

Response to Reviewer #1's comments:

General:

This is well-written and a good presentation of a new model for downwelling long wave. As a reader I kept looking for but did not find the motivation for developing the model. I suggest it would improve the paper to explain the motivation.

Response:

Thank you for your valuable feedback. We agree that clarifying the motivation for developing this model is essential. In short, the main motivation is to address the limitations of existing satellite-based and reanalysis products, which often struggle with accuracy under cloudy-sky conditions and have lower spatial resolution. Our new model is designed to overcome these challenges, providing improved accuracy across both clear- and cloudy-sky conditions, and at finer spatial and temporal resolutions. This makes the model more suitable for localized and high-resolution studies.

Here is the revised version of the first two paragraphs:

“The downward longwave radiation (Rl) at the ocean surface is the thermal infrared (4–100 μm) radiative flux emitted by the entire atmospheric column over the ocean surface (Yu et al., 2018). The ocean-surface Rl is a critical component of the heat flux across the ocean–atmosphere interface, [shaping the climate state of both the atmosphere and the ocean \(Caniaux, 2005; Fasullo et al., 2009; Fung et al., 1984\)](#). Accurate estimates of Rl are essential for studying air–sea interactions and for improving our understanding of climate and oceanic systems.

Although the ocean-surface Rl is measured at most buoy sites, the available ocean-surface Rl measurements cannot meet the needs of various applications because of the small number of buoys currently employed (especially moored buoys) and their sparse distribution across global oceans. Another way to get the Rl at the ocean surface is by using satellite-based or model reanalysis products. The ocean-surface Rl from satellite-derived products, such as the International Satellite Cloud Climatology Project (ISCCP) (Rossow & Zhang, 1995; Young et al., 2018) and Clouds and the Earth's Radiant Energy System Synoptic Radiative Fluxes and Clouds (CERES/SYN1deg) (Doelling et al., 2013; Rutan et al., 2015), is usually generated using satellite data and a radiative transfer model, which simulates the radiative transfer interactions of light absorption, scattering, and emission through the atmosphere with the input of given atmospheric parameters. However, radiative transfer models are not widely used in practice due to their complexity and the difficulties associated with collecting all essential inputs. The ocean-surface Rl provided in model reanalysis products, such as the fifth generation of the European Centre for Medium-Range Weather Forecasts atmospheric reanalysis of the global climate (ERA5) (Hersbach et al., 2020) and the Modern-Era Retrospective analysis for Research and Applications, Version 2 (MERRA2) (Gelaro et al., 2017), is produced by assimilating various observations into an atmospheric model to get the optimal estimates of the state of the atmosphere and the surface (Gelaro et al., 2017). Previous studies indicated that Rl estimates from satellite-based products are generally in better agreement with buoy measurements than those obtained from reanalysis products (Pinker et al., 2014; Pinker et al., 2018; Thandlam & Rahaman, 2019). However, applications of the ocean-surface Rl from these two kinds of products are limited due to their coarse spatial resolutions (most of them are coarser than 1°), limited periods (especially satellite-based products) ([Xu et al., 2023; Zeng et al., 2020](#)), and discrepancies in accuracy and consistency (Cronin et al., 2019).

To overcome these limitations, many parameterization and empirical models for estimating ocean-surface Rl that can easily be implemented in practical use have been established during the past few decades

