

PRODUCING SATELLITE-DERIVED IRRADIANCES IN COMPLEX ARID TERRAIN

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ABSTRACT

This paper describes a methodology to correct satellite-derived irradiances over complex terrain. The correction applies to satellite models using visible images from geostationary satellites.

1. DESCRIPTION OF CURRENT MODEL

The authors recently proposed a new semi-empirical model for deriving global and direct irradiances from the visible channel of weather satellites [1]. This model was a logical evolution of earlier work by [2, 3].

In its simplest description the model resumes to the modulation of clear sky -- global and direct -- envelopes as a function of satellite-derived **cloud indices**. The clear sky envelopes are locally adjustable for regional/seasonal turbidity and ground elevation.

For a given time/location, a cloud index is derived from image's pixel brightness in relation to the local pixel's **dynamic range** -- i.e., the possible range of pixel brightness at the considered location, with the darkest pixels corresponding to clear

conditions and the brightest to cloudy conditions. Pixel dynamic range varies as a function of location and time because of ground reflectivity (albedo), ground bi-directional -- specular -- reflectivity, the presence of snow cover, and the satellite sensor's calibration. In its operational version, the model maintains individual, unique dynamic ranges for all locations. Dynamic ranges evolve over time and are derived from the flux of incoming pixel counts at each location (see Fig. 1). This approach allows

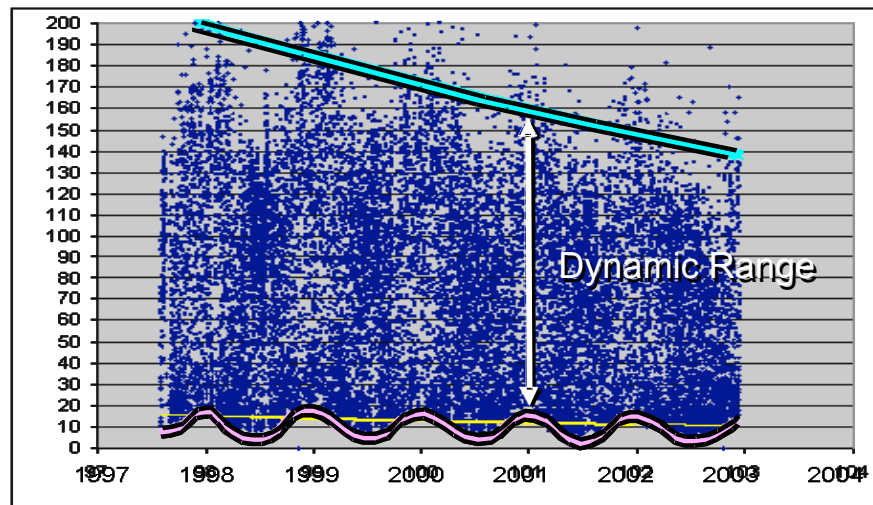


Fig 1: Evolution of dynamic range at a sample location. Each dot represents a pixel count corrected for solar incidence. The upper bound reflects the decay of satellite calibration. The lower bound reflects seasonal variability. Both upper and lower bounds are derived from the history of pixel count at that location.

the model to dispense with absolute satellite calibration [1] and to account for seasonal and geographical changes in ground reflectivity. Further, access to external information on ground snow cover [1] allows the model to also account for snow in its dynamic range management.

Ground specular reflectance had been identified as the main source of model inaccuracy in arid regions. The model addresses the bi-directional reflectance issue by deriving daily/seasonal specular reflectance signatures for each individual location. These signatures are based upon the history of pixel brightness at each considered location, by looking at the variation of observed dynamic ranges' lower bound as a function of time-of-year and time of day [1].

2. LIMITATIONS OF CURRENT MODEL IN COMPLEX AND/OR ARID TERRAIN

The model was tested against ground truth stations located in the arid western US and was found to perform adequately [1]. However, these stations, namely, Daggett, CA, Albuquerque, NM and Burns, OR, are not located in extreme ground reflectance environments.

We define extreme ground reflectance environments as

- (1) very high specularity regions such as salt beds found throughout the southwestern US and Mexico and in many other arid regions of the world, and/or
- (2) the juxtaposition of bright and dark areas, such as arid plains and forested ridges (see Fig.2)

In the first case, the specular signature imbedded in the current model proved to be sometimes insufficient, leading to small but significant underestimations, particularly noticeable on the direct irradiance (DNI) component. In the second case, slight satellite navigation uncertainties may induce very large errors. Indeed, because location-specific dynamic ranges are maintained by the flux of incoming pixel counts at that location, a satellite navigation error may at times "throw in" dark pixels (from a forested ridge) into otherwise bright ground regions. These few dark pixels lower the local dynamic range's lower bound, resulting in large model underestimation, because most clear occurrences are mistaken for cloudy conditions when contrasting the clear bright ground pixel against the darker dynamic range. This process is illustrated in Fig. 3.

