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2 Surface Shortwave Radiative Fluxes derived from the US Air Force  
3 Cloud Depiction Forecast System World-Wide Merged Cloud Analysis  
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5 Rachel T. Pinker,<sup>a</sup> Wen Chen,<sup>a</sup> Yingtao Ma,<sup>a</sup> Sujay Kumar,<sup>b</sup> Jerry Wegiel,<sup>b, c</sup> Eric Kemp<sup>b, d</sup>  
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7 <sup>a</sup> Department of Atmospheric and Oceanic Science, University of Maryland, College Park, MD

8 <sup>b</sup> Hydrological Sciences Lab, NASA GSFC, Greenbelt, MD

9 <sup>c</sup> Science Applications International Corporation, McLean, VA

10 <sup>d</sup> Science Systems and Applications, Inc, Lanham, MD  
11

12 Corresponding Author: pinker@atmos.umd.edu

13 ABSTRACT: We present a global scale evaluation of surface shortwave (SW↓) radiative fluxes  
14 as derived with cloud amount information from the US Air Force Cloud Depiction Forecast System  
15 (CDFS) II World-Wide Merged Cloud Analysis (WWMCA) and implemented in the framework  
16 of the NASA Land Information System (LIS). Evaluation of this product is done against ground  
17 observations, a satellite-based product from the Moderate Resolution Imaging Spectroradiometer  
18 (MODIS), and several reanalysis outputs. While the LIS/US Air Force (USAF) product tends to  
19 overestimate the SW↓ fluxes when compared to ground observations and satellite estimates, its  
20 performance is comparable or better than the following reanalysis products: ERA5, CFSR and  
21 MERRA-2. Results are presented using all available observations over the globe and  
22 independently for several regional domains of interest. When evaluated against ground  
23 observations over the globe the bias in the LIS/USAF product at daily time scale was about 9.34  
24  $\text{Wm}^{-2}$  and the rms was 29.20  $\text{Wm}^{-2}$  while over the USA the bias was about 10.65  $\text{Wm}^{-2}$  and the  
25 rms was 35.31  $\text{Wm}^{-2}$ , respectively. The sample sizes used were not uniform over the different  
26 regions and the quality of both ground truth and the outputs of the other products may vary  
27 regionally. It is important to note that the LIS/USAF is a Near-Real-Time (NRT) product of  
28 interest for potential users and as such fills a need that is not met by most products. Due to latency  
29 issues, the level of observational inputs in the NRT product is less than in the reanalysis data.  
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31 KEYWORDS: USAF surface radiation; evaluation of LIS/USAF surface radiation; comparison

32 of LIS/USFA radiation with re-analysis model data.

33

34 SIGNIFICANCE STATEMENT: We evaluate a current scheme to produce surface radiative  
35 fluxes in the NASA Land Information System (LIS) framework as driven with cloud amount  
36 information from the US Air Force Cloud Depiction Forecast System (CDFFS) II World-Wide  
37 Merged Cloud Analysis (WWMCA). The LIS/USAF product is provided at Near-Real-Time  
38 (NRT) and as such, fills a need that is not met by most products. Information used for evaluation  
39 are ground observations, Moderate Resolution Imaging Spectroradiometer (MODIS) satellite-  
40 based estimates, and independent outputs from several reanalysis. Since the various LIS products  
41 are used by the hydrometeorology community, this manuscript should be of interest to the users  
42 of the LIS/US Air Force (AF) information on surface radiative fluxes.

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#### 44 **1. Introduction**

45 Modeling land surface processes at global scale at high spatial resolutions is challenging.  
46 Efforts to do so have progressed gradually from models with time-fixed soil moisture to Bucket  
47 Models (Manabe 1969) with time- and space-varying soil moisture, to Big-Leaf models (Deardorff  
48 1978) with explicit vegetation treatment, to the development of more sophisticated models  
49 including hydrological, biophysical, biochemical, ecological processes. Examples are the  
50 pioneering work of Sellers et al. (1986) who introduced the simple biosphere model (SiB), the  
51 Biosphere-Atmosphere Transfer Scheme (BATS) (Dickinson et al. 1993), the simplified Simple  
52 Biosphere model (SSiB) (Xue et al. 1991), and the Mosaic Model (Koster and Suarez 1992). The  
53 integration of land surface simulations, observation, and analysis methods to accurately determine  
54 land surface energy and moisture states led to such accomplishments as the 25 km Global Land  
55 Data Assimilation System (GLDAS) (Rodell et al. 2004) and the 12.5 km North American Land  
56 Data Assimilation System (NLDAS) (Mitchell et al. 2004). The NASA Land Information System  
57 (LIS) (Kumar et al. 2006, 208a, 2008b, 2013) represents a step forward by taking advantage of  
58 technological improvements in implementing Land surface Models (LSMs) at high spatial  
59 resolution and by enabling land data assimilation (Arsenault et al. 2018). Consequently, the NASA  
60 LIS became a widely used land data assimilation system that runs several LSMs with observation-  
61 based on meteorology and remote sensing data to generate high quality estimates of land surface  
62 conditions. In LIS, land surface and atmosphere are linked to each other over a variety of time

63 scales through the exchanges of water, energy, and carbon. An accurate representation of land  
64 surface processes is critical for improving models of the boundary layer and land/atmosphere  
65 coupling at all spatial and temporal scales and over heterogeneous domains. Configurations of LIS  
66 are used in operational environments at various agencies, including the US Air Force. Establishing  
67 the quality of the radiative forcing fields in LIS and their standing in respect to those from other  
68 well-established reanalysis models is a critical step in the development of improved  
69 representations of surface energy and water budget partitions.

70 The primary objective of this study is to evaluate a current NRT scheme in the LIS  
71 framework that produces surface radiative fluxes as driven with cloud information from the US  
72 Air Force Cloud Depiction Forecast System (CDFFS) II World-Wide Merged Cloud Analysis  
73 (WWMCA) (D'Entremont et al., 2016) (LIS/USAF). This can serve as a basis for evaluating future  
74 modifications of the LIS/USAF product such as replacing the cloud amount information with  
75 Fields of Cloud Optical Depth (COD) from the same WWMCA system. For all the products used  
76 in this study the evaluation is done against ground observations at available sites. The primary tool  
77 used for comparisons at global scale is a satellite-based inference scheme described in Wang and  
78 Pinker (2009) with subsequent modifications as detailed in section 3.0. The inference scheme is  
79 driven with cloud optical parameters from the MODIS instrument on Terra and Aqua that are  
80 similar in nature to those that are generated by the US Air Force WWMCA product. The  
81 performance of the MODIS satellite product was first established against ground observations.

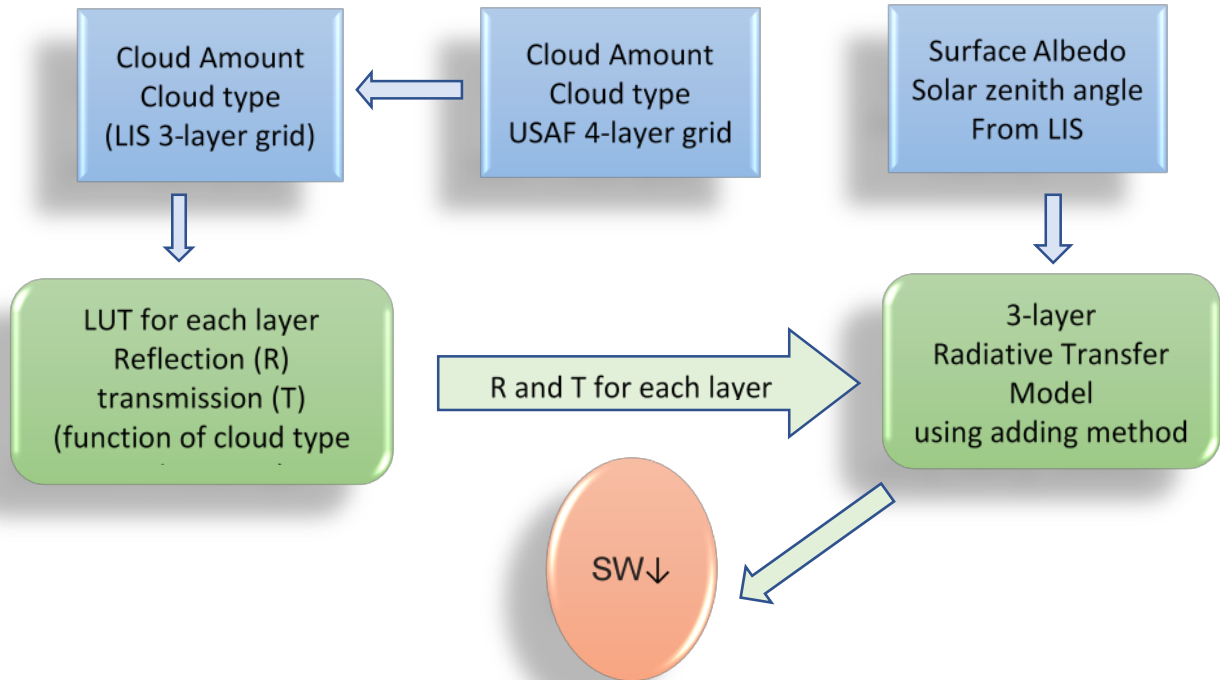
82 In section 2 we describe the current LIS/USAF scheme to derive surface  $SW\downarrow$  fluxes as  
83 driven with information on clouds from the WWMCA; in section 3 described is the UMD MODIS  
84 SW scheme; in section 4 we introduce the independent data used for comparison; results are shown  
85 in section 5 and a discussion and summary are provided in section 6.

86

## 87 **2. Basics of the radiative model in the Air Force configuration of LIS**

88 The methodology to derive surface  $SW\downarrow$  radiative fluxes in the LIS/USAF version is based  
89 on information from the Air Force cloud products using the approach described in Shapiro (1972).  
90 It is a statistical model tracing solar radiation through a reflecting and absorbing medium where  
91 the atmosphere is composed of  $n$  homogeneous layers. A flowchart illustrating the various steps  
92 is provided in **Figure 1**.

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Fig. 1. A flowchart of the Shapiro (1976) model as implemented in the LIS/USAF NRT scheme.

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The major components of the LIS/USAF scheme is a LUT for layer transmittance and reflectance and a 3-layer 2-flux radiative transfer solver based on adding method. The input WWMCA 4-layer cloud information is first converted to Shapiro's 3-layer setup. The layer transmittances and reflectances of each layer are determined based on the layer cloud type and amount. Together with a surface albedo and solar angles as used in LIS modeling, the  $SW\downarrow$  can be computed analytically. Since the total solar radiation reaching the ground and reflected to space can be measured routinely, given a suitably sizeable series of such measurements under a variety of cloud conditions, the layer reflectivity and transmissivity can be estimated by a simple least-squares procedure. The downward flux of radiation leaving any layer is equal to the fractional transmission of that layer times the downward flux of radiation reaching that layer from above plus the fractional reflection of that layer times the upward flux of radiation reaching that layer from below. The system can be solved explicitly for radiation reaching the ground as a function of

110 the vertically incident radiation and known or assumed reflection and transmission coefficients for  
111 each of the  $n$  layers and the ground surface with an assigned transmission and reflection  
112 coefficients for each cloud type. It can be used for any combination of cloud and cloud-free layers.  
113 Information on cloud amount and types is provided by the USAF World-Wide Merged Cloud  
114 Analysis (WWMCA) outputs (D'Entremont et al. 2016). As stated in Shapiro (1972) the approach  
115 is deliberately kept simple; however, the structure of the model permits progressive refinement. In  
116 this study, the 3-hourly averaged USAF product that covers the region bounded by:  $59.875^\circ$  S  
117  $\sim 89.875^\circ$  N,  $179.875^\circ$  W  $\sim 179.875^\circ$  E at  $0.25^\circ$  resolution ( $1440 \times 600$  points) has been used.

