–240 K. Over the 2012–2016 period, the Meteosat-7/MVIRI exhibits the highest error as discussed previously. It is also to be noted that the MSG/SEVIRI imager shows a relatively high negative bias (-1.39 K) to be compared to the other instruments whose biases are mainly in the range -0.5 – 0.5 K over the entire period. While the calibration issues only explain the bias between Meteosat-7 and ScaRaB, the differences of temperature between ScaRaB and the other geostationary platforms can be explained by both a poor calibration and some differences in the SRF. The results of the intercalibration and spectral normalization are shown in Fig. 6(b). Over the entire period and for all the geostationary platforms, the decadal

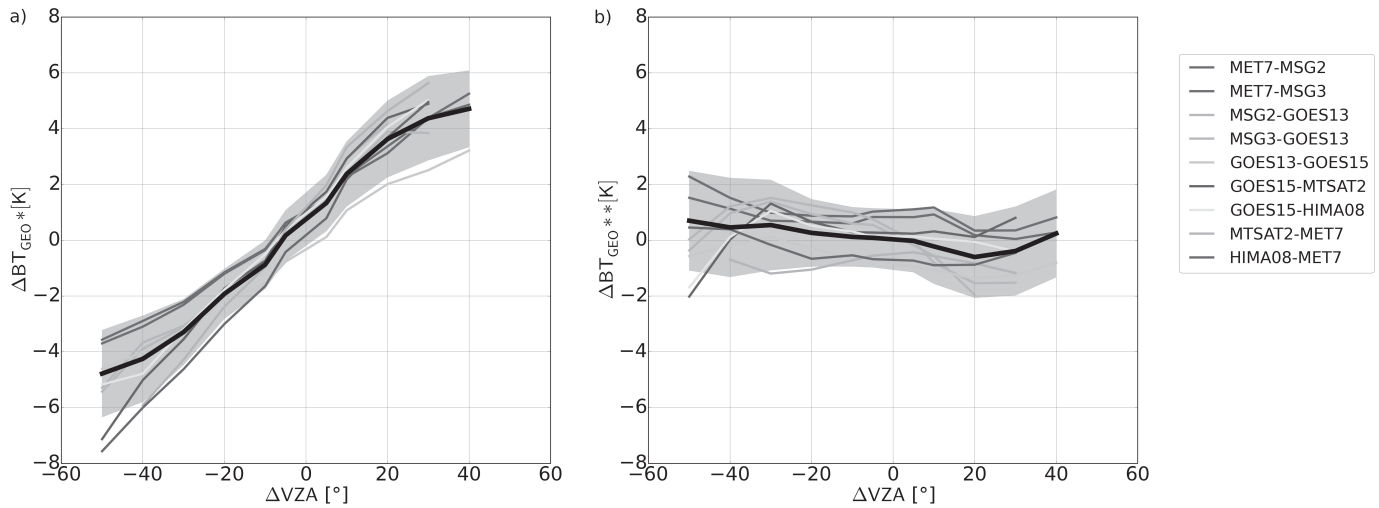


Fig. 9. Variation of the BT bias according to the VZA differences between pairs of geostationary platforms observing common areas and for BT_{ScaRaB} in the range [180–235 K]. (a) Before VZA corrections. (b) After VZA corrections. The BT bias and its standard deviation for all the pairs of geostationary platforms are represented respectively by the black line and the filled area in gray over the 2012–2016 period for all the ten days.

468 residual bias indicates a very low and stable bias, around 0 K,
 469 with a standard deviation lower than 0.08 K between BT_{ScaRaB}
 470 and BT_{GEO} . The corrections are then directly applied to the
 471 original geostationary data.