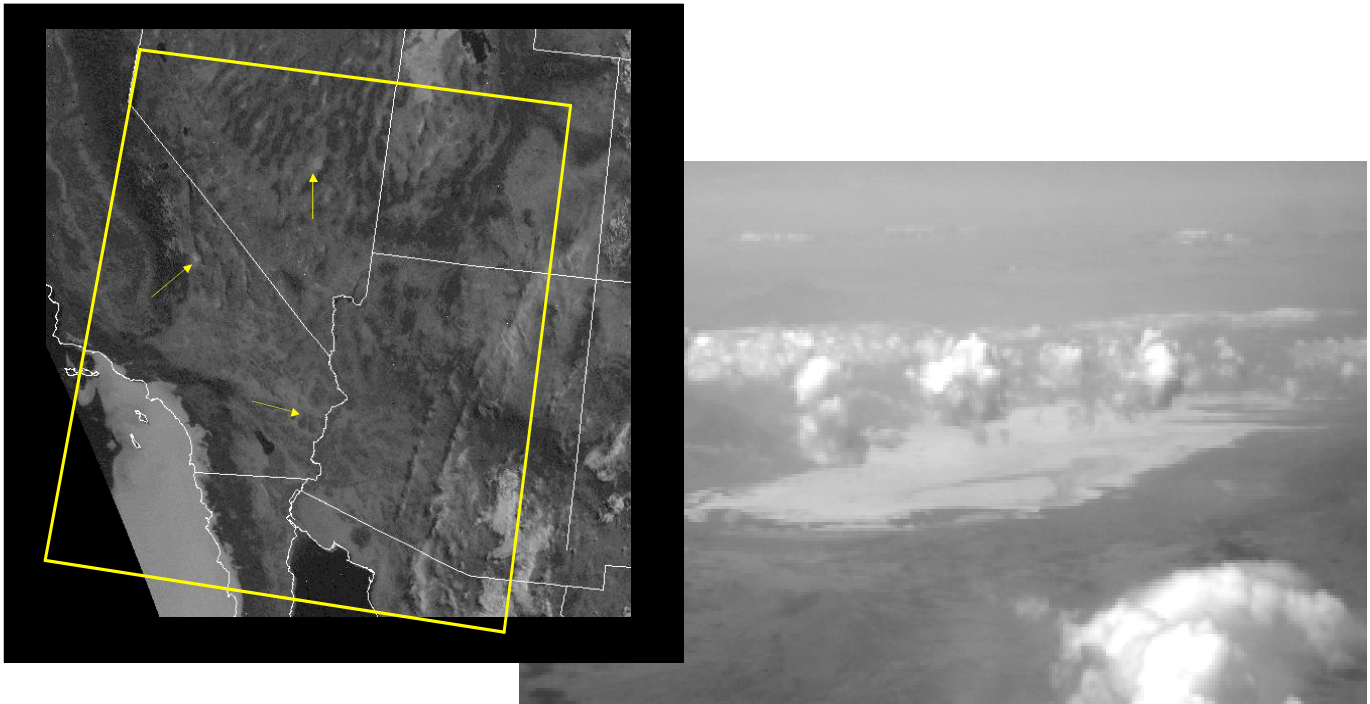


Fig. 2: Satellite view of the southwest US showing complex ground reflectivity (three extreme cases shown by arrows), and airplane view of a Nevada Salt bed