(Bignami et al., 1995; Josey, 2003; Zapadka et al., 2001; Zapadka et al., 2020). Most of the commonly used R_l estimation models were established using the relationship between R_l and the relevant meteorological variables (i.e., air temperature, humidity, column integrated water vapor (IWV), and cloud parameters) or oceanic parameters (i.e., bulk sea surface temperature), which are usually obtained from in situ measurements or model simulations (Li & Coimbra, 2019; Li et al., 2017; Paul, 2021). However, a significant limitation is that many of these models were originally developed for land surfaces and applied directly to the ocean, assuming similar atmospheric conditions. (Bignami et al., 1995; Clark et al., 1974; Frouin et al., 1988; Josey, 2003). This assumption introduces the uncertainty in R_l estimates, as water vapor profiles differ significantly between land and ocean surfaces (Bignami et al., 1995). Even those models specifically designed for ocean surfaces are often based on limited regional data, raising concerns about their robustness when applied globally. (Bignami et al., 1995; Josey, 2003; Zapadka et al., 2001). For example, Josey (2003) proposed a model for R_l estimation at mid-to-high latitude seas with a satisfactory validation accuracy, but this new model performed worse over tropical seas with a tendency to underestimate R_l by up to 10–15 W/m². Furthermore, most existing R_l estimation models only work under clear-sky conditions, which are especially rare over ocean surfaces. Additionally, most of these models only derive R_l at instantaneous scales, yet the R_l at the daily scale is more preferred across a range of applications.

Given these challenges, there is a clear need for a new, easily implemented model capable of providing accurate R_l estimates at the global ocean surface. Such a model should function effectively under all-sky conditions, offer flexibility in temporal scales (e.g., instantaneous and daily), and be robust enough for global application. Addressing these gaps would provide a valuable tool for improving our understanding of air–sea interactions and contribute to more accurate climate and oceanic models. More details about the existing R_l estimation models are given in Section 2.”

Comment #1:

What is the need for a new model? Why not use satellite based long wave radiation at the surface (As Pinker et al describe - this seems to be more accurate than the model you developed.)? Why not use surface radiation from ERA5 reanalysis or from ECMWF model - you import cloud information from ERA5, so why not just take surface radiation from that model?

Response:

Thank you for this insightful question. The need for a new model arises from several limitations inherent in the existing satellite-based and reanalysis products. While satellite-derived longwave radiation products, such as those described by Pinker et al., generally perform well, they tend to have good accuracy under clear-sky conditions. However, their performance degrades under cloudy-sky conditions, where accurate cloud base temperature and height estimation remains challenging (Zeng et al., 2024). To address this, our model incorporates two cloud parameters—total column cloud liquid water and total column cloud ice water—which help improve the estimation of longwave radiation under cloudy conditions.

Similarly, while products like ERA5 provide surface radiation, they are often generated at coarse spatial resolutions, making them less suitable for local-scale studies or applications requiring finer granularity. Our new model addresses these challenges by providing high accuracy under both clear- and cloudy-sky conditions and across different temporal scales (hourly and daily). The model offers a consistent and computationally efficient method to estimate downward longwave radiation at finer resolutions.

Moreover, we conducted comparisons with ERA5 and CERES longwave radiation products, and our model consistently outperformed these products, including in land regions, where it also surpassed GLASS-AVHRR, ERA5, and CERES_SYN1deg_Ed4A products (Chen et al., 2024).

Comment #2:

Years ago, absent satellite or model-based surface radiation, folks needed a model such as your to estimate surface long wave to force, for example, an ocean model. But now people use model or satellite fields. So will anybody utilize your new model? What was the purpose in developing it?

Response:

Thank you for this comment. While satellite and model-based surface radiation products are widely used today, our model addresses a gap in applications that require higher spatial resolution and greater accuracy for estimating longwave radiation. The proposed model is designed to meet the needs of the atmospheric science and remote sensing communities (Yang et al., 2023; Jiao & Mu, 2022; Zeng et al., 2020; Chen et al., 2024; Zapadka et al., 2020), where parameterization techniques are commonly employed for generating longwave radiation products.

The computational efficiency and high accuracy of our model make it especially well-suited for operational ocean-atmosphere heat budget studies, regional climate modeling, and real-time applications, where existing longwave radiation products are subject to delays. By addressing the limitations of satellite and reanalysis products—such as coarse spatial resolution or reduced accuracy under cloudy conditions—our model provides a valuable and practical tool for researchers and practitioners who need improved, more timely estimates of downwelling longwave radiation.

