Mapping land surface radiation and energy budget from the AVIRIS and MASTER data

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- The land surface responds to climate variability and also modulates the climate through the exchange of energy, matter and momentum.
- Funded by the NASA HyspIRI program, we have developed algorithms to estimate a series of surface radiation budget components from the airborne AVIRIS and MASTER data.

Objectives

Quantification of the variations in land surface radiation and energy budget over different land cover types in response of climate variability from the AVIRIS and MASTER data to support the development of the HyspIRI mission.

- Mapping the surface radiation and energy budget components from both AVIRIS and MASTER data.
 Algorithm development/refinement
 Algorithm and data validation
 - □ Mapping of surface radiation and energy budget components
- 2. Quantifying the variations in surface energy budget of different surface types.
 - □ Mapping land cover types from AVIRIS and MASTER data
 - Assessing variations in those surface radiation and energy budget components of different cover types under various climate conditions
 - □ Addressing a set of scientific questions using these datasets

NASA Grant, Characterizing land surface energy budget under varying climatic conditions from the AVIRIS and MASTER data, Program Manager: Woody Turner, 2012-2015









He, et al, (2014). Analysis of global land surface albedo climatology and spatial-temporal variation during 1981-2010 from multiple satellite products. *Journal of Geophysical Research: Atmospheres, 119, 10281-10298*

Land surface albedo trend in the past decades



Trends in January albedo (per decade): GLASS: 0.0029*** GEWEX: 0.0047 ISCCP: N/A CLARA-SAL: 0.0002

Trends in July albedo (per decade): GLASS: -0.0013*** GEWEX: -0.0053 ISCCP: -0.0086*** CLARA-SAL: N/A

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Long-term trend in surface albedo over the Northern Hemisphere

(He, et al., 2014. JGR)

Direct estimation of surface albedo



Liang, S., (2003), A direct algorithm for estimating land surface broadband albedos from MODIS imagery, *IEEE Trans. Geosci. Remote Sen.*, 41(1):136-145;

Liang, S., J. Stroeve and J. Box, (2005), Mapping daily snow shortwave broadband albedo from MODIS: The improved direct estimation algorithm and validation, *Journal of Geophysical Research.* 110 (D10): Art. No. D10109.

VIIRS shortwave albedo product



Validation results of 16-day mean albedo from VIIRS BRDF LUT (left), and MODIS (right), using data from 2012 non-snow seasons (May-September) at seven SURFRAD sites.

Wang, D., Liang, S., He, T., Yu, Y. 2013. Direct Estimation of Land Surface Albedo from VIIRS Data: Algorithm Improvement and Preliminary Validation. *JGR-Atmosphere*. *118*, 12577-12586, doi: 10.1002/2013jd020417.



Spectral vs. angular information in broadband albedo direct estimation



Hyperspectral information is MORE important than angular information in surface broadband albedo estimation for snow-free surfaces

Albedo estimation accuracy and view zenith angle (solar zenith angle=35°)

Simulated surface albedo

Variable			Value		
View zenith angle (°)	0	15	30	45	60
RMSE	0.0155	0.0100	0.0091	0.0104	0.0147
R ²	0.9181	0.9505	0.9493	0.9261	0.8596

He, T., S. Liang, D. Wang, and Q. Shi, (2014). Estimation of high-resolution land surface shortwave albedo from AVIRIS data. *IEEE JSTARS*, *7*, 4919-4928

Validation of AVIRIS albedo estimates



Validation of surface shortwave albedos at sites from (a) AmeriFlux network and (b) UCI network

He et al. 2013; He et al. 2014



Mapping surface albedo: AVIRIS vs. Landsat



Bias: 0.002 0.45 RMSE: 0.032 0.4 R²: 0.762 0.35 AVIRIS SW albedo 0.2 0.25 0.2 0.15 0.1 (C) 0.05 0.05 01 0.15 0.25 0.45 0.5 0.2 0.3 0.35 04 Landsat SW albedo

Shortwave albedo estimations from: (a) Landsat TM on Aug 18th, 2010; (b) AVIRIS on Aug 26th, 2010 using the stepwise regression algorithm; and (c) scatter plot. Image is centered at 43.08°N, 89.41°W in Madison, WI, USA.



Estimation of shortwave net radiation



Kim, H.Y., & Liang, S. (2010). Development of a hybrid method for estimating land surface shortwave net radiation from MODIS data. *Remote Sensing of Environment*, 114, 2393-2402

Algorithm refinement for Landsat data



Comparison between instantaneous and daily SSNR (in W/m²) using the method of water vapor correction and in situ measurements at six AmeriFlux sites from 2003-2005.

