

Article

Radiation Component Calculation and Energy Budget Analysis for the Korean Peninsula Region

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Abstract: In this study, a radiation component calculation algorithm was developed using channel data from the Himawari-8 Advanced Himawari Imager (AHI) and meteorological data from the Unified Model (UM) Local Data Assimilation and Prediction System (LDAPS). In addition, the energy budget of the Korean Peninsula region in 2016 was calculated and its regional differences were analyzed. Radiation components derived using the algorithm were calibrated using the broadband radiation component data from the Clouds and the Earth's Radiant Energy System (CERES) to improve their accuracy. The calculated radiation components and the CERES data showed an annual mean percent bias of less than 3.5% and a high correlation coefficient of over 0.98. The energy budget of the Korean Peninsula region was -2.4 Wm^{-2} at the top of the atmosphere (R_T), -14.5 Wm^{-2} at the surface (R_S), and 12.1 Wm^{-2} in the atmosphere (R_A), with regional energy budget differences. The Seoul region had a high surface temperature (289.5 K) and a R_S of -33.4 Wm^{-2} (surface emission), whereas the Sokcho region had a low surface temperature (284.7 K) and a R_S of 5.0 Wm^{-2} (surface absorption), for a difference of 38.5 Wm^{-2} . In short, regions with relatively high surface temperatures tended to show energy emission, and regions with relatively low surface temperatures tended to show energy absorption. Such regional energy imbalances can cause weather and climate changes and bring about meteorological disasters, and thus research on detecting energy budget changes must be continued.

Keywords: Himawari-8 AHI; UM LDAPS; energy budget; Korean Peninsula region; energy imbalance

1. Introduction

Global warming refers to the increase in mean global temperatures caused by the increase in greenhouse gasses, which started since the dawn of industrialization at the end of the 19th century. Over the past 100 years, mean global temperatures have increased by approximately 0.6 K, and global warming is accelerating [1–3] because of increased positive radiative forcing [4] caused by increases in greenhouse gases in human-centered ecosystems [5] and land changes caused by urbanization [6]. This increase in global mean temperatures causes changes in the energy budget and leads to global energy imbalances. Such a rise in the mean global temperature causes more energy to be emitted from the surface, especially in low altitudes. Subsequently, energy (surplus) that is neither absorbed into the atmosphere nor emitted into space, moves to high latitudes where energy is relatively lower [7]. This process will result in the weather and climate changes. In short, climate change caused by global warming causes changes in ecosystems from polar regions to tropical regions [8,9] and induces natural disasters that threaten human lives [10] as well as the society and economy [11,12]. As such, continuous research and policies are required to monitor and respond to climate change [13–17].

The Earth's energy balance is the most important factor determining changes in climate and weather, and it is determined by radiation and distribution of radiation [18]. Changes in the global energy budget from the past until the present can be studied using the results of space-based observations, and currently, the global energy budget can be analyzed in detail [19]. In early 1959, the energy absorbed by white and black objects attached to the Explorer-7 satellite was converted into heat and a thermal equilibrium equation was used to calculate global radiation, thus enabling global energy budget observations. Later in the 1970s, the global radiation balance was measured by the Nimbus-7 satellite, and these data became an important component of climate change research, and they were used in analyzing the results of climate models, which were in the initial development phase at that time. Since the 1980s, Earth Radiation Budget (ERB) sensors have been developed to perform more detailed research on the radiation budget, and measurements of both solar radiation reflected by the Earth and emitted longwave radiation have been made possible. Later, sensors known as the Earth Radiation Budget Experiment (ERBE) were installed on the National Oceanic and Atmospheric Administration 9 (NOAA-9) and 10 satellites to perform continued research on the ERB [7]. After the development of ERB, high-resolution satellite sensors, such as ERBE, Scanner Radiometer for Radiation Budget (ScaRaB) [20], and the Clouds and the Earth's Radiant Energy System (CERES) [21–23], were developed. CERES provides radiative fluxes calculated with highly accurate cloud detection [24], and is currently used in energy budget research and various other research. CERES data can be used to improve not only global but also regional climate predictions, and thus, continued data monitoring and accumulation are required [6,25]. With an error rate of less than 2% over 10 years [26], the accuracy of CERES data is higher than that of other radiation data [19,27–29]. From the past until the present, numerous studies have used satellite data to extensively research the global energy budget and uncertainty in radiation components [18,19,25,30–33]. Climate change caused by changes and imbalances in the energy budget are a very important issue not only globally but also regionally. However, analyses on this issue are limited and contain systematic errors because most data on the regional energy budget analysis are not detailed and have a resolution of tens or hundreds of kilometers [34].

This study applied high resolution data to calculate the energy budget of the Korean Peninsula region and demonstrates changes in the regional energy budget and regional energy imbalances. For this purpose, spectral data from the Himawari-8 Advanced Himawari Imager (AHI) [35], as well as meteorological data, sensible heat flux (SHF), and latent heat flux (LHF) from the Local Data Assimilation and Prediction System (LDAPS) analysis field were used. In addition, an algorithm was developed to calculate shortwave radiation components. The algorithm uses the Himawari-8 AHI's shortwave and longwave spectrals to calculate the reflected shortwave radiation (RSR) and outgoing longwave radiation (OLR), which are radiation components at the top of the atmosphere, and the LDAPS' meteorological data to calculate downward shortwave radiation (DSR), absorbed shortwave radiation (ASR), downward longwave radiation (DLR), and upward longwave radiation (ULR), which are radiation components at the surface. To improve their accuracy, the radiation components calculated by the developed algorithm were calibrated with the broadband radiation components of the CERES sensor installed on a polar orbit satellite. Based on these results, this study produced an energy budget for the Korean Peninsula region and analyzed changes and regional differences in the energy budget. Section 2 describes the research data in this study. Section 3 describes the radiation component calculations and calibration with CERES data. Section 4 shows the verification of calculated radiation components and the results of the energy budget analysis. Finally, Section 5 presents a summary and conclusions.

2. Data

The radiation components required for calculating the energy budget are as follows: ISR (incoming solar radiation from outside the atmosphere measured at the top of the atmosphere), RSR (radiation reflected into space from the surface, atmosphere, and clouds), DSR (radiation which goes through the

atmosphere and reaching the Earth's surface), ASR (the net radiation, excluding radiation reflected to the atmosphere and space by albedo at the surface), ULR (radiation absorbed by the surface and emitted back to the atmosphere and space), DLR (radiation absorbed by the atmosphere and clouds and emitted back towards the surface), OLR (radiation emitted from the surface, atmosphere, and clouds toward space), and SHF and LHF (non-radiative fluxes). This study used satellite data and the analysis field of a numerical weather prediction (NWP) model to calculate the energy budget.

AHI channel data from the Himawari-8 geostationary satellite [35] were used to calculate RSR (AHI 1–6 channels) and OLR (AHI 8 and 15 channels). Himawari-8 was launched in 2014 and floats over the equator at 140.7°E and 0°N, observing a hemisphere area, which includes the Korean Peninsula region. The AHI sensor has very similar channel characteristics as the Advanced Meteorological Imager (AMI) of South Korea's next-generation satellite, the Geostationary Korea Multi-Purpose satellite 2A (GK-2A, scheduled to launch in 2018) [36]. It has 6 shortwave channels and 10 longwave channels and provides data of 2 km × 2 km resolution at 10-min intervals [37]. This sensor provides radiance observed through a narrow band; therefore, the broadband radiation component was calculated through an empirical method (refer to Section 3).

The radiation components on the surface were calculated by an empirical method using LDAPS data of the Unified Model (UM) [38]. This model was introduced by the United Kingdom Meteorological Office (UKMO) and is currently being used by the Korea Meteorological Administration (KMA). The model provides data on 86 variables regarding isobaric surface and surface for each 1.5 km × 1.5 km grid cell at 3-h intervals, which are sufficient for the purpose of the study. Similar numerical weather prediction models, such as the National Centers for Environmental Prediction (NCEP) Final (FNL) (<https://rda.ucaredu/datasets/ds083.2>) and ERA-Interim [39], provide the required meteorological data but their spatiotemporal resolution is not sufficient (6-h intervals, 1° × 1°). Therefore, they were not used as input data.

In this study, changes in the energy budget of the Korean Peninsula region (122–132°E, 33–43°N; Figure 1) from 1 January 1 to 31 December 2016 were analyzed. This period was selected for the purpose of collecting latest data (Himawari-8 AHI data collection was possible from August 2015) required for energy budget calculations. In addition, the temperature of the Korean Peninsula region in that year was 286.6 K, approximately 1 K higher than the mean temperature from 1981 to 2010 (30 years); however, the precipitation (1272.5 mm), number of days with precipitation (109 days), and cloud cover (53%) were similar to normal year mean [40,41]. Hence, it was thought that the data selected would show the typical energy budget of the Korean Peninsula region. For the radiation component calculations, data were produced eight times per day at 3-h intervals starting at 0000 UTC based on the temporal resolution of LDAPS data. Some Himawari-8 AHI data are missing for this time, and if shortwave radiation or longwave radiation component data were missing two times or four times, respectively, out of eight times, the day was excluded from the analysis period. As such, January 14, 17, 20, 21, and 24; February 9, 10, and 27; April 10, 17, 18, and 19; May 11; June 21; September 18; and December 24 were excluded from the analysis (for a total of 350 days). In addition, the spatial grid was created based on Himawari-8 AHI's 2 km × 2 km grid.

CERES is a next-generation ERB sensor, and it is the only sensor on currently operating satellites that provides shortwave (0.2–5 µm) and longwave (5–100 µm) broadband radiation [26]. Therefore, the radiation components calculated by the algorithm developed in this study were compared to broadband radiation data from CERES. CERES is installed on the Terra, Aqua, and Suomi National Polar-Orbiting Partnership (NPP) platforms. The data used for verification and calibration of the calculated radiation components were obtained from the Terra CERES single scanner footprint (SSF) Level2 Edition3A (Ed3A) [42], which observes the Korean Peninsula region in 16-day cycles. Terra CERES data were used because it passes the Korean Peninsula around 1200 local standard time (LST) when the maximum radiation occurs (nighttime verification data for longwave radiation were collected at around 0000 LST). In the verification and calibration, the spatial resolution of the calculated data was set at 20 km × 20 km to match that of CERES data. The radiation components were calculated at 3-h intervals and interpolated to

match the CERES observation times. For the calibrated radiation components, an analysis was performed on the CERES SYN1d Ed3A, which provides daily mean data at a spatial resolution of $1^\circ \times 1^\circ$, and the daily mean time series data, and then the calculation accuracy was evaluated.

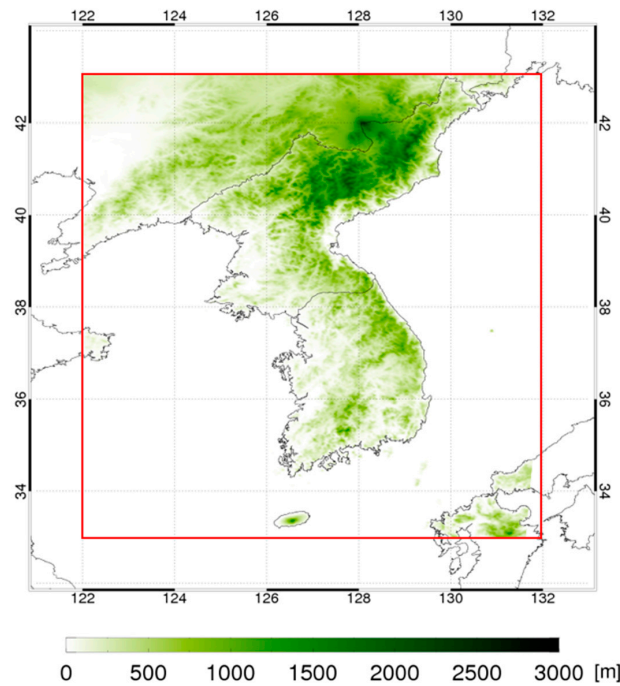


Figure 1. Korean Peninsula region (red box) and altitude distribution for energy budget calculation.