Reference

- Chen, Y., Jiang, B., Peng, J., Yin, X., & Zhao, Y. (2024). Evaluation of the Surface Downward Longwave Radiation Estimation Models over Land Surface. *Remote Sensing*, 16(18), Article 18. <https://doi.org/10.3390/rs16183422>
- Yang, J., Hu, J., Chen, Q., & Quan, W. (2023). Parameterization of downward long-wave radiation based on long-term baseline surface radiation measurements in China. *Atmospheric Chemistry and Physics*, 23(7), 4419–4430. <https://doi.org/10.5194/acp-23-4419-2023>
- Jiao, Z.-H., & Mu, X. (2022). Global validation of clear-sky models for retrieving land-surface downward longwave radiation from MODIS data. *Remote Sensing of Environment*, 271, 112903. <https://doi.org/10.1016/j.rse.2022.112903>
- Zeng, Q., Cheng, J., & Dong, L. (2020). Assessment of the Long-Term High-Spatial-Resolution Global LAnd Surface Satellite (GLASS) Surface Longwave Radiation Product Using Ground Measurements. *IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing*, 13, 2032–2055. <https://doi.org/10.1109/JSTARS.2020.2992472>
- Zeng, Q., Cheng, J., Sun, H., & Dong, S. (2024). An integrated framework for estimating the hourly all-time cloudy-sky surface long-wave downward radiation for Fengyun-4A/AGRI. *Remote Sensing of Environment*, 312, 114319. <https://doi.org/10.1016/j.rse.2024.114319>

Zapadka, T., Ostrowska, M., Stoltmann, D., & Krężel, A. (2020). A satellite system for monitoring the radiation budget at the Baltic Sea surface. *Remote Sensing of Environment*, 240, 111683.
<https://doi.org/10.1016/j.rse.2020.111683>

Response to Reviewer #2's comments:

General: *This study uses data from 65 moored buoys, supplemented with satellite and reanalysis data to evaluate eight existing models for downward longwave radiation R_l , as well as a new model developed here called “modnew”. The hourly and daily-averaged R_l estimated from modnew model have overall relatively low errors – very good news as these observations are not widely available. I applaud the authors for gathering so many historic LWR buoy measurements! What a job and what a resource.*

Response:

Thank you for your positive feedback and encouraging remarks regarding our study. We are delighted that you recognize the value of our effort in compiling and analyzing this extensive dataset of historic longwave radiation (LWR) buoy measurements. Your thoughtful and detailed comments have significantly enhanced the quality of our manuscript. Below, we provide a detailed response to each of your comments.

Comment #1:

My major concern though is with the universal application of the Pascal & Josey 2000 (PJ2000) longwave radiation (LWR) correction applied to all 65 buoy timeseries.

The PJ2000 LWR Correction = $(a + \lambda)R_{\text{solar}} + b(R_{\text{solar}})^2$,

with $a=0.00434$, $\lambda = 0.011$, $b= 1.72 \times 10^{-6}$. This is a large correction if $R_{\text{solar}} \sim 1000 \text{ W/m}^2$!

The polynomial terms involving a & b ($a \cdot R_{\text{solar}} + b \cdot R_{\text{solar}}^2$) are a correction for the differential heating of the dome and casing. The last sentence in PJ2000 is “We suggest that such a correction should be made in future analyses if the component temperatures are not logged in order to improve the accuracy of the measured longwave flux”. I have emphasized the “if” part here because most if not all of the OceanSITES (including all from NOAA/PMEL, e.g., GTMBA, KEO, Papa, ARC) LWR sensors are the 3 output Eppley sensors that measure case and dome temperature and correct for this effect. This correction has been done by the data provider and should not be done by the user. Essentially the authors here have double corrected these data.

Response:

Thank you for raising this critical concern regarding the application of the Pascal & Josey (2000) correction across all buoy datasets. We acknowledge that the correction's applicability depends on sensor-specific configurations. Our responses to the key points are as follows:

1. Prevalence of High R_{solar} Values

While $R_{\text{solar}} \sim 1000 \text{ W/m}^2$ can occur, such extreme values are rare in our datasets due to typical oceanic cloud cover. Only 1.1% of hourly R_{solar} data exceeded 1000 W/m^2 , limiting the correction's impact.