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### 119 **3. University of Maryland (UMD) MODIS SW $\downarrow$ model**

120 In the original version of the UMD MODIS model (Wang and Pinker 2009), a  $1^\circ$  surface  
121 SW $\downarrow$  for all sky is computed in seven spectral intervals (0.2–0.4, 0.4–0.5, 0.5–0.6, 0.6–0.7, 0.7–  
122 1.19, 1.19–2.38, and 2.38–4.0  $\mu\text{m}$ ) assuming a plane-parallel, vertically inhomogeneous, scattering  
123 and absorbing atmosphere. Water vapor absorption is parameterized following the methods of  
124 Ramaswamy and Freidenreich (1992) and Chou and Suarez (1999). Ozone absorption in the  
125 ultraviolet wavelengths and in the visible wavelengths is computed following the approach of  
126 Lacis and Hansen (1974). The single-scattering properties and vertical profiles of aerosols were  
127 derived from the Optical Properties of Aerosols and Clouds (OPAC) software package (Hess *et al.*  
128 *1998*). Five atmospheric aerosol vertical profiles (continental, desert, maritime, Arctic, and  
129 Antarctic) are used with the inference scheme. Cloud extinction coefficients, single-scattering  
130 albedos, and asymmetry factors are computed from the parameterizations of Edwards and Slingo  
131 (1996) for water clouds and from Chou et al. (2002) for ice clouds. Multiple scattering is dealt  
132 with by using the delta-Eddington approximation following the method of Joseph et al. (1976).  
133 Top-of-atmosphere solar spectral irradiance data are from MODerate resolution atmospheric  
134 TRANsmision3 (MODTRAN3). In the original MODIS inference scheme (Wang and Pinker  
135 2009), the spectral reflectance for snow was assumed to be 0.9 and 0.6 for the visible and near-  
136 infrared parts of the spectrum, respectively. In the updated version, the surface spectral reflectance  
137 in the presence of snow is derived from a combination of snow-cover percentage and the MODIS  
138 surface spectral reflectance products, which are provided as 5-yr (2000–04) climatological  
139 statistics (the underlying surface types are aggregated according to the International Geosphere–

140 Biosphere Program classification (Moody et al. 2007). The model was further modified to facilitate  
141 the use of new information that became available, such as:

- 142 ○ MISR Level 3 monthly aerosol product (MIL3MAE or MIL3MAN)
- 143 ○ MODIS level3 weekly snow and ice product (MOD10C2 & MYD10C2)
- 144 ○ MODIS level3 daily snow and ice product (MOD10C1 & MYD10C1)
- 145 ○ Sea Ice Concentrations from Nimbus-7 SMMR and DMSP SSM/I-SSMIS Passive  
146 Microwave Data, Version 1, including both daily and monthly data and covering both  
147 North and South hemisphere.
- 148 ○ Precipitable water from the National Centers for Environmental Prediction (NCEP)-  
149 Department of Energy (DOE) daily reanalysis II.
- 150 ○ MODIS Aerosol Cloud Water Vapor Ozone Daily L3 Global 1° CMG (MOD08\_D3 and  
151 MYD08\_D3)

152 Auxiliary data prepared at UMD include land and sea mask, surface type, surface elevation,  
153 cloud layer thickness model coefficients, averaged albedo maps. A flowchart illustrating the entire  
154 process is presented in Figures 1 of Wang and Pinker (2009).

155 While the basic idea of the Shapiro (1972) model is similar to the UMD MODIS model  
156 (both are based on adding method for vertical quadrature) there are major differences that can be  
157 summarized as follows: 1). Shapiro (1972) model (SM) assumes only 3 layers of atmosphere,  
158 while the MODIS model has more than 40 layers, depending on the locations of clouds.  
159 2). SM assumes that the whole solar spectral range is quasi-monochromatic or single band and gas  
160 absorption is crudely treated by choice of values assigned to the atmospheric layer absorptions,  
161 while the MODIS model has 7 bands and gas absorption is treated with a more detailed K-  
162 distribution method. 3). In SM, the quasi-monochromatic transmittance and reflectance of clouds  
163 are assigned based on climatological surface observations for various cloud types. The MODIS  
164 model has detailed parameterizations for the spectral cloud single scattering properties from  
165 Edwards and Slingo (1996) for water clouds and from Chou et al. (2002) for ice clouds. 4). Aerosol  
166 scattering and absorption, and molecular scattering are not explicitly included in the SM but are  
167 in the MODIS model. 5). While being based on the 2-stream adding method, the SM does not  
168 divide the radiation into direct and diffuse components. Radiation is considered direct before  
169 encountering clouds, and as diffuse when transmitted through clouds.

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171 **4. Independent Data used for comparison**

172 In addition to ground observations, we use satellite-based estimates and several well-  
 173 known reanalysis products to evaluate the LIS/USAF SW↓ fluxes. The ground data are of primary  
 174 importance in supporting the evaluation of all the other estimates used.

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 176 *a. SURFRAD/BSRN data*

177 The Baseline Surface Radiation Network (BSRN) is a project of the Data and Assessments  
 178 Panel from the Global Energy and Water Cycle Experiment (GEWEX) under the umbrella of the  
 179 World Climate Research Programme (WCRP) (Ohmura et al. 1998; Driemel et al. 2018) and as  
 180 such is aimed at detecting important changes in the Earth's radiation field at the Earth's surface  
 181 which may be related to climate changes. In 2004 the BSRN was designated as the global baseline  
 182 network for surface radiation for the Global Climate Observing System (GCOS). The BSRN  
 183 stations also contribute to the Global Atmospheric Watch (GAW). Since 2011 the BSRN and the  
 184 Network for the Detection of Atmospheric Composition Change (NDACC) have reached a formal  
 185 agreement to become cooperative networks. Twenty-four stations (**Table 1**) are available over the  
 186 period 10/01/2013 to 08/31/2015 and used in this study. For several years the Surface Radiation  
 187 (SURFRAD) Network (Augustine et al. 2000, Augustine et al. 2005; Augustine et al. 2013) was  
 188 operated independently over the US. More recently, it became a part of the BSRN. Data can be  
 189 downloaded from <ftp://aftp.cmdl.noaa.gov/data/radiation/surfrad/>. Instrument information can be  
 190 found at <https://www.esrl.noaa.gov/gmd/grad/surfrad/overview.html>. The downloaded data are  
 191 one-minute data, and written in ASCII format. Before the comparisons the data are processed to  
 192 daily averages. Missing values are filled by the closest values as a function of solar zenith angle.

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 194 Table 1. Global BSRN sites used in the evaluation of SW↓ from the various products.

Station full name	Abbreviation	Location	Lat	Lon	Surface type	Topography type	Rural/Urban II
Bondville	BON	Illinois, U.S.	40.07	-99.37	grass	flat	rural
Desert Rock	DRA	Nevada, U.S.	36.63	-116.02	desert	flat	rural
Fort Peck	FPK	Montana, U.S.	48.31	-105.10	grass	flat	rural

Goodwin Creek	GWN	Mississippi, U.S.	34.25	-89.87	grass	hilly	rural
Penn. State Univ.	PSU	Pennsylvania , U.S.	40.72	-77.93	cultivated	mountain valley	rural
Sioux Falls	SXF	South Dakota, U.S.	43.73	-96.62	grass	hilly	rural
Table Mountain	TBL	Colorado, U.S.	40.12	-105.24	grass	flat	rural
Alice Springs	ASP	Australia, Northern Territory	-23.80	133.89	grass	flat	rural
Cocos Island	COC	Australia, Cocos (Keeling) Islands	-12.19	96.84	grass	flat	rural
Darwin Met Office	DWN	Australia	-12.42	130.89	grass	flat	rural
Brasilia	BRB	Brazil, Brasilia City	-15.60	-47.71	Concrete/shrub	flat	rural
Petrolina	PTR	Brazil	-9.07	-40.32	Concrete/shrub	flat	rural
São Martinho da Serra	SMS	Brazil	-29.44	-53.82	Concrete/grasses	flat	rural
Cabauw	CAB	Netherlands	51.97	4.93	grass	flat	rural
Camborne	CAM	United Kingdom	50.22	-5.32	grass	flat	rural
Carpentras	CAR	France	44.08	5.06	cultivated	hilly	rural
Cener	CNR	Spain, Navarra	42.82	-1.60	asphalt	mountain valley	urban
Izaña	IZA	Spain, Tenerife	28.31	-16.50	rock	mountain top	rural
Lindenberg Palaiseau, SIRTA Observator y	LIN  PAL	Germany  France	52.21  48.71	14.12  2.21	cultivated  concrete	hilly  flat	rural  urban



Payerne	PAY	Switzerland	46.82	6.94	cultivated	hilly	rural
Sonnblick	SON	Austria	47.05	12.96	rock	mountain top	rural
Xianghe	XIA	China	39.75	116.96	desert, rock	flat	rural
Gobabeb	GOB	Namibia, Namib Desert	-23.56	15.04	desert gravel	flat	rural
Tamanrasset	TAM	Algeria	22.790 3	5.5292	desert, rock	flat	rural

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196 *b. ARM/SGP C1 site*

197 The Southern Great Plains (SGP) atmospheric observatory was the first field measurement  
 198 site established by the Atmospheric Radiation Measurement (ARM) user facility (*Stokes and*  
 199 *Schwartz* 1994). This observatory is the world's largest and most extensive climate research  
 200 facility (<https://www.arm.gov/capabilities/observatories/sgp>). The Central location (C1) is 36.61°  
 201 N, 97.49° W. The data are available from <https://www.arm.gov/capabilities/observatories/sgp>

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203 *c. The European Centre for Medium-Range Weather Forecasts (ECMWF) Reanalysis v5 (ERA5)*

204 The European Centre for Medium-Range Weather Forecasts (ECMWF) Reanalysis v5 (ERA5) is  
 205 the fifth generation of the European Centre for Medium-Range Weather Forecasts (ECMWF) re-  
 206 analysis for the global climate and weather for the past 4 to 7 decades (Hersbach et al. 2018).  
 207 Currently data are available from 1950, split into Climate Data Store entries for 1950-1978  
 208 (preliminary back extension) and from 1979 onwards (final release plus timely updates). ERA5  
 209 replaces the ERA-Interim re-analysis. Re-analysis combines model data with observations from  
 210 across the world into a globally complete and consistent dataset. ERA5 provides hourly estimates  
 211 for a large number of atmospheric, ocean and land-surface quantities. The data are re-gridded to a  
 212 regular latitude/longitude grid of 0.25° for the re-analysis. In this study we use the "ERA5 hourly  
 213 data on single levels from 1979 to present". The data are available at:  
 214 <https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-single-levels?tab=overview>

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218 *d. The Climate Forecast System Reanalysis (CFSR)*

219 The Climate Forecast System Reanalysis (CFSR) (Saha et al. 2010, 2014) is a third-  
220 generation reanalysis product developed by NOAA National Centers for Environmental Prediction  
221 (NCEP). It is a global, high resolution, coupled atmosphere-ocean-land surface - sea ice system  
222 designed to provide the best estimate of the state of these coupled domains over this period. Here  
223 we used the 6-hourly product with a spatial resolution of  $0.5^{\circ} \times 0.5^{\circ}$ . The data are available at:  
224 <https://climatedataguide.ucar.edu/climate-data/climate-forecast-system-reanalysis-cfsr>

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226 *e. The Modern-Era Retrospective analysis for Research and Applications (MERRA)*

227 The Modern-Era Retrospective analysis for Research and Applications, Version 2  
228 (MERRA-2) (Gelaro et al. 2017) is a global atmospheric reanalysis developed by NASA's Global  
229 Modeling and Assimilation Office (GMAO) providing data from 1980 on. It replaces the original  
230 MERRA data because of the advances made in the assimilation system that enable assimilation of  
231 modern hyperspectral radiance and microwave observations, along with GPS-Radio Occultation  
232 datasets. It also uses NASA's ozone profile observations that began in late 2004. Additional  
233 advances in both the GEOS model and the GSI assimilation system are included in MERRA-2.  
234 The data center for MERRA-2 provides DOI and a full citation for all the MERRA-2 data. For the  
235 1 hourly radiation:  
236 Global Modeling and Assimilation Office (GMAO) (2015), MERRA-2 tavg1\_2d\_rad\_Nx: 2d,1-  
237 Hourly, Time-Averaged,Single-Level,Assimilation, Radiation Diagnostics V5.12.4, Greenbelt,  
238 MD, USA, Goddard Earth Sciences Data and Information Services Center (GES DISC), Accessed:  
239 [Data Access Date], 10.5067/Q9QMY5PBNV1T.  
240 The data are available at: <https://gmao.gsfc.nasa.gov/reanalysis/MERRA-2/>.