CERES Energy Balanced and Filled (EBAF) Ed4.0 [43,44] and Modern Era Retrospective Analysis for Research and Applications Version 2 (MERRA-2) M2T1NXFLX data [45] were used for comparing energy budgets. SHF and LHF data from MERRA-2 were used because CERES data do not provide them. MERRA-2 data include atmosphere reanalysis data from the 1980s and after, and they are used in conjunction with GEOS-5 data [46]. These data have several advantages for analysis in that their accuracy has been confirmed by [47] and [48] and their grid size is smaller than that of other global models [49]. The spatial resolutions of CERES EBAF and MERRA-2 data are $1^\circ \times 1^\circ$ and $0.625^\circ \times 0.5^\circ$, respectively. Therefore, the spatial resolutions were modified to match that of CERES for the analysis.

3. Calculation of Radiation Components and Energy Budget

Sections 3.1 and 3.2 describe the algorithm for calculating the shortwave/longwave radiation components needed to calculate the energy budget. In Section 3.3, the calculated radiation components are compared and calibrated. Section 3.4 describes the methods for calculating the energy budget at the top of the atmosphere, the surface, and the atmosphere.

3.1. Shortwave Radiation Components

3.1.1. ISR

To calculate the net radiation at the top of the atmosphere, ISR was calculated using the method developed by [50], as shown in Equations (1)–(3).

$$\Gamma = 2\pi(d_n - 1)/365 \quad (1)$$

$$E_0 = (r_0/r)^2 = 1.000110 + 0.034221 \cos \Gamma + 0.001280 \sin \Gamma + 0.000719 \cos 2\Gamma + 0.000077 \sin 2\Gamma \quad (2)$$

$$\text{ISR} = S_0 E_0 \cos \theta \quad (3)$$

Here, Γ is the day angle, d_n is the Julian day, E_0 is the eccentricity, θ is the solar zenith angle (SZA), and S_0 is the solar constant, which is 1361 Wm^{-2} [51].

3.1.2. RSR

RSR is ISR that enters from the top of the atmosphere and is reflected back out of the atmosphere by the surface, the atmosphere, and clouds. It was calculated using Himawari-8 AHI's shortwave channels (channels 1–6). The broadband albedo at the top of the atmosphere was calculated, and the RSR was calculated according to the sun zenith angle and eccentricity as in Equations (4) and (5). Here, the required regression coefficients were calculated via the linear relationship between the narrowband reflectance and the broadband reflectance for each channel [52]. In addition, because the atmosphere is anisotropic, a chart of coefficients for each direction was created, taking into account the sun zenith angle, viewing zenith angle, and relative azimuth between the sun and the satellite [53,54]. For the relationship between the narrowband reflectance and the broadband reflectance, the Santa Barbara DISORT Atmospheric Radiative Transfer (SBDART) [55] was used to simulate various atmosphere conditions that may be present in the actual atmosphere (Table 1), and the calculated results were used [56].

$$A = a_1(\theta, \Phi, \Omega)\rho_{0.47} + a_2(\theta, \Phi, \Omega)\rho_{0.51} + a_3(\theta, \Phi, \Omega)\rho_{0.64} + a_4(\theta, \Phi, \Omega)\rho_{0.86} + a_5(\theta, \Phi, \Omega)\rho_{1.61} + a_6(\theta, \Phi, \Omega)\rho_{2.26} \quad (4)$$

$$\text{RSR} = AS_0E_0 \cos \theta \quad (5)$$

Here, A is the broadband albedo at the top of the atmosphere; Φ is the viewing zenith angle (VZA); Ω is the relative azimuth angle (RAA); a_1 – a_6 are the regression coefficients according to the SZA, VZA, and RAA required for calculating A ; ρ is the narrowband reflectance; and the subscripts are the central wavelengths for each of Himawari-8 AHI's shortwave channels.

Table 1. Settings of the radiative transfer model for calculating the broadband albedo at the top of the atmosphere

Parameter	Values	N
Spectral Band (μm)	0.41–0.51, 0.47–0.55, 0.56–0.71, 0.81–0.91, 1.55–1.67, 2.19–2.32, 0.2–3.3	7
Atmospheric Profile	Tropical, Mid-Latitude Summer & Winter, Sub-Arctic Summer & Winter, US Standard	6
SZA ($^\circ$)	0, 5, 10, 15, 20, 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 75, 80, 85	18
RAA ($^\circ$)	0, 10, 20, 30, 40, 50, 60, 70, 80, 90, 100, 110, 120, 130, 140, 150, 160, 170, 180	19
Aerosol Optical Thickness (AOT)	Rural, Urban, Oceanic, Tropospheric: VIS 1, 5, 10, 15, 20 km	20
Cloud Optical Thickness (COT)	2, 4, 8, 16, 32, 64, 128	7
Cloud Height (km)	2, 4, 6, 8, 10, 12, 14, 16	8
Surface Albedo	0.0, 1.0, Water (spectral range mean: 0.04), Vegetation (0.28), Sand (0.32), Snow (0.43)	6

3.1.3. DSR and ASR

DSR is the shortwave radiation which through the Earth's atmosphere and arriving at the surface, excluding RSR, which is reflected at the top of the atmosphere. ASR is the net shortwave radiation excluding the radiation reflected to the atmosphere or outside the atmosphere due to surface albedo (α) [57]. ASR is defined as $(1 - \alpha) \times \text{DLR}$, but surface albedo is one of the meteorological data variables with high uncertainty. Consequently, its values widely vary depending on the time and location, and it is difficult to use highly accurate data at each moment because of limitations in the provided data [58]. Therefore, in this study, surface albedo data were not used, and the linear relationship between RSR and total precipitable water (TPW) was used to calculate DSR (Equations (6)–(8)) and ASR (Equations (9)–(11)) [59,60]. Here, the radiative transfer model was simulated and the relationships between radiation and meteorological variable were determined, as shown in Table 2. The created regression coefficients (b_0 – b_5 and c_0 – c_5) are shown in Table 3.

$$\text{DSR} = \text{offset} + \text{slope} \times \text{RSR} \quad (6)$$

$$\text{offset} = b_0 \cos \theta - b_1 \ln(\text{TPW}) - b_2 \quad (7)$$

$$\text{slope} = b_3 \cos \theta - b_4 \ln(\text{TPW}) - b_5 \quad (8)$$

$$\text{ASR} = \text{offset} + \text{slope} \times \text{RSR} \quad (9)$$

$$\text{offset} = c_0 \cos \theta - c_1 \ln(\text{TPW}) - c_2 \quad (10)$$

$$\text{slope} = c_3 \cos \theta - c_4 \ln(\text{TPW}) - c_5 \quad (11)$$

Table 2. Settings of the radiative transfer model for calculating DSR and ASR

Parameter	Values	N
Spectral Band (μm)	0.2–3.3	1
Atmospheric Profile	Tropical, Mid-Latitude Summer & Winter, Sub-Arctic Summer & Winter, US Standard	6
CSZA	0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1.0	10
TPW (cm)	0.2, 0.4, 0.8, 1.6, 3.2, 6.4	6
AOT	Rural, Urban, Oceanic, Tropospheric: 0.02, 0.04, 0.08, 0.16, 0.32, 0.64, 1.28	28
COT	2, 4, 8, 16, 32, 64, 128	7
Cloud Height (km)	2, 4, 6, 8, 10, 12, 14, 16	8
Cloud Effective Radius (μm)	Water Cloud: 8, 16, 32, 64 Ice Cloud: 16, 32, 64, 128	8
Surface Albedo	0.0, 1.0, Water (0.04), Vegetation (0.28), Sand (0.32), Snow (0.43)	6

Table 3. Regression coefficients for calculating DSR and ASR

	b_0	b_1	b_2	b_3	b_4	b_5
DSR	1.112×10^3	1.308×10^1	7.522×10^1	-8.829×10^{-2}	-4.931×10^{-3}	9.566×10^{-1}
	c_0	c_1	c_2	c_3	c_4	c_5
ASR	1.060×10^3	1.429×10^1	5.433×10^1	-8.412×10^{-2}	-4.700×10^{-3}	9.119×10^{-1}

3.2. Longwave Radiation Components

3.2.1. OLR

OLR is longwave radiation at the top of the atmosphere emitted by the Earth's surface, the atmosphere, and clouds. It was calculated using Himawari-8 AHI's longwave channels (channels 8 and 15). Based on the high correlation between narrowband irradiance and broadband irradiance, similar to that in RSR [61], narrowband radiance was converted to narrowband irradiance considering the atmosphere's anisotropy via Equations (12)–(15) [62,63], and the converted narrowband irradiance was converted into broadband irradiance [64]. The radiative transfer model was simulated using parameters as shown in Table 4, and the coefficients developed by [65] were used in the equations (Tables 5 and 6).

$$F = A(\theta)L(\theta) + B(\theta) \quad (12)$$

$$A(\theta) = k_0 + k_1(\sec \theta - 1) + k_2(\sec \theta - 1)^2 \quad (13)$$

$$B(\theta) = k_3 + k_4(\sec \theta - 1) + k_5(\sec \theta - 1)^2 \quad (14)$$

$$\text{OLR} = d_0 + d_1 F_{6.2} + d_2 F_{6.2}^2 + d_3 F_{6.2}^3 + d_4 \ln F_{12.4} + d_5 \ln F_{12.4}^2 \quad (15)$$

Here, F is the irradiance (the subscripts are the central wavelengths of Himawari-8 AHI's channels 8 and 15); L is the radiance; k_0 – k_5 are the regression coefficients for converting radiance to irradiance; and d_0 – d_5 are the regression coefficients for calculating OLR.

