Research Article

Characteristics of the Reanalysis and Satellite-Based Surface Net Radiation Data in the Arctic

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In this study, we compared four net radiation products: the fifth generation of European Centre for Medium-Range Weather Forecasts atmospheric reanalysis of the global climate (ERA5), National Centers for Environmental Prediction (NCEP), Clouds and the Earth’s Radiant Energy System Energy Balanced and Filled (EBAF), and Global Energy and Water Exchanges (GEWEX), based on ground observation data and intercomparison data. ERA5 showed the highest accuracy, followed by EBAF, GEWEX, and NCEP. When analyzing the validation grid, ERA5 showed the most similar data distribution to ground observation data. Different characteristics were observed between the reanalysis data and satellite data. In the case of satellite-based data, the net radiation value tended to increase at high latitudes. Compared with the reanalysis data, Greenland and the central Arctic appeared to be overestimated. All data were highly correlated, with a difference of 6–21 W/m² among the products examined in this study. Error was attributed mainly to difficulties in predicting long-term climate change and having to combine net radiation data from several sources. This study highlights criteria that may be helpful in selecting data for future climate research models of this region.

1. Introduction

The Arctic has a complex climate system with atmosphere–ocean–land interactions and lower-latitude forcing that occur on various temporal and spatial scales [1]; thus, this area is particularly vulnerable to climate change [2, 3]. The surface radiation budget (SRB) is a key parameter for understanding the polar regions, which have a high albedo and emissivity [4–7]. SRB can be used to characterize the net radiation, specifically, the balance between downward and upward shortwave solar radiation and longwave thermal radiation from the atmosphere and various surfaces [8]. It is effectively the remainder of the radiative energy on the Earth’s surface. Surface net radiation plays an important role in not only the energy cycle but also the ecosystem and carbon and water cycles [8–11]. Remote sensing of the radiative flux can provide information on a global scale, as well as point-scale data from ground observations [12]. Due to these advantages, numerous radiative flux products have been derived from satellite and multispectral sensor data over the last decade [12]. Recently, various radiative flux products of the Arctic from satellite data and reanalysis data have been made available.

Satellite-based data, such as Clouds and the Earth’s Radiant Energy System (CERES) [13], Global Energy and Water Exchanges Project (GEWEX) SRB [14], CloudSat [15], and Climate Monitoring Satellite Application Facility cLoud, Albedo and surface RAdiation dataset from Advanced Very High Resolution Radiometer (AVHRR) data-Edition 2 (CMSAF CLARA-A2) [16], provide information on radiative flux and radiative surface exchange processes. Reanalysis-based data, provided as absorbed solar radiation, includes...
the European Centre for Medium-Range Weather Forecasts (ECMWF) interim reanalysis (ERA-Interim) [17]; fifth-generation ERA (ERA5) [18]; National Centers for Environmental Prediction/National Center for Atmospheric Research (NCEP/NCAR) [19]; Modern-Era Retrospective Analysis for Research and Applications, version 2 (MERRA-2) [20, 21]; and the Japan Meteorological Agency’s Japanese 55-Year Reanalysis (JRA-55) [22].

Numerous studies have attempted to evaluate these radiation products; however, most were conducted at low to mid-latitudes and only a few focused on the Arctic [23–29]. Riihelä et al. performed an intercomparison of radiative components on the Arctic during the spring and summer of 2007 [29]. Seo et al. compared CERES synoptic TOA and surface fluxes and clouds (SYN) and GEWEX SRB Arctic data from March 2003 to December 2007 [30]. Both of these studies analyzed only satellite data. Research has been conducted to develop the Arctic Observation and Reanalysis Integrated System (ArORIS) that combines satellite, reanalysis, and in situ products in the Arctic [28]. Studies evaluating the Arctic’s net radiation have mostly been short term, on the order of 3 years or less, or are based on single-source data (satellite-based, reanalysis, or model data). Despite the many radiative products produced, comparative analysis by data type is lacking or has been carried out on accumulated data from the past. The analysis of recently updated data is still in progress.

