

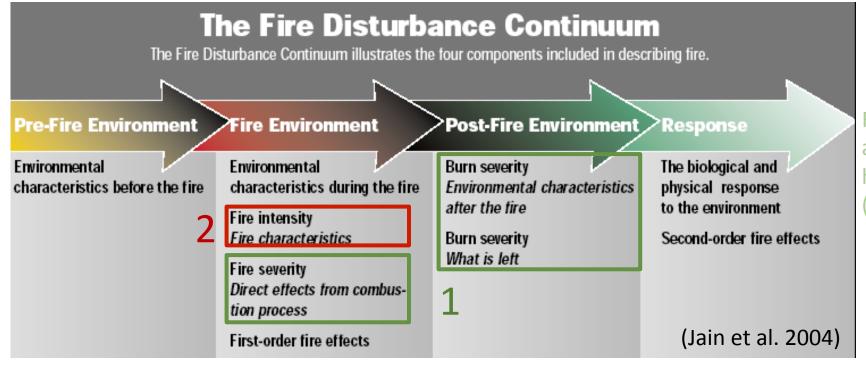
Sander Veraverbeke S.J. Hook, E.N. Stavros UCI JPL MASTER over King fire, 9.19.2014 Left: NIR-R-G, Right: MIR-NIR-R



### The fire continuum

'Wildfire, Fuel and Recovery' is one of HyspIRI's science application areas

Fuel type
Fuel composition
Fuel conditions
Drought stress



Recovery of vegetation and hydrological cycle (energy fluxes)

Landsat and MODIS have fueled many fire science studies

#### **IMPORTANT QUESTIONS**

What will HyspIRI provide that current sensors can not? Can we quantify potential improvements?

## Fire intensity and severity

Fire intensity: energy release from the combustion of organic matter (quantitative)



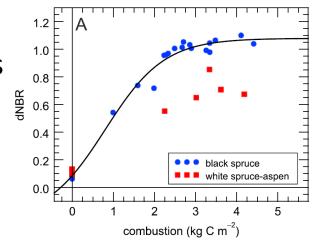
Fire severity: degree of environmental change caused by a fire Measured immediately after the fire (qualitative to semi-quantitative)

Burn severity: degree of environmental change caused by a fire *Measured longer-term after the fire (includes ecosystem responses)* (qualitative to semi-quantitative)

Severity may specifically refer to soil alteration, tree mortality, biomass consumption, etc.

Severity scales with emissions, important for post-fire management

Rogers, Veraverbeke et al. (2014)



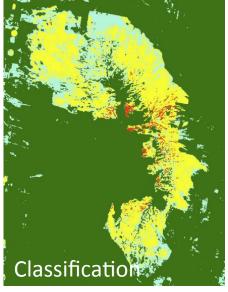


Landsat remote sensing of severity











# BAER BURNED AREA EMERGENCY RESPONSE

Normalized Burn Ratio:

NBR=NIR-SWIR/NIR+SWI

*dNBR*=*NBR*↓*pre* − *NBR*↓*post* 

dNBR is the operational method to provide information to BAER teams Often seen as the standard method to assess severity

#### But dNBR has several weaknesses:

- Relationships are specific per fuel type
- Saturation for high severity
- Has no biophysical meaning without calibration with field data
- Sensitive to soil brightness variations
- ...

## AVIRIS and MASTER capabilities to assess fire severity



What improvement is possible with imaging spectroscopy compared to traditional broadband VSWIR data?

Remote Sensing of Environment 154 (2014) 153-163

ELSEVIED

Contents lists available at ScienceDirect

#### Remote Sensing of Environment

journal homepage: www.elsevier.com/locate/rse



Assessing fire severity using imaging spectroscopy data from the Airborne Visible/Infrared Imaging Spectrometer (AVIRIS) and comparison with multispectral capabilities



Sander Veraverbeke <sup>a,b,\*</sup>, E. Natasha Stavros <sup>b</sup>, Simon J. Hook <sup>b</sup>



What improvement is possible from the synergy between VSWIR and MTIR data?