2. Data Quality and Pre-Calibration

We used "Highest Quality" data, which undergo pre-deployment calibration in the laboratory, as indicated by the quality code definitions (detailed below). While case and dome temperature corrections were not directly available to us, applying a global, consistent correction such as PJ2000 was considered a practical alternative to ensure uniform treatment across all datasets.

0 = Datum Missing.

1 = Highest Quality. Pre/post-deployment calibrations agree to within sensor specifications. In most cases, only pre-deployment calibrations have been applied.

2 = Default Quality. Default value for sensors presently deployed and for sensors which were either not recovered, not calibratable when recovered, or for which pre-deployment calibrations have been determined to be invalid. In most cases, only pre-deployment calibrations have been applied.

3 = Adjusted Data. Pre/post calibrations differ, or original data do not agree with other data sources (e.g., other in situ data or climatology), or original data are noisy. Data have been adjusted in an attempt to reduce the error.

4 = Lower Quality. Pre/post calibrations differ, or data do not agree with other data sources (e.g., other in situ data or climatology), or data are noisy. Data could not be confidently adjusted to correct for error.

5 = Sensor or Tube Failed.

3. Magnitude of Adjustment

Our analysis revealed that the correction resulted in differences of less than 3 W/m² for 89% of hourly-scale LWR data, within the observational uncertainty of 10 W/m². This suggests minimal over-correction impact.

4. Purpose of the Data

Importantly, the buoy data were not used to validate the LWR products but rather to develop the estimation model. Systematic differences, including potential over-corrections, are accounted for as part of the model's offsets, preserving the integrity of our conclusions.

We appreciate the opportunity to clarify this point and hope this addresses your concern.

Comment #2:

The lambda term, on the otherhand, is intended to correct for solar radiation leakage caused by pinholes or degradation of the dielectric coating on the sensor dome. It is not a universal problem and PJ2000 found that the lambda ranged from 0.007 to 0.024. The 0.011 lambda value is thus a middle of the road case. In some cases, the authors will be adding an error, while in other cases, they will be partially fixing it. Perhaps if the authors keep this correction, it might be also treated as an uncertainty estimate.

Alternatively, its more work, but one indication that solar radiation leakage is an issue is if there is a noontime peak in LWR during clearsky days. Perhaps this correction should be applied only in those cases?

Response:

We appreciate your insights into the variability of the lambda term correction. Here is our detailed response:

- The lambda term addresses solar radiation leakage, and its effect varies among sensors. While most sensors exhibit such leakage, the differences between $\lambda=0.007$, 0.011, and 0.024 resulted in minimal changes to LWR, with 78% and 72% of samples showing deviations less than 1 W/m² and 4 W/m², respectively. These values are below the instruments' uncertainty.

We have added the following sentence to the manuscript for clarification:

“The differences between $\lambda = 0.007$, 0.011, and 0.024 resulted in minimal changes to R_i , with 78% and 72% of samples showing deviations of less than 1 W/m² and 4 W/m², respectively. These values are below the sensors' uncertainty. Therefore, we used 0.011 for λ .” (Line 324-327)

- The potential noontime peak in LWR on clear-sky days was considered, but we find that the calibrated differences during non-peak periods remain within the instruments' uncertainty.

Given these results, we maintain that applying a consistent lambda correction introduces negligible bias and does not alter our conclusions.

Comment #3:

The bottom line is that I think the authors should go back and check which sensors are 3-output LWR and then redo the analysis without the a & b terms in the correction for those sites. The authors may also want to review their use of the lambda term correction.

Response:

Thank you for this valuable suggestion. We confirm that we used "high-quality" datasets consisting of raw data that had not been post-processed by the data providers. For these datasets, the Pascal & Josey (2000) correction was applied uniformly to ensure consistency.

Regarding the lambda term correction, we refer you to our detailed response to Comment #2, where we address its impact in our analysis.