472 To fully describe the impact of the ScaRaB calibration
 473 and spectral corrections, an independent validation has been
 474 developed. The bias of uncorrected and corrected BT_{GEO} is
 475 then computed in the range of 180–240 K every ten days
 476 between pairs of neighbored geostationary platforms observing
 477 common areas and sharing an equivalent VZA (Fig. 7). Indeed,
 478 for two adjacent geostationary satellites, the BT bias at the
 479 middle of the overlap region should average to zero, due to
 480 the similar geometry for both platforms. Fig. 7(a) shows the
 481 time series of decadal mean differences between uncorrected
 482 BTs of each pair of geostationary satellites. Each color
 483 represents a pair of geostationary platforms. Results indicate
 484 a scatter of the decadal bias between -4 and 4 K. The
 485 maximum bias occurs for the BT differences between
 486 MTSAT-1/2 and Meteosat-7, which averages at 2.9 K from
 487 the beginning of 2012 to June 2015. The bias between
 488 uncorrected BT_{MET-7} and $BT_{MSG-1/2/3}$ exhibits a relatively
 489 high bias (-1.62 K on average) and is distinguishable from
 490 the other pairs of geostationary satellites, in contrast to
 491 its relatively high seasonal variations. The distribution of
 492 the decadal initial bias for all the geostationary platforms
 493 over the 2012–2016 period [Fig. 7(c)] reveals a multimodal
 494 distribution, which averages at 0.25 K and displays a standard
 495 deviation of 1.65 K. Fig. 7(b) shows that the corrections
 496 applied to the geostationary IR observations improve the error
 497 as well as the disparity of the scatter plot over the whole
 498 period and for all the geostationary platforms. These results
 499 are confirmed by the Gaussian distribution of the decadal
 500 bias for corrected BT_{GEO} shown in Fig. 7(c), which averages
 501 at 0.19 K with a standard deviation of 0.87 K. The resulting
 502 geostationary satellite calibration residuals specifications are
 503 hence well within the 1 K limit previously mentioned and

demonstrate the importance to develop a correction procedure
 for the geostationary IR observations at cold temperatures.

D. Limb-Darkening Adjustment

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1) *Methodological Considerations:* To complete the homogenization procedure of the geostationary database, we focus now on the dependence of the BT_{GEO} on the VZA. This issue, also called limb darkening, corresponds to a decrease in the temperature as the VZA increases. The greater optical path length of the absorbing atmosphere, as the VZA increases, results in a larger atmospheric absorption. Indeed, a longer optical path length contains much more water vapor and ozone explaining the observation of colder temperatures [35], [37], [38]. Cloudy scenes imply a second mechanism in the VZA issues [35]. A geometric effect may also be involved when the sides of the clouds obstruct the Earth’s emitted radiation at a large VZA. Some studies on these geometric effects discussed the different configurations of cloud fields [39]. To prevent an erroneous analysis between meteorological situations, which occurred at nadir and at a large VZA, the BTs have to be limb-adjusted. Some studies have been carried out to limb adjust the IR observations for cloudy regions by establishing empirical limb correction functions, depending on the cosine of zenith angle from the radiative transfer model for low earth orbit platforms [39] and IR geostationary observations [40]. However, it has been shown [41] that the corrections developed by [39] underestimate the observed BTs. The variation of the radiance, according to the zenith angle, is approximated in the CLAUS data set, by applying a function of the cosine of the zenith angle. Another way to face the limb-darkening issues is to use observations from low earth orbit platforms. Limb-correction algorithms have been developed for microwave observations from the AMSU-A and are based on a physical-statistical methodology [37]. The limb-darkening problem has also been

