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3	Top of the Atmosphere Reflected Shortwave Radiative Fluxes from GOES-R
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12	Abstract. Under the GOES-R activity, new algorithms are being developed at the National Oceanic and
13	Atmospheric Administration (NOAA)/Center for Satellite Applications and Research (STAR) to derive
14	surface and Top of the Atmosphere (TOA) shortwave (SW) radiative fluxes from the Advanced Baseline
15	Imager (ABI), the primary instrument on GOES-R. This paper describes a support effort in the
16	development and evaluation of the ABI instrument capabilities to derive such fluxes. Specifically, scene
17	dependent narrow-to-broadband (NTB) transformations are developed to facilitate the use of observations
18	from ABI at the TOA. Simulations of NTB transformations have been performed with MODTRAN4.3
19	using an updated selection of atmospheric profiles and implemented with the final ABI specifications.
20	These are combined with Angular Distribution Models (ADMs), which are a synergy of ADMs from the
21	Clouds and the Earth's Radiant Energy System (CERES) and from simulations. Surface condition at the
22	scale of the ABI products as needed to compute the TOA radiative fluxes come from the International
23	Geosphere-Biosphere Programme (IGBP). Land classification at 1/6° resolution for 18 surface types are
24	converted to the ABI 2-km grid over the (CONtiguous States of the United States) (CONUS) and
25	subsequently re-grouped to 12 IGBP types to match the classification of the CERES ADMs. In the
26	simulations, default information on aerosols and clouds is based on the ones used in MODTRAN.

Comparison of derived fluxes at the TOA is made with those from the CERES-and/or the Fast Longwave and Shortwave Radiative Flux (FLASHFlux) data. An-satisfactory agreement between the fluxes was observed and possible reasons for differences have been identified; the agreement of the fluxes at the TOA for predominantly clear sky conditions was found to be better than for cloudy sky due to possible time shift in observation times between the two observing systems that might have affected the position of the clouds during such periods. Differences in assumed cloud properties can also lead to differences in the fluxes derived from the two instruments.

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1 **1** Introduction

37 When a new satellite is contemplated, the exact characteristics of the newly selected sensors are not fully 38 known; simulations of proposed sensors are also not readily available. Yet, there is a need to obtain a 39 priori information on the expected performance of the new instruments. This is usually accomplished by 40 using characteristics of instruments in closest resemblance to the proposed ones and performing 41 simulations that can provide insight on the expected performance of the new instrument. As such, an 42 evolutionary process can be expected precedes the final stage that is reported in this paper, and it did 43 precede activities reported in this manuscript. One of the objectives at The ultimate objective at 44 NOAA/STAR in respect to the utilization of observations from is to be able to derive shortwave (SW) radiative fluxes 45 from the Advanced Baseline Imager (ABI) is to be able to derive shortwave (SW) radiative fluxes from., To get to the 46 surface SW from TOA satellite observations, there are two generic approaches: 1) the direct approach and 47 2) the indirect approach. In the direct approach one uses all the necessary information needed for deriving 48 the surface fluxes (some of which can be derived from satellites). Implementation of sSuch an approach 49 is feasible, for instance, with observations from MODIS where there is which has a long history of product 50 availability and evaluation. Examples of such an approach using MODIS observation are illustrated in Wang and 51 Pinker (2009), Ma et al. (2016), Pinker et al. (2018), Pinker et al., (2017a), Pinker et al. (2017b), Niu and 52 Pinker, (2015). GOES-R is a new instrument and as yet, similar information to the one available from

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53	MODIS is not yet available. Therefore, the indirect approach is used where one starts from satellite	
54	information at the TOA and models the atmosphere and surface with best available inputs (which do not	
55	have to be based on ABI). Examples of such an approach are discussed in Pinker, Zhang and Dutton	_
56	(2005), Ma and Pinker (2012), and Zhang et al. (2019). The "indirect path method" is used at the Center	
57	for Satellite Applications and Research (STAR) (Laszlo et al., 2020) for deriving SW radiative fluxes	\swarrow
58	from satellite observations; it requires knowledge of the SW broadband (0.2 - 4.0 μ m) top of the	
59	atmosphere (TOA) albedo. The Advanced Baseline Imager (ABI) observations onboard of the NOAA	
60	GOES-R series of satellites provide reflectances reflectance in six narrow bands in the shortwave spectrum	
61	(Table 1); these must be first transformed into broadband reflectance (the narrow-to-broadband, NTB,	
62	conversion process), and then the broadband reflectance must be transformed into a broadband albedo	
63	(the ADM conversion process).	
64	During the pre-launch activity NTB transformations were developed based on theoretical radiative	
65	transfer simulations with MODTRAN-3.7 and 14 land use classifications from the International	
66	Geosphere-Biosphere Programme (IGBP) (Hansen et al., 2010). They were augmented with ADMs from	
67	(CERES) observed ADMs (Loeb et al., 2003) and theoretical simulations (Niu and Pinker, 2011) to	
68	compute TOA fluxes. The resulting NTB transformations and ADMs have been tested using proxy data	
69	and simulated ABI data. The proxy instruments used in the simulations include the GOES-8 satellite, the	
70	Advanced Very-High Resolution Radiometer (AVHRR) sensor on the Polar Orbiting satellites, the	
71	Spinning Enhanced Visible Infra-Red Imager (SEVIRI) sensor on the European METEOSAT Second	
72	Generation (MSG) satellites, and the Moderate Resolution Imaging Spectroradiometer (MODIS)	
73	instrument on the NASA Terra and Aqua Polar Orbiting satellites (Pinker et al., 2021, unpublished). For	
74	each of these satellites, the evaluation of the methodologies was done differently; some results were	

76 from the (ESA) Geostationary Earth Radiation Budget (GERB) satellite (Harries et al., 2005). The results

evaluated against ground observations while others, against TOA information from CERES as well as

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77 obtained provided an insight on the expected performance of the new ABI sensor. Those procedures have

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78 been subsequently updated and applied to the new ABI instrument once it was built and fully 79 characterized.

In this paper we describe activity in support of methodologies to derive surface shortwave (SW) radiative fluxes from the operational Advanced Baseline Imager (ABI) instrument on the GOES-R series of the NOAA geostationary meteorological satellites. We describe the physical basis and the development of the (NTB) transformations of satellite observed radiances and the bi-directional corrections to be applied to the broadband reflectance to obtain broadband TOA albedo. The methodology will be presented in section 2, <u>data used are described in section 3</u>, results in section <u>3-4</u> and a summary and discussion in section 5.

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88 2. Methodology

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90 The following two flowcharts (**Figs. 1 and 2**) describe the necessary steps to derive the NTB 91 transformations and the ADMs. Details of these two steps will follow.

92 The TOA narrowband and broadband reflectances can be calculated from the spectral radiances

 $\rho_{nb}(\theta_0, \theta, \phi) = \frac{\pi \int_{\lambda_1}^{\lambda_2} I(\lambda, \theta_0, \theta, \phi) G(\lambda) d\lambda}{\int_{\lambda_2}^{\lambda_1} \cos(\theta_0) S_0(\lambda) G(\lambda) d\lambda}$

simulated from MODTRAN 4.3 and the response functions of the satellite sensor as shown in equations

94 (1) and (2):

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(1)

(2)

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$$\rho_{bb}(\theta_0, \theta, \phi) = \frac{\pi \int_{4\mu m}^{4\mu m} I(\lambda, \theta_0, \theta, \phi) d\lambda}{\int_{0.2\mu m}^{4\mu m} \int_{0.2\mu m}^{0.2\mu m} \cos(\theta_0) S_0(\lambda) d\lambda}$$

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- 98 where ρ_{nb} is narrowband reflectance; ρ_{bb} is broadband reflectance; θ_0 : solar zenith angle; θ : view
- 99 (satellite) zenith angle; ϕ : relative azimuth angle;
- 100 I_{λ} : reflected spectral radiance; $S_0(\lambda)$: solar spectral irradiance;
- 101 G_{λ} : spectral response functions of satellite sensors; λ_{1} and λ_{2} are the spectral limits of the sensor spectral
- band. This approach is widely used in the scientific community as also implemented in the work of Loeb
- 103 et al (2005), Wielicki et al. (2008), Su et al. (2015) and Akkermans et al. (2020).
- As stated previously, the ADMs from CERES-based observations (Loeb et al., 20032005; Kato et al.
- 2015) were augmented with theoretical simulations (Niu and Pinker, 2011) to compute TOA fluxes. This
- was done since due to the fact that_ CERES observations at that time higher latitudes are were under sampled. or not existent. at higher latitudes.
- 108 The combined ADMs are developed for each angular bin by weighting the modeled and CERES ADMs
- 109 based on the number of samples used to derive the ADMs of each type (Niu et al., 2011). Specifically:

110
$$\overline{R}(\theta_0, \theta, \phi) = \frac{1}{m+n} \left(m \times R_{CERES}(\theta_0, \theta, \phi) + n \times R_S(\theta_0, \theta, \phi) \right)$$
(3)

- 111 $\overline{R}(\theta_{\alpha}, \theta, \phi)$: averaged ADMs at each angular bin;
- 112 R_{CERES} : anisotropic factor from CERES ADMs;
- 113 R_s : anisotropic factor from simulated ADMs;
- 114 *m* and *n*: observation numbers at angular bins for CERES and simulated ADMs.
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116 2.1 Selection of Atmospheric profiles for simulations

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118 We have selected 100 atmospheric profiles covering the globe and the seasons, to use as input for 119 simulations with MODTRAN4.3. A tool was developed to select profiles from a Training Data set known 120 as SeeBor Version 5.0 (https://cimss.ssec.wisc.edu/training data/) (Borbas et.al. 2005). Originally it 121 consisted of 15704 global profiles of temperature, moisture, and ozone at 101 pressure levels for clear 122 sky conditions. The profiles are taken from NOAA-88, and the European Centre for Medium-Range 123 Weather Forecasts (ECMWF) 60L training set, TIGR-3, ozone-sondes from 8 NOAA Climate Monitoring 124 and Diagnostics Laboratory (CMDL) sites, and radiosondes from the Sahara Desert during 2004. A 125 technique to extend the temperature, moisture, and ozone profiles above the level of existing data was 126 also implemented by the providers (University of Wisconsin-Madison, Space Science and Engineering 127 Center, Cooperative Institute for Meteorological Satellite Studies (CIMSS). Fig. 3 shows the selected 128 profile locations; each season includes 25 profiles.