Remote Sensing of Environment 124 (2012) 771-779



Contents lists available at SciVerse ScienceDirect

#### Remote Sensing of Environment

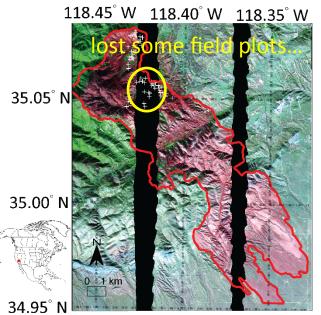
journal homepage: www.elsevier.com/locate/rse



Synergy of VSWIR (0.4–2.5  $\mu m$ ) and MTIR (3.5–12.5  $\mu m$ ) data for post-fire assessments

S. Veraverbeke <sup>a,\*</sup>, S.J. Hook <sup>a</sup>, S. Harris <sup>a,b</sup>





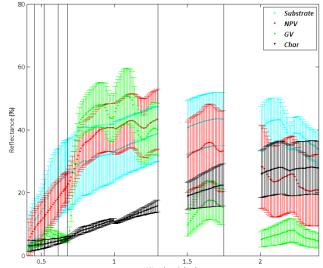






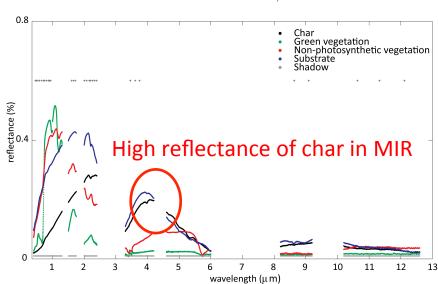
N 0 2 km

Fire severity field data in mixed shrub-conifer ecosystem



#### Spectra of:

- Charcoal
- Green veg
- NPV
- Substrates



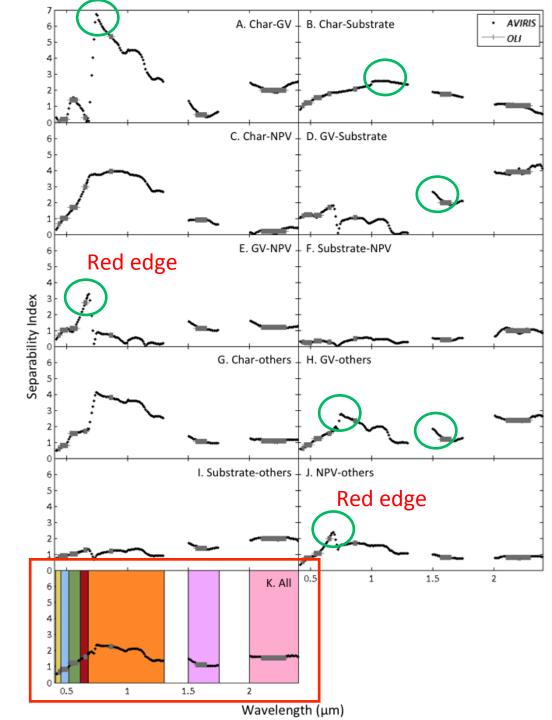


Separability index: assess discriminatory power of bands between different ground cover classes

$$SI_{i} = \frac{2}{m \times (m-1)} \sum_{z=1}^{m-1} \sum_{j=z+1}^{m} \frac{\left| \mu_{i,z} - \mu_{i,j} \right|}{\sigma_{i,z} + \sigma_{i,j}}$$

Some minor improvements in narrowband separability (e.g. red edge)

Does not highlight any narrow spectral features that are averaged out by broadbands





What improvement is possible with imaging spectroscopy compared to traditional broadband VSWIR data?



What improvement is possible from the synergy between VSWIR and MTIR data?

Approach: Multiple Endmember Spectral Mixture Analysis (MESMA)

Charcoal fraction is indicator of severity

A) Unmixing of simulated mixed spectra, performance assessment based on input fractions

#### 3 scenarios:

- 1) 'AVIRIS all bands'
- 2) OLI bands
- 3) 'AVIRIS multispectral' (best band per broadband region)

	Char			Green vegetation				NPV				
	a	b	$\mathbb{R}^2$	RMSE	a	b	$\mathbb{R}^2$	RMSE	a	b	R <sup>2</sup>	RMSE
AVIRIS (all)	0.97	0.02	0.69	0.12	1.01	0	0.95	0.05	0.97	0.01	0.84	0.09
OLI	0.77	0.05	0.46	0.16	1.00	0	0.88	0.08	0.79	0.07	0.40	0.15
AVIRIS (multispectral)	0.68	0.07	0.38	0.18	1.00	0	0.85	0.09	0.68	0.07	0.37	0.17