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242 **5. Results**

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244 *a. Issues related to homogeneity of data products*

245 Before conducting the comparison, all products are re-gridded (linear interpolation) to  $1^{\circ}$   
246 resolution and converted to daily values; they are cropped to the domain of  $59.5^{\circ}$  S~ $59.5^{\circ}$  N as  
247 used in LIS. The spatial matching is done by using the estimations (daily data) at the nearest points

248 for each site location. If the number of nearest points is more than 2, than the estimation is the  
249 mean values with the weights of latitude and longitude.

250 Several aspects of the comparison process itself can introduce errors that are difficult to  
251 estimate. For instance, each model was produced at different spatial and temporal scales. In the  
252 comparisons, all data were scaled to 1° spatial resolution and to daily time scales. LIS/USAF  
253 provides data averaged for each 3-hourly interval. The daily value is obtained by simply averaging  
254 the 3-hourly mean product for both ERA5 and MERRA-2. For CFSR, the daily values are obtained  
255 by averaging the 6-hourly mean products. The satellite UMD/MODIS product is based on two  
256 observations per day. The procedure to obtain a daily average is described in detail in Wang and  
257 Pinker (2009). It will be recaptured here briefly.

258 The diurnal variation of SW↓ is caused mainly by clouds and position of the Sun. The latter  
259 can be well described, but the diurnal variation of clouds is not readily available. Using MODIS  
260 observations from both Terra and Aqua to estimate SW↓, a difference between morning and  
261 afternoon fluxes was observed. Over most of the continents, SW↓ fluxes are larger in the morning  
262 than in the afternoon (over much of the oceans, the differences are reversed). Over land there are  
263 more clouds in the afternoon. The diurnal cycle of clouds drives the surface SW↓ cycle (Chen and  
264 Houze, 1997; Duvel, 1989; Gray and Jacobson, 1977). The combination of MODIS observations  
265 from Terra and Aqua provides an opportunity to construct realistic daily values. The daily average  
266 is computed by assuming that MODIS observations from Terra at 1030 LT and from Aqua at 1330  
267 LT represent the atmospheric conditions from sunrise to local noon and from local noon to sunset,  
268 respectively. The diurnal variation of incident fluxes will be dictated only by the incident solar  
269 flux at top of the atmosphere which is determined by the cosine of the solar zenith angle. The daily  
270 integration of radiative fluxes is reduced to the integration of the cosine of the solar zenith angle.  
271 Thus, the daily average radiative flux is calculated as follows:

$$Flux_{daily} = \left( \int_{Sunrise}^{Noon} Flux(\mu(t))dt + \int_{Noon}^{Sunset} Flux(\mu(t))dt \right) / 24 \text{ hours,}$$

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274 where Flux(μ(t)) is instantaneous radiative flux at time t with cosine of solar zenith angle μ(t).

275 Another issue related to the accuracy of SW↓ fluxes as derived from satellite observations is related  
276 to the nonlinearity of the relationship between radiance and flux. In most cases, provided are  
277 radiances averaged at a certain scale and these are used to compute the flux.

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279 *b. Evaluation against Ground Observations*

280 Observations from the BSRN network are available over numerous global sites. The  
281 ARM/SGP C1 site is considered a super site in terms of quality and scope of observations.  
282 Evaluation will be done using all available data. Since the performance of LIS/USAF product over  
283 different regions is of interest, the evaluation will also be presented independently over the US and  
284 Europe, where several observing sites are available. Results for Brazil, Australia, Africa and China  
285 (with a limited number of ground sites) will be provided in a Supplement.

286

287 1) EVALUATION USING ALL GLOBALLY DISTRIBUTED SITES

288 Evaluation of daily SW↓ from UMD/MODIS, LIS/USAF, ERA5, CFSR and MERRA-2  
289 against ground observations (all available sites as illustrated in **Figure 2**) during 10/01/2013 –  
290 08/31/2015 has been performed.

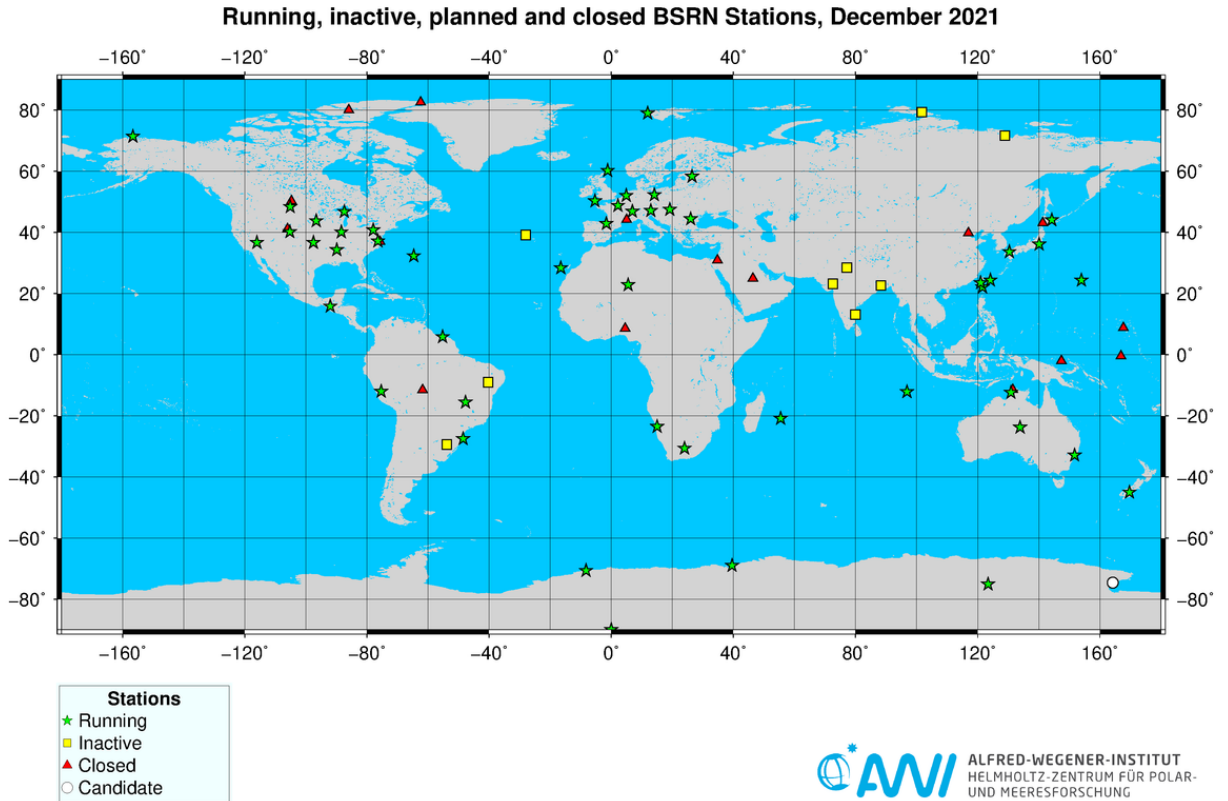


Fig. 2. Global distribution of BSRN sites.

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As seen from **Table 2** in best agreement with ground observations are the results from UMD MODIS and ERA5. In terms of lowest bias LIS/USAF are close to each other while the RMSE for LIS/USAF is much lower than CFSR. To get a better insight on possible reasons for the observed differences one could segregate the data by season, latitude or land use type. It is well known that cloud detection is not uniform for different cloud conditions and over different surface types (dark or bright). The movement of clouds within the interval of observations or prediction time steps has an impact on the results. The record length used in this study and the limited number of ground observations are not conducive to such separations. While for understanding differences, such analysis may be helpful, most users are interested in the overall agreement in deciding which data they prefer rather than seasonal or latitudinal differences.

To understand the reasons for differences among products is very difficult. While the key features of the LIDS/USAF and UMD MODIS have been discussed, the reasons for differences between the reanalysis products are numerous such the observing system, data assimilation (DA) system, model components and post processing system. As documented for ERA5 (<https://confluence.ecmwf.int/display/CKB/ERA5%3A+data+documentation>), it is produced

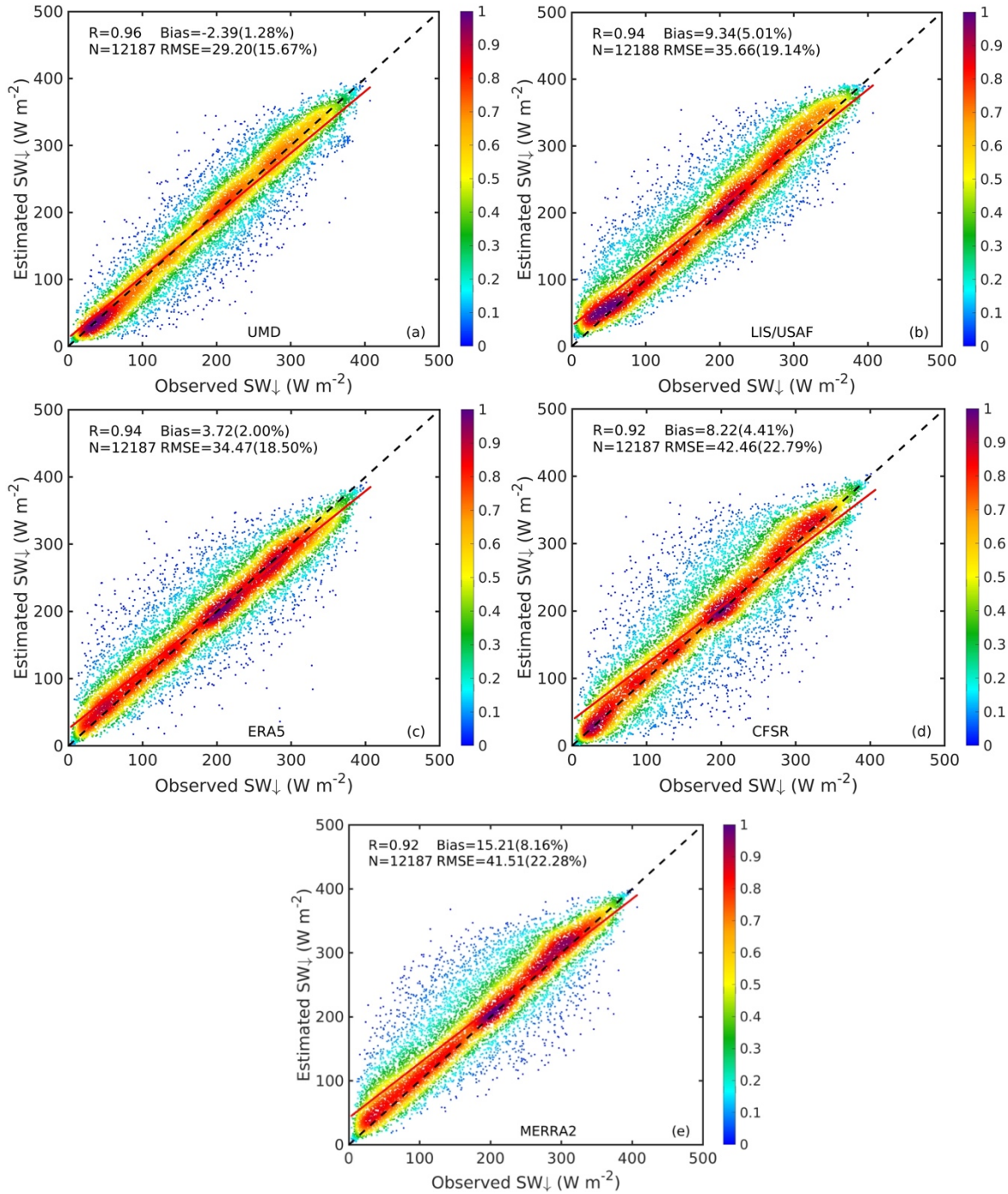
309 using 4D-Var data assimilation and model forecasts in CY41R2 of the ECMWF Integrated  
310 Forecast System (IFS), with 137 hybrid sigma/pressure levels in the vertical. The atmospheric  
311 model in the IFS is coupled to a land-surface model (HTESSEL) and an ocean wave model  
312 (WAM).