The SeeBor profiles are clear sky profiles. The top of the profiles is at 0.005 mb which is about 82.6 km.
We did an experiment to check the impact of reducing the number of levels for a profile (initially, we have used only 40 levels). In the experiment computed were radiances from profiles with 50 levels as well as radiances from profiles with 98 Levels. The difference between the two radiances (50 lev-98 lev) were below 5 % reaching 15 % around 2.5 µm. In the experiment we used the odd number levels starting from surface (plus the highest level) to reduce the number of profile levels.
Based on these experiments we have opted to keep all 98 profile levels.

136 The atmospheric profiles at each pressure level include temperature, water vapor and ozone. The surface 137 variables include surface skin temperature, 2 m temperature, land/sea mask, and albedo. We have 138 conducted a thorough investigation how the selected profiles represent the entire sample of 15704 profiles. 139 An example showing the comparison of temperature, humidity and ozone profiles is shown in Fig. 4. As 140 seen, there is a positive bias in the selected profile of temperatures due to their higher concentration at the 41 lower latitudes. A positive bias can be found at the lower levels while a negative bias is seen above 1 mb. 142 Since our domain of study is in such latitudes this selection should not have adverse effects on the 143 simulations.

144 2.2 Surface conditions

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Surface condition is one of the primary inputs into the MODTRAN simulations. The International Geosphere-Biosphere Programme (IGBP) land classification is used as data source (Hansen et al., 2010; Loveland et al., 2010). The dataset is at 1/6-degree resolution and includes 18 surface types. We have converted the 1/6° (~18.5 km) resolution to the ABI 2-km grid using the nearest grid method (Fig. 5). The surface type is fixed in time. The method for cloudy sky uses 4 surface types; these are also derived from 12 IGBP types (Table 2).

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153 2.3 Clear and cloudy sky simulations

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- 55 Under clear sky, multiple_scattering from aerosols is important. We have included 6 aerosol types (Table
- 156 3) to cover a range of possible conditions under clear sky. Aerosol models are selected based on the type
- 157 of extinction and a default meteorological range for the boundary-layer aerosol models as listed below:
- 158 Aerosol Type 1: Rural extinction, visibility = 23 km
- 159 Aerosol Type 4: Maritime extinction, visibility = 23 km
- 160 Aerosol Type 5: Urban extinction, visibility = 5 km
- 161 Aerosol Type 6: Tropospheric extinction, visibility = 50 km
- 162 Aerosol Type 8: Advective Fog extinction, visibility = 0.2 km
- 163 Aerosol Type 10: Desert extinction, visibility based on wind speed
- For the 6 aerosol types, the total number of MODTRAN simulations for each surface type is
- 462,000.288,000. It is obtained as follows: 6 aerosol types x 100 profiles x 770 angles.
- 166 When doing NTB simulation, we use all 6 types of aerosols. The Rural, Ocean, Urban and Fog aerosols
- 167 are distributed in the lower 0-2 km region. Tropospheric aerosol is distributed from 0 to 10 km tropopause.
- 168 The Rural, Ocean, Urban and Tropospheric aerosol optical properties have Relative Humidity (RH)
- 169 dependency. The Single Scattering Albedo (SSA) is given on 4 RH grids (0, 70, 80, 99) on a spectral grid

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 properties are given for 4 wind speeds (0, 10, 20, 30).

Simulations were performed for ABI for all the cloud cases described in **Table 3**. To merge cloud layers with atmospheric profiles we have followed the procedure as described in *Berk et al.* (1985, 1998), namely: "Cloud profiles are merged with the other atmospheric profiles (pressure, temperature, molecular constituent, and aerosol) by combining and/or adding new layer boundaries. Any cloud layer boundary within half a meter of an atmospheric boundary layer is translated to make the layer altitudes coincide; new atmospheric layer boundaries are defined to accommodate the additional cloud layer boundaries." 100% relative humidity is assumed within the cloud layers (default).

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180 2.4 Selection of angles

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The total number of angles used in the simulations is given in **Table 4**. The selected spectral grids for solar zenith angles, satellite view angles and <u>relative</u> azimuth angles are at Gaussian quadrature points, plus 0° to solar zenith angles (sza) and satellite viewing angles (vza) and 0° and 180° (forward and backward view) to the satellite relative azimuth angles. Solar angle and satellite view angle are referenced to target or surface for satellite simulation with 0° meaning looking up (zenith). <u>Relative a</u>Azimuth angle is defined as when the relative azimuth angle equals 180°, the sun is in front of observer.

The definitions of solar zenith angle and azimuth angle in this table corresponds to the definitions of MODTRAN but that is not the case for the satellite zenith angle. MODTRAN uses nadir angle as 180°satellite zenith angle, ignoring spherical geometry.

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192 2.5 Selection of optimal computational scheme

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- Computational speed is an issue for simulations that account for multiple scattering. MODTRAN4.3 provides three multiple scattering models (Isaacs, DISORT, and Scaled Isaacs) and three band models at
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resolutions (1 cm⁻¹, 5 cm⁻¹, and 15 cm⁻¹). The DISORT model (Stamnes et al., 1988) provides the most accurate radiance simulations but the runs are very time consuming. The Isaacs (Isaacs et al. 1987) 2stream algorithm is fast but oversimplified. The Scaled Isaacs method performs radiance calculations at a small number of atmospheric window wavelengths. The multiple scattering contributions for each method are identified and ratios of the DISORT and Isaacs methods are computed. This ratio is interpolated over the full wavelength range, and finally, applied as a multiple scattering scale factor in a spectral radiance calculation performed with the Isaacs method.

To optimize simulation speed and accuracy, we performed various sensitivity tests, including combinations of multiple scattering models, band resolution, and number of streams. **Table 5** lists simulation options and their corresponding calculation speed. The most computationally extensive option is DISORT 8-stream with 1 cm⁻¹ resolution which requires 930 seconds to finish one single run. The fastest is Scaled Isaacs with 15 cm⁻¹ resolution which only needs 6.67 seconds. Number of streams does not affect the Scaled Isaacs calculation speed. This is different from Isaacs and DISORT for which both stream number and band resolution have notable effects.

210 Based on results presented in **Table 5**, the efficient options (< 40 seconds) are Isaacs, DISORT 2-stream 211 with 15 cm⁻¹, DISORT 4-stream 15 cm⁻¹, and Scaled Isaacs all streams at all resolutions. Although the 212 ideal option is DISORT 8-stream with 1 cm⁻¹ resolution, there is a trade-off between speed and accuracy. 213 Fig. 6 compares DISORT simulated radiances at three band resolutions. We use two spectral ranges of $0.4 - 0.5 \,\mu\text{m}$ and $1.5 - 2.0 \,\mu\text{m}$ to illustrate the differences. Fig. 6 shows that the coarser band resolution 214 215 has smoothed out the radiance variations. The 15 cm⁻¹ has the smoothest curve among the three, and 1 216 cm⁻¹ shows more variations than the other two. Another (scientific) criteria for selecting the spectral 217 resolution is the ability to resolve/match the relative spectral response function (SRF) of a sensor. For 218 example, the SRFs of channels 1-6 of ABI are given at every 1 cm⁻¹.

219 Accordingly, we have chosen the 1 cm⁻¹ band model for the MODTRAN radiance simulations. Performed

220 were also radiance simulations from different multiple scattering models at 1 cm⁻¹ resolution. The whole

221 spectrum of $0.2 - 4 \,\mu\text{m}$ was separated to 14 sections so that the differences can be assessed clearly. For