Comment #4:

Most of my other comments are to help clarify text, figures, and tables.

With the analysis redone with corrections only applied to the subset of observations that do not already have the correction for heating & solar radiation leakage, and with the manuscript revised for clarity, I expect this will eventually make for a well-cited paper.

Comments to improve clarity:

Table 1 Caption: Add statement that Variables are defined in Table 2.

Response:

Thank you for your suggestion. We have added the requested clarification to the caption. The revised caption now reads:

"Eight Existing Models for Ocean-Surface R_t Estimation, with Variables Defined in Table 2." (Line 134)

Comment #5:

Table 1. Consider expanding this to also include Modnew model. I think this would help the reader find the equation and see its structure in relation to the other models.

Response:

Thank you for the suggestion. Table 1 is intended to present only the existing estimation models for comparison purposes. Since Modnew is a newly developed model introduced in this study, we have chosen to discuss it separately.

Comment #6:

Table 2 show all variables used in this study, including R_l , R_g , DSR_{toa} , CBH etc.

Response:

Thank you for your suggestion. We have updated Table 2 to include all the variables used in this study, such as R_i , R_g , and DSR_{toa} . (Line 273) However, we have excluded CBH as it was not utilized in our analysis.

Comment #7:

Table 3. In the text, it would be nice if you said where or who these 8 OceanSITES stations are. If 4 of these OceanSITES stations are from TAO (this needs to be clarified), then you only need to describe 4 stations, or even fewer groups as some of these stations (e.g. KEO and Papa) are from one group (NOAA Ocean Climate Stations). These smaller groups making these long OceanSITES time series would benefit from being named in this analysis.

Response:

Thank you for this suggestion. We have clarified the origin of the 8 OceanSITES stations in the manuscript by adding the following sentence:

"Eight sites from OceanSITES were utilized, specifically: OS_PAPA, OS_KAUST, OS_NTAS, OS_KEO, OS_ARC, OS_JKEO, OS_STRATUS, and OS_WHOTS." (Line 302)

This ensures transparency and acknowledges the specific sources of these long-term time series datasets.

Comment #8:

Figure 2 caption. What is being represented by the color bar in the left column? What are its units? In the right column, what are the error levels in the "box plots"?

Response:

Thank you for pointing this out. We have clarified the caption for Figure 2 as follows:

"In the left column, the color bar represents points per unit area. In the right column, the dots indicate the mean value of the ΔR_i (ME), while the vertical lines represent the standard error of the mean (SEM)." (Line 467-469)

Comment #9:

Figure 3 caption. What is being represented by the color bar in a and b? What are the units?

Response:

Thank you for your question. The information in the caption of Figure 3 aligns with the clarification provided in our response to Comment #8:

"In panels a and b, the color bar represents points per unit area." (Line 494)

Comment #10:

Figure 4 caption. Same as #6 comment. Also, what is the daytime vs. nighttime criteria?

Response:

For Figure 4, the color bar represents points per unit area.

We have already clarified this criteria in the manuscript as follows:

"The hourly samples used for independent validation were further divided into daytime ($R_g > 120 \text{ W/m}^2$) and nighttime conditions ($R_g \leq 120 \text{ W/m}^2$)" (Line 336-338)

Comment #11:

Figure 7 caption. Same as #6 comment.

Response:

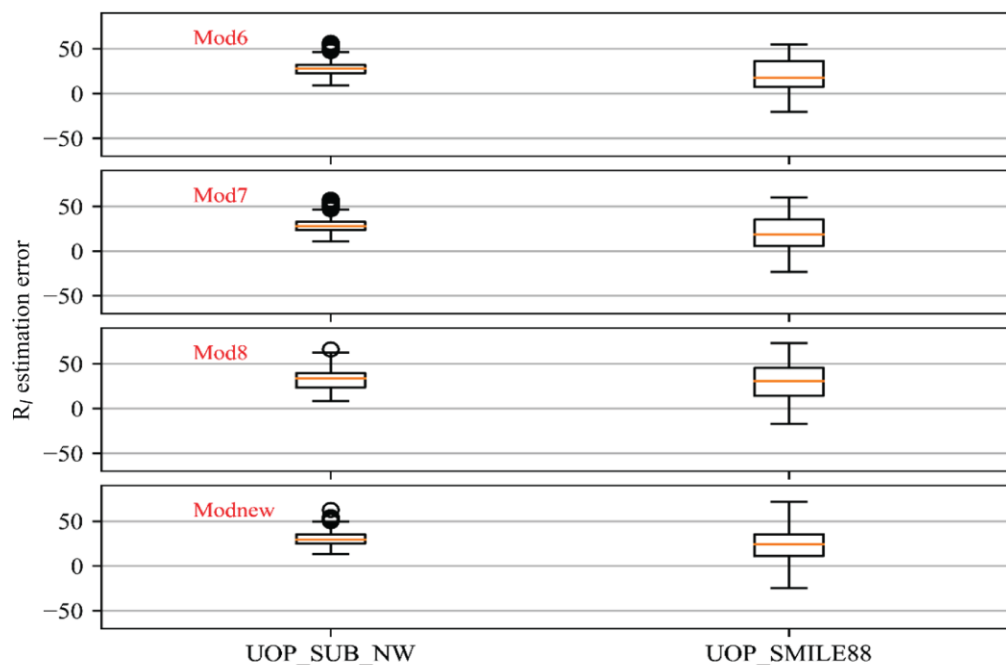
For Figure 7, the color bar represents points per unit area.

Comment #12:

Figure 11. Could you make the y-axis labeling more concise so that it is legible? Also please define what the different levels are in the box plots.

Response:

We have fixed the Figure 11. *The top edge, center, and bottom edge of the box represent the 75th, 50th (median), and 25th percentiles, respectively. The whiskers indicate the maximum and minimum values within 1.5 times the interquartile range (IQR), and the circles denote outliers. (Line 713-715)*



Comment #13:

Throughout the text, the observations are described as “screen-level”. What does this mean? I’ve never heard of this.

Response:

Thank you for pointing this out. In meteorology, "screen-level temperature" refers to the air temperature measured at a standard height of 2 meters above the ground.

Good, E. J. (2016). An in situ-based analysis of the relationship between land surface “skin” and screen-level air temperatures. *Journal of Geophysical Research: Atmospheres*, 121(15), 8801–8819. <https://doi.org/10.1002/2016JD025318>

Comment #14:

Line 41, “Although the ocean-surface Rl is routinely measured at most buoy sites...”. Unfortunately, this hasn’t been true. Of the 55 TAO sites, only 4 have routinely measured longwave radiation. This is being changed in response to the Tropical Pacific Observing System (TPOS) 2020 project (Kessler et al. 2021), but historically and currently, this statement is only true for OceanSITES bulk flux buoy stations.

Response:

Thank you for bringing this to our attention. We have revised the sentence to remove the word “routinely” to ensure accuracy. (Line 43)

Comment #15:

Line 52. “complicacy” is the wrong word I think. Perhaps “complexity” ?

Response:

Thank you for pointing this out. We have replaced “complicacy” with “complexity” to improve accuracy and readability. (Line 54)

Comment #16:

Line 81 “mid-high” --> “mid-to-high” ?

Response:

Thank you for your suggestion. We have updated “mid-high” to “mid-to-high” for improved clarity and correctness. (Line 88)

Comment #17:

Line 312 “At last” --> “In total” ?

Response:

Thank you for your suggestion. We have replaced “At last” with “In total”. (Line 334)

Comment #18:

Line 324 “On the contrary” --> “On the otherhand” ? or “In contrast”

Response:

Thank you for your suggestion. We have replaced “On the contrary” with “On the other hand”. (Line 346)

Comment #19:

Line 331. For regions where winds are weak, afternoon near surface stratification can cause the skin temperature to be quite a bit warmer than the bulk SST. This is mainly an issue in the tropics but can also matter in the summer elsewhere. See Cronin et al. (2024) or Clayson and Bogdanoff (2013)

Response:

Thank you for your insightful comment. While we acknowledge that near-surface stratification and skin temperature deviations from bulk SST can occur under low-wind conditions, especially in tropical regions and during summer, this specific issue is beyond the scope of our study, which focuses on downward longwave radiation (Rl) estimation.