313 The CFSRv2 (Saha et al., 2014) is produced by the second version of the NCEP Climate  
314 Forecast System (CFSv2) which uses 3D-Var DA system, and coupled atmosphere-ocean-land  
315 surface-sea ice system. The horizontal resolution is ~38 km (T382) with 64 levels in the vertical.  
316 The global ocean is  $0.25^\circ$  at the equator, extending to a global  $0.5^\circ$  beyond the tropics, with 40  
317 levels. The global land surface model has 4 soil levels and the global sea ice model has 3 levels.

318 The MERRA-2 (R. Gelaro et al., 2017) is produced with version 5.12.4 of the GEOS  
319 atmospheric data assimilation system (GSI 3D-Var). The key components of the system are the  
320 GEOS atmospheric model (Rienecker et al. 2008; Molod et al. 2015) and the GSI analysis scheme  
321 (Wu et al. 2002; Kleist et al. 2009). The model includes the finite-volume dynamical core of  
322 Putman and Lin (2007), which uses a cubed-sphere horizontal discretization at an approximate  
323 resolution of  $0.5^\circ \times 0.625^\circ$  and 72 hybrid-eta levels from the surface to 0.01 hPa. The analysis is  
324 computed on a latitude–longitude grid at the same spatial resolution as the atmospheric model  
325 using a 3DVAR algorithm based on the GSI with a 6-h update cycle and the so-called FGAT  
326 procedure for computing temporally accurate observation-minus-background departures. The  
327 analysis is applied as a correction to the background state using an IAU procedure (Bloom et al.  
328 1996). As such, to pinpoint the reasons for observed differences in the predicted  $SW\downarrow$  from these  
329 models is beyond the scope of this study.

330 In the following, independent evaluation over the US and Europe will be presented. Independent  
331 results over Brazil, Australia, Africa and China are presented in the **Supplement**.

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Fig. 3. Evaluation of daily SW<sub>↓</sub> from UMD/MODIS, LIS/USAF, ERA5, CFSR and

MERRA-2 against ground observations (all available sites) during the period of

10/01/2013 – 08/31/2015.

339 Table 2. Statistics of evaluation of daily SW↓ from UMD/MODIS, LIS3, ERA5,  
 340 CFSR and MERRA-2 against ground observations 10/01/2013 –  
 341 08/31/2015. Units: W m<sup>-2</sup>

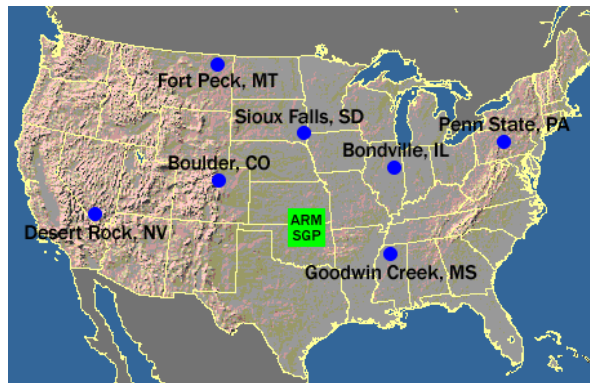
	<i>R</i>	<i>Bias</i>	(%)	<i>RMSE</i>	(%)	<i>N</i>
<b>UMD</b>	0.96	-2.39	1.28	29.20	15.67	12187
<b>LIS/USAF</b>	0.94	9.34	5.01	35.66	19.14	12188
<b>ERA5</b>	0.94	3.72	2.00	34.47	18.50	12187
<b>CFSR</b>	0.92	8.22	4.41	42.46	22.79	12187
<b>MERRA2</b>	0.92	15.21	8.16	41.51	22.28	12187

342

343 2) EVALUATION OVER THE U.S.

344 Observations from seven BSRN sites and one ARM/SGPC1 site are used for comparisons.  
 345 The BSRN stations are, “Desert\_Rock\_NV” (DRA, 36.63°N, 116.02°W),  
 346 “Penn\_State\_PA”(PSU ,40.72°N, 77.93°W), “Bondville\_IL” (BON, 40.06°N, 88.37°W),  
 347 “Goodwin\_Creek\_MS”(GWN, 34.25°N, 89.97°W), “Fort\_Peck\_MT” (FPK, 48.31°N, 105.10°W),  
 348 “Boulder\_CO” (TBL, 40.13°N, 105.24°W), “Sioux\_Falls\_SD” (SXF, 43.73°N, 96.62°W).

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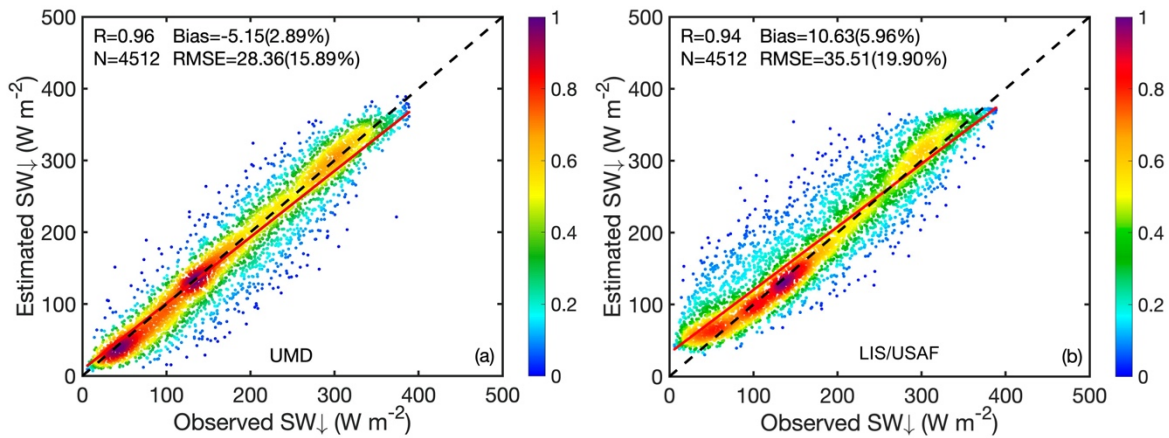
351 Fig. 4. SURFRAD sites and ARM/SGP location (Downloaded from SURFRAD website)

352

353 The site locations over the US are shown in **Figure 4**, results are shown in **Figure 5** and  
 354 statistics are summarized in **Table 3**. As seen the UMD/MODIS product performs best with  
 355 highest correlation *R* of (0.96), smallest *Bias* (-5.15 W m<sup>-2</sup>) and *RMSE* (28.36 W m<sup>-2</sup>). LIS/USAF  
 356 product performs better than the other reanalysis products. The *R* of LIS/USAF is 0.94 with *Bias*  
 357 of 10.63 W m<sup>-2</sup> and *RMSE* of 35.51 W m<sup>-2</sup>

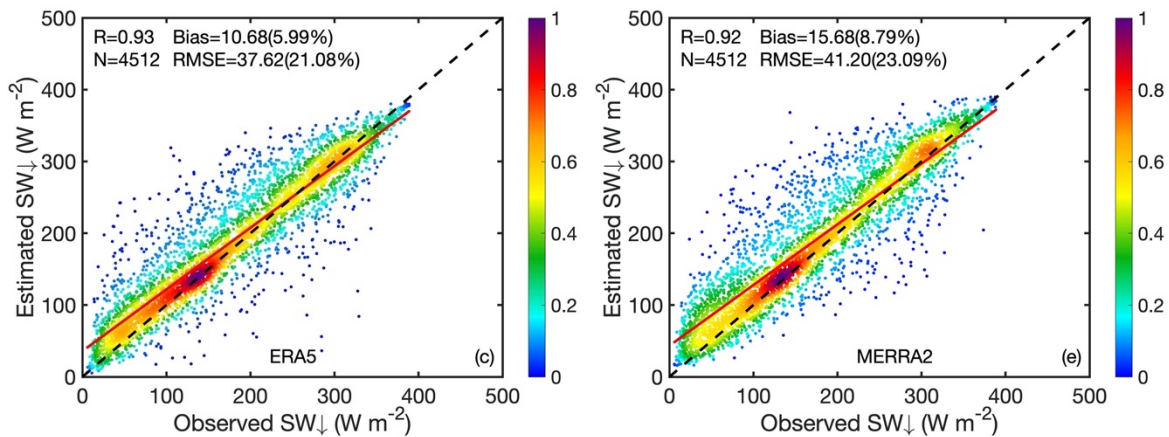


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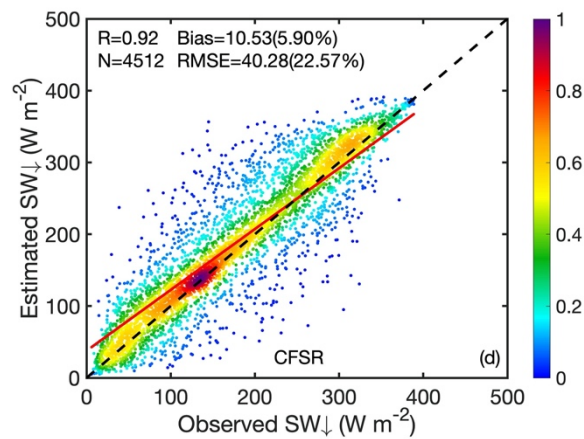


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363 Fig. 5. Evaluation of daily SW<sub>↓</sub> from UMD/MODIS, LIS/USAF, ERA5, CFSR and MERRA-2 against  
364 ground observations over U.S (7 SURFRAD sites and 1 ARM/SGPC1) during 10/01/2013 – 08/31/2015.