222	wavelength below 0.3 µm and beyond 2.5 no discernible differences were found among Isaacs, DISORT
223	2-, 4-, and 8-strem, and Scaled Isaac. The largest differences occurred in the spectral range of $0.4 - 1.0$
224	μ m. Scaled Isaac 8-stream follows DISORT 8-stream closely across the whole spectral range; the Scaled
225	Isaac method provided near-DISORT accuracy with the speed of Isaacs. Thus, the MODTRAN4.3
226	simulations for GOES-R ABI were set-up with Scaled Isaac 8-stream with 1 cm ⁻¹ band resolution.
227	For illustration, in Fig. 7 compared are radiances simulated by Isaac 2 stream, Scaled Isaac, and DISORT-
228	4 stream for the case of Relative Azimuthal Angle=1.9°, View Angle=76.3°, Solar Zenith Angle=87.2°.
229	The lines are differences between various settings and DISORT-8 stream (e.g. Isaacs minus DISORT-8).
230	Isaac has the least accuracy since it is oversimplified, 4-stream showed some improvements when
231	compared with Isaac while still has large differences for 0.4 μ m and is still computationally demanding.
232	Scaled Isaac provides the smallest differences between DISORT-8. Fig. 6 (lower) zoomed in to the large
233	difference area of 0.3-0.35 µm which indicates that Scaled Isaacs still provides satisfactory results.
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235	2.6 Regression methodologies
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236 237 238 239	We have derived coefficients of regression using a constrained least-square curve fitting methods of Matlab, "lsqnonneg", which can solve a linear or nonlinear least-squares (data-fitting) problem and produce non-negative coefficients. Non-negative coefficients avoid generating negative TOA flux, which is not a physically valid.
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 236 237 238 239 240 241 242 243 	We have derived coefficients of regression using a constrained least-square curve fitting methods of Matlab, "lsqnonneg", which can solve a linear or nonlinear least-squares (data-fitting) problem and produce non-negative coefficients. Non-negative coefficients avoid generating negative TOA flux, which is not a physically valid. To ensure that information from all channels is used and avoid the complex cross-correlation problem, it was opted to generate Narrow to Broad (NTB) coefficients for each ABI channel separately (using "lsqnonneg"). These channel specific NTB coefficients are applied to each channel to convert ABI narrow-band reflectance to extended band. The final broad-band TOA reflectance is taken as the
 236 237 238 239 240 241 242 243 244 	We have derived coefficients of regression using a constrained least-square curve fitting methods of Matlab, "lsqnonneg", which can solve a linear or nonlinear least-squares (data-fitting) problem and produce non-negative coefficients. Non-negative coefficients avoid generating negative TOA flux, which is not a physically valid. To ensure that information from all channels is used and avoid the complex cross-correlation problem, it was opted to generate Narrow to Broad (NTB) coefficients for each ABI channel separately (using "lsqnonneg"). These channel specific NTB coefficients are applied to each channel to convert ABI narrow-band reflectance to extended band. The final broad-band TOA reflectance is taken as the weighted sum of all 6-channel specific broad-band reflectance. The logic behind this approach is the
 236 237 238 239 240 241 242 243 244 245 	We have derived coefficients of regression using a constrained least-square curve fitting methods of Matlab, "lsqnonneg", which can solve a linear or nonlinear least-squares (data-fitting) problem and produce non-negative coefficients. Non-negative coefficients avoid generating negative TOA flux, which is not a physically valid. To ensure that information from all channels is used and avoid the complex cross-correlation problem, it was opted to generate Narrow to Broad (NTB) coefficients for each ABI channel separately (using "Isqnonneg"). These channel specific NTB coefficients are applied to each channel to convert ABI narrow-band reflectance to extended band. The final broad-band TOA reflectance is taken as the weighted sum of all 6-channel specific broad-band reflectance. The logic behind this approach is the assumption that the narrow-band reflectance from each channel is a good representative for a limited

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We have derived coefficients of regression using a non-constrained and constrained least square curve fitting methods of Matlab "stepwisefit" and "lsqnonneg". The first one does <u>is a stepwise</u> regression by adding terms to and removing terms from a multilinear model based on their statistical significance. It may give negative coefficients that results in a negative TOA flux, which is not a physically valid result. Subsequently, we have re-derived all the coefficients with "lsqnonneg" which can solve a linear or nonlinear least squares (data fitting) problem and produce non-negative coefficients.

255 To ensure that information from all channels is used and avoid the complex cross-correlation 256 problem, it was opted to generate Narrow to Broad (NTB) coefficients for each ABI channel 257 separately (using "lsqnonneg"). These channel specific NTB coefficients are applied to each channel 258 to convert ABI narrow-band reflectance to extended band. The final broad-band TOA reflectance is 259 taken as the weighted sum of all 6-channel specific broad-band reflectance. The logic behind this 260 approach is the assumption that the narrow-band reflectance from each channel is a good 261 representative for a limited spectral region centered around the middle of the channel and the total 262 spectral reflectance is dominated by the spectral region that contains the most solar energy.

To generate "separate-channel" NTB coefficients, each narrow-band ABI channel reflectance is converted to a reflectance $\rho_{bb,l}$ separately,

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$$\rho_{bb,i}(\theta_0,\theta,\phi) = c_{0,i}(\theta_0,\theta,\phi) + c_{1,i}(\theta_0,\theta,\phi) * \rho_{nb,i}(\theta_0,\theta,\phi)$$
(4)

where $\rho_{bb,i}$ is the band reflectance for an interval around each channel *i*; $c_{0,i}$ and $c_{1,i}$ are regression coefficients for channel *i*. These regression coefficients are derived separately for various combination of surface, cloud and aerosol types \dot{z} The total shortwave broad band (0.25 – 4.0µm) reflectance ρ_{bb}^{est} is obtained by taking the weighted sum of all 6 $\rho_{bb,i}$ reflectance

270
$$\rho_{bb}^{est}(\theta_0, \theta, \phi) = \sum_i \rho_{bb,i}(\theta_0, \theta, \phi) \frac{s_{0,i}}{s_i}$$

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Here, S_0 and $S_{0,i}$ are total solar irradiance and band solar irradiance for each channel, respectively. Band edges around the six ABI channels are: 49980,-18723, 18723-13185, 13185-9221, 9221-6812, 6812-

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273	5292 , 2500 cm ⁻¹ (0.2001-0.5341, 0.5341-0.7584, 0.7584-1.0845, 1.0845-1.4680, 1.4680-1.8896,)		Commented [RTP1]: Something seems missing here.
274	<u>1.8896-4.0000 μm).</u> The corresponding band solar irradiance values are 364, 360, 287, 168, 91, 87	\bigtriangledown	Formatted: Highlight
275	W m ⁻² . Fig. 8 shows the sensor response function (SRF) and locations of the six ABI channels.		Formatted: Not Superscript/ Subscript
276 277	Coefficients are generated for clear condition and 3 types of cloudy conditions. Comparison between ABI TOA flux and CERES products are shown in Figure Fig. 9. The "separate-channel" coefficients work		Formatted: Font: (Default) +Headings CS (Times New Roman), Complex Script Font: +Headings CS (Times New Roman)
278	well for predominantly clear sky (Fig.10), Differences are somewhat more scattered for cloudy cases.	\mathbb{N}	Formatted: Line spacing: 1.5 lines
279 280	The reason may be due to the fact that the ABI observation time and CERES product time do not match perfectly since cloud condition change quickly. <u>As discussed in Gristey et al. (2019) there are SW spectral</u>		Formatted: Font: (Default) +Headings CS (Times New Roman), Bold, Complex Script Font: +Headings CS (Times New Roman), Bold
281 282	reflectance variations for different cloud types. Possibly, for ABI bands some spectral variations associated with cloud variability are missed. It is important to have the correct cloud properties to be able		Formatted: Font: (Default) +Headings CS (Times New Roman), Complex Script Font: +Headings CS (Times New Roman)
283	to select correct ADM. Misclassification of cloud properties will therefore result in flux differences. They		Formatted: Font: Bold, Complex Script Font: Bold
284 285	also argue that ADMs have an uncertainty due to within-scene variability and within-angular bin variability leading to additional flux differences.		Formatted: Font: (Default) +Headings CS (Times New Roman), Complex Script Font: +Headings CS (Times New Roman)
286 287	· · · · · · · · · · · · · · · · · · ·		Formatted: Font: (Default) +Headings CS (Times New Roman), 12 pt, Complex Script Font: +Headings CS (Times New Roman), 12 pt
288 289	3.0 Data used		Formatted: Font: (Default) +Headings CS (Times New Roman), Complex Script Font: +Headings CS (Times New Roman)
290	3.1 Satellite data for GOES-16 and GOES17		
291			
292	The GOES Imager data <u>used (Table 6)</u> were downloaded from <u>https://www.bou.class.noaa.gov/</u> and the		Formatted: Font: Bold
293	SRF from https://ncc.nesdis.noaa.gov/GOESR/ABI.php		
294			
295	* The CODC data were not always available from CLASS and had to be obtained from NOAA/STAR		
296	temporary archives. Also, not all the required angular information needed for implementation of		
297	regressions was available online and had to be recomputed.		

298 3.2 Reference data from CERES and-FLASHFlux Level2 (FLASH_SSF) Version 3C

299

300	Near real-time CERES fluxes and clouds in the SSF format are available within about a week of
301	observation (Kratz et al., 2014). They do not use the most recent CERES instrument calibration and thus
302	contains some uncertainty. Before GOES data were transferred to the Comprehensive Large Array-data
303	Stewardship System (CLASS) system, the NOAA/STAR archive was holding new data for about a week.
304	Therefore, the initial evaluations had to be done only with data that overlapped in time. The CERES data
305	known as the FLASHFlux Level2 (FLASH_SSF) were available almost in real time and did overlap with
306	GOES. These data were downloaded from:
307	https://ceres.larc.nasa.gov/products.php?product=FLASHFlux-Level2
307 308	https://ceres.larc.nasa.gov/products.php?product=FLASHFlux-Level2 Due to these limitations the early comparison was done between ABI data as archived at NOAA/STAR
308	Due to these limitations the early comparison was done between ABI data as archived at NOAA/STAR
308 309	Due to these limitations the early comparison was done between ABI data as archived at NOAA/STAR and the FLASHFlux products. The archiving of GOES-R at the NOAA Comprehensive Large Array-data
308 309 310	Due to these limitations the early comparison was done between ABI data as archived at NOAA/STAR and the FLASHFlux products. The archiving of GOES-R at the NOAA Comprehensive Large Array-data Stewardship System (CLASS) started only in 2019_however, it contains data starting from 2017. Once

313

314 **3.3 Data preparation**

~	315	For the re-mapping, we adopted the ESMF re-gridding package. The detailed information can be found	Formatted: Line spacing: Double