In this study, we used data from the Arctic that has frequently been used in past research with recently updated radiation products. Among several reanalyzed data, ERA5 and NCEP data were used in this study. NCEP is a traditional reanalysis data developed in the 1990s, and the rest of the reanalysis data have been developed after the 2000s. ERA5 was selected because it is the most recently produced data and is calculated using various observation-based products as input data compared to past reanalysis data.

The purpose of this study was to assess the net radiation products over the Arctic using validation and intercomparison analyses. This paper is divided into two parts for validation and intercomparison analysis. The intercomparison analysis used four net radiation products: two reanalysis products, ERA5 and NCEP/NCAR, and two satellite-based products, CERES and GEWEX, for diversity. ERA5 and CERES Energy Balanced and Filled (EBAF) Edition 4.0 have been updated recently; NCEP/NCAR and GEWEX have been used for a longer period of time. This research is expected to provide some clarity regarding the accuracy of net radiation products and the relationship between product characteristics and comparative data from the Arctic for future modeling.

2. Materials and Methods

2.1. Data

2.1.1. ERA5. ERA5 is the fifth-generation ECMWF atmospheric reanalysis product of the global climate. ERA5 data are produced by Copernicus Climate Change Service (C3S) and provide the most recent atmospheric reanalysis of ECMWF. ERA5 is based on the new numerical weather product (IFS Cycle 41r2, 2016 version) [18]. Compared to ERA-Interim, ERA5 offers improved spatial resolution, from 79 km to 31 km, and a higher output frequency, from every 6 hours (h) to hourly sampling from 1979 [31]. Therefore, ERA5 is a higher-resolution model, as a replacement for ERA-Interim [32, 33]. Atmospheric data assimilation has been upgraded to a 12-hour 4D-Var ensemble, and advanced climate data are used for model input [33]. Observation data such as satellite data, weather station, buoy, and radiosonde have been used on an average of 0.75 million per day in 1979, when data assimilation begins, and 24 million in recent years.

C3S provides ERA5 products with a spatial resolution of 0.25° × 0.25° in spherical coordinates. We used the monthly averaged EAR5 radiative flux data to match the temporal resolution provided by EBAF. Monthly average data refers to the entire month and is generated from all hourly data for the month. ERA5 is provided in joule per square meter, not in a common unit; it is necessary to convert the unit for comparing with other radiation products. In the case of monthly mean data, it is converted into watt per square meter by dividing 86400 seconds (24 hours).

2.1.2. NCEP. The NCEP/NCAR is a reanalysis product [19]. The spatial resolution of NCEP comprises a T62 Gaussian grid (~1.91°), and temporal resolution is available at several levels, e.g., daily and monthly [34]. We selected monthly mean data to match the temporal resolution of EBAF data for the comparison. Radiative transfer schemes are based on the method of Lacis and Hansen [35] in the shortwave range and that of Schwarzkopf and Fels [36] in the longwave range [37].

2.1.3. EBAF. CERES EBAF is an all-sky surface flux product. Monthly data are presented as the monthly mean, with a spatial resolution of 1° on a uniform grid. The EBAF data used in this study were acquired over the coverage period from March 2000 to March 2018, using the Surface Radiation Edition 4.0 data product (May 2017 release). The EBAF product is calculated using EBAF top-of-atmosphere (TOA) flux, temperature/humidity profiles based on CERES SYN1deg-Month/SYN1deg-3hour, and cloud vertical profiles obtained from CALIPSO/CloudSat [38]. The EBAF product includes monthly surface upward/downward shortwave and longwave irradiances constrained by CERES-derived TOA irradiance [39]. CERES EBAF TOA Edition 4.0 was computed by Loeb et al. [40]. Kato et al. [39] determined the uncertainty in downward and upward shortwave irradiances of the North Pole to be 14 and 16 W/m², respectively, using this product; both downward and upward longwave irradiances were 12 W/m².