Table 4. Settings of the radiative transfer model for calculating OLR

Parameter	Values	N
Spectral Band (μm)	5.44–7.03, 11.18–13.65, 3.3–100	3
Atmospheric Profile	Tropical, Mid-Latitude Summer & Winter, Sub-Arctic Summer & Winter, US Standard	6
SZA ($^\circ$)	0, 5, 10, 15, 20, 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 75, 80, 85	18
COT	2, 4, 8, 16, 32, 64, 128	7
Cloud Height (km)	0, 2, 4, 6, 8, 10, 12, 14, 16	9
Temperature, H ₂ O, CO ₂ , O ₃ , etc.	Default	

Table 5. Regression coefficients of each channel for converting radiance into irradiance

Channel	k_0	k_1	k_2	k_3	k_4	k_5
8	2.672	7.049×10^{-1}	-4.051×10^{-2}	1.002×10^{-1}	-1.444×10^{-1}	8.956×10^{-3}
15	3.018	1.115×10^{-1}	-4.938×10^{-3}	4.505×10^{-2}	8.430×10^{-2}	-9.793×10^{-3}

Table 6. Regression coefficients for converting narrowband irradiance to OLR

d_0	d_1	d_2	d_3	d_4	d_5
8.008×10^1	1.663	3.706×10^{-1}	-5.837×10^{-2}	-2.483×10^1	2.828×10^1

3.2.2. DLR

DLR is longwave radiation emitted from clouds and the atmosphere that arrives at the surface. It can be calculated using its high correlation with weather variables such as the temperature near the surface and cloud information, such as total cloud cover, cloud emissivity, and cloud base temperature [66–70]. This study used the method presented by [71], in which DLR is calculated empirically using air temperature, total cloud cover, and water vapor pressure (Equation (16)). However, directly using their method resulted in discontinuous DLR calculations at lowland and mountain sites and underestimations of DLR due to total cloud cover. Therefore, the method was modified using the relationship between air temperature and total cloud cover to calibrate the coefficients, as in Equation (17). Different results were obtained using the modified method, as shown in Figure 2. Here, the data used in this study (total 350 days) were compared with CERES DLR. Using the method of [71], the obtained Bias, RMSE, and R were -24.54 Wm^{-2} , 27.34 Wm^{-2} , and 0.98, respectively. The results were improved to -0.14 Wm^{-2} , 12.93 Wm^{-2} , and 0.99 using the modified method.

$$\text{DLR} = \sigma T_a^4 \left\{ 1 - X_s \exp \left(-\frac{Y_s e_a}{T_a} \right) \right\} (1 + Z_s CF^2) \quad (16)$$

$$\text{DLR} = \sigma T_a^4 \left\{ 1 - e_1 \exp \left(-\frac{e_2 e_a}{T_a} \right) \right\} \left\{ 1 + (e_3 - e_4 T_a) CF^2 \right\} \quad (17)$$

Here, the X_s , Y_s , and Z_s coefficients are respectively 0.35, 10.0 K hPa^{-1} , and 0.0035 in lowland areas with an altitude of less than 212 m, and they are 0.43, 11.5 K hPa^{-1} , and 0.0050 in mountainous areas. T_a is the air temperature; e_a is the water vapor pressure; CF is the cloud cover; and e_1 – e_4 are the coefficients for calculating DLR. e_1 – e_4 are the coefficients from when the smallest difference exists between the CERES DLR and the results of calculating Equation (17) using MERRA-2 air temperature and water vapor pressure and CERES total cloud cover, as shown in Table 7.

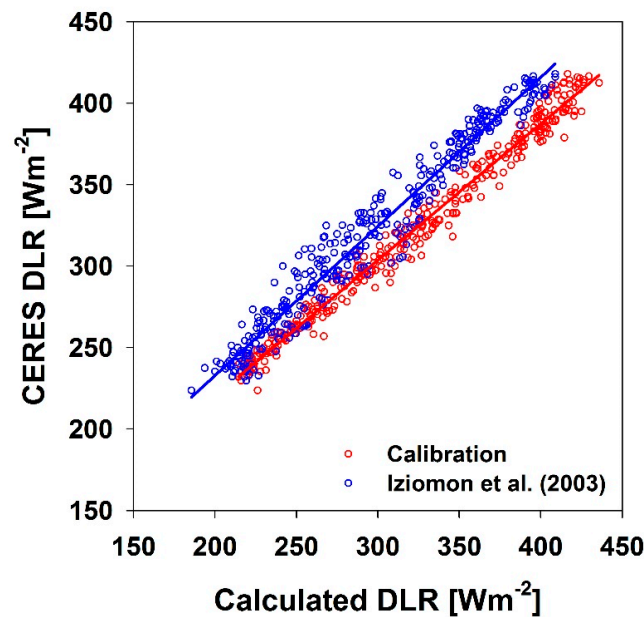


Figure 2. Scatter plot of calculated DLR (red: calibrated result in this study, blue: [71] method) and CERES DLR.

Table 7. Regression coefficients for calculating DLR.

e_1	e_2	e_3	e_4
0.390	1.049×10^1	3.396	0.011

3.2.3. ULR

ULR is DLR reflected from the surface and longwave radiation emitted from the surface. It can be calculated using Equation (18).

$$\text{ULR} = \varepsilon \sigma T_s^4 + (1 - \varepsilon) \text{DLR} \quad (18)$$

Here, ε is the surface emissivity; T_s is the surface temperature; and σ is the Stefan–Boltzmann constant ($5.67 \times 10^{-8} \text{ Wm}^{-2} \text{ K}^{-4}$). The surface emissivity for calculating ULR is distributed within the range of 0.9 and 1, depending on the properties of the surface. It is a core component in calculating ULR [72], but it involves high uncertainty depending on the surface properties and time and location. Furthermore, if it is assumed to be a constant (long-term mean), a maximum difference of over 10% with the observed surface emissivity may result [73,74]. Therefore, in this study, an algorithm to calculate ULR is presented without using surface emissivity which has a relatively higher uncertainty than other meteorological data, as with ASR.

Infrared atmospheric transmittance (T_{LW}) is defined as a ratio of ULR and OLR (Equation (19)), and ULR can be expressed as Equation (20). That is, ULR can be calculated even without surface emissivity data by using OLR and infrared atmospheric transmittance.

$$T_{LW} = \text{OLR} / \text{ULR} \quad (19)$$

$$\text{ULR} = \text{OLR} / T_{LW} \quad (20)$$

When the results of the simulation for calculating OLR from Section 3.2.1 were used, atmospheric transmittance showed a linear relationship with OLR as shown in Figure 3, and the slope showed other features according to surface temperature. Ultimately, the relationship between the three variables can be described by Equation (21). In this study, only OLR and surface temperature were used to calculate ULR.

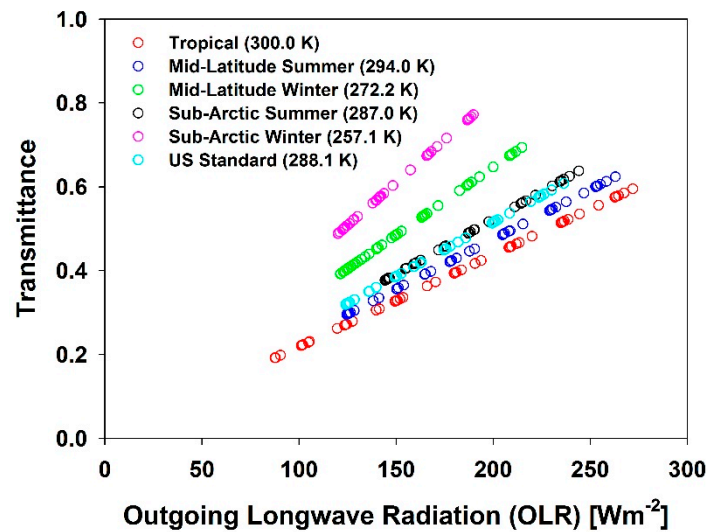


Figure 3. Scatter plot of infrared atmospheric transmittance according to surface temperature and OLR.

$$ULR = OLR / [f_0 + OLR\{f_1 \exp(f_2 T_s)\}] \quad (11)$$

Here, f_0 – f_2 are the regression coefficients for calculating the ULR, as shown in Table 8.

Table 8. Regression coefficients for calculating ULR.

f_0	f_1	f_2
-6.654×10^{-5}	0.171	−0.015

3.3. Calibration Using CERES and MERRA-2 Data

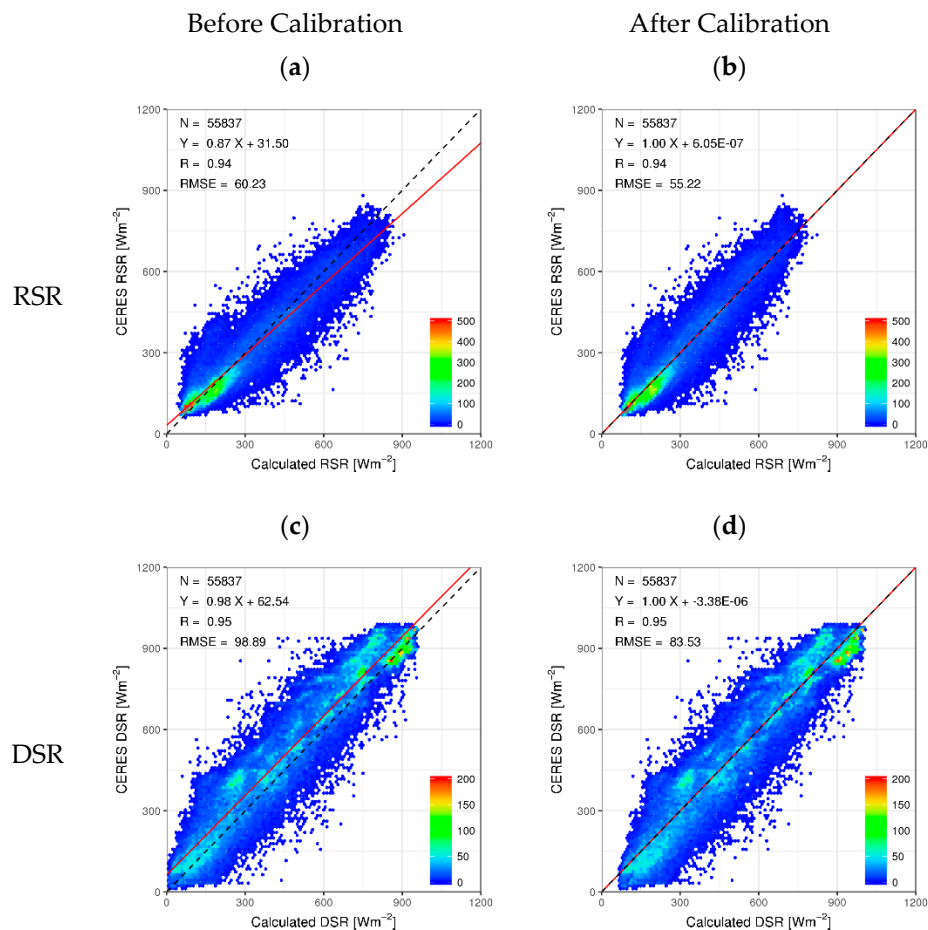
The radiation components calculated in this study were compared to the radiation of Terra CERES SSF Level2 Ed3A. Terra CERES observes the Korean Peninsula region at approximately 16-day intervals. The example shown in Table 9 was set up to perform the analysis. Shortwave radiation components were observed only during the daytime, and thus only data from 0230 UTC were used. Longwave radiation components were observed during both daytime and nighttime, and thus data from 0230 UTC and 1330 UTC were used. Here, cases with missing Himawari-8 AHI data were excluded from the analysis.

For the calculated radiation components, data were generated at 3-h intervals starting at 0000 UTC. Therefore, data from 0000 UTC and 0300 UTC or 1200 UTC and 1500 UTC were interpolated to match the times of the verification data, and the analysis was performed. Furthermore, the calculated radiation components were calibrated using a linear relationship ($Y = AX + B$). The calibration was performed using highly accurate data from long-term observations to compensate for nonlinearity and various unconsidered problems that may arise in the empirical method, the calculation accuracy could be improved [75,76]. The before and after calibration results for each radiation component showed that the difference between the after calibration results and the CERES data was smaller than the difference between the before calibration results and the CERES data, as shown in Figures 4 and 5. In the case of shortwave radiation, the radiation was large, and there was a large difference in the radiation irrespective of the existence of clouds; thus, the RMSE was larger than that of longwave radiation. That is, interpolation was performed based on the spatiotemporal resolution of CERES, and thus, the RMSE was found to be large. Nevertheless, the calibrated radiation components showed a high concentration along the 1:1 line. The slopes were all calibrated to 1, and the radiation showed a high correlation coefficient with the CERES radiation at over 0.94.