- 316 <u>at:</u>
- 317 <u>http://earthsystemmodeling.org/regrid/</u>

318	For an ideal situation, the ABI high-resolution TOA SW fluxes should be mapped into the CERES	
319	footprint for validation as suggested by the Reviewer. However, there are reasons that make it difficult to	
320	do so. For example, the case 12/26/2019 UTC 19. There can be more than 18000 pixels in a single swath	
321	of the SSF, when constrained to U.S. Different pixels have different times. Neglecting the seconds, there	
322	are still more than 30 mins differences (this changes case by case) between the first pixel and the one at	
323	the end and this brings up a time matching time issue. But if remapping the SSF to ABI, we can set up a	
324	unique time for ABI (ABI is at 5 min intervals) and then constrain the region and the time range of SSF.	
325	Both remapping the ABI to SSF and remapping SSF to the ABI bring up spatial matching errors as	
326	recognized by the scientific community. In Fig. 11, we show the SSF before re-gridding (Figs 11 (a) &	_
327	(b)) and after re-gridding (Figs. 11 (c) and (d)). The fluxes after re-mapping CERES SSF to the ABI	
328	resolution resemble well the original mapping. Another consideration is the computational efficiency of	
329	re-mapping the curvilinear tripolar grid to unconstructed grid. For large arrays, it is more efficient to	
330	remap the unconstructed grid to the curvilinear tripolar grid.	
331		
332	The CERES FLASHFlux_SSF data are re-gridded to match ABI spatial resolution by bi-linear	
333	interpolation method from the Earth System Modeling Framework (ESMF) package. The full description	
334	of the package can be found via http://earthsystemmodeling.org/regrid/#overview. The time difference	
335	between CERES FLASHFlux_SSF and GOES-16 data must be less than ±5 min. e.g., if the GOES-R	
336	scanning time is 18:51, then the seripts search the FLASHFLUX points between 18:46~18:56, and use	

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337	the re gridding method mentioned above to remap the FLASHFLUX to the GOES R (2 km) domain.	
338	Several cases will be illustrated.	
339	The statistics are based on all available points in overlap area. No outliers are removed. All sky, clear sky	
340	only, and cloudy only are compared for dates randomly selected. The hour was selected when both GOES-	
341	16 and GOES 17 had overlap with CERES FLASHFlux_SSF (Aqua/Terra) data. The coefficients for	
342	GOES-17 were obtained by replacing the GOES-16 spectral response function (SRF) by the GOES-17	
343	SRF. All the regressions have been repeated for GOES-17. The GOES-17 SRF was downloaded from	
344	https://ncc.nesdis.noaa.gov/GOESR/ABI.php. Simultaneous evaluation for both satellites was performed.	
345	The evaluations against the CERES FLASHFlux_SSF data is at footprint scale and covers one hour. The	
346	GOES-16 and 17 CONUS data have 5 min intervals, and there are 12 cases in one hour; this requires to	
347	test each case independently to find the best time match with CERES FLASHFlux_SSF.	
348		
349	4.0 Results	
350		
351	4.1 Comparison between ABI TOA fluxes to those from CERES SSF and/or FLASHFlux	
352	The CERES Single Scanner Footprint (SSF) is a unique product for studying the role of clouds, aerosols,	Formatted: Font: 12 pt, Complex Script Font: 12 pt
353	and radiation in climate. Each CERES footprint (nadir resolution 20-km equivalent diameter) on the SSF	
354	includes reflected shortwave (SW), emitted longwave (LW) and window (WN) radiances and top-of-	
355	atmosphere (TOA) fluxes from CERES with temporally and spatially coincident imager-based radiances,	
356	cloud properties, and aerosols, and meteorological information from a fixed 4-dimensional analysis	
357	provided by the Global Modeling and Assimilation Office (GMAO). Each file in this data product	

.

358	contains one hour of full and partial-Earth view measurements or footprints at a surface reference level.
359	Detailed information can be found via https://ceres.larc.nasa.gov/data/#ssf-level-2.
360	The FLASHFLUX is in footprint format thus it is a variable in time [flux (time)].
361	In the matching, points that fall in the ±5 min interval of the GOES R scanning time are used using
362	bilinear interpolation method to get the values for GOES R domain (e.g., if the GOES R scanning time
363	is 18:51, then the scripts search the FLASHFLUX points between 18:46-18:56, and use bilinear
364	interpolation method to do the remapping to GOES-R (2 km) domain). A case for 2019/12/26 (doy 360)
365	UTC 19:36 is illustrated in Figs. 1011-134. Statistical summaries from an extended number of cases are
366	presented in Table 7, and cover all four seasons.
367	The derivation and evaluation of TOA radiative fluxes as simulated for any given instrument are quite

36/ 368 challenging. In principle, there is a need to account for all possible changes in the atmospheric and surface 369 conditions one may encounter in the future. Yet, to know what these conditions are at the time of actual 370 observation when there is a need to select the appropriate combination of variables from the simulations, 371 is a formidable task. Therefore, error can be expected due to discrepancies between the actual conditions 372 and the selected simulations and these are difficult to estimate. The approach we have selected is based 373 on high-quality simulations using a proven and accepted radiative transfer code (MODTRAN) of known 374 configurations and a wide range of atmospheric conditions. We have also selected the best available 375 estimates of TOA radiative fluxes from independent sources for evaluation. However, the matching 376 between different satellites in space and time is challenging. In selecting the cases for evaluation, we have 377 adhered to strict criteria of time and space coincidence as described in section 3.3.

We have conducted several experiments to select an appropriate regression approach to the NTB transformation ensuring that non-physical results are not encountered. Based on the samples used in this study (Table 7) the differences found for Terra and GOES-16 were in the range of -0.5-(-12.10) for bias

and 43.28-82.09 for standard deviation; for Terra and GOES-17 they were 10.81-48.17 and 70.25-109.19,

382 respectively. For Aqua and GOES-16 they were 7.02-29.66 and 45.55-109.08 respectively while for Aqua

383 and GOES-17 they were 0.19-26 and 53.08-94.90, respectively (all units are W m⁻²). The evaluation

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process revealed the challenges in undertaking such comparisons. Both estimates of TOA fluxes (CERES and GOES) do no account for seasonality in the land use classification; the time matching for the different satellites is important and limits the number of samples that can be used in the comparison. Based on the results of this study recommendation for future work include the need to incorporate seasonality in land use and spectral characteristic of the various surface types. Possible stratification by season in the regressions could also be explored.

390

391 4.2 Causes for differences between ABI and CERES TOA fluxes

392 **4.2.1 Differences in surface spectral reflectance**

393

394 In the MODTRAN simulations we use the spectral reflectance information on various surface types as 395 provided by MODTRAN. MODTRAN version 4.3.1 contains a collection of spectral surface reflectance 396 dataset from the Moderate Spectral Atmospheric Radiance and Transmittance (MOSART) model 397 (Cornette et al., 1994) and others from Johns Hopkins University Spectral Library (Baldridge et al., 2009). 398 When doing simulation, we call the built-in surface types and use the provided surface reflectance. As 399 such, the spectral dependence of the surface reflectance used in the simulations and matched to the 400 CERES surface types may not be compatible with the classification of CERES. Also, seasonal changes 401 in surface type classification can invtotroduce errors dute to changes in the spectral surface reflectances 402 for different surface types (Fig. 145).,

403

404 4.2.2 Issues related to surface classification

405

Another possible cause for differences between the TOA fluxes is the classification of surface types as
 originally identified by the IGBP and used in the simulations. No seasonality is incorporated in the surface
 type classification and the impact can be illustrated in the following case study while such variability is
 part of the CERES observations.² Simulation results for surface type 8 (open shrub) have been checked

410	in depth. The average simulated broad band reflectance is around 0.2. The regression residual for this
411	surface type is reasonably small for sun angle <80 degrees, namely, the fitted broad band reflectance is
412	very close to the simulated broad band reflectance. This would indicate that the regressions are
413	performing properly. However, when we applied the regression coefficient to the GOES 16 ABI
414	observations, the calculated TOA broad band reflectance was around 0.45, which seemed too high. To
415	explain why the coefficient for channel 6 for "open shrub" was high we illustrate the filter function for
416	channel 6 and spectral albedos for open shrub, desert, woody savanna and grassland in Fig. 14.
417	In Fig. 15 we show the TOA fluxes for the entire domain using the original IGBP classification (open

shrub) in the area of interest and subsequent replacement with a desert surface. Due to seasonal changes
in surface properties, "Desert" classification may be more appropriate for the surface type at the time of
the observations. This would indicate the need for introducing seasonal variability in the classification of
surface types before one selects the representative NTB transformations.

422

423 4.2.3 Issues related to match-up between GOES-R and CERES

424

Both Terra and Aqua have sun-synchronous, near-polar circular orbits. Terra is timed to cross the equator from north to south (descending node) at approximately 10:30 am local time. Aqua is timed to cross the equator from south to north (ascending node) at approximately 1:30 pm local time. The periods for Terra and Aqua are 99 and 98 minutes, respectively. Both have 16 orbits per day. CERES on Terra and Aqua optical FOV at nadir is 16 x 32 or 20 km resolution. Terra passes CONUS during 03-06 UTC (US night time), 16-20 UTC (US day time), and Aqua passes CONUS during 07-11 UTC (US night time), 18-22 UTC (US day time).

432 Both Terra and Aqua have an instantaneous FOV values at SWATH level. There is no

433 perfect overlap, temporally or spatially with ABI data. The ABI radiance and cloud data are on a regular

434 grid of 2*2 km over CONUS at each hour. To use CERES data for evaluation of ABI, there is a need to

435 perform collocation in both time and space.

