

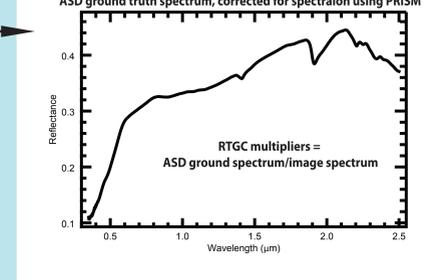
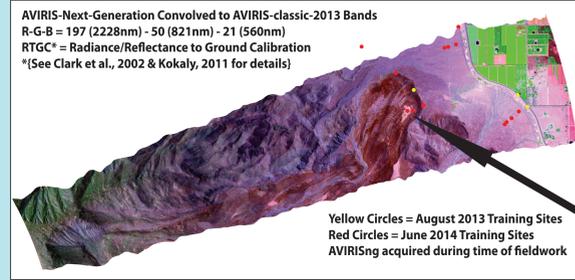
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Example Field Calibration Training Site Used for RTGC Correction of Imagery

Introduction

The wealth of image data that will be produced during the course of the HypsIRI mission creates the need for new software tools and processing algorithms that can be used to rapidly derive land surface material maps from standardized higher level VSWIR (380 – 2500 nm) data products. One potentially useful processing routine is the Material Identification and Characterization Algorithm (MICA) developed at the USGS as a module within the Processing Routines in IDL for Spectroscopic Measurements (PRISM) software. This program can be used as a simple plug-in for the ENVI (ENvironment for Visualizing Images) image processing system. MICA employs continuum removal and linear regression to compare observed spectral absorption features (e.g. such as those in HypsIRI pixel spectra) to those diagnostic of reference materials contained within a spectral library or measured separately in the field or laboratory; as such, it builds on the legacy of the USGS Tetracorder algorithm. PRISM is designed to be "user friendly" and has been applied to mineral detection over large areas such as Afghanistan.

In this study, we show and compare the results of MICA mineral mapping around the Salton Sea area of Southern California, using AVIRIS-classic, AVIRIS-next-generation, Hyperion and AVIRIS-classic convolved to both 30-m and 60-m simulated HypsIRI datasets. Examples of minerals mapped successfully using MICA includes: calcite and aragonite dominating lacustrine deposits left by the receding Lake Cahulla; montmorillonite, illite and/or muscovite found as coatings on quartz grains in wind-blown aeolian deposits; and variable mineralogy of alluvial fans depending on their source bedrock compositions and surface processes of modification. For example, some alluvial fans are dominated by muscovite, chlorite and/or epidote depending on the amount of unaltered granite relative to mylonitized granite and schists found in bar-n-swale materials. Younger and more active fans tend to be dominated by weathering clays such as montmorillonite, which forms abundant soil crusts after major floods. Montmorillonite often mixes with kaolinite and muscovite in alluvial fans derived from hydrothermally altered source volcanic rocks. Older alluvial fans tend to be dominated by desert varnish coatings containing nanocrystalline forms of hematite, mixed with montmorillonite and/or illite. Additional Fe₂₊, Fe₃₊, serpentine minerals, buddingtonite and gypsum were mapped in other bedrock areas, as well as ancient and modern geothermal deposits. We also highlight differences between mineral and vegetation maps caused by variations in the spatial-, spectral- and radiometric-resolution characteristics of the sensor data being input to MICA. Finally, we also assess data quality issues by comparing spectra from these sensors with in situ measurements of our study area collected using an ASD (Analytical Spectral Device) field spectrometer during August 2013 and June 2014.