2.1.4. GEWEX. GEWEX SRB data provide several temporal resolutions of 1-hour, 3-hour, daily, and the monthly mean. We used monthly mean data to better coincide with the temporal resolution of EBAF for the comparison. The spatial resolution is 1° for uniform grid data. The coverage period of the data was from July 1983 to December 2007. The satellite-based data are calculated from readings obtained from the
AVHRR sensor mounted on the National Oceanic and Atmospheric Administration satellite [41]. Radiative transfer schemes use the Pinker and Laszlo [42] method in the shortwave range and that of Fu et al. [43] in the longwave range. Commonly used cloud information was obtained from the International Satellite Cloud Climatology Project [29]. GEWEX monthly mean data for shortwave irradiance shows a root mean square (RMS) accuracy of 23.34 W/m² and a bias of −5.22 W/m². In the longwave range, the RMS accuracy was 11.1 W/m², with a bias of 0.9 W/m² from 1998 to 2007 [44, 45]. All accuracy values were calculated on a global scale.

Table 1 presents a summary of the spatial radiative flux products used in this study.

Table 1: Summary of radiative flux products used in this study.

| Dataset | Coverage          | Temporal resolution | Spatial resolution | Based on                            | Version           |
|---------|-------------------|---------------------|--------------------|-------------------------------------|-------------------|
| ERA5    | 1979.01-2019.03   | Monthly mean        | 0.25° × 0.25°      | Reanalysis                          | Monthly averaged reanalysis |
| NCEP    | 1948.01-2018.12   | Monthly mean        | T62 Gaussian grid  | Reanalysis                          | Mean monthly NMC reanalysis |
| GEWEX   | 1983.07-2007.12   | Monthly mean        | 1° × 1°            | Satellite                           | LW: rel3.1         |
|         |                   |                     |                    |                                     | SW: rel3.0         |
| EBAF    | 2000.03-2018.03   | Monthly mean        | 1° × 1°            | Satellite                           | Surface Ed4.0      |

Figure 1: Locations of the six ground observation sites: Alert (ALE), Ny Alesund (NYA), Tiksi (TIK), Barrow (BAR), ETH/Greenland Summit (SMT), and North Slope of Alaska (NSA). The red circle indicates the Arctic Circle (poleward 65° N).

2.2. Method. This research is divided into two parts: (1) validation of the net radiation products using ground observation data and (2) an intercomparison among the products in an attempt to better understand the characteristics of the data (Figure 2). In the research flow chart, the yellow square is the validation part and the green box is the intercomparison part. Validation is performed through three analyses after collocation with measured data using spatial data. And it analyzes the characteristics of the differences of each data through five intercomparison analyses.

In the validation part of the study, the four net radiation products were compared using ground observation data. Collocation of the spatial data was carried out using a minimum distance method. The method of the minimum distance found the nearest location using the great-circle distance. The distance between each point is obtained through

\[ d = r \Delta \sigma \]

\[ \Delta \sigma = 2 \arcsin \left( \sqrt{\frac{1}{2} \sin \left( \frac{\Delta \text{lat}}{2} \right) \cos \text{lat}_1 \cos \text{lat}_2 \left( \frac{\Delta \text{lon}}{2} \right) } \right). \]

\[ (1) \]
d refers the distance between ground observation data and spatial product. Find and compare the point where this distance is the minimum. r refers to the radius of sphere, Δ refers to the central angle, and lat_1 and lat_2 refer to the latitude at ground observation and net radiation products to be compared. The absolute difference between longitude and latitude between two points is denoted by Δlon.