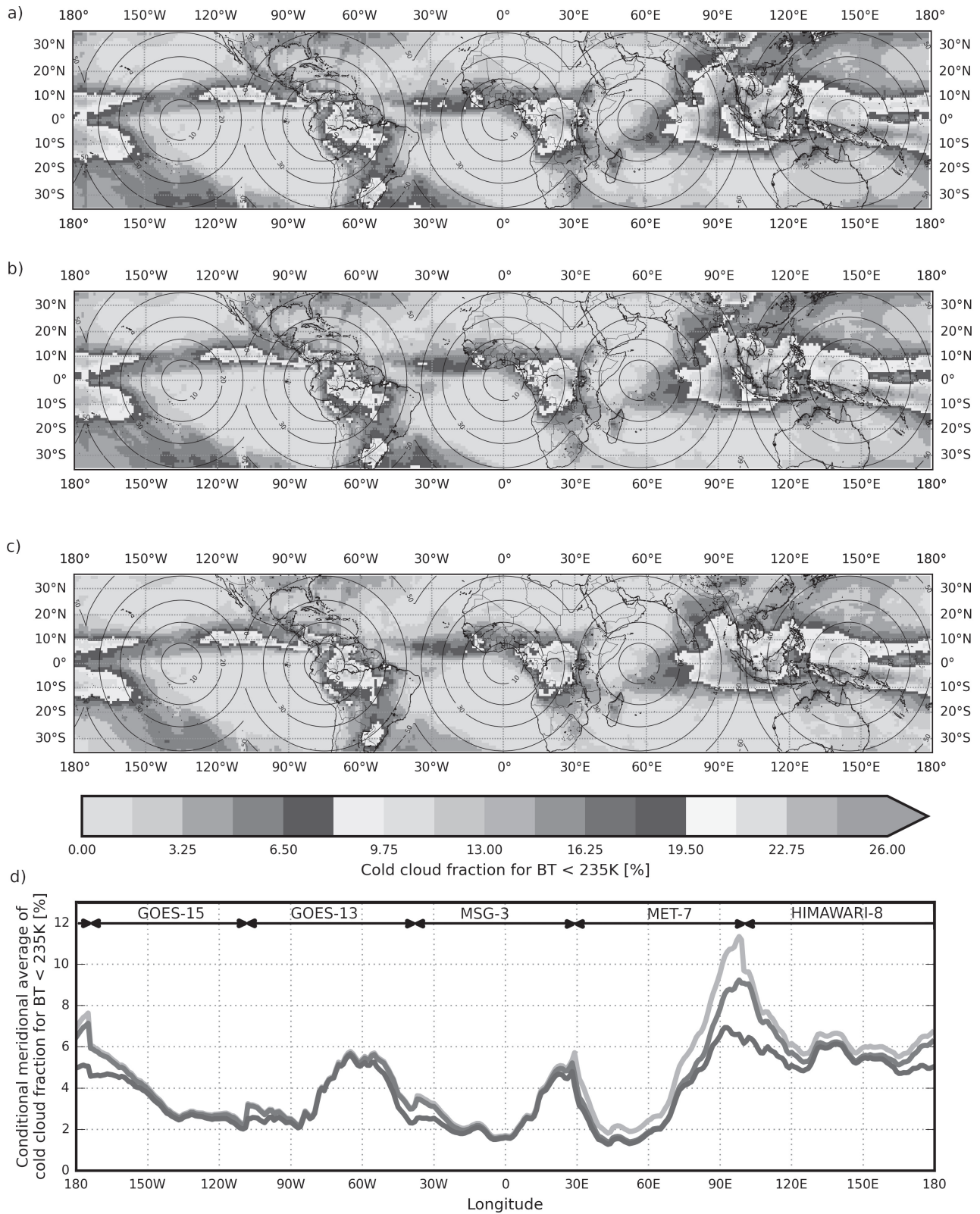


Fig. 10. Map of the mean cold cloud fraction (%) for $BT_{GEO} < 235\text{ K}$ over 2016 (a) by using the initial BT_{GEO} before any corrections, (b) by using the calibrated and spectrally corrected BT_{GEO} , and (c) after applying the VZA corrections on BT_{GEO} . The black circles represent the different levels of VZA. (d) Conditional meridional average of the cold cloud fraction for $BT_{GEO} < 235\text{ K}$ before any corrections (orange line), after applying the calibration and spectral corrections (blue line) and after applying the VZA corrections (red line).

538 explored by comparing the radiances from the geostationary
 539 platforms with observations from polar orbiter platforms [40].
 540 Some limb-darkening corrections have been computed from

collocated pairs of GOES platforms (GOES-8/GOES-10) 541
 and Meteosat platforms (Meteosat-5/Meteosat-7) [35]. They 542
 proposed a two-step correction. First, a correction table of 543

544 the BT_{GEO} as a function of VZA is applied to BT_{GEO} only
 545 for observations in the tropics. For mid-latitude regions,
 546 an additional correction is applied to tackle the latitudinal
 547 and seasonal dependence of BT_{GEO} . This dependence is
 548 explained by a smaller contrast between the cloud top and
 549 the Earth's surface in mid-latitudes than over the tropics.
 550 This limb-darkening correction is used in GridSat and ISCCP
 551 data sets [4]. However, this methodology is likely to suffer
 552 from limitations to be used to homogeneously limb adjust a
 553 fleet of geostationary satellites. Indeed, the limb adjustment,
 554 developed in the 2000s and empirically derived from a
 555 combination of a specific and limited set of instruments,
 556 may also be biased for other and new generation of
 557 geostationary instruments [12]. Moreover, the computation of
 558 the limb-darkening correction is limited by the range of VZA,
 559 due to the combination of the fixed location of geostationary
 560 satellites inducing few observations of cold BT_{GEO} .