Table 9. CERES example used in the verification of each calculated radiation component and observation time

Shortwave Radiation (RSR, DSR, ASR)			Longwave Radiation (OLR, DLR, ULR)	
Month	Day	Time	Day	Time
1	4, 20	0230 UTC	4, 20	0230, 1330 UTC
2	5, -		5, -	
3	8, 24		8, 24	
4	2, -		2, 18	
5	4, 20		4, 20	
6	5, -		5, 21	
7	7, 23		7, 23	
8	6, 22		6, 22	
9	7, 23		7, 23	
10	2, 18		2, 18	
11	3, 19		3, 19	
12	5, 21		5 *, 21	

* Excluded from analysis because Himawari-8 AHI data at 0230 UTC were missing. - Missing case.

**Figure 4.** Cont.

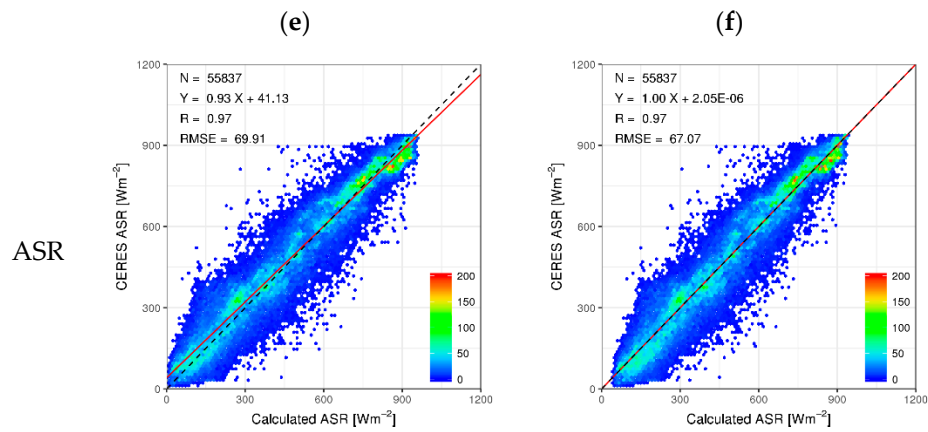


Figure 4. Density scatter plot of calculated shortwave radiation components and CERES shortwave radiation before (a,c,e) and after (b,d,f) calibration. Here, N is the total number of samples used for verification. The black dash line is the 1:1 line and the red line is the regression line. The unit of RMSE is Wm^{-2} .

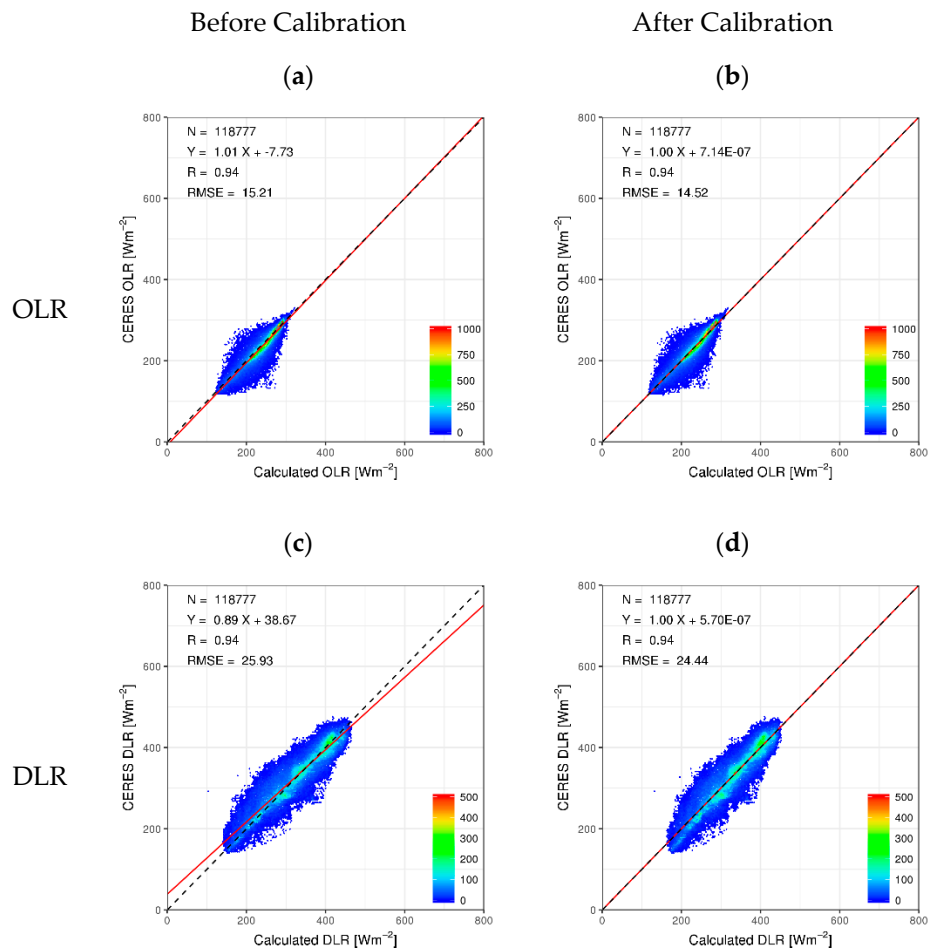


Figure 5. Cont.

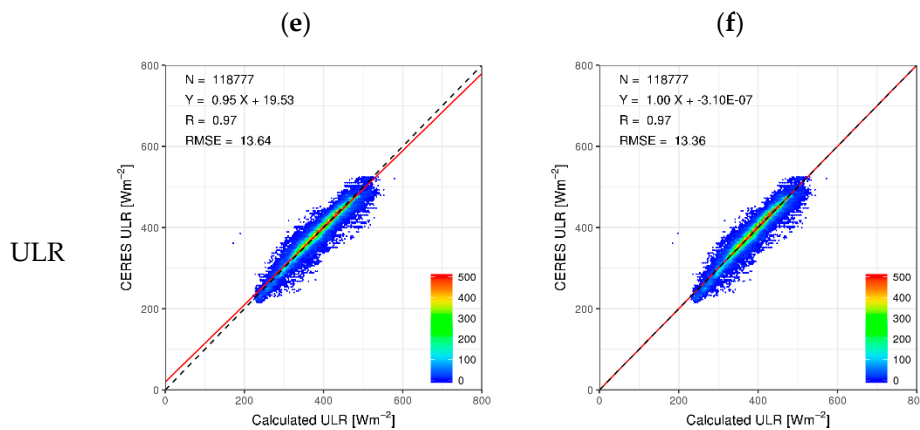


Figure 5. Density scatter plot of calculated longwave radiation components and CERES longwave radiation before (a,c,e) and after (b,d,f) calibration. Here, N is the total number of samples used for verification. The black dash line is the 1:1 line and the red line is the regression line. The unit of RMSE is Wm^{-2} .

In the calculation of the energy budget at the surface, large fluxes of LDAPS SHF and LHF were found to be emitted during spring and winter due to the effect of the Kuroshio current, as shown in Figure 6. LHF showed a different trend than SHF due to the effects of typhoons, which moved northward near the Korean Peninsula during summer and fall (e.g., Lionrock at the end of August and Namtheun at the beginning of September). For verifying SHF and LHF, values from a real ship and remote sensing can be compared, but observation data is very rare, and remote sensing includes a relatively large flux of calculation errors; therefore, a comparison was normally made with model results [77]. However, when SHF and LHF calculated from model results were compared to the real observed values, SHF included $\sim 25\%$ uncertainty and LHF included 10% uncertainty [25,78]. Therefore, in order to improve the accuracy of the LDAPS SHF and LHF, they were calibrated with MERRA-2, which performs global data assimilation. Before calibration, SHF and LHF had RMSE values of 8.56 Wm^{-2} and 15.70 Wm^{-2} , respectively, when compared to the MERRA-2 data, and after calibration their RMSE improved to 4.35 Wm^{-2} and 9.62 Wm^{-2} .

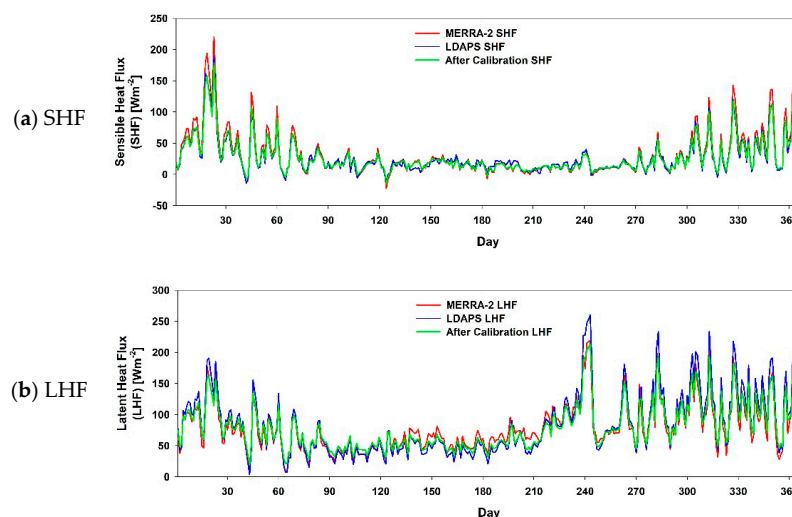


Figure 6. Daily mean (a) SHF and (b) LHF of MERRA-2 and LDAPS in 2016 before and after calibration.

3.4. Energy Budget

The radiation components described in Sections 3.1 and 3.2 and LDAPS SHF and LHF were used to calculate the energy budget as described below. Equation (22) is the net radiation at the top of the

atmosphere (R_T), which can be shown as the difference between the incoming solar radiation ($SW^\downarrow$) and the reflected shortwave radiation at the top of the atmosphere ($SW^\uparrow$) and the emitted longwave radiation ($LW^\uparrow$).

$$R_T = SW^\downarrow - SW^\uparrow - LW^\uparrow = \text{ISR} - \text{RSR} - \text{OLR} \quad (12)$$

The net radiation at the surface is defined by the change in thermal energy over time, as shown in Equation (23), and it can be considered as the difference in shortwave net radiation and longwave net radiation (R_N) and the difference in SHF and LHF [79,80]. Following the law of conservation of energy, the net radiation at the surface (R_S) achieves an energy balance as in Equation (26), and it can be expressed as Equation (27).

$$dQ/dt = R_N - \text{SHF} - L_e E \quad (13)$$

$$R_N = (1 - \alpha)SW^\downarrow - \epsilon(\sigma T_s^4 - LW^\downarrow) = \text{ASR} + \text{DLR} - \text{ULR} \quad (14)$$

$$E = \text{LHF}/L_e \quad (15)$$

$$R_N - \text{SHF} - L_e E = 0 \quad (16)$$

$$R_S = \text{ASR} + \text{DLR} - \text{ULR} - \text{SHF} - \text{LHF} \quad (17)$$

Here, L_e is the evaporative latent heat of the water vapor.

The difference between the net radiation at the top of the atmosphere and the net radiation at the surface is expressed as the net radiation in the atmosphere (R_A), as in Equation (28).

$$R_A = R_T - R_S \quad (18)$$

4. Results

4.1. Verification of Calculated Radiation Components

For the calibrated radiation components, a 2016 daily mean time series analysis was performed for the Korean Peninsula region before calculating the energy budget, as shown in Figures 7 and 8. Figure 7 shows the time series distribution of shortwave radiation components, revealing a large amount of ISR and RSR on the Korean Peninsula in the summer season. RSR varied depending on total cloud cover. In August (Julian day 214–244), the total cloud cover was 46%, which was less than the mean total cloud cover in summer and fall, which was 57%. Therefore, RSR was low. In contrast, DSR and ASR showed the opposite trend. There were large longwave radiation components in summer and fall when the surface temperature was high, and the radiation became smaller in spring and winter (Figure 8). OLR showed no significant variations throughout the year, but there were differences in DLR and ULR according to the temperature near the surface. The calculated radiation components showed a similar time series distribution as the CERES radiation components, and the differences between CERES and calculated radiation are shown in Table 10.