We emphasize the importance of addressing such stratification effects in regions with consistently low wind speeds, where autonomous ship-of-opportunity radiometer systems are particularly useful. However, in our dataset, 83% of the samples were observed under wind speeds exceeding 4 m/s, meaning the

conditions for significant stratification effects were rare. Therefore, the applied correction remains valid for the majority of our dataset.

Furthermore, **only Mod6** incorporates SST as a model parameter, and its influence is moderated by the Stefan-Boltzmann constant ($\sigma=5.67\times 10^{-8}$). As a result, any potential deviations between skin and bulk SST have a minimal impact on the model's performance and do not affect the overall conclusions of our study.

We appreciate the suggested references for further exploration and recognize their relevance to studies focused specifically on stratification effects.

Comment #20:

Line 369, this should reference Table 2, not Table 1.

Response:

Thank you for catching this oversight. We have corrected the reference to point to Table 2 instead of Table 1. [\(Line 391\)](#)

Comment #21:

Line 403. What is CBH ? Perhaps this needs to be included in Table 2.

Response:

Thank you for your comment. In meteorology, CBH typically stands for Cloud Base Height, which refers to the height of the base of a cloud layer above the ground. CBH was not utilized in our study. Therefore, we have chosen not to include it in Table 2.

Comment #22:

Section 4.2.1 Clear sky, is very short. Section 4.2 is Model comparison results, but this 4.2.1 has no results/analysis. How do we interpret these results? Or will that be discussed later? Please let the reader know.

Response:

Thank you for your comment. Section 4.2.1 serves as a subtitle and is further divided into two subsections: 4.2.1.1 (clear sky hourly scale) and 4.2.1.2 (clear sky daily scale). The results and analysis for clear sky conditions are presented within these subsections.

Comment #23:

Line 519. "On the contrary" --> "In contrast"

Response:

Thank you for your suggestion. We have replaced "On the contrary" with "In contrast". [\(Line 547\)](#)

Comment #24:

Paragraph 2 of the Conclusion. Could you use some more words to describe the physical dependencies of the model? This paragraph relies too heavily upon variable names which may be unfamiliar to some readers.

Response:

Thank you for this excellent suggestion. We have expanded the paragraph in the conclusion to include more detailed descriptions of the physical dependencies of the model. The revised text is as follows:

“In this study, the newly developed Modnew model estimates all-sky ocean-surface downward longwave radiation (RL) by incorporating key atmospheric and cloud parameters: screen-level air temperature (Ta), relative humidity (RH), fractional cloud cover (C), total column cloud liquid water (clw), and total column cloud ice water (ciw). Ta governs the thermal radiation emitted by the atmosphere, as described by the Stefan–Boltzmann law. RH modifies the atmospheric emissivity by representing the water vapor content. C quantifies the cloud's overall presence, while clw and ciw capture the thermal contributions of liquid and ice clouds, respectively, enabling a more accurate characterization of cloud radiative effects.” (Line 746-755)

Comment #25:

Paragraph 2 of the Conclusion. Please also clarify how satellite data must be used to run the Modnew

Response:

Thank you for this valuable suggestion. We have expanded the conclusion to clarify the role of satellite data in running the Modnew model. The revised text is as follows:

“The Modnew model relies on specific atmospheric and cloud-related parameters for accurate RL estimation. While inputs such as Ta and RH are commonly obtained from in situ measurements, critical cloud-related parameters (i.e. clw and ciw) are typically derived from satellite products or reanalysis datasets, such as ERA5. These parameters are essential for capturing the radiative properties of clouds, which in situ measurements alone cannot reliably provide. Therefore, satellite data or reanalysis products are indispensable for supplying these inputs.” (Line 755-761)