561 It appears important, for this article, to tackle limb-
 562 darkening issues on geostationary satellite per geostationary
 563 satellite over the full time period by using a common reference,
 564 such as the IR observation from the ScaRaB instrument.
 565 This correction focuses on cold clouds monitored by the IR
 566 imagery of geostationary satellites previously intercalibrated
 567 and normalized into a $10.5\text{--}12.5\ \mu\text{m}$ IR equivalent BT, called
 568 BT_{GEO}^* . For this purpose, we perform for each GEO platform
 569 a collocation procedure between the regular lon/lat 0.5° and
 570 corrected IR geostationary data, presented previously, and the
 571 ScaRaB/L2B IR observation over the entire 2012–2016 period.
 572 The collocation procedure is applied under some criteria.
 573 Where the zenithal angles of ScaRaB do not exceed 20° ,
 574 the geostationary zenith angles can range from 0° to 70° . The
 575 collocation is also performed with a time delay of less than
 576 10 min between the GEO and the ScaRaB observations and for
 577 BT_{ScaRaB} , which does not exceed 235 K for standard deviations
 578 lower than 2 K.

579 GEO-ScaRaB match-ups are binned into geostationary
 580 zenith angle intervals of 2° and for zenith angles ranging
 581 from 20° to 70° . For GEO VZAs lower than 20° , a unique
 582 limb correction is performed. Moreover, to ensure a relevant
 583 sample size on each 2° zenith angle bin, the regressions are
 584 applied annually. For each zenith angle interval, the collocated
 585 BT_{GEO}^* and BT_{ScaRaB} are binned and averaged for every
 586 BT_{ScaRaB} interval of 5 K, from 180 to 235 K. To filter corrupted
 587 collocated points, the binning procedure is performed when
 588 at least ten samples are present. The limb correction then
 589 consists of applying annual regressions between the binned
 590 GEO and ScaRaB data for every 2° zenith angle bin and
 591 for each geostationary platform. A second-order polynomial
 592 regression has been preferred over the linear regression, due
 593 to slight improvements in the minimization of the residuals.

594 2) *Results*: Fig. 8(a)–(c) shows the variation of the initial
 595 bias between each GEO platform and ScaRaB, according
 596 to the VZA for different bins of ScaRaB temperatures and
 597 over the 2012–2016 period. First, one can observe a negative
 598 increase in the bias as the VZA increases, whatever the range
 599 of BT. For VZA lower than 30° , the absolute bias seems to be
 600 lower than 1 K, whatever the BT_{ScaRaB} is. However, we can
 601 observe drastic differences in the bias evolution depending on

the BT_{ScaRaB} bin. Indeed, at warmer BT_{ScaRaB} [225–235 K],
 the negative increase in the bias according to the VZA is more
 pronounced than for colder values of BT_{ScaRaB} [190–200 K].
 For a VZA at 40° , the bias averages at -0.63 K for BT_{ScaRaB} ,
 ranging from 190 to 200 K and for all the GEO platforms,
 while the bias averages at -2.9 K for BT_{ScaRaB} greater than
 225 K and for a similar VZA. It is also to be noticed that
 the disparity of the bias among GEO platforms increases with
 the VZA. Thus, at a 60° zenith angle and for the 225–235 K
 range, the bias varies from -7.4 to -9.5 K for GOES-15 and
 Meteosat-7, respectively. Such large errors have been previ-
 ously reported for an older Meteosat first-generation satellite
 (Meteosat-4) compared to NOAA/AVHRR reference [27].
 Note that the uncertainty does not increase much with the BT
 but increases with the VZA. Fig. 8(d)–(f) shows the results
 of the limb-darkening corrections according to the variation
 of the zenith angle for all the geostationary platforms and for
 different ranges of BT_{ScaRaB} . Over the whole range of ScaRaB
 BTs, the residual biases average at 0 K, regardless of the
 geostationary platform and the zenith angles. These results
 demonstrate the capability of the limb-darkening correction
 methodology we have developed. An independent validation
 is provided in Fig. 9 by comparing pairs of geostationary
 platform observations. It shows the variation of the decadal
 biases of BT_{GEO}^* between pairs of geostationary satellites,
 according to the difference of their VZAs (ΔVZA), before
 and after applying limb-darkening corrections. A ΔVZA of
 0° means that two adjacent geostationary platforms observe
 the same region with an equivalent VZA. On the contrary,
 a large ΔVZA indicates that a given geostationary platform
 observes a region with a nadir zenith angle, while the adjacent
 geostationary platform monitors the same region with a limb
 zenith angle. Before applying the zenith angle corrections,
 we can observe an increase of BT_{GEO}^* bias for all the
 platforms from ~ -5 to ~ 5 K as the ΔVZA moves from
 -50° to 40° [Fig. 9(a)]. The standard deviation, on its side,
 varies between 0.86 K for a low ΔVZA and 1.57 K for
 large ΔVZA . Results indeed indicate a larger disparity of
 the BT_{GEO}^* biases between pairs of geostationary satellites
 for large ΔVZA . After applying the zenith angle correction
 [Fig. 9(b)], the BT_{GEO}^{**} bias is relatively stable and averages
 at 0.09 K with a standard variation of 1.43 K, regardless of
 the variation of ΔVZA . The standard deviation varies from
 1.05 to 1.78 K for low and large ΔVZA , respectively. Note
 that when the limb-darkening issue is clearly improved and
 corrected, the standard deviation for a low VZA is a little
 larger than before applying the VZA correction.