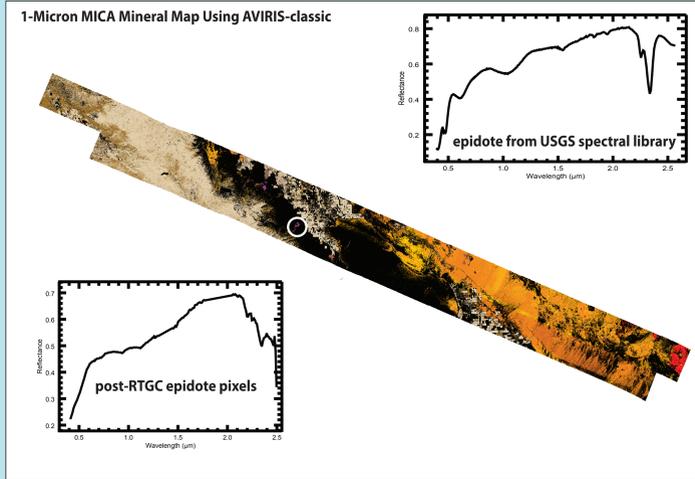
MICA Mapping Results Color Legend*

*Based on the command file constructed to produce mineral assessment maps for Afghanistan (see Kokaly et al., 2013) modified for this study by the inclusion of the minerals aragonite and nontronite, as well as a green microlite mineral discovered during the course of this study area in parts of the Santa Rosa Mylonite Zone that will be added later.

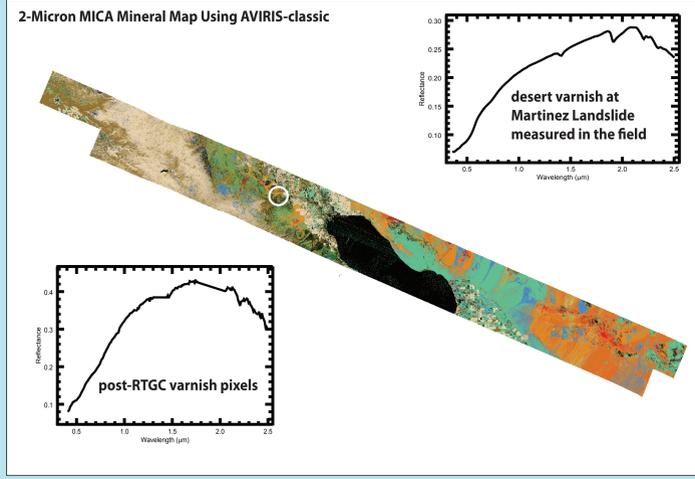
1-m micron thematic map				MICA 1-µm summary image				Reference spectrum title			
Class value in map	Color in map	Red	Green	Class names in MICA summary image	Class value in summary image	Color in summary image	Red	Green	Blue	Reference spectrum title	
Not classified	0	0	0	Not Classified	0	0	0	0	0	not applicable	
Hematite, nanocrystalline	1	255	0	Hematite	1	255	0	0	0	Nanohematite BR03-1482 W R IR FB	
Hematite, fine-grained	2	215	0	Hematite	2	215	0	0	0	Hematite FE2602 W R IR FB	
Hematite, medium-grained	3	175	0	Hematite	3	175	0	0	0	Hematite GDS27 W R IR FB	
Hematite, coarse-grained	4	140	0	Hematite	4	140	0	0	0	Hematite_Coast_Qtz BR03-1250 W R IR FB	
Iron hydroxide	5	255	200	Fe hydroxide	5	255	200	0	0	Fe Hydroxide SC93-106 anorth W R IR FB	
Goethite, fine-grained	6	255	170	Goethite	6	255	170	0	0	Goethite_Thin_WR222 W R IR FB	
Goethite, medium-grained	7	240	130	Goethite	7	240	130	0	0	Goethite MPCM43-B FinGradj W R IR FB	
Goethite, coarse-grained	8	215	100	Goethite	8	215	100	0	0	Goethite MPCM43-B FinGradj W R IR FB	
Goethite and jarosite	9	223	190	Goethite	9	223	190	0	0	Lepidocrocite GDS08 (S) W R IR FB	
Jarosite	10	218	112	Jarosite	10	218	112	214	17	Jarosite, anhydrous BR03-34A2 W R IR FB	
Fe ²⁺ type 1	11	0	48	Fe ²⁺ type 1a	11	0	48	255	19	Jarosite GDS99 K 2002_Syn W R IR FB	
Fe ²⁺ type 2	12	20	132	Fe ²⁺ type 2a	12	20	132	255	19	Nontronite BR03-38 Pyrrhotite W R IR FB	
Fe ²⁺ type 3	13	150	230	Fe ²⁺ type 3a	13	150	230	20	20	Diopside NMN1885-160 Pyrrhotite W R IR FB	
Fe ²⁺ type 4	14	125	170	Fe ²⁺ type 4a	14	125	170	20	20	Schweitzermanite BR03-1 W R IR FB	
				Fe ²⁺ type 4b	15	130	210	0	0	Acid_Min_Dr_Assemb-Fe ²⁺ W R IR FB	
				Fe ²⁺ type 4c	16	125	170	0	0	Acid_Min_Dr_Assemb-Fe ²⁺ W R IR FB	
				Fe ²⁺ type 4d	17	125	170	0	0	Doeberl_Varnish GDS28A Rhy W R IR FB	