Three common statistical methods were used for the validation [11, 52]: the correlation coefficient (R), root mean square error/difference (RMSE/RMSD), and mean bias error (MBE). The R-value here indicates the correlation between observation data and spatial data of the net radiation products; closer to 1 (−1) indicates a strong positive (negative) relationship with ground observation or comparison data. R > 0 indicates that there is a positive correlation, R < 0 indicates a negative correlation, and R = 0 indicates that the data being compared are uncorrelated. RMSE is associated with the weight of the maximum error [53]; this is sometimes referred to as RMSD between values being modeled. MBE indicates systematic differences [52]. Positive (negative) MBE represents an overestimation (underestimation). The statistical characteristics of ground observation data were analyzed. In addition, the validation grid of net radiation data was analyzed using a box plot owing to the different spatial resolutions. The box plot represents the information as a univariate distribution with the advantages of an approximate range and the ability to accommodate greater variability.

The second part of the study involved an intercomparison of the spatial net radiation products over the Arctic Circle, to analyze the differences among products used in actual research in the area. The study period was designated as March 2000 to December 2007, where all four datasets existed, thus over an approximately 8-year period. The intercomparison part of the study was further divided into four parts. First, the entire Arctic Circle was compared and analyzed as a large region from 65°N poleward. Second, a correlation analysis was conducted among the products. Third, the R-value, MBE, and RMSD were calculated. Finally, the results were configured in matrix form to intuitively analyze the correlation relationships among the radiation products for the Arctic Circle region. Individual data points were compared using the average net radiation in the Arctic Circle.

In the second part, the meridional distribution and seasonal cycle parts were analyzed separately. For analysis of the meridional distribution, we calculated the zonal mean at 2° intervals, as this corresponds to the spatial resolution of NCEP (with a resolution greater than 1°). The distribution of net radiation for each datum was analyzed by averaging over an approximately 8-year period. The intercomparison part of the study was further divided into four parts. First, the entire Arctic Circle was compared and analyzed as a large region from 65°N poleward. Second, a correlation analysis was conducted among the products. Third, the R-value, MBE, and RMSD were calculated. Finally, the results were configured in matrix form to intuitively analyze the correlation relationships among the radiation products for the Arctic Circle region. Individual data points were compared using the average net radiation in the Arctic Circle.

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| Site   | Location (latitude, longitude) | Observation site data source                                                                 | Data periods    |
|--------|-------------------------------|-----------------------------------------------------------------------------------------------|-----------------|
| TIK    | 71.59, 128.92, Elev: 48 m     | WCRP/GEWEX Baseline Surface Radiation Network (BSRN)                                            | 2011.04–2016.12 |
| SMT    | 72.60, 321.58, Elev: 3216 m  | Swiss Federal Institute, ETH broadband radiometer measurements                                  | 2011.01–2014.06 |
| NYA    | 78.93, 11.93, Elev: 11 m     | WCRP/GEWEX Baseline Surface Radiation Network (BSRN)                                            | 2000.03–2018.03 |
| NSA    | 71.32, 203.39, Elev: 8 m     | DOE Atmospheric Research Measurement (ARM) Program                                                | 2000.03–2018.03 |
| ALE    | 82.45, 297.49, Elev: 200 m  | WCRP/GEWEX Baseline Surface Radiation Network (BSRN)                                            | 2004.07–2014.03 |
| BAR    | 71.32, 203.39, Elev: 8 m     | WCRP/GEWEX Baseline Surface Radiation Network (BSRN)                                            | 2000.03–2017.08 |

Table 2: Overview of the ground observation data.
distributions and the range of values. Scatter density plot analysis was performed with high accuracy, in connection with the validation part. Finally, we analyzed the differences in spatial distributions in matrix form to identify the characteristics of the products with respect to the Arctic region.