437 5.0 Summary

438

439 Critical elements of an inference scheme for TOA radiative flux estimates from satellite observations are:
440 1) transformation of narrowband quantities into broadband ones;

2) transformation of bi-directional reflectance into albedo by applying Angular Distribution Models 441 442 (ADMs). In principle, the order in which these transformations are executed is arbitrary. However, since 443 well established, observation-based broadband ADMs derived from the Clouds and the Earth's Radiant 444 Energy System (CERES) project already exist, the logical procedure is to do the NTB transformation on 445 the radiances first, and then apply the ADM. This is the sequence that has been followed here. While the 446 road map to accomplish above objectives seems well defined, reaching the final goal of having a stable 447 up-to-date procedure for deriving TOA radiative fluxes from a new instrument like the ABI on the new 448 generation of GOES satellites is quite complicated. The process of preparing for the usefulness of a new 449 satellite sensor needs to be done in advance, Since the final configuration of the instrument becomes 450 known at a much later stage. As such; the evaluation of the new algorithms is in a fluid stage for a long 451 time. Agreements early evaluation or disagreement with know-"ground truth" is not fully 452 informativeconclusive about on_the performance of the new algorithms, to estimate desired geophysical 453 parameters. Additional complication is related to the lack of maturity of basic information needed in the 454 implementation process, such as a reliable cloud screened product which in itself is in a process of 455 development and modifications. The "ground truth", namely, the CERES observations are also 456 undergoing adjustments and recalibration. As such, the process of deriving best possible estimates of 457 TOA radiative fluxes from ABI underwent numerous iterations to reach its current status. An effort was made to deal the best way possible with the fluid situation. All the evaluations against CERES were 458 459 repeated once the ABI data reached stability and were archived in CLASS and we used the most recent auxiliary information. The prominence of certain issues surfaced from this study itself. One example is 460 461 the sensitivity to land classification which currently is static. Another issue is related to the representation

- 462 of real time aerosol optical depth which is important under clear sky conditions. It is believed that only
- 463 now when NOAA/STAR has a stable aerosol retrieval algorithm, it would be timely to address the aerosol
- 464 issue in the estimation of TOA fluxes under clear sky.
- 465

- 466 Data availability. The data are available upon request from the corresponding author.
- 467 Author contributions. The investigation and conceptualization were carried out by RTP, IL and JD. YM
- and WC developed the software. RTP prepared the original draft. All authors contributed to the writing,
- 469 editing and review of the publication.
- 470 Competing interests. The authors declare that they have no conflict of interest.
- 471 Disclaimer. Publisher's note: Copernicus Publications remains neutral with regard to jurisdictional claims
- 472 in published maps and institutional affiliations.

473 Acknowledgements. We acknowledge the benefit from the use of the numerous data sources used in this-474 study. These include the Clouds and the Earth's Radiant Energy System (CERES) teams, the Fast

475 Longwave and Shortwave Radiative Flux (FLASHFlux) teams, the

476 University of Wisconsin-Madison, Space Science and Engineering Center, Cooperative Institute for 477 Meteorological Satellite Studies (CIMSS) for providing the SeeBor Version 5.0 data 478 (https://cimss.ssec.wisc.edu/training data/, and the final versions of the GOES Imager data were

downloaded from <u>https://www.bou.class.noaa.gov/</u>. Several individuals have been involved in the early stages of the project whose contribution led to the refinements of the methodologies. These include M.

481 M. Woncsick and Shuyan Liu. We thank the anonymous Reviewers for a very thorough and constructive

482 comments that helped to improve the manuscript. We thank the Editor Sebastian Schmidt for overseeing

483 <u>the disposition of the manuscript.</u>484

Financial support. This research was supported by NOAA/NESDIS GOES-R Program under grants
 5275562 1RPRP_DASR and 275562 RPRP_DASR_20 to the University of Maryland.

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Tables

Table 1. Relevant information for the derivation of SW fluxes from selected satellites:

Cehannel information and spectral bands for ABI.

ABI Band #	<u>Central</u> <u>wavelength (</u> بسر <u>)</u>	Spectral band (µm)	Formatted Table	
1	VIS 0.47	0.45-0.49	Field Code Changed	
2	VIS 0.64	0.60-0.68		
3	VIS <u>NIR</u> 0.86	0.847-0.882		
4	NIR 1.38	1.366-1.380		
5	NIR 1.61	1.59-1.63		
6	NIR 2.26	2.22-2.27		

578

579

581 Table 2. Surface classification description for IGBP 18 types, IGBP 12 types, CERES clear sky 6

types, and NTB c	cloudy sky 4 types			
ICDD(19, trm co)	ICDD (12 tymes)	CERES clear-sky	NTB cloudy-sky	
IGBP (18 types)	IGBP (12 types)	<u>(6 types)</u>	<u>(4 types)</u>	
Evergreen				
<u>Needleleaf</u>	Needleleaf Forest			
Deciduous				
<u>Needleleaf</u>				
Evergreen Broadleaf	Broadleaf Forest	Mod-High Tree/Shrub		
Deciduous Broadleaf			Land	
Mixed Forest	Mixed Forest	-		
Closed Shrublands	Closed Shrub	-		
Woody Savannas	Woody Savannas	-		
<u>Savannas</u>	<u>Savannas</u>			
Grasslands				
Permanent Wetlands	Grasslands	Low-Mod Tree/Shrub		
<u>Tundra</u>	Orassiands	Low-Wod Tree/Sillub		
<u>Croplands</u>	Croplands			
Open Shrublands	Open Shrub			
<u>Urban and Built-up</u>	Open Shrub	Dark Desert	Desert	
Bare Soil and Rocks	Barren and Desert	Bright Desert		
Snow and Ice	Snow and Ice	Snow and Ice	Snow and Ice	
Water Bodies	Ocean	Ocean	Water	

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.

Table 2. Surface classification description for IGBP 18 types, IGBP 12 types, CERES clear sky 6

types, and NTB cloudy sky 4 types

IGBP (18 types)	IGBP (12 types)	CERES clear-sky	NTB cloudy-sky		
IGBF (16 types)	tobr (12 types)	(6 types)	(4 types)		
Evergreen	Needleleaf Forest				
Needleleaf	reculeicur r orest				
Evergreen Broadleaf	Broadleaf Forest	-			
Deciduous	Needleleaf Forest	- Mod-High Tree/Shrub			
Needleleaf	reculeicur r orest	Wod-Ingh Tree/Sillub	Land		
Deciduous Broadleaf	Broadleaf Forest	-			
Mixed Forest	Mixed Forest	1			
Closed Shrublands	Closed Shrub				
Open Shrublands	Open Shrub	Dark Desert			
Woody Savannas	Woody Savannas	Mod-High Tree/Shrub			
Savannas	Savannas				
Grasslands	Grasslands	- Low Mod Tree/Shrub			
Permanent Wetlands	Grassianus	Low Woo Hee/Shrub			
Croplands	Croplands	-			
Urban and Built up	Open Shrub	Dark Desert	Desert		
Cropland Mosaics	Croplands	Low-Mod Tree/Shrub	Land		
Snow and Ice	Snow and Ice	Snow and Ice	Snow and Ice		
Bare Soil and Rocks	Barren and Desert	Bright Desert	Desert		
Water Bodies	Ocean	Ocean	Water		
Tundra	Grasslands	Low-Mod Tree/Shrub	Land		

Parameter	Clear condition	Cloudy condition
	6 aerosol types	3 cloud types
Aerosol or cloud type	(rural, maritime, urban,	(cirrus, stratocumulus, altostratus)
	tropospheric, fog, desert)	
	Typical VIS (km) values for	Cirrus: [0, 0.8, 1.2, 1.8, 3.2]
	each aerosol types (no OD grid	Stratocumulus: [0, 0.8, 1.2, 1.8,
	for each aerosol type).	3.2, 5.8, 8.2, 15.8, 32.2, 51.8,
Optical depth (OD)	Rural: 23, maritime: 23, urban:	124.2]
	5, tropospheric: 50, fog: 0.2,	Altostratus: [0, 15.0, 30.0, 50.0,
	desert: (default VIS for wind	80.0]
	speed 10m/s)	
Surface type	12 IGBP surface types	4 types (Water, Land, Desert,
Surface type		Snow/Ice)

591 Table 3. The various classes for which NTB coefficients are generated.

596 Table 4. Angles used in simulations. To be consistent with what is presented in the

597 ABI Shortwave Radiation Budget (SRB) Algorithm Theoretical Basis Documents (ATBD) (Laszlo

598 et al, 2018) the additional angles used in the simulations are not given in this Table.

Angle Type	Angles	Formatted Table
Solar Zenith Angle [°]	0.0, 12.9, 30.8, 41.2, 48.3, 56.5, 63.2, 69.5, 75.5, 81.4, 87.2	
Satellite Zenith Angle [°]	0.0, 11.4, 26.1, 40.3, 53.8, 65.9, 76.3	
Azimuth Angle [°]	0.0, 1.9, 10.0, 24.2, 44.0, 68.8, 97.6, 129.3, 162.9, 180	

Table 5. N	IODTRAN	l sim	ulatior	n speed	test	(CPI	J MH	Iz 2099	.929).

Algorithm	Stream	Band Resolution (cm ⁻¹)	Speed (~seconds)
Isaacs	2	1	40
DISORT	2	1, 5, 15	280, 70, 30
	4	1, 5, 15	560, 120, 40
	8	1, 5, 15	930, 300, 110
Scaled	2	1, 5, 15	30, 10, 6.67
Isaac	4	1, 5, 15	30, 10, 6.67
	8	1, 5, 15	30, 10, 6.67

609						
	Short Name	Long Name	MODE	ABI-Channel	Scan Sector	Spatial Resolution
	RadC	L1b Radiance	M6	C01-C06	CONUS	5000x3000
	AODC	L2 Aerosol	M6		CONUS	2500x1500
	ACMC	L2 Clear Sky	M6		CONUS	2500x1500
		Masks				
	ACTPC	L2 Cloud Top	M6		CONUS	2500x1500
		Phase				
	CODC*	L2 Cloud	M6		CONUS	2500x1500
		Optical Depth				

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Table 6. Details on data used as input for calculations.