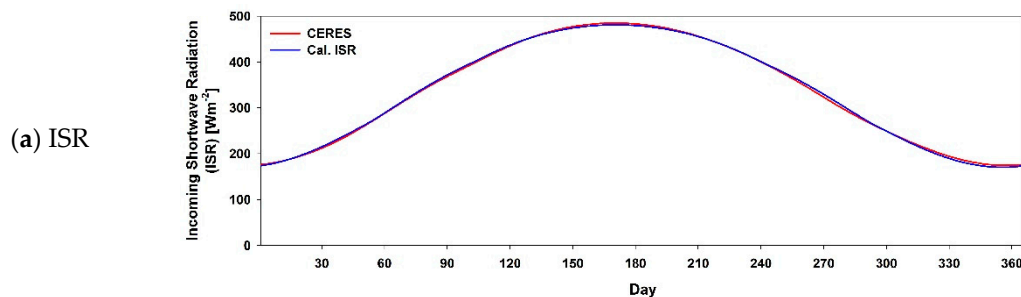


Figure 7. Cont.

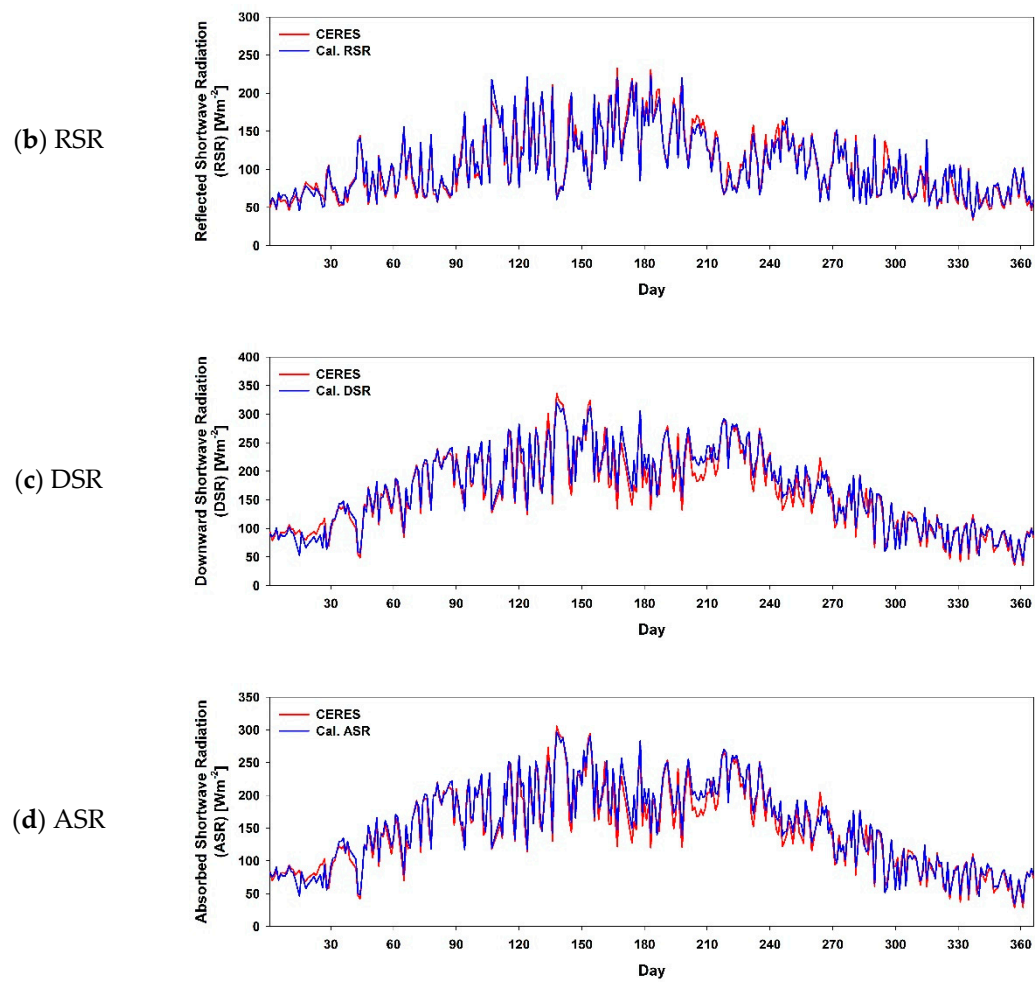


Figure 7. Daily mean time series distribution of CERES and calculated (a) ISR; (b) RSR; (c) DSR; and (d) ASR.

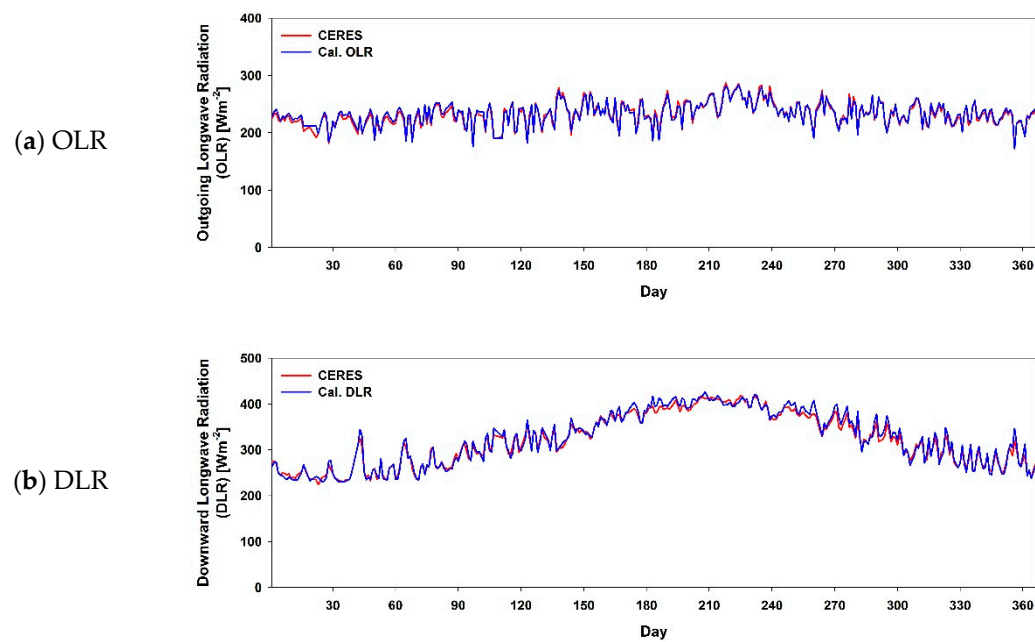


Figure 8. Cont.

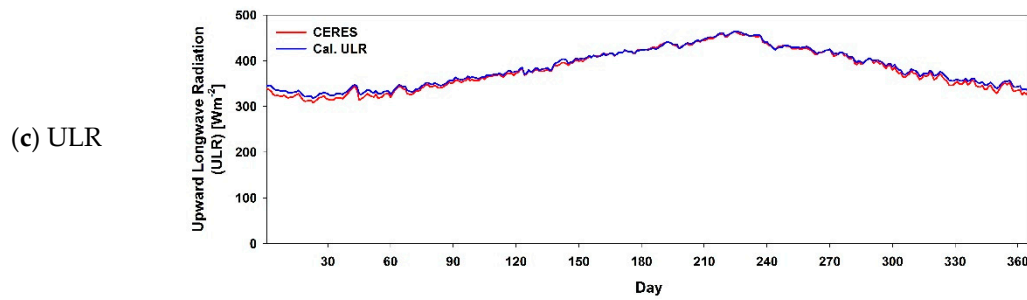


Figure 8. Daily mean time series distribution of CERES and calculated (a) OLR; (b) DLR; and (c) ULR.

Table 10. Bias, RMSE and correlation coefficient (R) of CERES and calculated radiation components using 2016 daily mean data. The mean, bias, and RMSE units are Wm^{-2} .

	CERES Mean	Bias	RMSE	R
ISR	337.47	-1.86×10^{-5}	3.13	1.000
RSR	105.99	−1.06	9.12	0.978
DSR	164.46	5.10	15.12	0.978
ASR	149.00	5.26	13.61	0.980
OLR	232.86	0.94	4.41	0.978
DLR	321.52	3.26	8.66	0.992
ULR	381.46	4.27	5.62	0.998

ISR was very similar to the radiation observed by the Total Irradiance Monitor (TIM) onboard NASA's Glory satellite [81], and the radiation component calibrated in this study showed a percent bias ($= \text{bias} / \text{mean} \times 100$) of less than 3.5%. The RSR and OLR showed a bias of approximately 1 Wm^{-2} , but this was calculated based on real observed satellite data, and thus, its calculation precision was higher than that of other radiation components. The radiation components calculated using meteorological data showed variations according to the accuracy of the input data, leading to large differences with CERES data [58,73,82]. In the case of DSR and ASR, the broadband albedo at the top of the atmosphere (A), which was used as input data, was 0.31. This value is the same as that of CERES, but the LDAPS TPW (1.69 cm) was 0.26 cm smaller than CERES TPW (1.95 cm). That is, the reduction in radiation due to TPW was underestimated, and thus, DSR and ASR were larger than that of CERES data during summer when TPW is high (Figure 7c,d). The bias and RMSE of DSR with the CERES data were 12.16 Wm^{-2} and 20.10 Wm^{-2} , respectively, and those of ASR were 12.13 Wm^{-2} and 18.42 Wm^{-2} , which were larger than in other seasons. In the case of DLR and ULR, there were differences in the calculated radiation according to the temperature near the surface (air/surface). For DLR, there was no difference between the calculated mean air temperature (285.56 K) and the MERRA-2 value (285.26 K). However, for ULR, there was a difference of approximately 1.25 K between the calculated surface temperature (286.97 K) and the CERES value (285.73 K). This difference was largest during winter when the CERES ULR and the calculated ULR had bias and RMSE of 8.72 Wm^{-2} and 9.04 Wm^{-2} , respectively, which were larger than those during other seasons (Figure 8c). In this study, the calculated radiation values were calibrated to achieve high-accuracy, high-resolution energy budget. However, the calculated radiation values contain uncertainty and inaccuracies due to the differences in the input data described here. Therefore, the input data must be improved through future research.

In addition to satellite data, DSR and DLR data observed from a pyranometer (CMP21) and pyrgeometer (CGR3) at the Gangneung-Wonju National University (GWNU) observatory (128.9°E , 37.8°N) [83] were used to verify for calculated DSR and DLR in this study. CERES data were also compared with GWNU observation data. As a result, both the CERES and the calculated DSR and DLR showed a time series distribution as shown in Figure 9, and it was similar to the distribution of the GWNU observation data, and the statistical analysis results are shown in Table 11. The correlation coefficients with observed data were similar or higher for the calculated results than CERES, and bias

and RMSE were approximately 10 Wm^{-2} less than CERES. Such a difference in the regional radiation is thought to have occurred because the grid size provided in the CERES data is relatively larger, at $1^\circ \times 1^\circ$. Therefore, the radiation data with $2 \text{ km} \times 2 \text{ km}$ grid, which was suggested by this study for a regional analysis, seems to be more appropriate than the CERES data.