3. Results and Discussion

3.1. Validation with Ground Observation Data

3.1.1. Overview of the Validation. Figure 3 shows histogram plots of the differences between ground observation data and individual net radiation products. ERA5 data were more accurate than the other products, with a coefficient of determination ($R^2$) of 0.88, an RMSE of 19.02 W/m$^2$, and a low MBE of −0.26 W/m$^2$. Compared to the other products, the difference was concentrated at 0, indicating a low variance. In contrast, NCEP was observed to have the lowest accuracy among the products, with an $R^2$ value of 0.64, an RMSE of 45.49 W/m$^2$, and an MBE of $-8.14$ W/m$^2$; in this case, the data appeared to be widely scattered, as evidenced by the high MBE and variance (standard deviation: 44.76 W/m$^2$). EBAF showed an $R^2$ of 0.86, an RMSE of 23.96 W/m$^2$, and an MBE of $-2.67$ W/m$^2$, with a standard deviation of 23.81 W/m$^2$ as the second most accurate product. GEWEX was only available until 2007. Thus, the number of samples was small compared to the other data. Only the mean of the difference was observed in GEWEX, with an $R^2$ value of 0.78, an RMSE of 32.01 W/m$^2$, an MBE of 10.92 W/m$^2$, and a standard deviation of 30.09 W/m$^2$. Overall, the most accurate model was ERA5, followed by EBAF, GEWEX, and NCEP.

The validation results for ground observation data are shown as Table 3. TIK and SMT locations have data going
back to 2007, preventing a comparison with GEWEX. EBAF showed a significantly lower correlation at SMT than other sites. ERA5 indicated high accuracy with a strong correlation, low RMSE, and an MBE approaching 0.

3.1.2. Analysis of the Net Radiation Distribution. The validation pixel of each data point was analyzed due to the different grid sizes of each product. The statistical distribution of the grid used for validation was confirmed, as shown in Figure 4; observation data are represented by a black box in the figure. Overall, ERA5 showed a denser distribution than NCEP, for which the distribution was scattered considerably. However, the interquartile range (IQR) for the two products was similar. Thus, the values of the net radiation were similarly distributed. In SMT, which is a homogeneous surface, the IQR of most data was observed to be small. However, in the case of heterogeneous surface condition as NYA, the IQR was determined according to the grid to validation. The NCEP with the lowest spatial resolution was observed to have the largest IQR, and the IQR of ERA5 with the high spatial resolution was observed to be small.

The statistical characteristics of each observation site are as follows. SMT showed an average net radiation of $-12.45 \text{ W/m}^2$ (standard deviation: $11.73 \text{ (25.72\%)} \text{ W/m}^2$). This site had a short-duration coverage of just 3 years. SMT is located in the center of Greenland, and it has a homogeneous surface compared with the other sites. The data variation was also smaller at SMT compared with the other sites. BAR showed an average net radiation of $21.58 \text{ W/m}^2$ (standard deviation: $60.72 \text{ (25.57\%)} \text{ W/m}^2$). Therefore, the data variation was larger than in other regions. The other ground observation sites, TIK, NYA, NSA, and ALE, showed average net radiation values of $19.34$, $5.41$, $16.16$, and $-1.99$, W/m$^2$, respectively, with standard deviations of $51.58 \text{ (30.09\%)}$, $48.28 \text{ (26.38\%)}$, $56.87 \text{ (24.29\%)}$, and $46.53 \text{ (25.43\%)} \text{ W/m}^2$.

Table 3: Validation of the four net radiation products by ground observation site (six sites total).

| Site | ERA5 | NCEP | EBAF | GEWEX |
|------|------|------|------|-------|
| TIK  | $R$  | 0.96 | 0.86 | 0.97 |
|      | RMSE (W/m$^2$) | 14.92 | 36.77 | 22.22 |
|      | MBE (W/m$^2$) | 5.22 | 5.05 | -4.41 |
| SMT  | $R$  | 0.94 | 0.88 | 0.63 |
|      | RMSE (W/m$^2$) | 8.45 | 23.25 | 20.78 |
|      | MBE (W/m$^2$) | -5.76 | -10.47 | 18.61 |
| NYA  | $R$  | 0.94 | 0.91 | 0.93 |
|      | RMSE (W/m$^2$) | 26.95 | 65.62 | 35.99 |
|      | MBE (W/m$^2$) | -18.52 | -4.03 | -7.11 |
| NSA  | $R$  | 0.97 | 0.83 | 0.96 |
|      | RMSE (W/m$^2$) | 14.39 | 37.59 | 16.30 |
|      | MBE (W/m$^2$) | 6.61 | -10.15 | -2.51 |
| ALE  | $R$  | 0.94 | 0.84 | 0.96 |
|      | RMSE (W/m$^2$) | 21.41 | 28.94 | 17.64 |
|      | MBE (W/m$^2$) | 13.79 | -9.34 | 4.79 |
| BAR  | $R$  | 0.97 | 0.81 | 0.96 |
|      | RMSE (W/m$^2$) | 14.52 | 41.29 | 19.1 |
|      | MBE (W/m$^2$) | 2.83 | -13.59 | -6.13 |