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613	Table 7. Statistical summary for all selected cases intercompared at instantaneous time
614	scale.

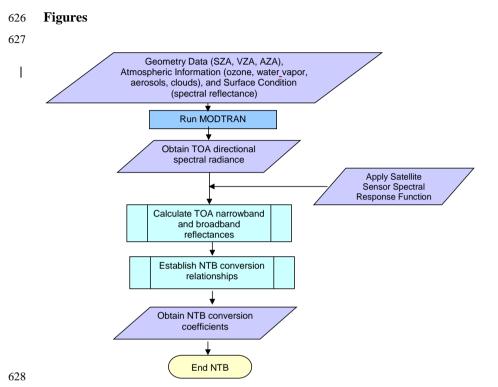
<u>Case</u>	<u>CERES</u>	<u>GOES-</u> <u>R</u>	<u>Corr</u>	<u>Bias</u>	<u>Std</u>	<u>RMSE</u>	N
<u>07/31</u>		<u>G16</u>	<u>0.82</u>	<u>0.81</u>	<u>69.81</u>	<u>69.81</u>	<u>0.22 x10⁶</u>
<u>2019</u>	<u>Terra</u>	<u>G17</u>	<u>0.87</u>	<u>29.13</u>	<u>90.10</u>	<u>94.70</u>	<u>1.78 x10⁶</u>
<u>UTC</u>		<u>G16</u>	<u>0.76</u>	<u>33.87</u>	<u>117.43</u>	<u>122.22</u>	<u>1.58 x10⁶</u>
<u>19</u>	<u>Aqua</u>	<u>G17</u>	<u>0.78</u>	<u>31.53</u>	<u>129.42</u>	<u>133.21</u>	<u>0.29 x10⁶</u>
00/12	<u>Terra</u>	<u>G16</u>	<u>0.87</u>	<u>-17.37</u>	<u>81.72</u>	<u>83.54</u>	<u>0.13x10⁶</u>
<u>09/13</u> 2019		<u>G17</u>	<u>0.71</u>	<u>47.09</u>	<u>108.73</u>	<u>118.48</u>	<u>1.73x10⁶</u>
<u>UTC</u> <u>20</u>	<u>Aqua</u>	<u>G16</u>	<u>0.76</u>	<u>18.22</u>	<u>108.50</u>	<u>110.02</u>	<u>1.46x10⁶</u>
		<u>G17</u>	<u>0.73</u>	<u>25.14</u>	<u>81.95</u>	<u>85.72</u>	<u>0.53x10⁶</u>
00/21	<u>Terra</u>	<u>G16</u>	<u>0.85</u>	<u>6.78</u>	<u>66.66</u>	<u>67.00</u>	<u>0.35x10⁶</u>
<u>09/21</u> <u>2019</u> <u>UTC</u> <u>19</u>		<u>G17</u>	<u>0.83</u>	<u>26.41</u>	<u>87.64</u>	<u>91.57</u>	<u>1.75x10⁶</u>
	<u>Aqua</u>	<u>G16</u>	<u>0.82</u>	<u>29.66</u>	<u>105.09</u>	<u>109.20</u>	<u>1.67x10⁶</u>
		<u>G17</u>	<u>0.76</u>	<u>6.03</u>	<u>94.70</u>	<u>94.89</u>	<u>0.15x10⁶</u>
<u>09/30</u> 2019 UTC 19	<u>Terra</u>	<u>G16</u>	<u>0.88</u>	<u>4.49</u>	<u>64.79</u>	<u>64.94</u>	<u>0.40x10⁶</u>
		<u>G17</u>	<u>0.80</u>	<u>19.35</u>	<u>86.41</u>	<u>88.55</u>	<u>1.74x10⁶</u>
		<u>G16</u>	<u>0.80</u>	<u>19.87</u>	<u>100.45</u>	<u>102.40</u>	<u>1.69x10⁶</u>
	<u>Aqua</u>	<u>G17</u>	<u>0.72</u>	<u>2.71</u>	<u>91.79</u>	<u>91.83</u>	<u>0.12x10⁶</u>

<u>10/23</u>	<u>Terra</u>	<u>G16</u>	<u>0.86</u>	<u>5.84</u>	<u>51.44</u>	<u>51.77</u>	<u>0.35x10⁶</u>
2019	Tena	<u>G17</u>	<u>0.87</u>	<u>22.47</u>	<u>70.25</u>	<u>73.76</u>	<u>1.75x10⁶</u>
<u>UTC</u>	Aqua	<u>G16</u>	<u>0.89</u>	<u>17.10</u>	<u>75.95</u>	<u>77.85</u>	1.67×10^{6}
<u>19</u>		<u>G17</u>	<u>0.78</u>	<u>8.98</u>	<u>72.52</u>	<u>73.07</u>	<u>0.15x10⁶</u>
<u>11/08</u>	<u>Terra</u>	<u>G16</u>	<u>0.87</u>	<u>-0.5</u>	<u>43.28</u>	<u>43.28</u>	<u>0.35x10⁶</u>
<u>2019</u>		<u>G17</u>	<u>0.82</u>	<u>17.18</u>	<u>71.27</u>	<u>73.31</u>	<u>1.75x10⁶</u>
<u>UTC</u>	<u>Aqua</u>	<u>G16</u>	<u>0.90</u>	<u>10.08</u>	<u>71.27</u>	<u>71.98</u>	<u>1.67x10⁶</u>
<u>19</u>		<u>G17</u>	<u>0.68</u>	<u>1.53</u>	<u>47.55</u>	<u>47.58</u>	<u>0.15x10⁶</u>
11/24	<u>Terra</u>	<u>G16</u>	<u>0.79</u>	<u>7.98</u>	<u>49.10</u>	<u>49.75</u>	<u>0.35x10⁶</u>
<u>11/24</u> <u>2019</u>		<u>G17</u>	<u>0.87</u>	<u>14.10</u>	<u>78.35</u>	<u>79.61</u>	<u>1.76x10⁶</u>
<u>UTC</u> <u>19</u>	Aqua	<u>G16</u>	<u>0.82</u>	<u>7.63</u>	<u>58.68</u>	<u>59.17</u>	<u>1.67x10⁶</u>
		<u>G17</u>	<u>0.65</u>	<u>0.19</u>	<u>63.14</u>	<u>63.14</u>	<u>0.15x10⁶</u>
<u>12/26</u> <u>2019</u> <u>UTC 19</u>	Terra	<u>G16</u>	<u>0.88</u>	<u>5.24</u>	<u>53.28</u>	<u>53.54</u>	<u>0.35x10⁶</u>
		<u>G17</u>	<u>0.76</u>	<u>11.26</u>	<u>73.95</u>	<u>74.80</u>	<u>1.76x10⁶</u>
	<u>Aqua</u>	<u>G16</u>	<u>0.83</u>	<u>9.79</u>	<u>58.90</u>	<u>59.56</u>	1.67×10^{6}
		<u>G17</u>	<u>0.73</u>	<u>0.85</u>	<u>52.53</u>	<u>52.54</u>	<u>0.15x10⁶</u>

Case	CERES	GOES- R	Corr	Bias	Std	RMSE	N
	Terra	G16	0.87	-12.10	82.09	82.98	0.13x10 6
09/13 2019		G17	0.71	4 8.17	108.19	118.42	1.73x10
UTC 20	Aqua	G16	0.76	17.38	109.08	110.45	1.46x10
20		G17	0.73	26.00	81.96	85.98	0.53x10 4
	T	G16	0.85	6.78	66.66	67.00	0.35x10 ¢
09/21 2019	Terra	G17	0.83	26.41	87.64	91.57	1.75x10 ¢
UTC 19	Aqua	G16	0.82	29.66	105.09	109.20	1.67x10 ¢
		G17	0.76	6.03	94.70	94.89	0.15x10 ¢
	Terra	G16	0.88	4.4 9	64.79	64.94	0.40x10
09/30 2019		G17	0.80	19.35	86.41	88.55	1.74x10
UTC 19	Aqua	G16	0.81	19.99	99.98	101.96	1.67x10 4
		G17	0.70	1.22	94.90	94.91	0.12x10 ¢
	Terra	G16	0.86	5.84	51.44	51.77	0.35x10 6
10/23 2019		G17	0.87	22.47	70.25	73.76	1.75x10 6
UTC 19	Aqua	G16	0.89	17.10	75.95	77.85	1.67x10 6
		G17	0.78	8.98	72.52	73.07	0.15x10 6
11/08	Terra	G16	0.87	-0.5	4 3.28	4 3.28	0.35x10 6
2019		G17	0.82	17.18	71.27	73.31	1.75x10 6

621	Table 7 Statistical summa	ry for all selected	Leases intercompared at	instantaneous time
Ψ <u>2</u> 1	Tuble 7. Statistical Summa	if y for all selected	e cuses intercompared at	instantaneous time
622	seale.			

