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# **Recent progress in land remote sensing: algorithms and products**

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### EDITORIAL

### Recent progress in land remote sensing: algorithms and products

During the past half century, since the launch of the first meteorological satellite in April 1960, satellite remote sensing technologies for observing the Earth have made great advances. Today, more than 200 Earth observing satellites are in obit and substantial progress has been achieved in extracting land surface information from many of their observations.

The National Aeronautics and Space Administration (NASA)'s Earth Observing System marks a new era of land remote sensing. One of its distinguishing features is the ability to generate high-level land products from most satellite missions. Thus, users can obtain not only satellite imagery with appropriate geometric and radiometric calibration and correction, but also many biogeophysical parameters that characterize land surface conditions. Generating high-level biogeophysical products has been driven by several factors. First, although huge quantities of satellite observation data from different missions have been generated, many ordinary users are unable to process them effectively. Second, different types of satellite data require sophisticated radiative transfer models and inversion algorithms to extract accurate land surface information. Third, many applications require a set of common preprocessing and analysis that should not be repeatedly done by individual users. Therefore, researchers with special skills and knowledge and a central processing facility have produced different high-level land products.

The user communities have expressed their need for high-level land products. For example, the Global Climate Observing System (GCOS) identified 50 Essential Climate Variables (ECVs) to support the work of the United Nations Framework Convention on Climate Change and the Intergovernmental Panel on Climate Change (GCOS [2010\)](#page-5-0). All ECVs are technically and economically feasible for systematic observation, but many land surface-related ECVs can be monitored at different spatial scales by satellite remote sensing. These include surface radiation budget, snow cover, glaciers and ice caps, ice sheets, albedo, land cover (including vegetation type), fraction of absorbed photosynthetically active radiation, leaf area index (LAI), above-ground biomass, and soil moisture.

The instantaneous estimates of geophysical variables retrieved in near real-time using only information from the past satellite observations are called Environmental Data Records (EDR). Many EDRs are being generated by the operational agencies, such as National Oceanic and Atmospheric Administration (NOAA). For scientific research, science-quality time series data records are known as Earth System Data Records.

Most land products are generated from specific mission-oriented satellite data. Since each satellite has a limited life span, many biogeophysical parameters cover only limited periods of time. However, long-term products are needed for climate change and environmental change studies. A Climate Data Record (CDR) is defined by the National Research Council as a time series of measurements of sufficient length, consistency, and continuity to determine climate variability and change (NRC [2004\)](#page-5-0). Efforts are currently under way to generate CDRs, such as NOAA's National Climatic Data Center satellite CDR program ([http://www.ncdc.noaa.gov/cdr/operationalcdrs.html\)](http://www.ncdc.noaa.gov/cdr/operationalcdrs.html).

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To assess the progress and identify the future research directions of land remote sensing, the First International Conference on Land Remote Sensing: Algorithm and Product was held in Chengdu, China, 30 May through 1 June 2012. More than 150 people attended this conference which featured eight sessions: land surface albedo, skin temperature and emissivity, LAI, biomass, incident solar radiation, soil moisture, and energy budget and phenology. Selected papers will be published in three special issues: The International Journal of Digital Earth, The Journal of Mountain Science, and The Journal of Remote Sensing.

The eight papers selected for this special issue are outlined below.

Liang et al. ([2013\)](#page-5-0) describe an effort for producing long-term land products from multiple satellite data in China. They have released five Global LAnd Surface Satellite (GLASS) products, including LAI, shortwave broadband albedo, longwave broadband emissivity, incident short radiation and photosynthetically active radiation (PAR). The first three products cover from 1981 to at present (was from 1981–2010) at 1–5 km and 8-day resolutions, longest similar products in the world. This paper also demonstrates the high quality and accuracy of the GLASS products through validation and also identifies potential applications and some analysis examples of these products.

Liu et al. ([2013\)](#page-5-0) present further details of the GLASS albedo product, which originally covered 1981–2010 but has now been extended to present. The paper introduces the algorithms and then analyzes the quality of the GLASS albedo product. It further demonstrates that the GLASS albedo product is spatially and temporally continuous and self-consistent with a comparable accuracy with the Moderate Resolution Imaging Spectroadiometer (MODIS) albedo product.

Huang et al. ([2013\)](#page-5-0) provide the validation results of the GLASS shortwave radiation product using surface measurements in arid and semi-arid regions of China. A more sophisticated validation scheme has been developed to reduce the impact of many other factors. Results indicate that the GLASS shortwave radiation products are considerably accurate over most parts of arid and semi-arid regions in China, but in complex terrain areas the products may need further refinements.

Cheng and Liang [\(2013](#page-4-0)) describe the algorithms and validation results of the GLASS longwave emissivity product from the Advanced Very High Resolution Radiometer visible and near infrared data during 1985–1999. The retrieved emissivity values were first compared with values derived from the MODIS data, then with those derived by the normalized difference vegetation index (NDVI)-threshold method. The results indicate the GLASS longwave emissivity product is highly consistent for the entire period.

Dong et al. [\(2013](#page-5-0)) validate the GLASS longwave emissivity product with multi-point field measurement data using infrared radiometry in China. Very few ground measurements of surface longwave emissivity were available in the remote sensing community. The GLASS longwave emissivity product was validated over bare surfaces with field measurement data from sand samples collected at many pseudo-invariant sand dune sites in western/northwestern China. These results indicate that the GLASS emissivity product is very accurate over arid and semi-arid regions, and for most deserts such as the Taklimakan Desert.

The GLASS products have been demonstrated to be more accurate than the MODIS corresponding products. Besides the new inversion algorithms compared to the MODIS algorithms typically developed before 2000 with constant refinements, further preprocessing of the MODIS surface reflectance product was the key. Tang et al. [\(2013](#page-5-0)) introduce a new cloud detection algorithm based on the relative stability of ground reflectance and <span id="page-4-0"></span>the quick variations of reflectance result from cloud cover. It shows the improvements over the MODIS algorithm.

Land surface temperature (LST) is another key variable in radiation and energy budgets. Wu et al. ([2013\)](#page-5-0) present a new spatial-temporal fusion method for retrieving LST at high spatial and temporal resolutions. The method was tested and assessed quantitatively using Landsat TM/ETM+, MODIS, and the GOES Imager data. The validation results indicate that the proposed multi-sensor fusion method performs well.

A paper by Sun et al. ([2013\)](#page-5-0) presents an improved Penman–Monteith method for estimating evapotranspiration (ET) of land surfaces. It incorporates a soil moisture index, derived from the improved surface temperature–vegetation cover feature space. Validation results indicate that it improves the most over sparse vegetation covers.

The land remote sensing community is faced with continuing challenges. Immense amounts of data available from satellite observations offer promise yet also require efficient and effective inversion algorithms. Considerable investments were put into developing physical models to understand surface radiation regimes in early years, but few people now are developing realistic and computationally simplified surface radiative transfer models mostly suitable for inversion of land surface variables from satellite data. Such models are urgently needed. State-of-the-art remote sensing inversion is well advanced, but the use of regularization methods by incorporating a-priori knowledge and integrating multiple-source data from different spectra and instruments deserves further research. More efforts are needed to develop various land CDRs. The good news is that the second phase of the GLASS project, which aims to produce a total of 12 GLASS products including evapotranspiration, and gross primary productivity, has begun. Validating, intercomparing and eventually integrating multiple products of the same biogeophysical variable are also needed.

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