3.2. Intercomparison of Net Radiation Products

3.2.1. Comparison Matrix of the Arctic Circle. In the intercomparison of four net radiation products, a confusion matrix was used to highlight the differences among data expressed in terms of the $R$-value, RMSD, and MBE (Figure 5). All products showed strong relationships with the other three products, with correlation coefficients exceeding 0.9. The MBE was positive, based on x-axis data. GEWEX appeared to overestimate the net radiation compared with the other products. The RMSD varied over a wide range, from 6.16 to 21.19 W/m$^2$. The lowest RMSD corresponded to a comparison between EBAF and ERA5 but was highest between GEWEX and NCEP. In the reanalysis data, the secondary low in the RMSD was 9.60 W/m$^2$; the difference from the largest to the smallest was in the following order: EBAF/NCEP, GEWEX/EBAF, and GEWEX/ERA5. GEWEX showed the largest difference in comparison with satellite-derived data, such as EBAF. Thus, all data showed a high correlation; however, the MBE and RMSD indicated high variation, despite similar patterns in the data. Thus, the differences associated with the variation in data and the range of

Journal of Sensors
observing values appeared to strongly affect the agreement among radiation products.

3.2.2. Comparisons of Meridional Distribution and Seasonal Cycle. To understand the characteristics of net radiation data for latitude, we analyzed the 2° zonal mean for each of the four products; this was chosen based on the spatial resolution of NCEP data. In a previous study, the same method was used to analyze meridional characteristics [54]. Figure 6(a) shows the net radiation distribution according to latitude. GEWEX data were 5~10 W/m² higher than other data. Net radiation was highest in the order of GEWEX, EBAF, ERA5, and NCEP. All data showed a decreasing pattern between latitudes 60°N and 78°N. However, above 78°N,
differences among the distributions emerged. ERA5 showed a poleward decrease in net radiation. In NCEP net radiation data, a decreasing pattern was observed initially closer to 60°; however, an increasing pattern developed at latitude 80°N. The net radiation distribution of EBAF data by latitude was divided into three ranges. In first range from 60°N to 78°N, the net radiation of EBAF showed a decreasing trend, similar to ERA5 and NCEP. However, the EBAF distribution rose from latitude 78°N to 86°N. Above 86°N, the EBAF net radiation pattern again showed a reduction. Similarly, the GEWEX net radiation exhibited a decreasing pattern up to latitude 78°N. From 78°N to 84°N, the net radiation for GEWEX remained relatively the same, regardless of latitude.

Figure 6(b) shows the seasonal cycle distribution. Except for summer, GEWEX indicate overestimation compared to for net radiation products, which was consistent with MBE in the confusion matrix and the 2° zonal mean by latitude. In all data, the maximum value of net radiation corresponded to summer in July, and the minimum value coincided with winter in December. This pattern is similar to the variation in temperature and the number of hours of daylight [55]; thus, this characteristic is caused by the polar night, because incoming solar radiation is weak or nonexistent in the autumn and winter seasons. Thus, the seasonal variation of the net radiation can be divided into two periods: from April to September and from October to March. The period from April to September includes the melting season and summer. During this time, differences in the data were reduced. Notably, EBAF and ERA5 distributions were similar. The second period from October to March included the freezing season and winter. During this period, the differences among the radiation products increased and the net radiation recorded increased in the order from the largest of GEWEX, EBAF, ERA5, and NCEP.