UTC 19	Aqua	G16	0.90	10.08	71.27	71.98	1.67x10⁶
		G17	0.68	1.53	47.55	47.58	0.15x10⁶
11/24 2019 UTC 19	Terra	G16	0.79	7.98	49.10	49.75	0.35x10⁶
		G17	0.87	14.10	78.35	79.61	1.76x10 6
	Aqua	G16	0.82	7.63	58.68	59.17	1.67x10⁶
		G17	0.65	0.19	63.14	63.14	0.15x10 6
12/26 2019 UTC 19	Terra	G16	0.89	7.6	52.79	53.33	0.35x10 6
		G17	0.77	10.81	73.1 4	73.93	1.76x10 6
	Aqua	G16	0.83	7.02	59.16	59.58	1.67x10 6
		G17	0.73	-1.09	53.08	53.09	0.15x10⁶



629 Figure 1. Flowchart of the NTB transformations illustrating the main processing sections.

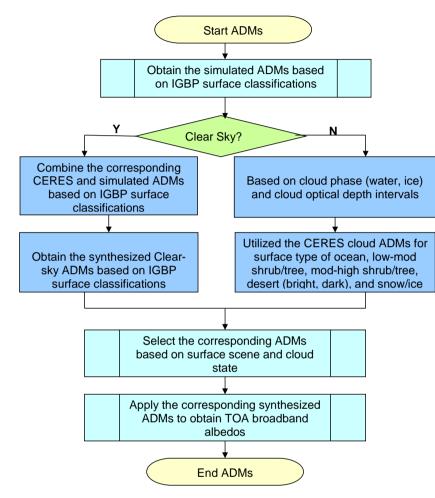


Figure 2. Schematic illustration of the logic employed to synthesize modeled and observed ADMs.

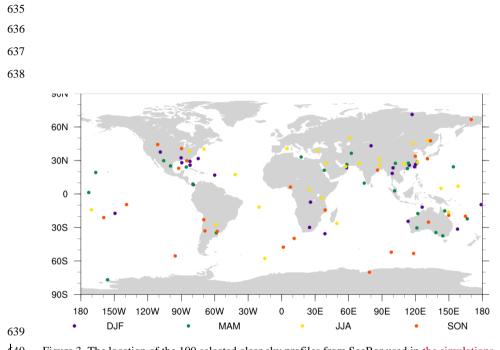
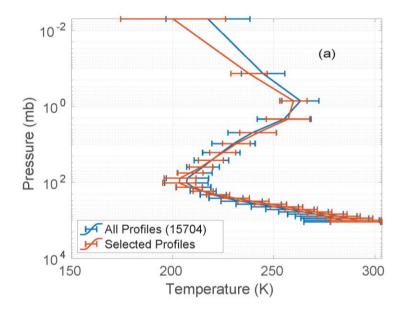
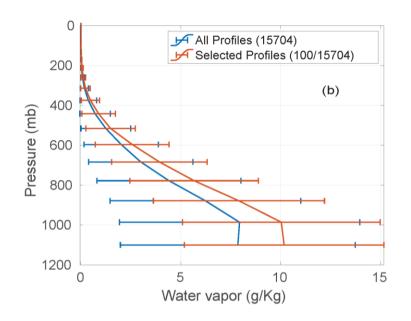
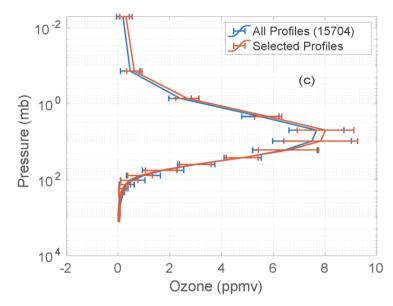


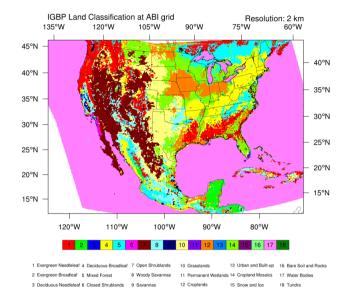
Figure 3. The location of the 100 selected clear sky profiles from SeeBor used in <u>the simulations</u>.







645
646
647 Figure 4. Profile statistics of: (a) temperature; (b): water vapor; (c) ozone <u>for</u> the entire available
648 sample and <u>for</u> the reduced sample used in this study. Error bar is 1 standard deviation. (logarithmic
649 scale).
650



653 Figure 5. Re-mapped IGBP surface classifications over the CONUS at 2-km ABI grid.

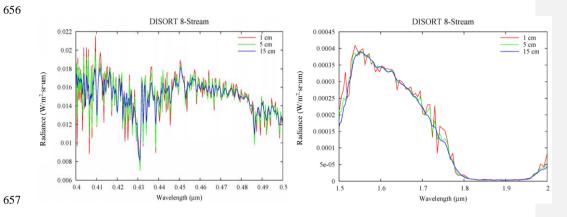
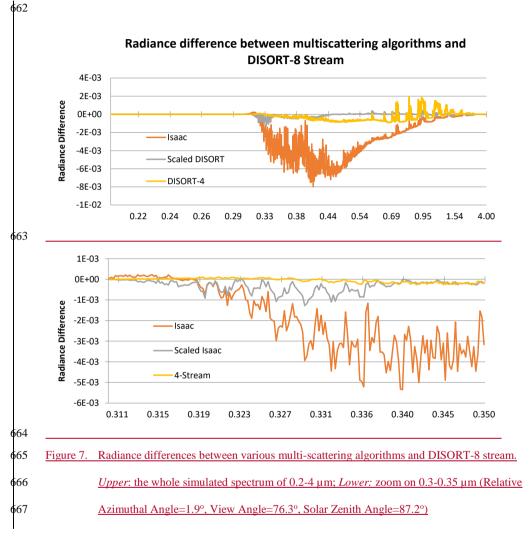
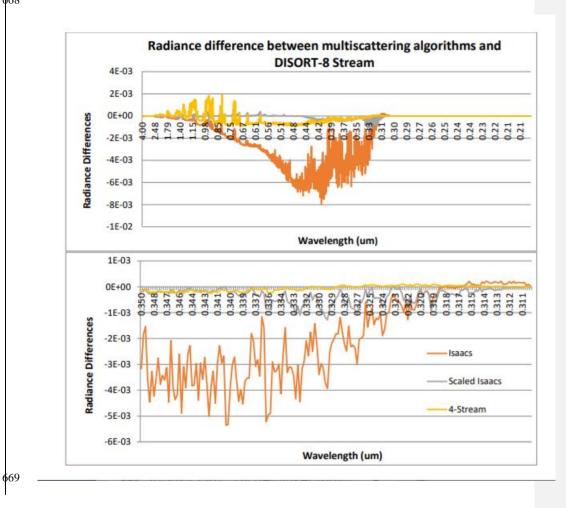


Figure 6. Simulated Radiances from DISORT 8-stream (with 1, 5, and 15 cm⁻¹ resolution band model for spectral range of $0.4 - 0.5 \,\mu$ m (left) and $1.5 - 2.0 \,\mu$ m (right).





- \$70 Figure 7. Radiance differences between various multi scattering algorithms and DISORT 8 stream.
- 671 Upper: the whole simulated spectrum of 0.2 4 μm; Lower: zoom on 0.3 0.35 μm (Relative Azimuthal
- 672 Angle=1.9°, View Angle=76.3°, Solar Zenith Angle=87.2°).
- 673

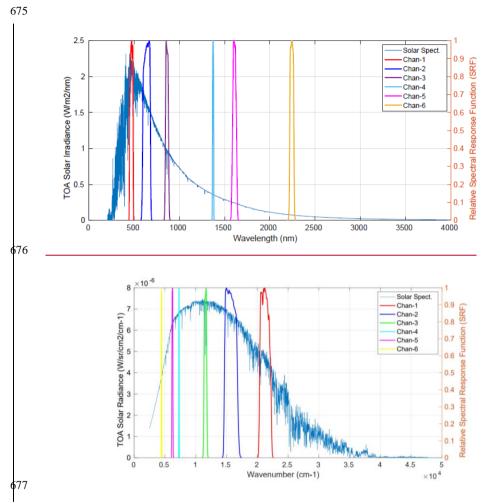
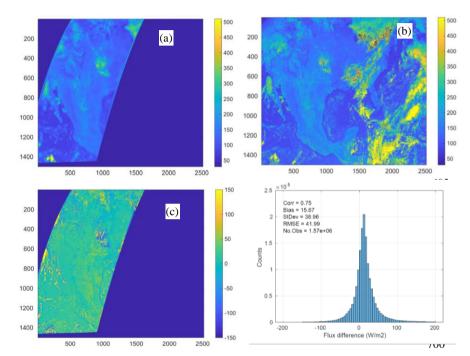
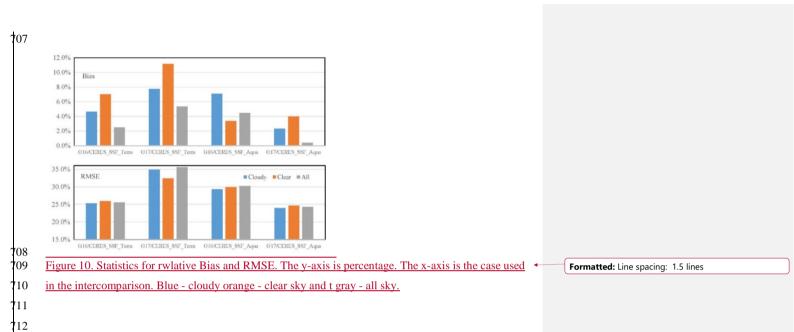
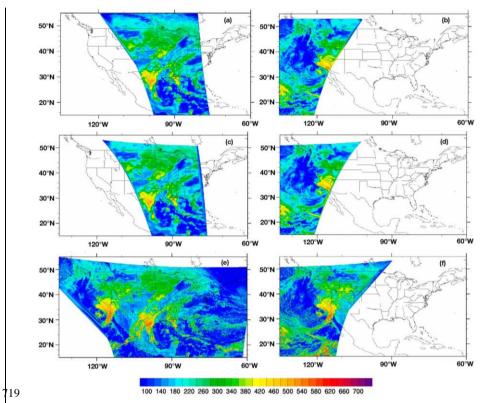


Figure 8. Locations of the six ABI channel SRFs. X-axis is wavenumber. Y-axis is solar irradiance.