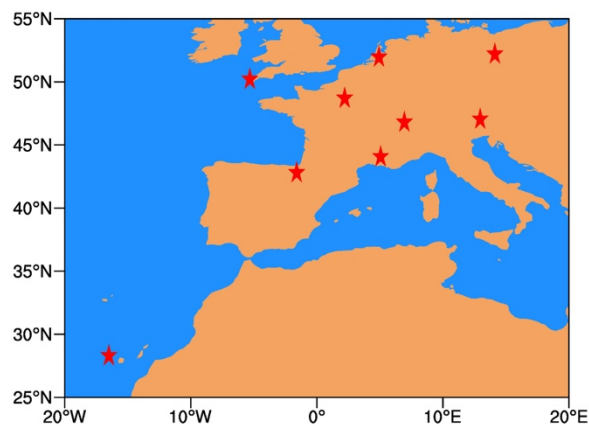
365

366 Table 3. Statistics of evaluation of daily SW↓ from UMD/MODIS, LIS/USAF,  
 367 ERA5, CFSR and MERRA-2 against ground observations over U.S. during  
 368 10/01/2013 – 08/31/2015.

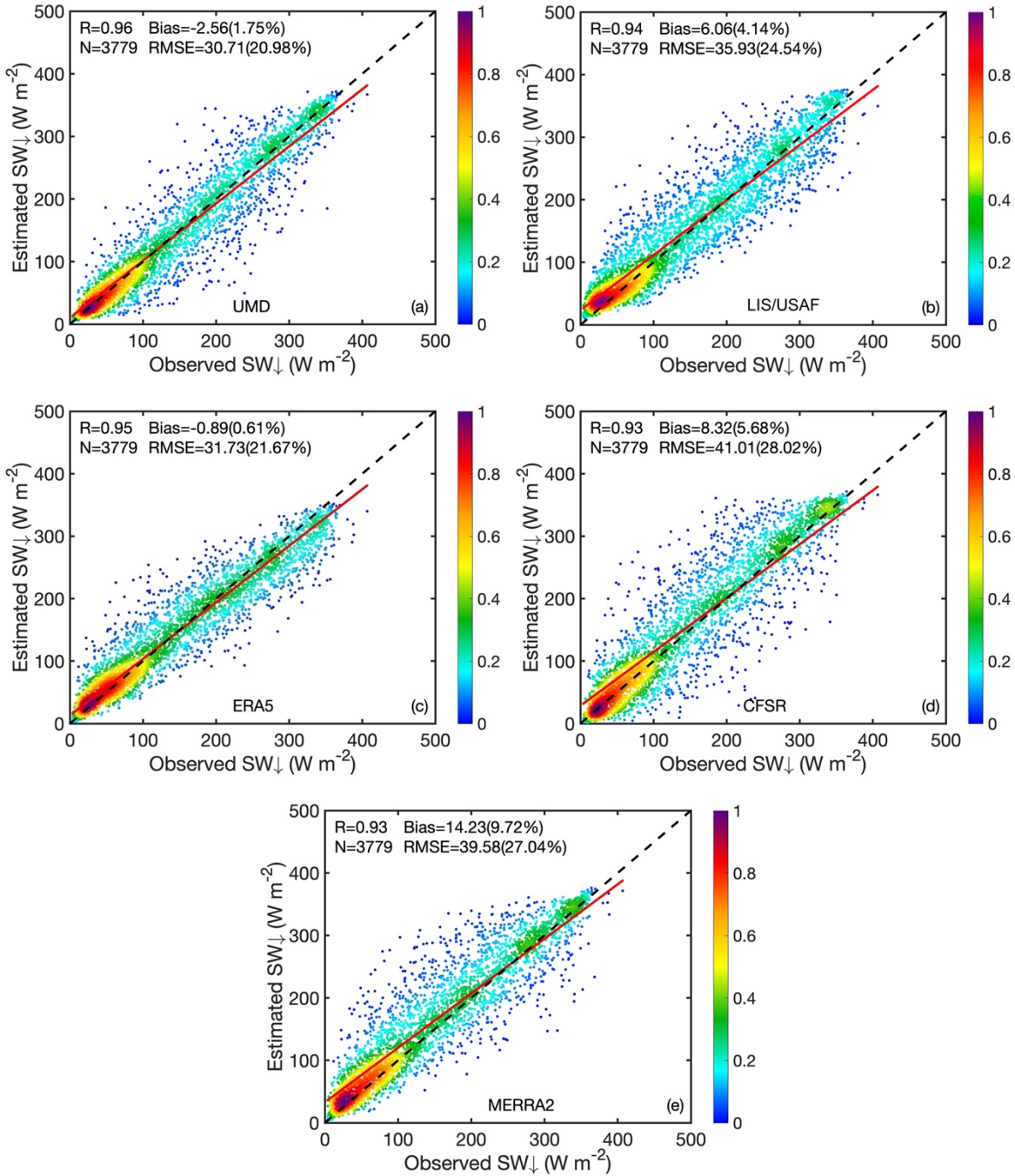
	<i>R</i>	<i>Bias</i>	(%)	<i>RMSE</i>	(%)	<i>N</i>
<b>UMD</b>	0.96	-5.15	2.89	28.36	15.89	4512
<b>LIS/USAF</b>	0.94	10.63	5.96	35.51	19.9	4512
<b>ERA5</b>	0.93	10.68	5.99	37.62	21.08	4512
<b>CFSR</b>	0.92	10.53	5.90	40.28	22.57	4512
<b>MERRA2</b>	0.92	15.68	8.79	41.20	23.09	4512

369  
 370 3) EVALUATION OVER EUROPE

371 Nine BSRN sites have been used for the Europe area. The locations of these sites are  
 372 shown in **Figure 6**. The evaluations of the daily SW↓ from UMD/MODIS, LIS/USAF, ERA5,  
 373 CFSR and MERRA-2 was conducted against the merged data of the nine sites for the study  
 374 period as shown in **Figure 7**. The LIS/USAF product still performed better than the others. The  
 375 *R* is 0.93, the *Bias* is 6.06 W m<sup>-2</sup>, and *RMSE* is 35.93 W m<sup>-2</sup>. The performance of CFSR and  
 376 MERRA2 are comparable to each other.



377  
 378 Fig. 6. Locations of 9 BSRN sites in Europe.  
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Fig.7. Evaluation of daily  $SW_{\downarrow}$  from UMD/MODIS, LIS/USAF, ERA5, CFSR and MERRA-2 against ground observations over Europe during 10/01/2013 – 08/31/2015.

386 Table 4. Statistics of evaluation of daily SW↓ from UMD/MODIS, LIS3, ERA5,  
 387 CFSR and MERRA2 against ground observations over Europe during  
 388 10/01/2013 – 08/31/2015.

	<i>R</i>	<i>Bias</i>	(%)	<i>RMSE</i>	(%)	<i>N</i>
<b>UMD</b>	0.96	-2.56	1.75	30.71	20.98	3779
<b>LIS3</b>	0.94	6.06	4.14	35.93	24.54	3779
<b>ERA5</b>	0.95	-0.89	0.61	31.73	21.67	3779
<b>CFSR</b>	0.93	8.32	5.68	40.01	28.02	3779
<b>MERRA2</b>	0.93	14.23	9.72	39.58	27.04	3779

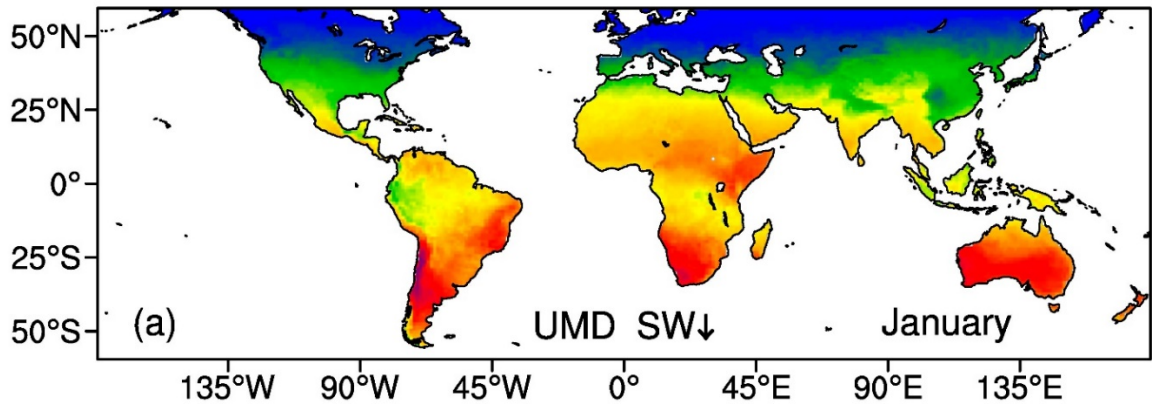
389  
 390 *d. Comparison of LIS/USAF SW↓ with independent products at global scale*

391 8. EVALUATION OVER the GLOBE

392 The averaged SW↓ from LIS/USAF WWMCA, ERA5, CFSR, MERRA-2 for January during  
 393 10/01/2013-08/31/2015 were compared against UMD/MODIS. As shown in **Figures 8-9**, the  
 394 distribution pattern and the averaged values of the SW↓ for January are similar in North America,  
 395 Europe and Australia. Differences are noted mainly in South America, Africa and Asia. **Figure 10**  
 396 shows the frequency distribution of these differences. The reanalysis products tend to overestimate  
 397 the SW↓ fluxes when compared to satellite observation for January, and most of the differences  
 398 are less than 20 W m<sup>-2</sup>. Statistics are shown in **Table 5**. The correlation coefficients (*R*) between  
 399 the reanalysis products and satellite observation are over 0.9 with positive *Bias* (≤15.1 W m<sup>-2</sup>).  
 400 The root mean-square errors (*RMSE*) are in the range of 31.8~43.9 W m<sup>-2</sup>.

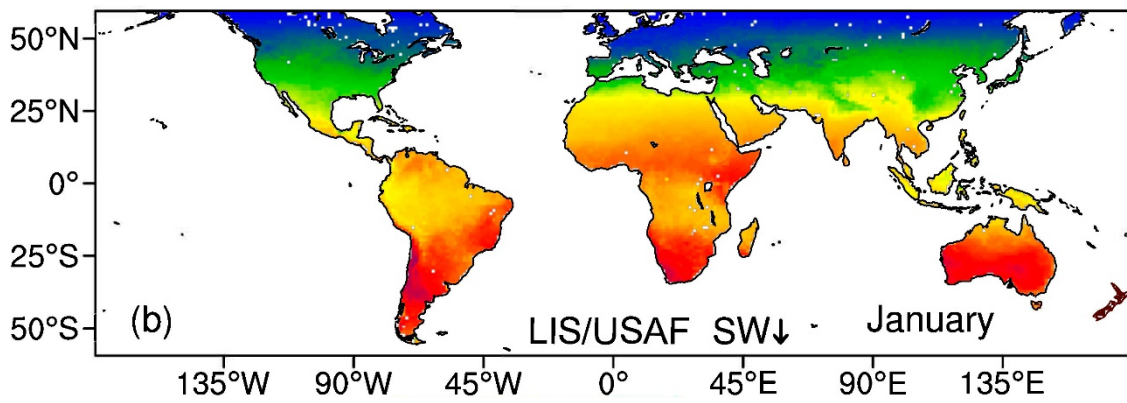
401 Used is the Student's t-test to test the null hypothesis that the sample means are from the  
 402 same population (i.e. H0: ave1=ave2). **p** is the significance which is two tailed and uses the  
 403 incomplete beta function to calculate the probability. It will range between zero and one. If **p** less  
 404 than significance level, then the null hypothesis is rejected and the alternative hypothesis is  
 405 accepted. In our case, we assume that UMD/MODIS has the same average values as the LIS/USAF  
 406 or ERA5 or CFSR or MERRA2 and the significance level is 0.1.