Results-1: AVIRIS-classic MICA mineral maps compared to 60-m HypsIRI-convolved

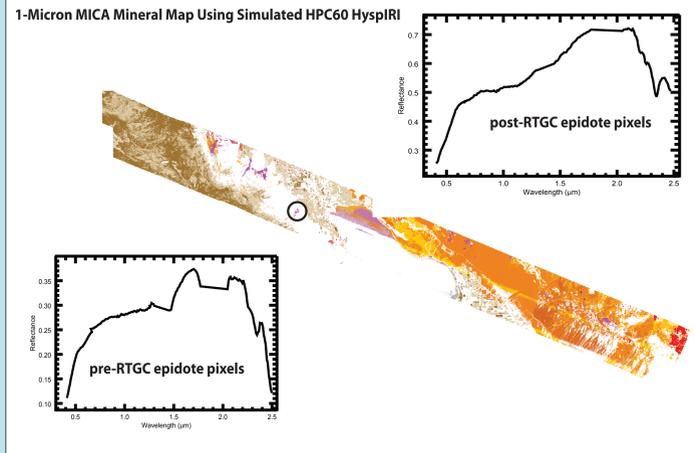
1-Micron MICA Mineral Map Using AVIRIS-classic



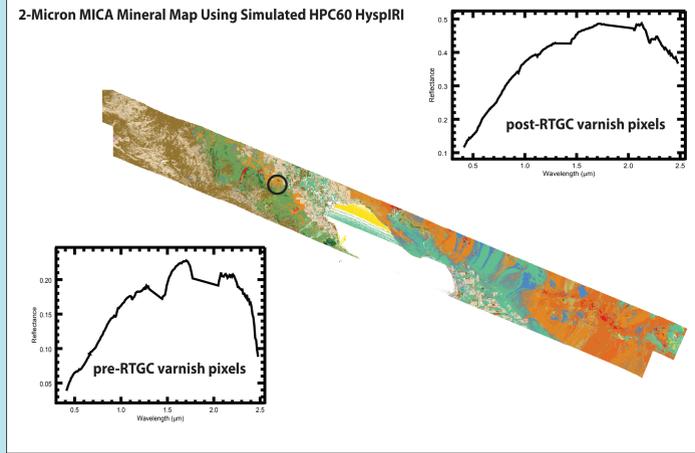
2-Micron MICA Mineral Map Using AVIRIS-classic