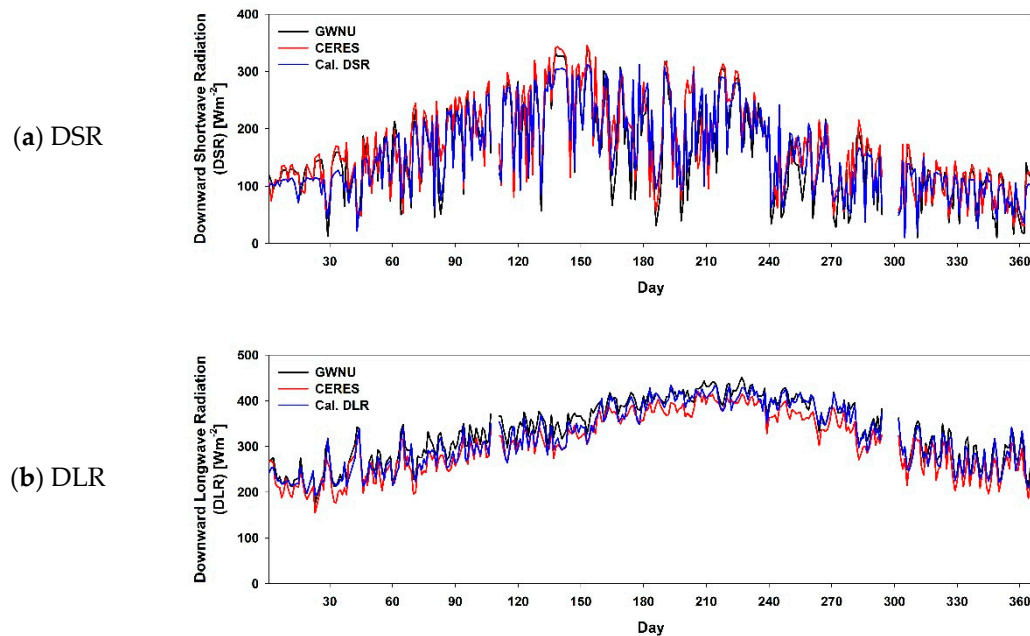


Figure 9. Daily mean time series distribution of GWNU observatory, CERES, and calculated (a) DSR and (b) DLR. Observation at GWNU observatory are missing in Julian day 108–110 and 295–301.

Table 11. Bias, RMSE, and correlation coefficient (R) of DSR and DLR observed at the GWNU observatory and CERES and calculated DSR and DLR. The mean, bias, and RMSE units are Wm^{-2} .

		Observation Mean	Bias	RMSE	R
DSR	CERES	156.32	12.94	37.20	0.900
	Calculation		2.94	25.66	0.946
DLR	CERES	334.67	−28.22	30.97	0.981
	Calculation		−12.68	18.55	0.979

4.2. Energy Budget of the Korean Peninsula Region

As ISR passes through the Earth's atmosphere and arrives at the surface, it is absorbed or reflected by clouds, etc., and DSR, which also arrives at the surface, is partially reflected by surface albedo. ASR, which excludes reflected radiation from DSR, is absorbed by the surface and heats the surface. The heated surface emits ULR, and SHF and LHF occur due to variations in the temperature and water vapor near the surface. Furthermore, RSR, which is reflected by the surface, atmosphere, and clouds, and the emitted OLR achieve a balance with the ISR (the energy balance between ISR, RSR, and OLR actually variations periodically [84]). The energy absorbed by the atmosphere in this process is the difference between ISR, RSR, and ASR, and this energy is emitted as DLR or OLR [79,80]. In addition, the difference between DSR and ASR can be shown as Upward Shortwave Radiation (USR). From the results of calculating the radiation components in this study, the energy budget for the Korean Peninsula region could be obtained, as shown in Figure 10. The calculated results showed radiation components and an energy budget similar to CERES and MERRA-2. When the radiation components calculated in this study were compared with long-term mean global radiation components and energy budgets [25,85], RSR was approximately 2.7 Wm^{-2} larger and ISR and OLR

were approximately 5.0 Wm^{-2} smaller. In polar regions, ISR is very small, approaching 0 Wm^{-2} ; therefore, even though the annual mean total cloud cover of the Korean Peninsula region was 47.4%, which is smaller than the global mean of 67.5%, its RSR was larger than the global mean. Conversely, equatorial regions have a large ISR, and the surface and atmosphere temperatures are high; thus, ISR and OLR are large. As a result, ISR and OLR in the Korean Peninsula region were smaller than the global mean. RSR was 2.7 Wm^{-2} larger and ISR and OLR were approximately 5.0 Wm^{-2} smaller [86]. Nevertheless, the Korean Peninsula lies in a region where a relative energy balance is achieved at the top of the atmosphere with a R_T of -2.4 Wm^{-2} . Regarding radiation components at the surface, shortwave radiation DSR was 16.5 Wm^{-2} , and ASR was 9.0 Wm^{-2} , whereas longwave DLR was 15.1 Wm^{-2} , and ULR was 13.2 Wm^{-2} . In tropical regions, the absorbed and emitted radiation at the surface is very large, so the Korean Peninsula region's surface radiation component values were smaller than the global mean. Conversely, SHF+LHF was approximately 5.0 Wm^{-2} larger than the global mean because the Korean Peninsula region is affected by the Kuroshio current, and it has high sensible heat and latent heat [77,87]. Because of this radiation balance, a negative net radiation (energy emission) occurred, with an R_S of -14.5 Wm^{-2} . This energy is absorbed into the atmosphere or emitted outside of the atmosphere. The energy absorbed in the atmosphere is used during land and sea energy exchange/transport [19,88] or it increases the temperatures of regions that have a relatively low temperature distribution and is transported to polar regions such that the global mean energy budget achieves a balance [7].

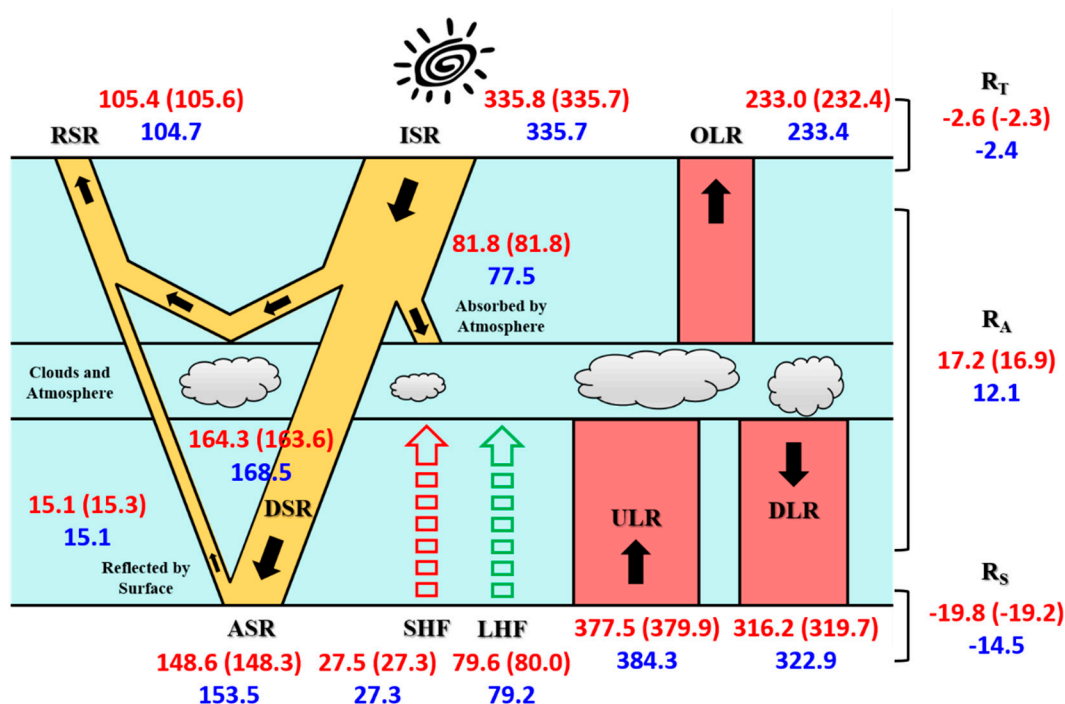


Figure 10. Annual mean radiation components and energy budget of the Korean Peninsula region in 2016. Red: Radiation components and energy budget calculated using CERES and MERRA-2 data (values in parentheses are the results of analyzing only 350 days of data used in this study); Blue: radiation components and energy budget calculated in this study.

The annual mean R_T , R_S , and R_A of the Korean Peninsula region for 2016 are shown in Figure 11. At lower latitudes, ISR and OLR are larger than RSR and the surface albedo of the land is larger than that of the sea, and thus RSR of the land becomes larger than that of the sea. Regarding R_T , the positive net radiation becomes large in low-latitude seas, and the negative net radiation becomes large in high-latitude land. R_S was similar to the distribution of SHF+LHF, and there was a particularly

strong negative net radiation of over 80.0 Wm^{-2} in the South Sea and East Sea regions, which are affected by the Kuroshio current. In the West Sea region, the Yellow Sea current—which is derived from the Kuroshio current—flows in, but as the ocean current circulates, the cold current flows from the Liaodong Peninsula and moves along the west coast (refer to Figure 11b). Therefore, the seawater temperature is low and a positive net radiation occurs. As such, the R_A in Figure 11c,d showed energy absorption in regions with a large temperature variation in different seasons and higher surface temperature, and conversely it showed energy emission in regions with a small temperature variation and lower surface temperature. That is, R_A was shown positive net radiation in metropolitan regions (e.g., Seoul (127.0°E , 37.6°N)), and negative net radiation was shown in rural regions, where the surface temperature was low (e.g., Gangneung (128.9°E , 37.8°N)) (refer to Section 4.3). Energy imbalances that arise from differences in the energy budget due to such regional weather and climate characteristics lead to mesoscale energy circulation [89], and they can bring about severe weather (e.g., drought, heavy rainfall, and snow) through changes in the weather and climate [2,3].

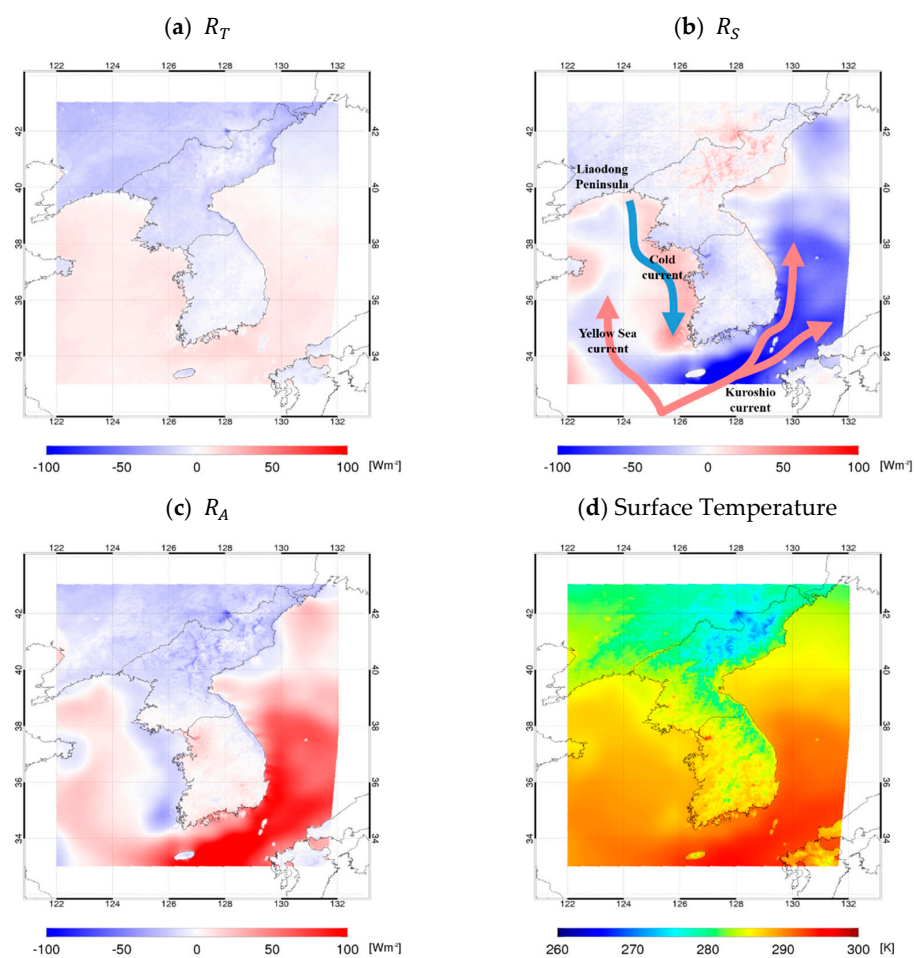


Figure 11. Annual mean energy budget for 2016 at (a) the top of the atmosphere; (b) the surface; and (c) in the atmosphere; and (d) surface temperature distribution.