407 All the **p** values are equal or larger than 0.1. Therefore, we can assume that the samples are similar  
 408 to each other.

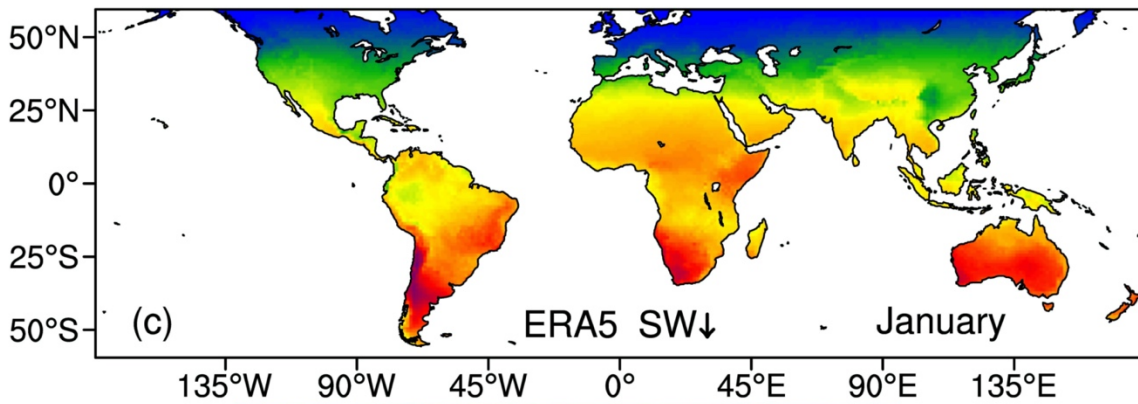


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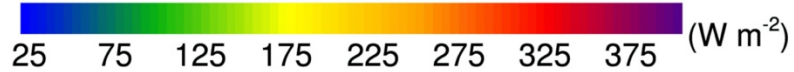
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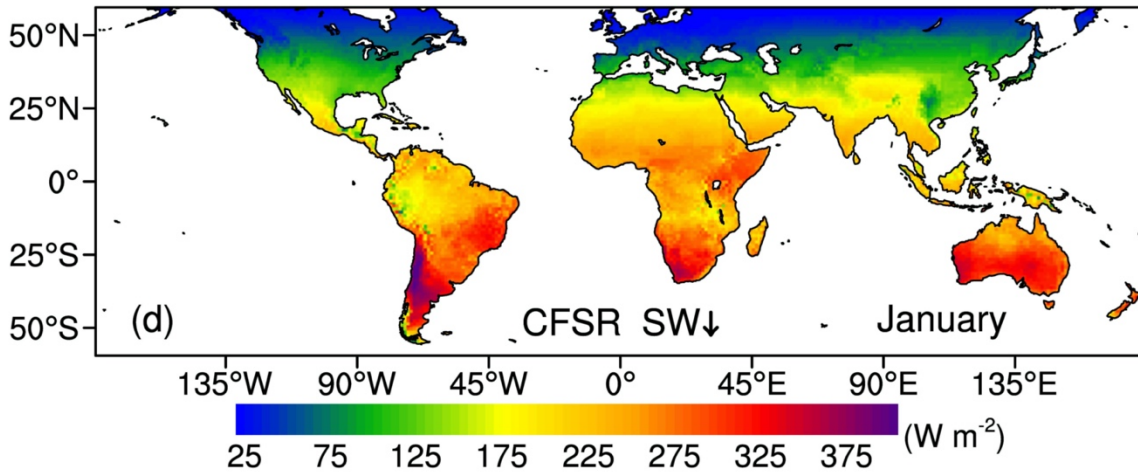


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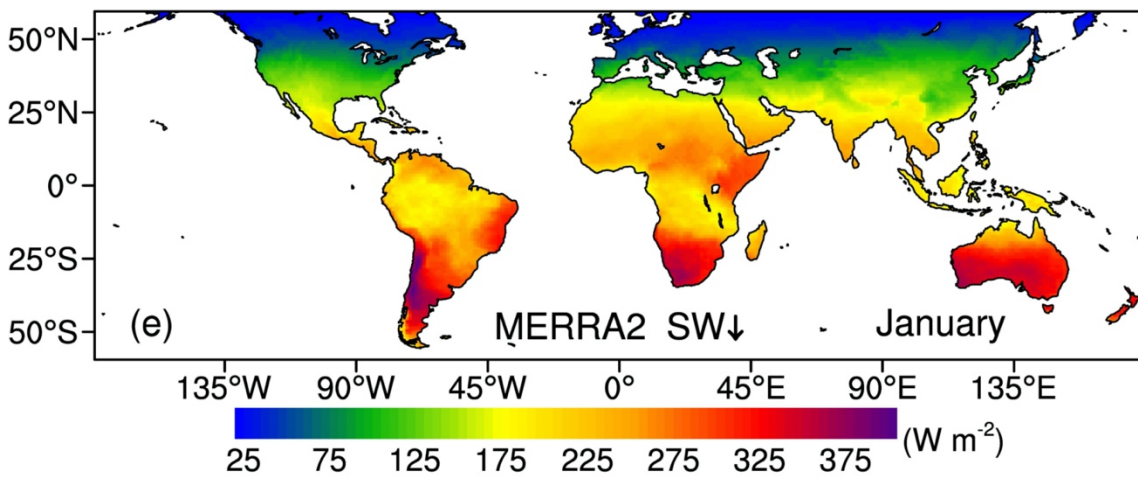


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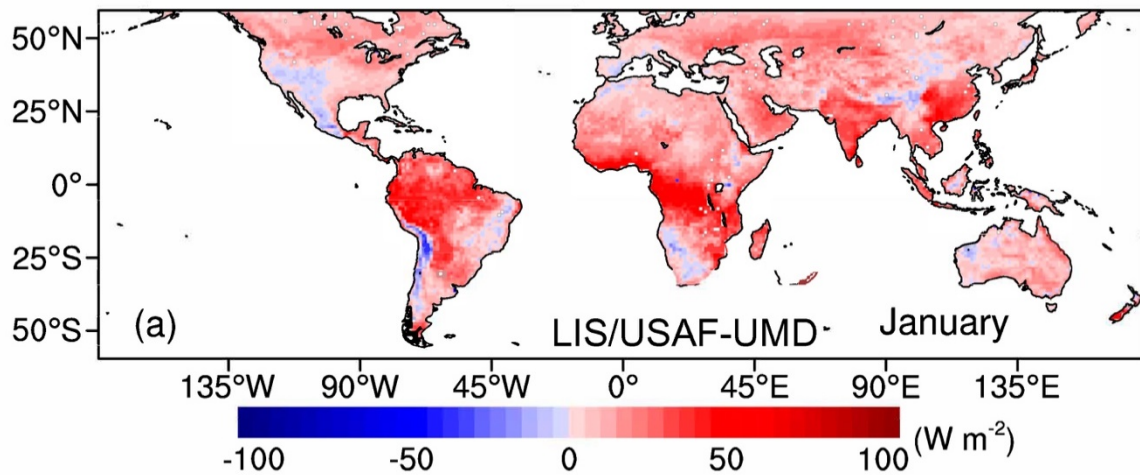
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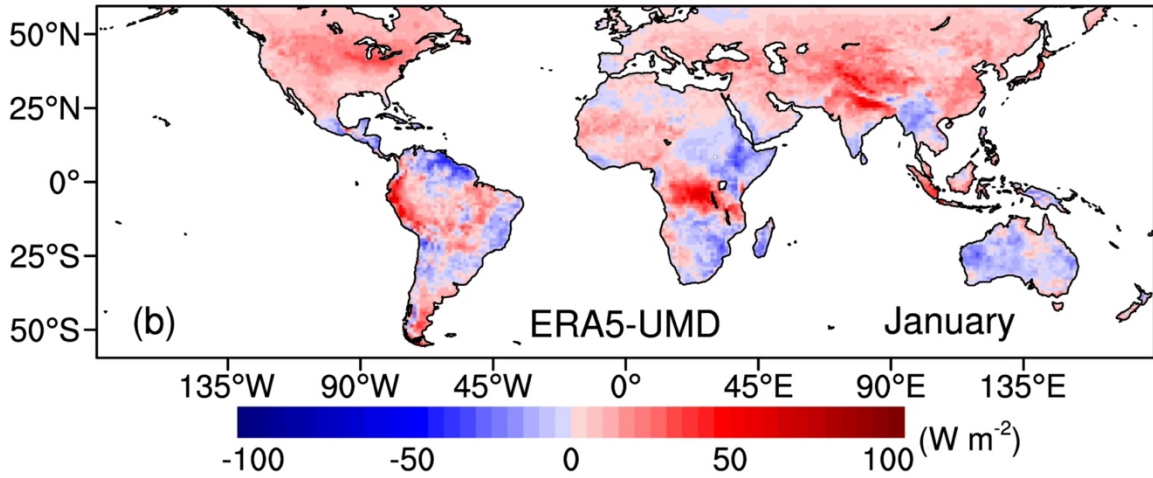
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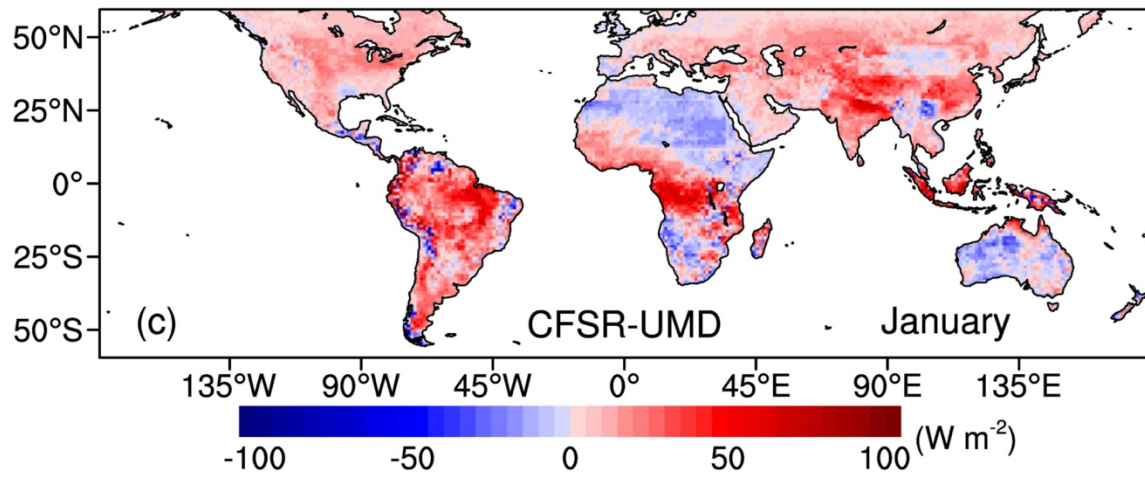
Fig. 8. Averaged daily SW↓ from LIS/USAF, ERA5, CFSR, MERRA-2 and UMD/MODIS for January during 10/01/2013-08/31/2015.