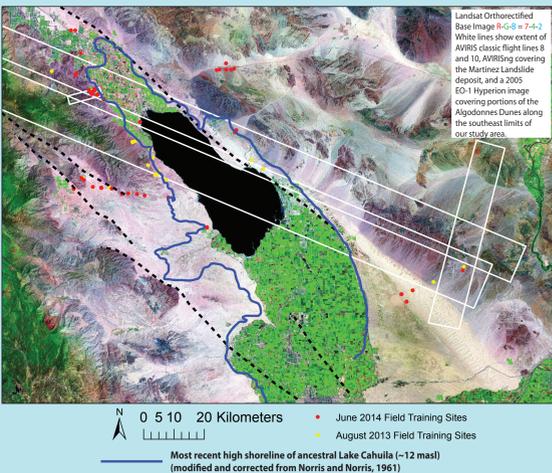
1-Micron MICA Mineral Map Using Simulated HPC60 HypsIRI



2-Micron MICA Mineral Map Using Simulated HPC60 HypsIRI



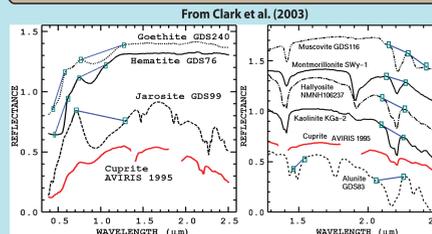
Salton Sea, Southern California Study Area, Datasets and Methods



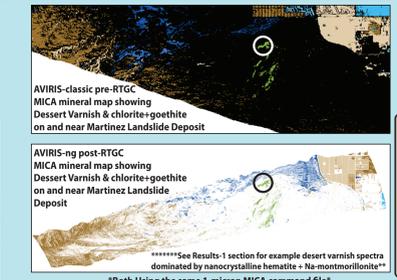
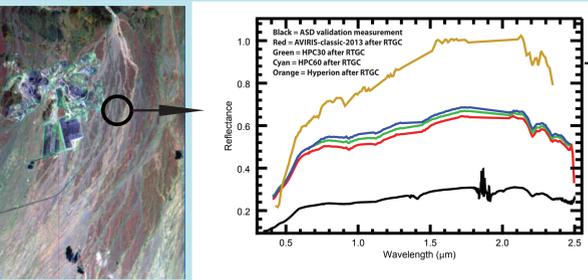
The overarching goals of this project is to evaluate HypsIRI-like datasets for mapping minerals that can be used to: 1) understand flood hazards in desert alluvial fans washes, and 2) to discriminate older pavement surfaces from more recently active channels that are more prone to flooding.

We used MICA as a first cut effort at rapid classification of the most dominant minerals in each pixel of AVIRIS-classic, AVIRIS-ng and Hyperion imaging spectrometer data, and subsequent HypsIRI convolved datasets. MICA is loosely based on the Tetracorder algorithm described by Clark et al (2003) and uses continuum-removal, least squares linear regression and analysis of absorption feature band depths to compare observed image spectral features with those from minerals and other substances compiled in spectral libraries (e.g. Clark et al., 2007).

MICA mineral maps helped to guide our fieldwork plans and to select spectral training sites for ASD measurements, calibration and validation purposes, as highlighted in this poster. MICA mineral maps have helped us to: 1) understand the complexity of mixing within alluvial fan bar-n-swale environments and between interacting alluvium, lake and wind-blown sediments; 2) locate spectral endmembers for key minerals indicative of age (e.g. 1-micron desert varnish) and recent flood activity (e.g. 2-micron montmorillonite); 3) gauge the role of rock alteration and related surface processes in understanding potential hazards related to landslides and debris flows; and 4) how to make effective use of accompanying TIR spectral emissivity measurements as shown on the left.



Results-2: How far have we come & where are we going: Hyperion, HypsIRI, AVIRIS-cl, AVIRIS-ng



References Cited

Clark, R.N., Swayze, G.A., Livo, K.E., Kokaly, R.F., King, T.V., Dalton, J.B., Vance, J.S., Rockwell, B.W., Hoefen, T., and McDougal, R.R., 2002. Surface reflectance calibration of terrestrial imaging spectroscopy data: A tutorial using AVIRIS. In Proceedings of the 10th Airborne Earth Science Workshop, JPL Publication 02-1.

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