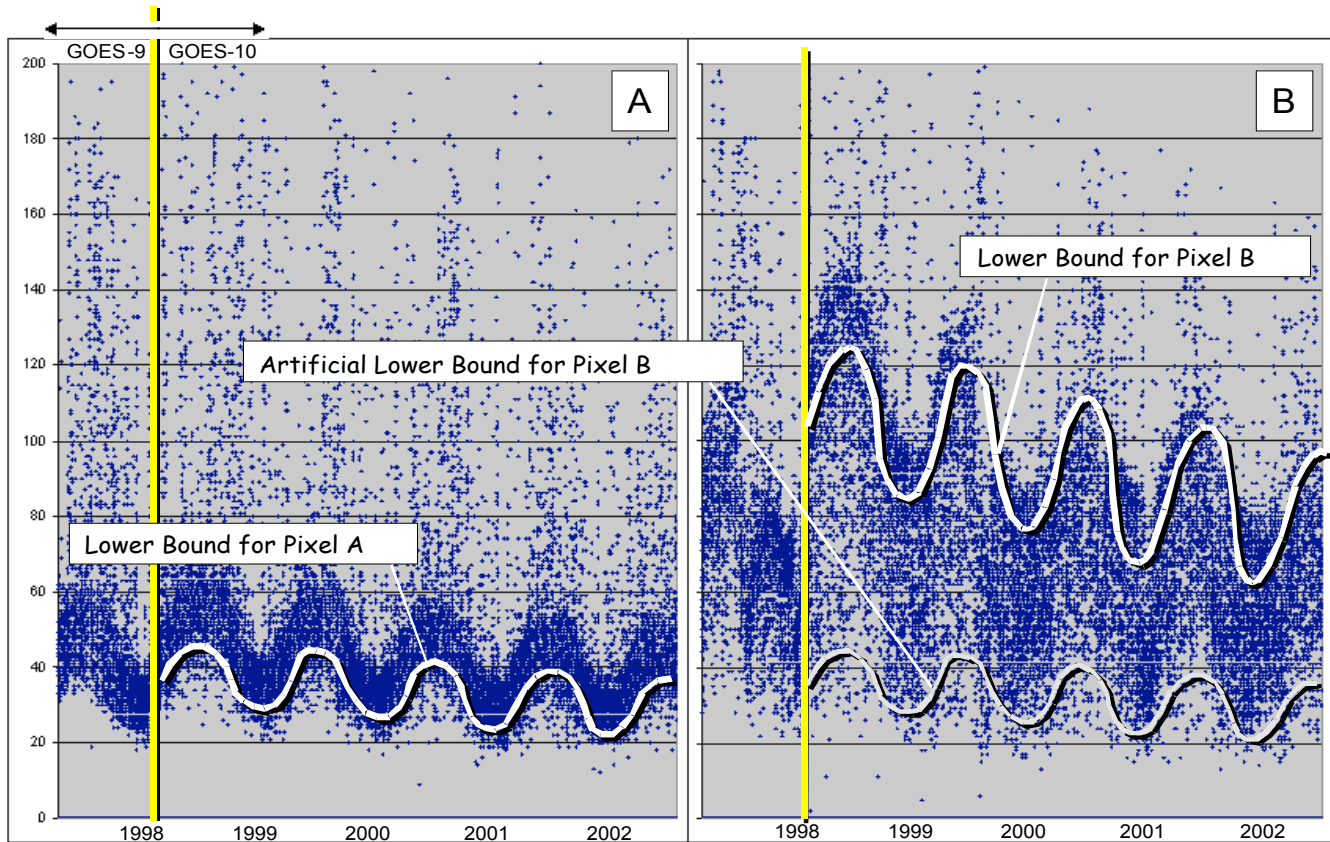


Fig. 3: Illustration of dynamic ranges for two neighboring pixels near Death-Valley, California. One of the pixels (A) has a dark albedo and the other (B) a very high albedo. Routine satellite navigation uncertainties result in an artificial decrease of pixel B's lower bound, resulting in cloud index overestimation and irradiance underestimation at that location.

3. PROPOSED SOLUTION

We developed a two-step addition to the model in an attempt to address these shortcomings.

Step-1: Clear sky calibration

By definition, the lower bound of the dynamic range corresponds to clear sky global (GHI) and direct (DNI) irradiances. At some locations, this clear sky limit may not be achieved by the model, because of an underestimated specular signature, or because of the contamination of the lower bound by neighboring darker pixels. In the first case the limit may be reached only at some hours in the day. In the second case, the limit may not be reached at all. In order to force the model to reach this limit, we postulate that for each monthly period, GHI and DNI are bound to reach their clear sky limit at least "n" times for each daylight hour. The value of "n" may be adjusted to reflect prevailing regional insolation conditions. For very clear/arid regions in the SW US, a value in the 5-10 range was found to be conservative.

This assumption allows us to derive a set of correction coefficients for each pixel, each time-of-day and each month, defined as the ratio between the clear sky (DNI or GHI) value and the n^{th} highest achieved value for each hour in the considered monthly period, and thus to produce a calibrated time series.