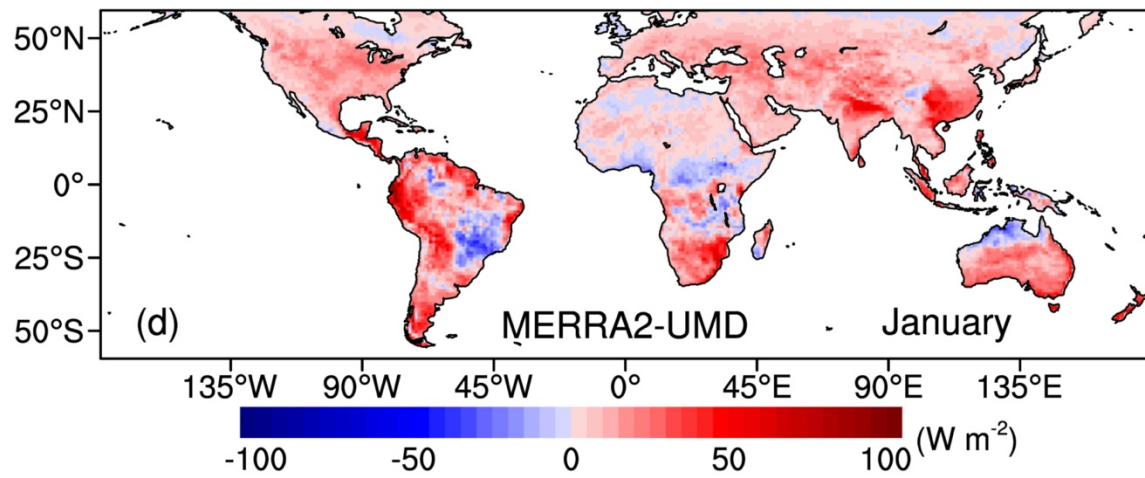
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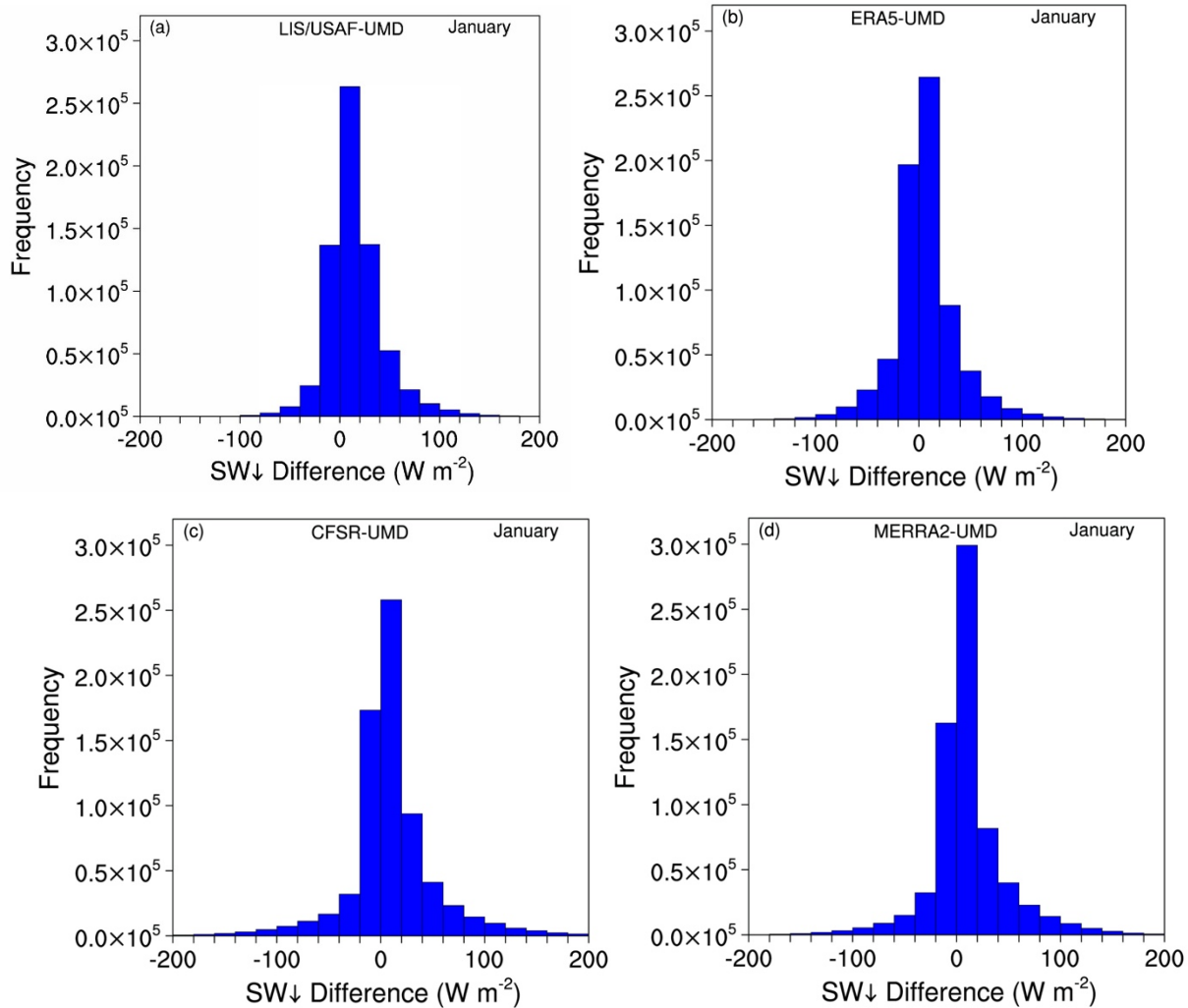
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Fig. 9. Averaged daily  $\text{SW} \downarrow$  difference between LIS/USAF, ERA5, CFSR, MERRA-2 and UMD/MODIS for January during 10/01/2013-08/31/2015.

427



428

429

430

431 Fig.10. Distribution of daily SW ↓ difference between LIS/USAF, ERA5, CFSR, MERRA2 and  
432 UMD/MODIS for January during 10/01/2013-08/31/2015

433

434 The averaged SW↓ from LIS/USAF, ERA5, CFSR, MERRA-2 for July over the study

435 period were also compared against UMD/MODIS and their differences are shown in **Figure 11**.

436 The re-analysis products of LIS/USAF, CFSR and MERRA-2 tend to overestimate the SW↓ fluxes

437 for July when compared with the UMD/MODIS product, especially in Asia. The frequency

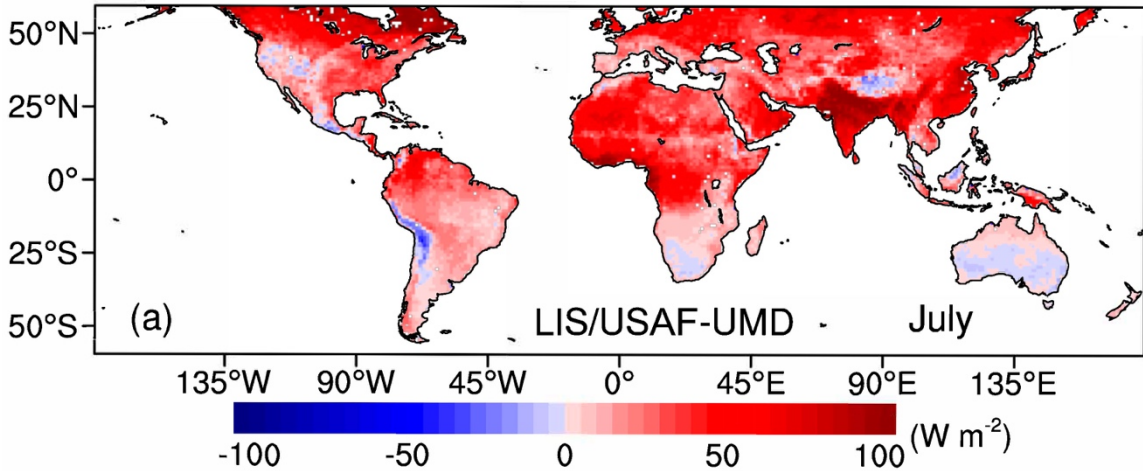
438 distribution of the differences (**Figure 12**) also shows such tendency except for ERA5. Most of

439 the differences are within  $\pm 20$  W m<sup>-2</sup>. The correlation coefficients between the reanalysis and

440 satellite observation are over 0.8. All of the reanalysis products have positive bias ( $\leq 35.0$  W m<sup>-2</sup>)

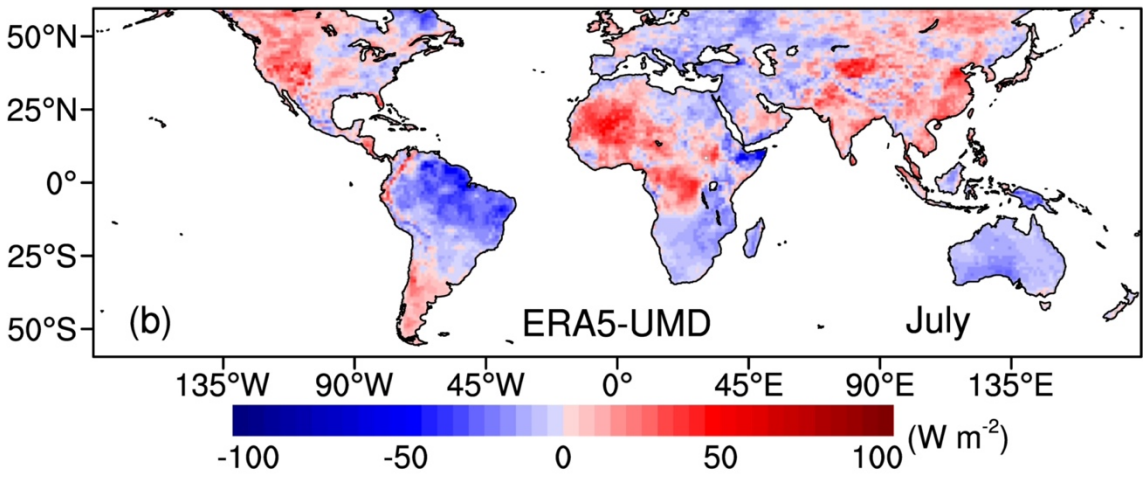
441 and the *RMSEs* are in between 41.6~59.7 W m<sup>-2</sup>.