Wang, D., & Liang, S. (2014). Mapping High-Resolution Surface Shortwave Net Radiation From Landsat Data. *Ieee Geoscience and Remote Sensing Letters, 11*, 459-463



Estimating shortwave net radiation from AVIRIS



He, T.,.S. Liang, D. Wang, Q. Shi, and M. Goulden. 2015. Estimation of highresolution land surface net shortwave radiation from AVIRIS data: Algorithm development and preliminary results, *RSE*, *167*, 20-30



Advantages of hyperspectral information in net shortwave radiation estimation



Conversion of cumulated radiation from AVIRIS bands to shortwave radiation (W/m²)

Method B



Statistics of NSR direct estimation based on simulation data

RMSE (W/m ²)		Solar zenith angle (°)							
		20	25	30	35	40	45		
	0	19.396	19.422	19.394	19.290	19.088	18.768		
View	5	19.412	19.434	19.429	19.358	19.192	18.906		
zenith	10	19.369	19.403	19.397	19.346	19.210	18.954		
angle (°)	15	19.276	19.319	19.312	19.267	19.150	18.921		
	20	19.159	19.186	19.184	19.130	19.026	18.815		
\mathbf{p}^2		Solar zenith angle (°)							
ĸ	20	25	20	25	40	15			
		20	25	50	33	40	45		
	0	0.975	0.973	0.970	<u> </u>	0.963	43 0.957		
View	0 5	0.975 0.975	0.973 0.973	0.970 0.970	33 0.967 0.967	0.963 0.962	43 0.957 0.957		
View zenith	0 5 10	0.975 0.975 0.975	0.973 0.973 0.973	0.970 0.970 0.970 0.970	0.967 0.967 0.967	0.963 0.962 0.962	43 0.957 0.957 0.956		
View zenith angle (°)	0 5 10 15	0.975 0.975 0.975 0.976	0.973 0.973 0.973 0.973	0.970 0.970 0.970 0.971	0.967 0.967 0.967 0.967	0.963 0.962 0.962 0.962	43 0.957 0.957 0.956 0.957		





Impacts of water vapor estimation uncertainty on AVIRIS shortwave net radiation direct estimation (He et al. 2015, RSE)



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AVIRIS water vapor estimation has an overestimation compared with MODIS product. Net radiation estimation W/m² is more accurate using AVIRIS-derived water vapor.



Comparison of ground measurements AVIRIS downward radiation (a) and net radiation from Method A (b) and Method B (c) estimates (W/m²) at AmeriFlux sites.

Two methods had similar estimation accuracies. N is smaller in (b) than that in (c) because surface albedo estimates were not available under cloud/shadow conditions.



Combined VSWIR and TIR to estimate clear-sky all-wave net radiation

- Previous methods to estimate all-wave net radiation
 - Component-based, summation of all the components
 - VSWIR: shortwave fluxes
 - TIR: long wave fluxes
 - Errors add up
- New method:
 - Combine VSWIR and TIR data to directly estimate allwave net radiation as one integral quantity.

Wang, D., Liang, S., He, T., & Shi, Q. (2015). Estimating clear-sky all-wave net radiation from combined visible and shortwave infrared (VSWIR) and thermal infrared (TIR) remote sensing data. *Remote Sensing of Environment, RSE, 167*, 31-39.



Validation: net radiation

- Seven SURFRAD stations.
- One year measurements.





700

800

Map from MASTER data





Preliminary Result of Land Cover Mapping

- Method: Maximum Likelihood Classification (MLC)
- Data: AVIRIS band10-60 surface reflectance





RGB: Band 53,30,20 (850 nm, 650 nm, 550 nm)









Mapping ET from MASTER and ancillary data



Validating latent heat flux at AmeriFlux sites





Summary

- Fine spectral and spatial resolution in both shortwave (VSWIR) and longwave (TIR) spectrum provides the scientific community an unprecedented opportunity to map the surface radiation and energy budget over heterogeneous landscapes.
- Hyperspectral AVIRIS data can be used for estimating land surface shortwave albedo and net radiation accurately.
- Clear-sky all-wave net radiation can be directly estimated with improved accuracy from the synergic use of VSWIR and TIR data.
- Hyperspectral data provide ample information for land cover classification.
- ET can be estimated with high accuracy at most sites. Additional algorithm tuning is needed to further improve the estuation.