3.2.3. Comparison with Distribution of Data per Grid. The confusion matrix analysis was performed by averaging the Arctic Circle region as one region. The Arctic has a highly seasonal cycle, as indicated in all four radiation products. However, using the 2° zonal mean, we confirmed differences in latitude patterns within the Arctic Circle. Figure 7 shows a histogram of the four net radiation products. ERA5 showed an average of 4.74 W/m² and a standard deviation of 51.39 W/m² for a total of 13,392,000 samples. In this case, the standard deviation accounted for 14.68% of the total value. NCEP had an average of 1.77 W/m² and a standard deviation of 58.58 W/m² for a total of 214,272 samples; here, the standard deviation accounted for 14.44% of the total value. EBAF produced an average of 10.07 W/m² and a standard deviation of 50.51 W/m² for a total of 837,000 samples; in this case, the standard deviation accounted for 15.04% of the total value. ERA5 produced an average of 20.01 W/m² and a standard deviation of 51.08 W/m² for a total of 837,000 samples; in this case, the standard deviation accounted for 14.68% of the total value. The NCEP product had the widest data distribution, whereas GEWEX was more concentrated. Satellite data also indicated a higher mean compared to that of reanalysis data.

Figure 8 shows the monthly density scatter of net radiation for NCEP, EBAF, and GEWEX based on ERA5; here, the x-axis represents ERA5 and the y-axis represents comparative data in the figure. Most of the net radiation values were distributed over a low range (<30 W/m²). All data had a $R^2$ above 0.9, indicating a strong relationship with ERA5.
With respect to ERA5, the differences among NCEP, EBAF, and GEWEX data were $-1.40 \pm 20.43$, $2.98 \pm 17.21$, and $13.80 \pm 22.10$ W/m$^2$, respectively. NCEP data were lower than the 1 : 1 line in range, where values below zero were distributed. In addition, the data indicated a more widespread distribution of net radiation. EBAF data were similar to ERA5 data, as indicated by the regression line fit. The regression line of GEWEX was generally higher than the 1 : 1 line; thus, GEWEX data tended to show higher values than ERA5. The density scatter distribution of GEWEX-ERA5 was more similar to EBAF data than NCEP.

3.2.4. Differences in Spatial Distribution. We analyzed the spatial differences in net radiation between NCEP, EBAF,
and GEWEX in the Arctic Circle, based on ERA5 data, as shown in Figure 9; the red shading indicates an overestimation of the corresponding data, and the blue shading indicates underestimation. In the case of NCEP, a mixture of overestimation and underestimation is shown in the figure. The net radiation differences were smaller over oceans than continents, with overestimation in the east and underestimation in the west. Regarding land areas, Canada and North Greenland showed a strong underestimation in net radiation, whereas South Greenland was overestimated. In the case of EBAF, less of a difference was observed compared with the other two datasets. Thus, substantial discrepancies were evident among the radiation products in Greenland. Also, the central Arctic area exhibited an overestimation of the net radiation, which was consistent with the 2° zonal mean by latitude. EBAF data indicated an increase in net radiation from 78°N to 86°N, whereas in the ocean adjoining the Central Arctic, a small underestimation was evident. In the case of GEWEX, the entire area was overestimated, with the exception of a few areas. The north coastline of Greenland and East Greenland Sea showed an underestimation in net radiation. These areas were underestimated in all comparisons; thus, ERA5 always estimated a high net radiation value in the area. These regional characteristics should be noted when using ERA5 data. Different regional characteristics were observed in GEWEX and EBAF. The two products are the same satellite-based data, but there is a difference in the radiative transfer scheme between the two products. The difference in the net radiated energy of the two data may occur primarily due to the difference in the method of calculating the radiation data. There are also differences in the cloud data used. CERES EBAF uses MODIS data and SAR-based CALIPSO/CloudSat data, and GEWEX uses ISCCP DX data. In Figures 9(b) and 9(c), both EBAF and GEWEX data are overestimated compared to ERA5 data in the area above about 80°N, which is the area where CALIPSO/CloudSat observations are missing. The two satellite-based data show the same characteristics in the region where CALIPSO/CloudSat data is excluded. Therefore, differences in satellite-based net radiation may occur due to differences in input cloud data.