IV. DISCUSSION AND CONCLUSION

In summary, a level 1c IR GEOing data set is introduced.
 This data set is a consistent $10.5\text{--}12.5\ \mu\text{m}$ IR equivalent BT
 data set, with a homogeneous recalibration, a $0.04^\circ \times 0.04^\circ$
 spatial resolution, a 30 min common time resolution, and a
 correction for limb effect. The data set covers the 2012–2016
 period. The global homogeneity of the IR GEOing data set,
 regardless of the variation of VZA, is then characterized by a
 standard deviation of 1.43 K within any of two geostationary
 satellites.

As discussed in the Introduction section, global geostationary IR observations provide useful resources to carry out studies on convective systems. Cold cloud tracking algorithms usually delineate cold clusters by applying a 235 K threshold on IR geostationary data. To evaluate the impact of the geostationary IR data homogenization on high cold clouds, we compute the cold cloud fraction. The fraction is determined by applying a 235 K threshold on the initial BT_{GEO} before any corrections, on the spectrally adjusted and calibrated BT_{GEO} and finally on the VZA corrected IR data for the 2016 period and over the entire tropics. The computations are performed by using the selected configuration of the geostationary fleet and the map of cold cloud fraction provided on a $1^\circ \times 1^\circ$ grid [Fig. 10(a)–(c)]. Before corrections, a local maximum of cold cloud fraction reaching 26% is seen in the Meteosat-7 area over the west coast of the Indo–China peninsula. However, the calibration correction allows to attenuate this fraction which falls to 22%. Also, note that this region is observed by Meteosat-7 with a VZA of 47° . After the limb-adjustment procedure, the value of cold cloud fraction for this specific region decreases to 17%. The maximum of cold cloud fraction is now located over the warm pool and is observed by the HIMAWARI-8 platform. One can also see that the local maximum of cold cloud fraction over the Tibetan plateau is strongly attenuated with all the corrections. Before applying the homogenization procedure, one can observe some steps of the conditional meridional average of cold cloud cover at the boundary of Meteosat-7 and HIMAWARI-8 and between HIMAWARI-8 and GOES-15, reaching 1.65% and 1.47%, respectively [Fig. 10(d)]. The combined corrections improve these issues and show a cold cloud cover exhibiting a smoother transition from one platform to another. Fig. 10(d) shows similar cold cloud covers for low VZA between a unique calibration/spectral correction and the combined calibration/spectral and VZA corrections, while one can observe a decrease in cold cloud cover for large VZA between the two corrections.

The extension of the current database beyond 2016 is under consideration. The configuration of the fleet, nevertheless, drastically changes in 2017 with the end of operation of Meteosat-7 and the arrival of MSG-1 in February on a shifted position, although not fully covering the Indian Ocean. Tests are needed to explore the sensitivity of the database to this new configuration. The use of INSAT-3D is also contemplated as a better way to bridge MSG-1 and HIMAWARI-8 data. GOES-R has become operational and the stream of GOES-16 replaces GOES-13 in December 2017. Similarly, GOES-17 is now operational as the new GOES west coverage replacing the GOES-15 platform from February 2019.

Up to the end of 2018, the Megha-Tropiques mission has been operated nominally granting the possibility of extending the present effort up to that time. While the present work relies on the ScaRaB instrument onboard Megha-Tropiques, it can easily be applied to alternative IR reference observations, from hyperspectral sounders, for instance. This article indicates that the final calibrated and limb-adjusted IR observations for $BT < 240$ K can be homogeneous throughout the GEOring

with less than 1.5 K standard deviation, and that further future efforts should strive for such, or better, accuracy.

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