Step-2: Removal of singularities

Even after application of the above correction, we found that there remained micro regions of model underestimation. We believe that in these small, highly reflective regions, the model approaches its clear sky limit often enough, because of mis-navigated neighboring darker pixels, so that the clear sky calibration correction is underestimated. These micro-regions errors become visible as singularities when average DNI values are mapped (see Fig. 3). Our approach for this second step correction is to use the average maps themselves as the instrument of data correction. For both the DNI and GHI components, the corrective process scans the monthly average maps in latitude and longitude to detect

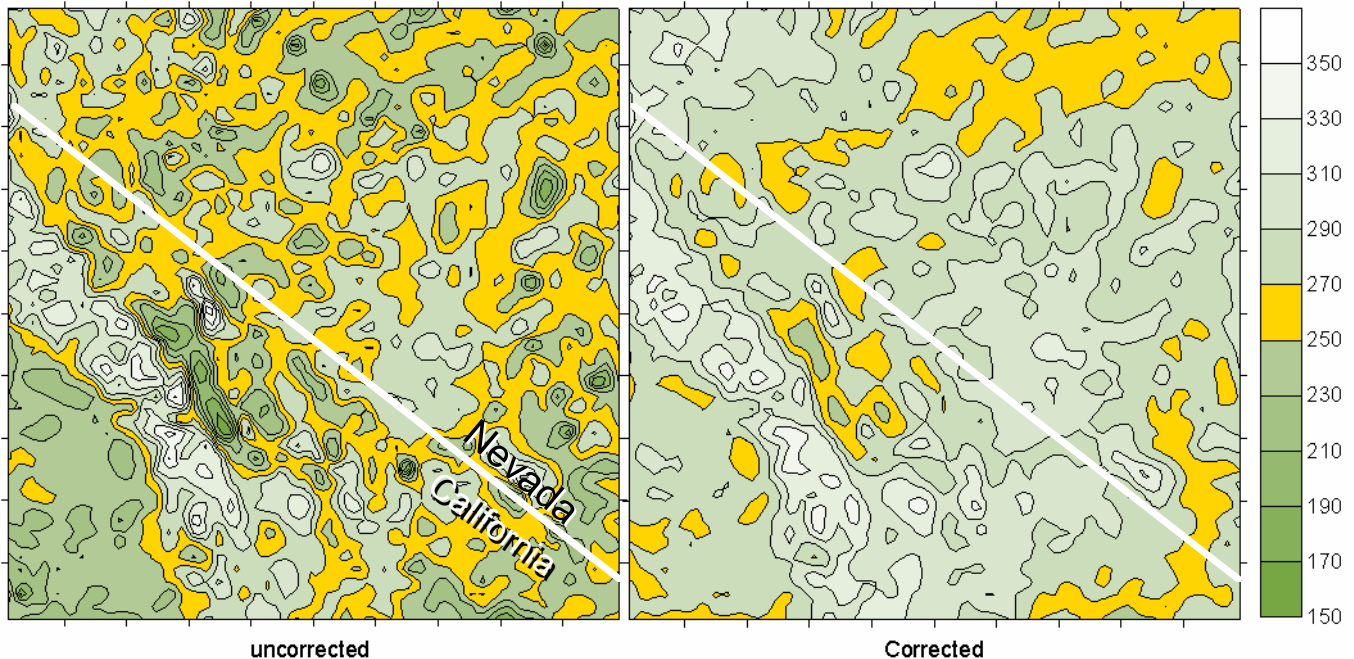


Fig. 4: Comparing uncorrected and corrected monthly averaged direct irradiance maps in a 400 x 400 km region straddling California and Nevada (average Wm^{-2})

pixel-to-pixel irradiance variation. If a threshold -- currently set at 2.5% for monthly average DNI and 1.5% for GHI for two neighboring locations (distant of 10 km) -- is exceeded, the model reduces the singularity by returning an average of the neighboring pixels. In this process, a monthly averaged map without singularities (fig. 3) is constructed. If necessary, time series at the corrected locations may be generated using a secondary model previously developed by the authors [4]. This secondary model is designed to simulate a time series of global and direct irradiance from the knowledge of (1) an existing time series -- in the present case the uncorrected time series -- and (2) the monthly average clearness index (kt' [5]) difference between the uncorrected and the corrected monthly maps (see Fig. 3).

4. DISCUSSION

We have presented a robust, straightforward two-step approach to correct irradiance estimated from weather satellites' visible channel, in cases where terrain reflectivity and texture limit the model ability to perform reliably. The first step -- clear sky calibration -- typically results in correction of less than 5% for global and less than 10% for direct in the most difficult terrain conditions. The second step may result in higher corrections, but only for a very limited number of pixel locations.

As an alternate approach we are exploring using the IR channel in addition to the visible channel in these difficult locations

Validation: the initial model has been thoroughly validated for several climatically distinct locations [1]. However, the current secondary approach has not been tested -- primarily because no ground truth stations have been deployed in problematic micro-regions. So until this is done, model results in these circumstances should be viewed with a degree of caution. At the very least, the proposed method provides a means of assessing data quality by gauging the magnitude of the correction applied to any given pixel.

5. ACKNOWLEDGEMENT

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