Monthly changes in R_T , R_S , and R_A are shown in Figures 12–14. R_T showed a positive net radiation in the spring and summer months, which received high ISR, and it showed a negative net radiation in fall and winter (Figure 12). Similar to R_T , R_S showed a positive net radiation in spring and summer and a negative net radiation in fall and winter (Figure 13). In the South Sea and East Sea regions, SHF+LHF increased during spring and winter under the effect of the Kuroshio current, and thus a strong negative net radiation was observed. Figure 14 shows the distribution of R_A through

the difference between R_T and R_S . During spring and summer, land areas, which showed a relatively high temperature distribution, had positive net radiation; and during fall and winter, sea areas had a positive net radiation.

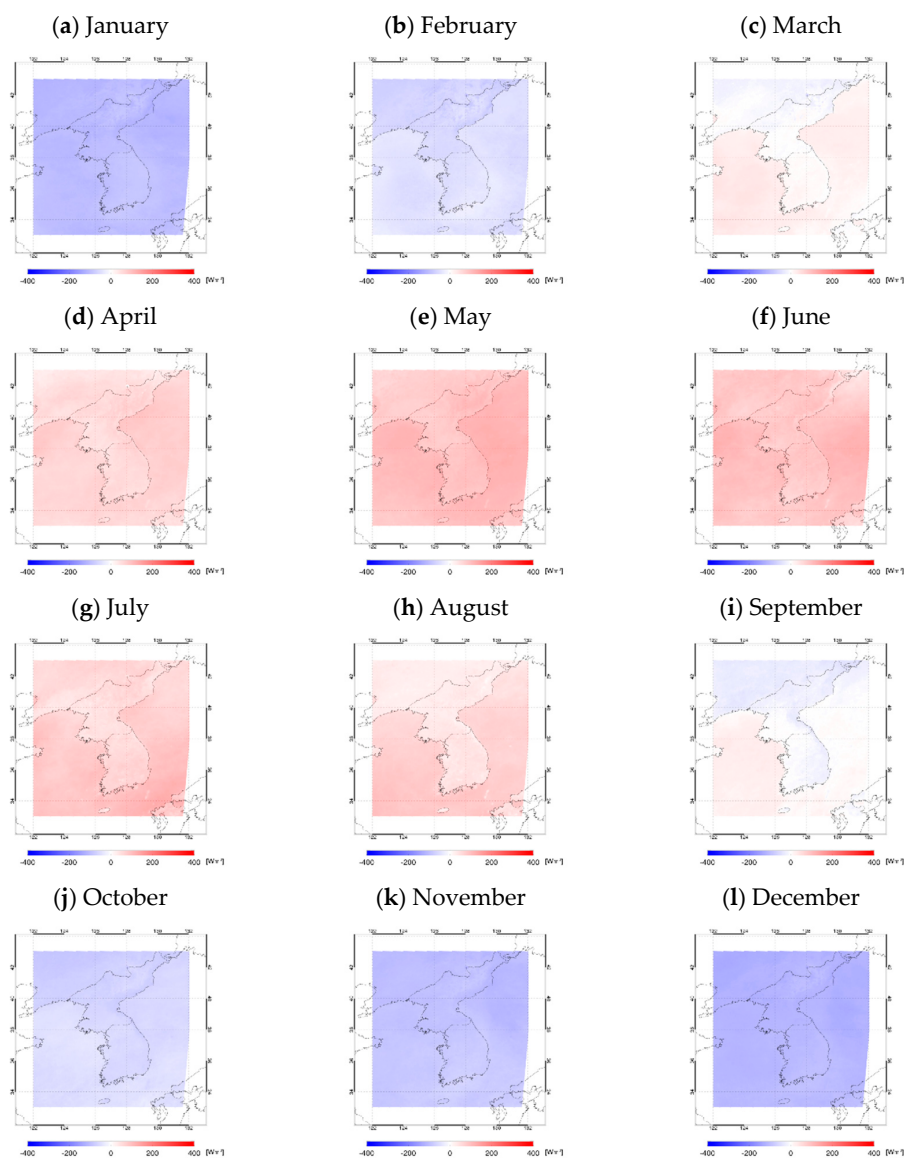


Figure 12. Distribution of monthly mean energy budget (R_T) at the top of the atmosphere for the Korean Peninsula region in 2016.

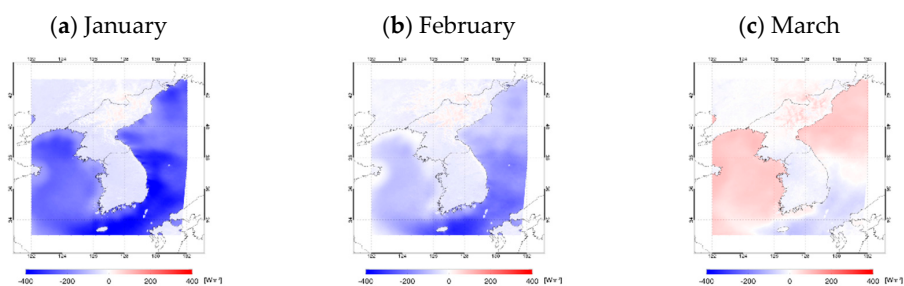


Figure 13. Cont.

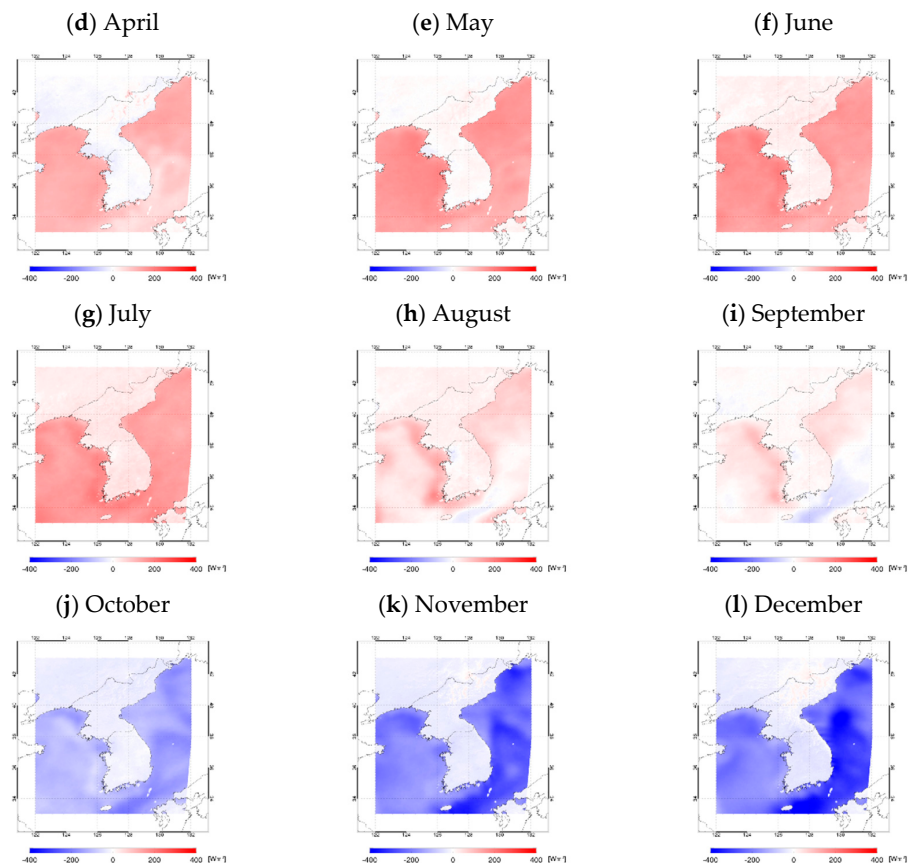


Figure 13. Distribution of monthly mean energy budget (R_s) at the surface for the Korean Peninsula region in 2016.

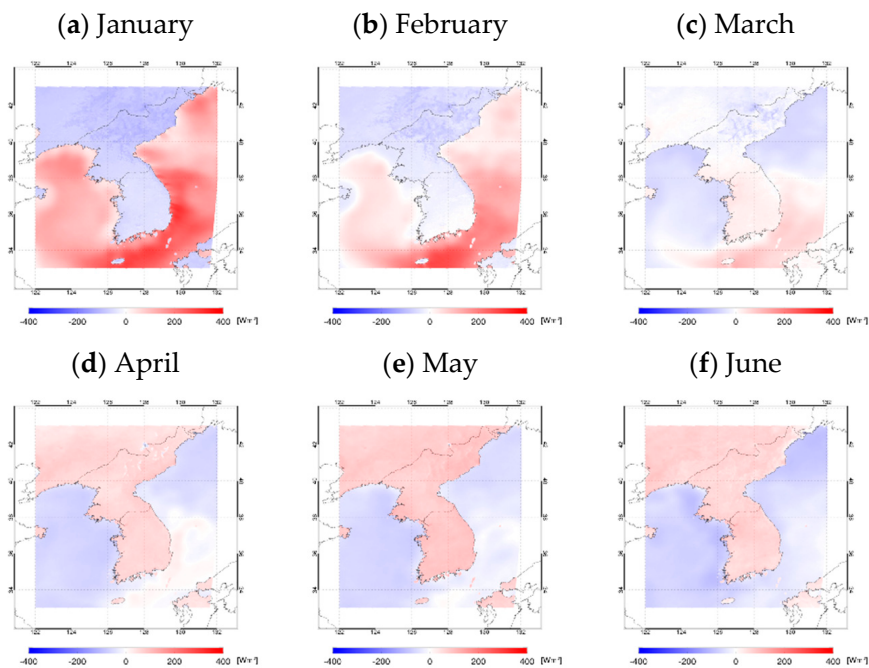


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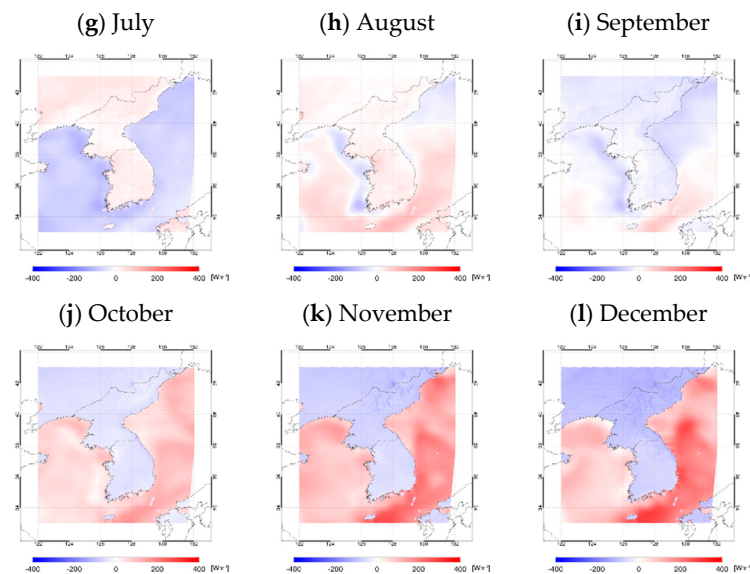


Figure 14. Distribution of monthly mean energy budget (R_A) in the atmosphere for the Korean Peninsula region in 2016.