4. Summary and Conclusions

In this study, we evaluated the surface net radiation from reanalysis (ERA5 and NCEP) and satellite-based (EBAF and GEWEX) products in the Arctic region. The maximum validation period was from March 2000 to March 2018, and the product comparison period was from March 2000 to December 2007, during which all data overlapped. The overall results for the validation process indicated ERA5 as having the highest accuracy of the four products. The quantitative accuracies of the four products were as follows: ERA5, $-0.26 \pm 19.02$; EBAF, $-2.37 \pm 23.96$; GEWEX, $10.92 \pm 32.01$; and NCEP, $-8.14 \pm 45.49$ W/m². ERA5 and EBAF showed similar accuracies with respect to CERES EBAF (RMSE: 20.93, MBE: 2.40), with higher uncertainty at higher latitudes.
[11]. The difference in accuracy may be due to spatial resolution. ERA5, the smallest valid grid, showed a distribution of values similar to the observation data. In the case of NYA, this particular observation site is located on an island (Svalbard, a Norwegian island); here, the grid size was larger, resulting in a larger IQR. The reason for the lower accuracy in ERA5 than ERA5 can be attributed to the low accuracy of the CERES inversion for the snow surface and the uncertainty in the matchup near the coastal line due to the low spatial resolution [56].

The reason for focusing on these differences is that the Arctic is covered in ice; thus, the ice albedo effect is important. In particular, net radiation is connected to the ice-albedo feedback mechanism, making this feedback a key parameter of climate change in the Arctic [6, 57]. When analyzing climate change, differences in model accuracy and patterns of change are due to differences in the net radiation. Thus, it is essential to monitor and improve the assessments of surface energy budgets to better understand and predict climate change [28]. Continuous research in this area is essential for a process-level connection between the model and observations [58].

The purpose of this study was to attempt to better understand the accuracy of net radiation products in high-latitude areas and the differences among products. This helps when selecting a product for climate change modeling. Based on the results from this study, it is possible to analyze long-term energy changes in the Arctic by selecting the optimized net radiation energy.

The validation results were limited to ground observation sites covered with ice. Since the Arctic Circle is mainly composed of oceans, validation in sea ice is an essential part. Therefore, it is necessary to add the net radiation validation of sea ice using drift data in the future. In this study, differences occurred between net radiated products and the characteristics of the differences were analyzed. However, it was not possible to find out what factors caused such differences. Therefore, future research will need to consider the factors responsible for creating differences in the net radiation products. There are many different things that cause differences in net radiation. It is necessary to study what factors cause the difference in net radiation from the results of this study.

The radiative energy component is expected to play a key role in projecting accurate representations and climate change trends for the Arctic region. Changes in net radiation can help predict variables that cause physical processes in the Arctic such as Arctic sea ice. Based on the results of this study, it can be extended and utilized in long-term energy change observation, sea ice prediction model, and climate change modeling.

It is necessary to analyze which data is most appropriate for analyzing climate change and how the difference between the radiative products has an impact on climate change analysis. A radiative flux such as net radiation can be used as a key parameter to determine the energy balance of the Arctic. These results will enable accurate observation and prediction of Arctic climate and help to identify the causes of recent abnormal climates in connection with the Arctic midlatitude region.

**Data Availability**

The data used to support the findings of this study are included within the article.

**Conflicts of Interest**

The authors declare that there is no conflict of interest regarding the publication of this paper.

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