- Figure 9. Comparison of TOA flux from ABI and CERES based FLASHFlux for 2017/11/25, 17:57Z.
- 702 (a) CERES Terra product; (b): results with "separate-channel" coefficients. (c): difference (ABI-
- CERES); (d): histogram of ABI-CERES differences (this is the only case illustrated in this paper with
- 704 <u>data from FLASHFlux)</u>-
- 705
- 706

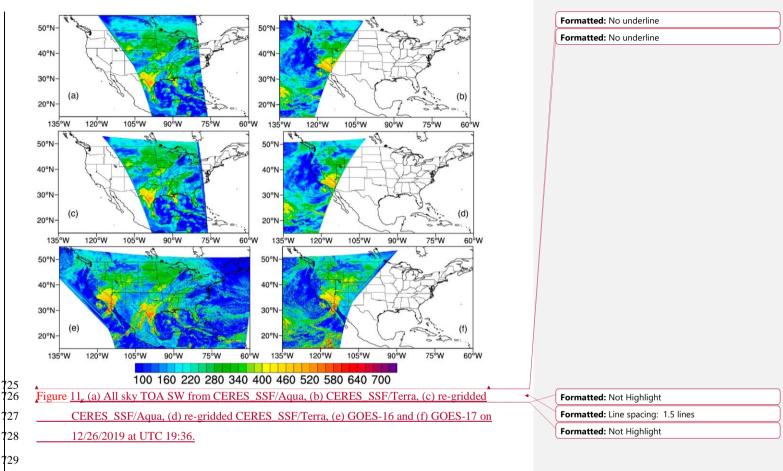


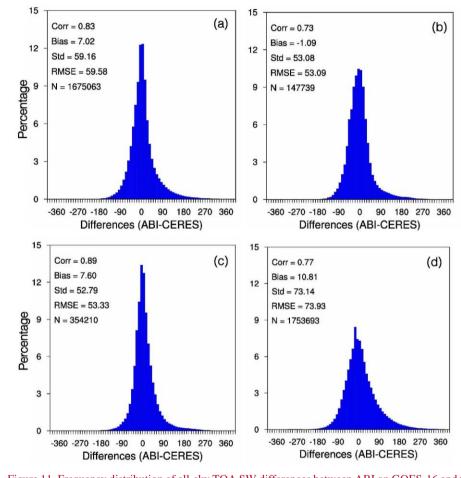




gridded CERES FLASHFlux/Aqua (c), CERES FLASHFlux/Terra GOES-16 (d) and GOES-17 (f) on

723 12/26/2019 at UTC 19:36.

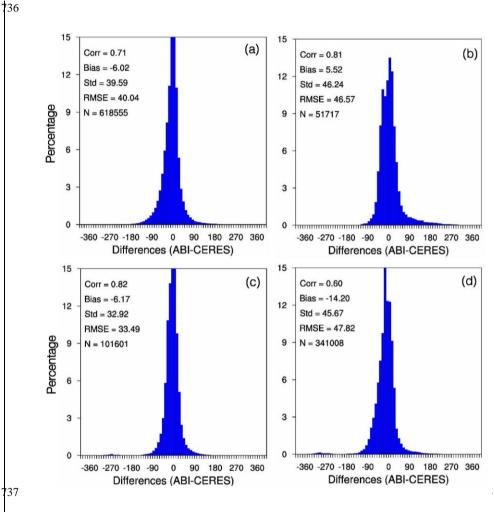




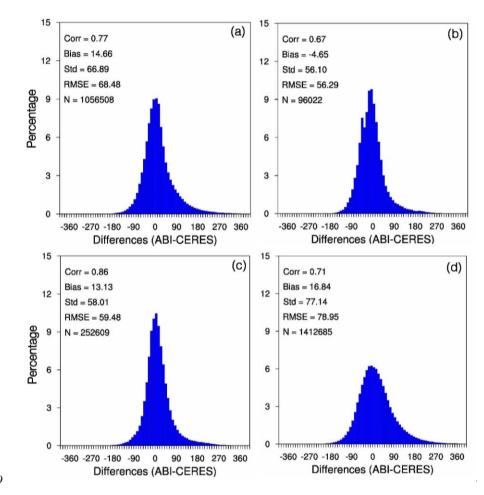




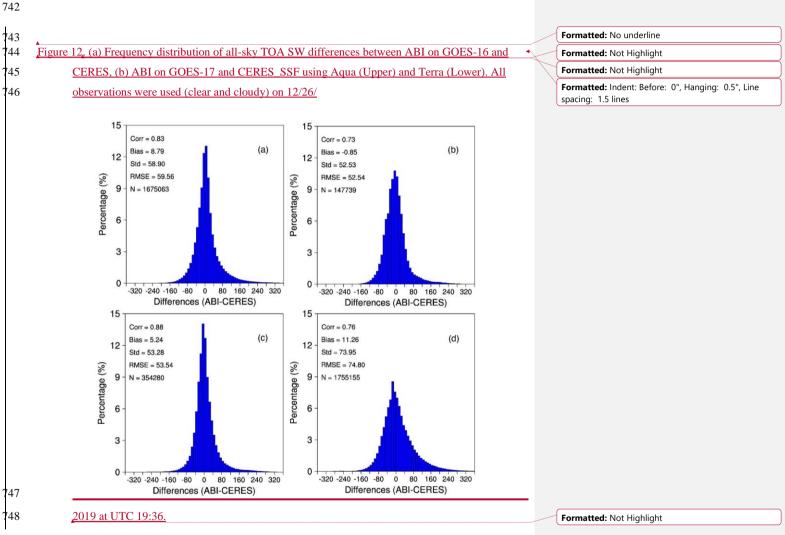
were used (clear and cloudy) on 12/26/2019 at UTC 19:36.

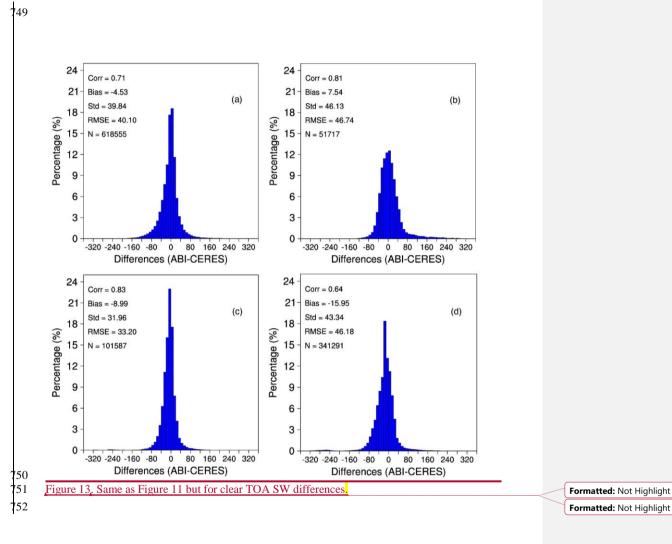


738 Figure 12. Same as Figure 11 but for clear TOA SW differences.

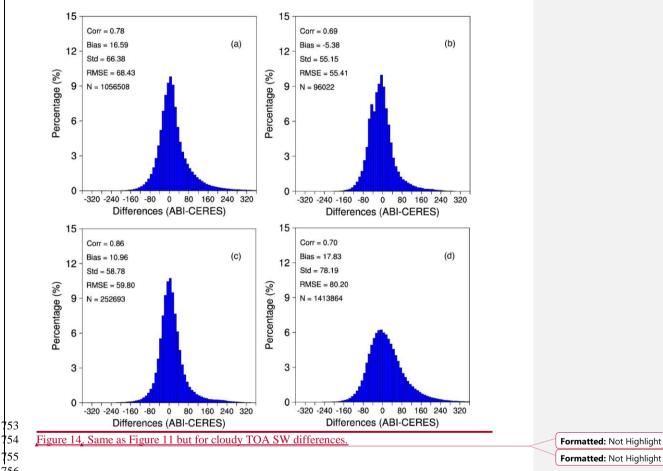


740 Figure 13. Same as Figure 11 but for cloudy TOA SW differences.









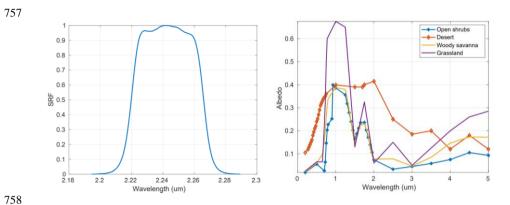
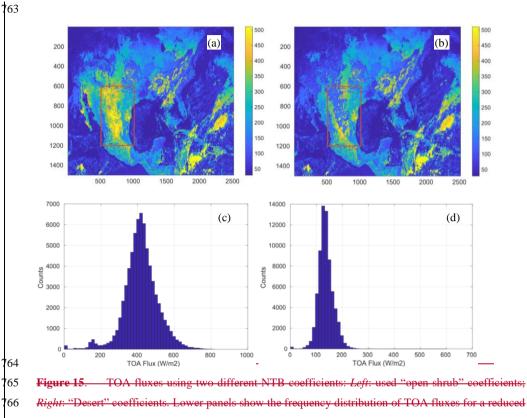


Figure 1415. Left: Sensor response function for ABI channel 6; *Right:* Spectral albedo for desert and open shrubs. Desert albedo value is much higher than open shrubs at 2.2 μm.



domain (over Mexico in the orange boxes) that includes the open shrub/desert classification. Case time

- 768 stamp is 2017/11/25 17:32Z.

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