4.3. Regional Energy Budgets within the Korean Peninsula

At the scale of a country such as South Korea, the energy budget of the Korean Peninsula region showed different regional characteristics. In particular, the emission and absorption of R_S/R_A varied among metropolises and areas outside of them. In metropolitan regions such as Seoul, the surface temperature was high and the temperature changes were large due to the high heat capacity of man-made structures and the heat island effect; thus, R_S/R_A showed a negative/positive net radiation (emission/absorption) [90,91]. Regions such as Gangneung lie adjacent to the Taebaek Mountains (refer to Figure 1) on the west side and the ocean on the east side, which form complex weather conditions, and the cold air of northeasterly winds flows in from the air masses of the Okhotsk Sea, lowering the temperature and reducing temperature changes; thus, R_S/R_A showed a positive/negative net radiation (absorption/emission). In order to analyze such regional differences, the Korean Peninsula was divided into the metropolitan regions (red) of Seoul, Busan, Daegu, Daejeon, and Gwangju; three regions on the left side of the Taebaek Mountains (the Yeongdong region in blue) including Gangneung, the East Sea, and Sokcho; and the island regions (green) of Jeju and Ulleung, as shown in Figure 15. With this classification, the mean energy budget corresponding to the area of each city was found and used in the analysis.

The energy budget for each city is shown as R_T , R_S , and R_A in Figure 16. The distribution of R_T was similar in all cities, but island regions had a relatively large amount of cloud cover compared to other regions (the Korean Peninsula mean is approximately 45.2%, and the island region mean is approximately 52.4%). Consequently, there was a large radiation of RSR (the Korean Peninsula mean was approximately 113.1 Wm^{-2} , and the island region mean was approximately 123.2 Wm^{-2}), and R_T was smaller than that of the Korean Peninsula monthly mean (Table 12). In the Ulleung region, the mean cloud cover was small in May and June at 20.7% and 34.9%, respectively, and thus the R_T value was high. The Yeongdong region is located at a higher latitude than other regions, and thus its ISR was smaller (the Korean Peninsula mean was approximately 341.4 Wm^{-2} , and the Yeongdong region mean was approximately 335.7 Wm^{-2}). Therefore, R_T was smaller than that of the Korean Peninsula mean (Table 12). Figure 16b shows the R_S distribution, which, unlike R_T , showed clear regional characteristics. R_S was smaller than the mean of the Korean Peninsula in metropolitan regions, similar to the mean in island regions and larger than the mean in the Yeongdong region. Metropolitan regions emitted more energy than other regions because the surface temperatures were high (metropolitan region mean: 288.3 K, Yeongdong region mean: 285.5 K, island region mean: 286.9 K). The variations

in surface temperature were small in the island and Yeongdong regions due to the effects of the ocean, and they had a similar or larger mean R_S than the Korean Peninsula. In particular, the Yeongdong region absorbed approximately 21.1 Wm^{-2} more energy than the metropolitan regions. The Seoul region (mean surface temperature: 289.5 K), and the Sokcho region (mean surface temperature: 284.7 K) showed the maximum and minimum net radiation with a difference of 38.5 Wm^{-2} (Table 12). In short, the higher the region's local surface temperature, the more energy tended to be emitted from the surface, and the lower the region's surface temperature, the more energy tended to be absorbed.

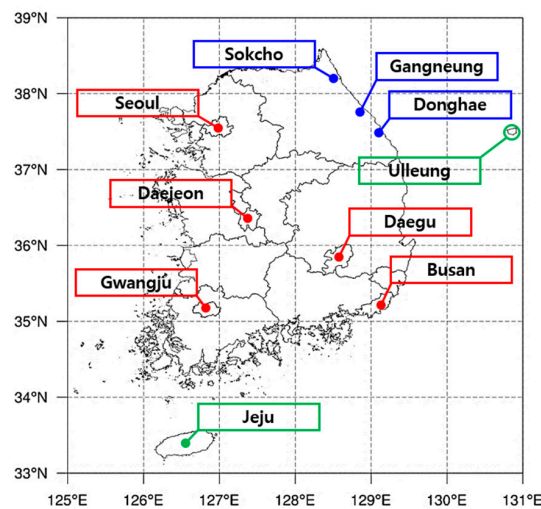


Figure 15. Metropolitan, Yeongdong, and island region cities for studying energy budgets by region in the Korean Peninsula.

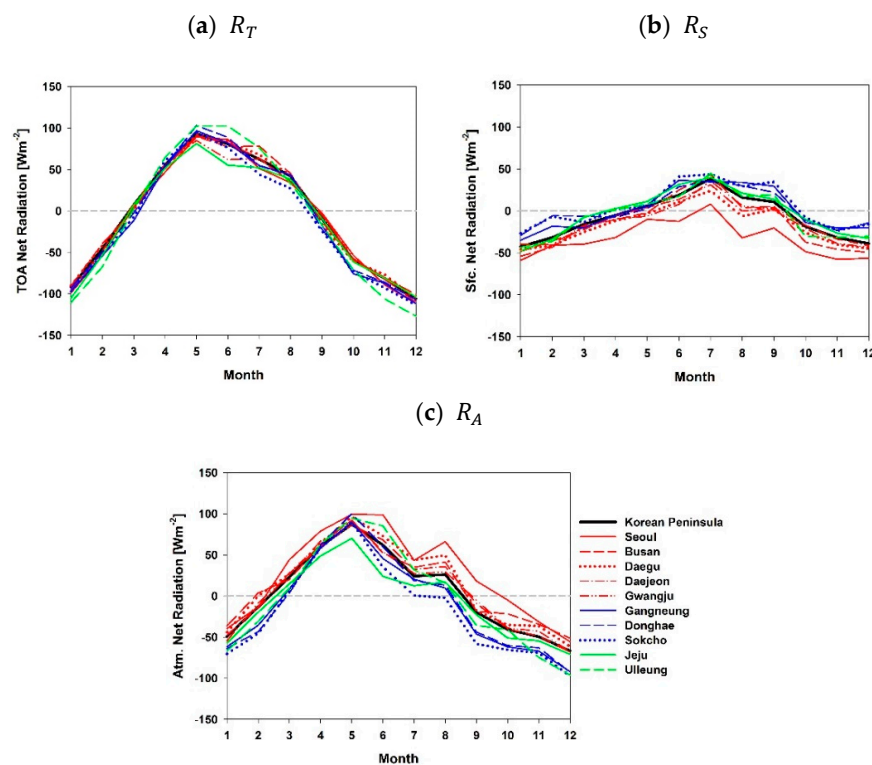


Figure 16. Monthly mean energy budget time series for the Korean Peninsula and cities by region in 2016 at (a) the top of the atmosphere; (b) the surface; and (c) in the atmosphere.

Table 12. Energy budgets for the Korean Peninsula, metropolitan, Yeongdong, and island regions at the top of the atmosphere (R_T), the surface (R_S), and the atmosphere (R_A) (unit: Wm^{-2}).

	Korean Peninsula	Metropolitan	Yeongdong	Island
R_T	−4.30	−5.74	−10.26	−10.32
R_S	−7.94	−18.51 (Seoul: −33.41)	2.62 (Sokcho: 5.04)	−4.51
R_A	3.65	12.77 (Seoul: 25.33)	−12.88 (Sokcho: −13.13)	−5.82

5. Summary and Conclusions

This study calculated the energy budget for the Korean Peninsula in 2016 as part of research dealing with climate change attributable to changes and imbalances in energy budgets, which are related to the issue of global warming. Satellite data (Himawari-8 AHI) and meteorological data (LDAPS) were used to calculate the energy budget. An algorithm was developed for calculating the shortwave radiation components required to calculate the energy budget such as RSR, DSR, and ASR, as well as longwave radiation components, such as OLR, DLR, and ULR. The calculations were made using an empirical method in which the results of a radiative transfer model and the linear relationship between weather variables were applied. The proposed method does not require surface albedo or emissivity data, which involve high uncertainty. The radiation components were calculated using meteorological data, which have low uncertainty and can easily be acquired/collected. To improve the nonlinearity and accuracy of the calculated radiation components, they were calibrated with CERES broadband radiation, and the results showed a percent bias under 3.5% and a high correlation coefficient of over 0.98 between CERES and calibrated radiation components.

Regarding the calculated energy budget, the Korean Peninsula had an R_T of -2.4 Wm^{-2} and is located in a region with a relatively balanced energy budget at the top of the atmosphere. With an R_S of -14.5 Wm^{-2} , this region has a large flux of SHF and LHF due to the Kuroshio current and typhoons in the summer and fall. Therefore, energy was emitted outside and within the atmosphere. The energy absorbed in the atmosphere supplied energy to regions with relatively low surface temperatures or transported excess energy to the polar regions such that a global energy budget balance was achieved [7]. Clear differences in energy budgets caused by regional weather and climate characteristics were observed in metropolises and regions outside them. The Seoul region, which has high surface temperatures, showed the highest energy emission (33.4 Wm^{-2}), and the Sokcho region showed energy absorption (5.0 Wm^{-2}). Such differences in local energy budget cause mesoscale energy circulation and affect weather and climate changes [2,3,89]; therefore, continued observation and analysis must be performed to predict future regional climate changes [6,25].

This study used satellite data and NWP data with a high spatiotemporal resolution to analyze energy budgets. In particular, the obtained data are more detailed than those provided by most climate models because the spatiotemporal resolution of the calculated data was $2 \text{ km} \times 2 \text{ km}$ at 3-h intervals. Climate change models show large differences in their results according to their input meteorological data and model settings [19]. Furthermore, they have limitations in predicting local climate predictions and contain systematic errors because they provide data with large intervals (e.g., $2.5^\circ \times 2.5^\circ$) [34]. Therefore, in order to study the local energy budget of a region such as the Korean Peninsula, it is necessary to perform extensive research on using satellite data and performing highly accurate radiation calculations.

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Abbreviations

AHI	Advanced Himawari Imager
ASR	Absorbed Shortwave Radiation
CERES	Cloud and Earth’s Radiant Energy System
DLR	Downward Longwave Radiation
DSR	Downward Shortwave Radiation
ISR	Incoming Solar Radiation
LDAPS	Local Data Assimilation and Prediction System
LHF	Latent Heat Flux
MERRA-2	Modern Era Retrospective Analysis for Research and Applications Version 2
OLR	Outgoing Longwave Radiation
RSR	Reflected Shortwave Radiation
SHF	Sensible Heat Flux
ULR	Upward Longwave Radiation

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