442

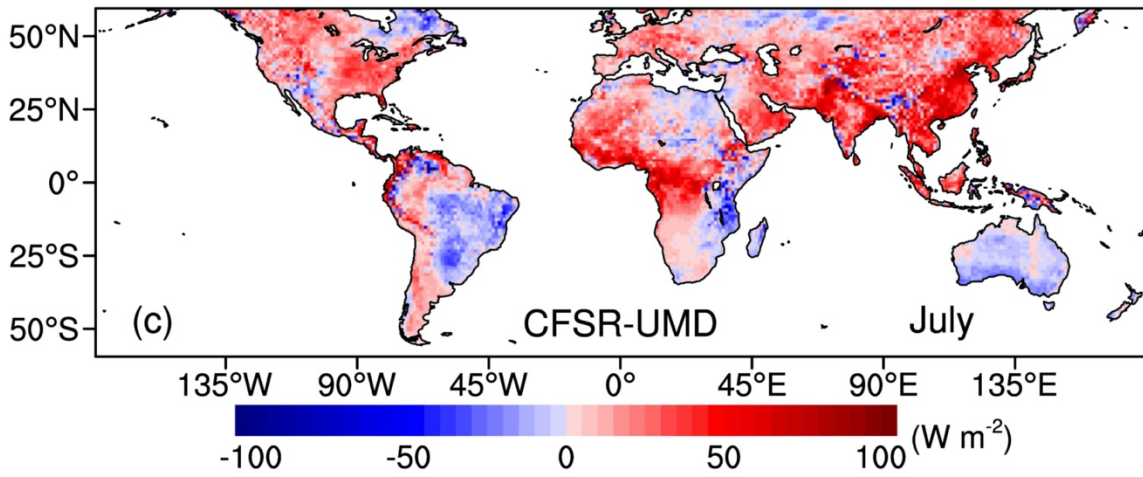
443



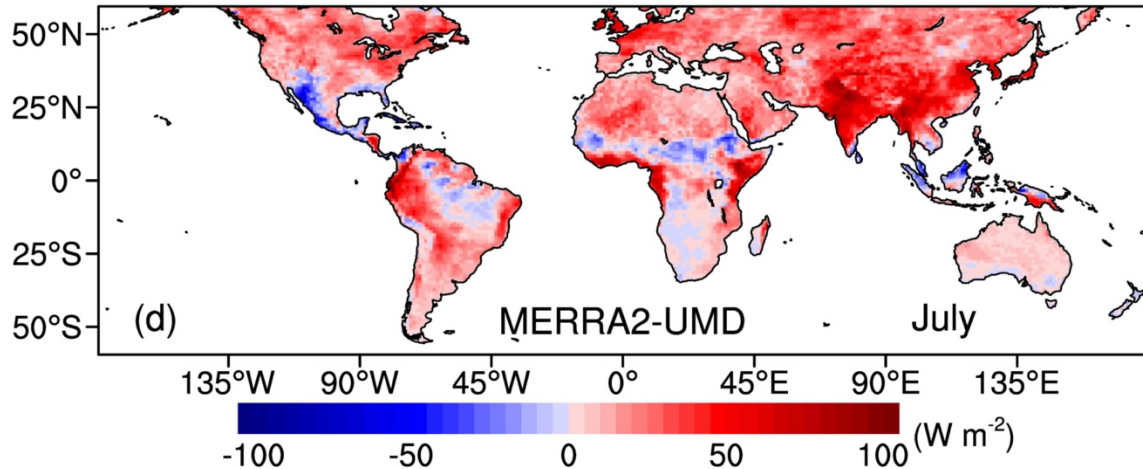
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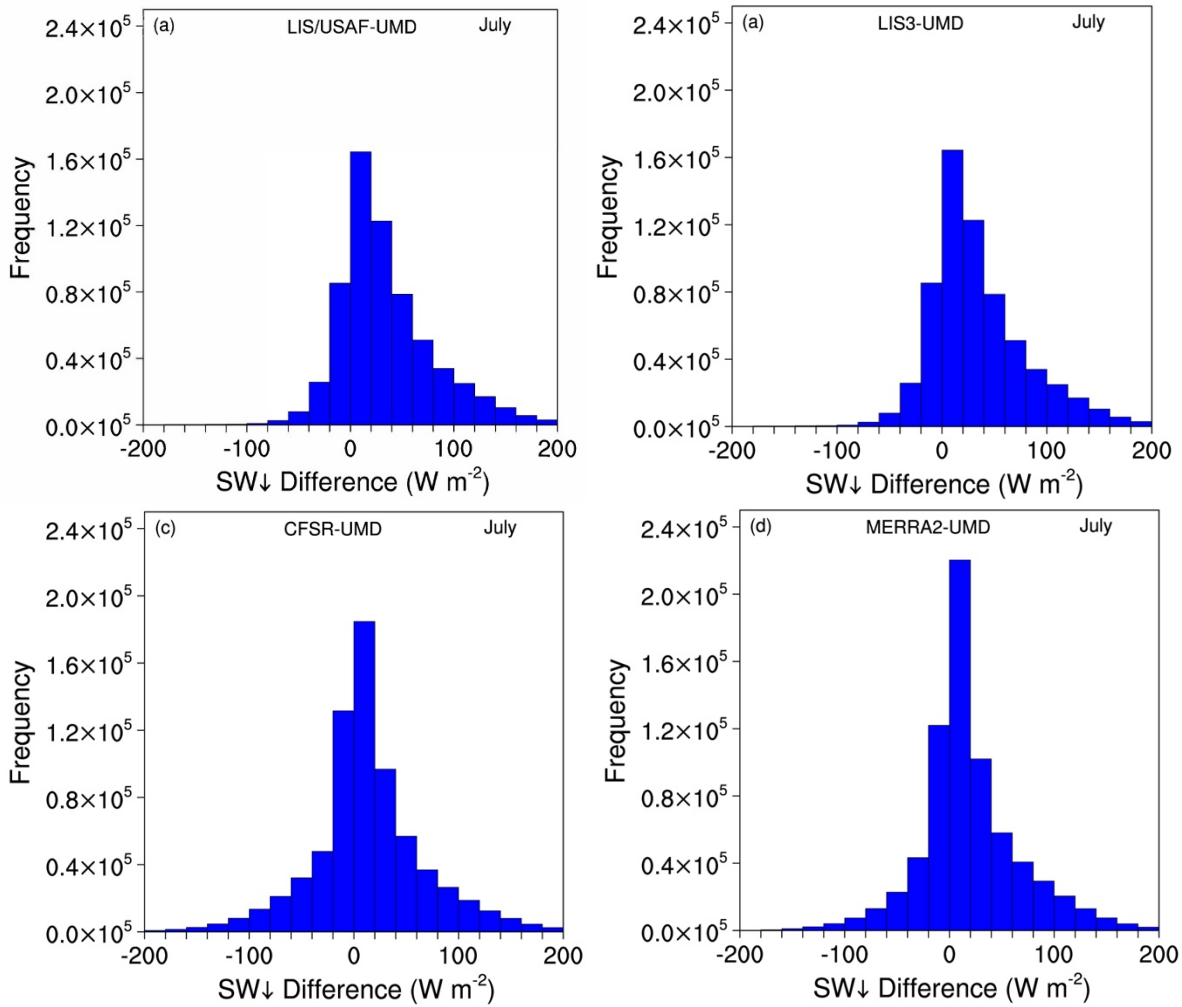
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448  
 449 Fig. 11. Averaged daily SW↓ difference between LIS/USAF, ERA5, CFSR, MERRA-2 and  
 450 UMD/MODIS for July during 10/01/2013~08/31/2015.  
 451



452

453

454 Fig. 12. Distribution of daily SW↓ difference between ERA5, CFSR, MERRA-2, LIS/USAF and  
 455 UMD/MODIS for July during 10/01/2013~08/31/2015.

456  
 457 Table 5. Statistics of evaluations of daily SW↓ for January and July against UMD/MODIS  
 458 for the entire study area from 10/01/2013~08/31/2015. Here **p** is the significance.

UMD	January						July					
	<i>R</i>	<i>Bias</i>	%	<i>RMSE</i>	%	<i>p</i>	<i>R</i>	<i>Bias</i>	%	<i>RMSE</i>	%	<i>p</i>
LIS3	0.96	15.1	9.8	31.8	20.7	0.2	0.83	35.0	15.2	59.7	25.9	0.1
ERA5	0.96	6.4	4.2	30.9	20.1	0.3	0.88	0.7	0.3	40.6	17.6	0.3
CFSR	0.92	10.1	6.6	43.9	28.6	0.2	0.81	13.6	5.9	56.1	24.3	0.2
MERRA2	0.94	10.2	6.6	38.3	25.0	0.2	0.83	19.1	8.3	51.6	22.4	0.2

459

## 460 6. Discussion and Summary

461

462 As stated in Kumar et al. (2006) land surface modeling seeks to predict the terrestrial  
 463 water, energy, and biogeochemical processes by solving the governing equations at the  
 464 earth/atmosphere interface. LSMs typically require several types of inputs states such as states  
 465 known as “forcing” such as information on clouds. Using these inputs, LSMs can predict surface  
 466 fluxes providing a realistic representation of the transfer of mass, energy, and momentum  
 467 between a vegetated surface and the atmosphere. One of the important boundary conditions to  
 468 the LSMs is SW↓ radiation. From the global scale comparisons, it became evident that most  
 469 models have problems to predict this parameter correctly in certain climatic regions and models  
 470 differ seasonally. For instance, during January, USAF shows overestimates primarily over S.  
 471 America, equatorial Africa, India and China and underestimation over North Africa. CFSR also  
 472 shows overestimates over India and China, equatorial Africa but underestimates over North  
 473 Africa. ERA5 overestimates over the Himalayas and sub-equatorial Africa but differences with  
 474 UMD/MODIS are much smaller than those seen in LIS/USAF. MERRA2 also tends to over-  
 475 estimate over the Himalayas and China but shows a mixture of over-estimation and under-  
 476 estimation over S. America. Notable differences between the models are seen over Australia. In  
 477 July, there seems to be a systematic overestimation by LIS/USAF over most of the globe while  
 478 the other models alternate between over-estimation and under-estimation. It should be noted that  
 479 as yet, there is no full agreement between available estimates of cloud amounts (Wonsick et al.,  
 480 2009). Some inference schemes to derive surface radiative fluxes use information on cloud

481 optical depth rather than on cloud amount but again, the methodologies how to derive such  
482 information from satellite observations differ (Wang and Pinker, 2008; Platnick et al. 2017)

483 Another accuracy issue in SW↓ fluxes as derived from satellite observations is related to  
484 the nonlinearity of the relationship between radiance and flux. In most cases, provided are  
485 radiances averaged at a certain scale and these are used to compute the flux. An example that  
486 illustrates this issue is the International Satellite Cloud Climatology Project (ISCCP) product  
487 (Rossow and Schiffer 1991) that is widely used to produce surface fluxes. For instance, what is  
488 known as the ISCCP D1 product provides spectral SW radiances at the top of the atmosphere at  
489 2.5° spatial resolution. There exists also what is known as the ISCCP DX product which is sampled  
490 at 30 km. An experiment was conducted (Ma and Pinker, 2012) to compute the SW↓ from ISCCP  
491 D1 and from ISCCP DX (which was first aggregated to 0.5° resolution). When the 0.5° product  
492 was upscaled to 2.5° and compared to the 2.5° derived directly from the ISCCP D1 product,  
493 differences were found when compared to ground observations. The 0.5° product upscaled to 1°  
494 had a bias of -0.5 Wm<sup>-2</sup> while the one from the ISCCP D1 had a bias of 5.7 Wm<sup>-2</sup>.

495 The MODIS products are also available at about 5 km resolution. Based on the findings  
496 reported in Ma and Pinker (2009) it is hypothesized that if the SW↓ fluxes were to be produced at  
497 that scale and upscaled to any of the resolutions used in comparison, the agreement with ground  
498 observations would improve. Another potential of improvement is to better represent the diurnal  
499 cycle of the MODIS SW↓ products. This study is a first attempt of its kind to obtain a  
500 comprehensive evaluation of the LIS/USAF SW↓ fluxes. It was shown that overall, at global scale  
501 the LIS USAF model tends to overestimate the surface SW↓ fluxes. It was also learned that  
502 compared to major re-analyses products over different climatic regions the LIS/USAF model  
503 performed frequently better than several of the reanalysis products when evaluated against satellite  
504 and ground observations.

505

506

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516 Global Modeling and Assimilation Office (GMAO), NASA/GSFC; and NCEP Reanalysis data are  
517 provided by the NOAA/OAR/ESRL PSD, Boulder, Colorado, USA. The BSRN/SURFRAD data  
518 are provided by the NOAA Earth System Research Laboratory, Global Monitoring Division  
519 (<https://www.esrl.noaa.gov/gmd/grad/surfrad/>).

## 520 521 Data Availability Statement

522 <https://bsrn.awi.de/>

523

524 <https://www.arm.gov/capabilities/observatories/sgp>

525

526 <https://climatedataguide.ucar.edu/climate-data/climate-forecast-system-reanalysis-cfsr>

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528 <https://earthdata.nasa.gov/>

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530 [https://gmao.gsfc.nasa.gov/reanalysis/MERRA-2/data\\_access/](https://gmao.gsfc.nasa.gov/reanalysis/MERRA-2/data_access/)

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533

534 Shortwave flux estimates from MODIS: Upon Request.

535

536

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