'AVIRIS – all bands' consistently outperformed multispectral scenarios

'AVIRIS – multispectral' and OLI performed similar

#### 3 scenarios:

- 1) VSWIR only
- 2) MTIR only
- 3) VSWIR-MTIR

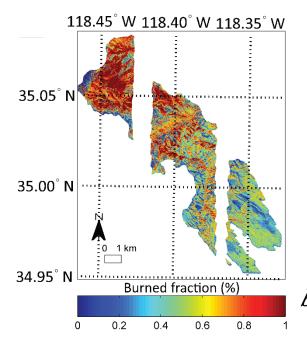
	Char				GV				NPV				
Spectral region	a	b	$R_{\rm adj}^2$	RMSE	a	b	$R_{\rm adj}^2$	RMSE	a	b	$R_{\rm adj}^2$	RMSE	
VSWIR MTIR VSWIR-MTIR	0.00 0.03 0.00	0.98 0.86 1.01	0.88 0.31 0.93	0.08 0.20 0.06	0.01 0.01 0.01	0.99 0.92 0.98	0.91 0.22 0.92	0.08 0.23 0.07	0.01 0.07 0.01	0.94 0.80 0.97	0.83 0.34 0.90	0.10 0.20 0.08	

Synergy between VSWIR and MTIR data performed better than VSWIR or MTIR alone

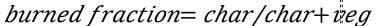


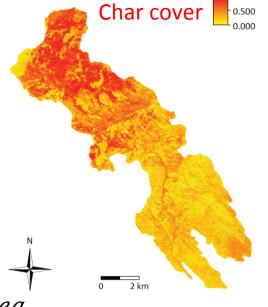


#### B) Unmixing of image data, performance assessment based on field data

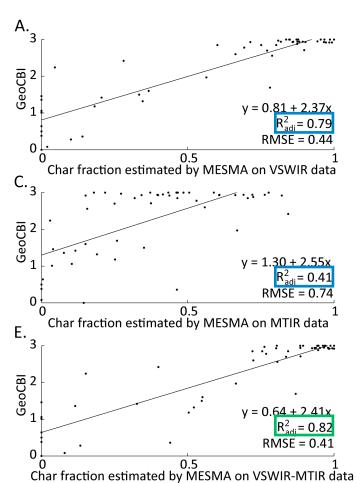


Field measure	GeoCBI					
Scenario	a	b	$\mathbb{R}^2$			
AVIRIS (all)	3.05	-0.05	0.86			
OLI	4.74	-2.07	0.65			
AVIRIS (multispectral)	3.59	-0.93	0.65			





Synergy between VSWIR and MTIR data performed better than VSWIR or MTIR alone



'AVIRIS – all bands' consistently outperformed multispectral scenarios

'AVIRIS – multispectral' and OLI performed similar

### Conclusions on fire severity

Significant improvement are possible in fire severity mapping by using imaging spectroscopy, and by including MTIR bands

Improved sensitivity to burned surfaces from IS did not come from narrow spectral features that broadbands average out

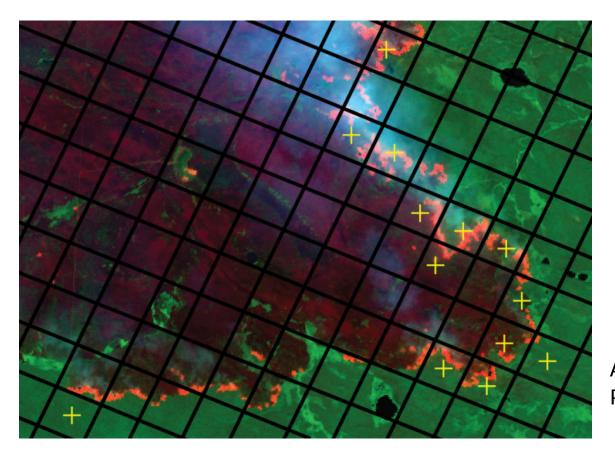
- Improved sensitivity to burned surfaces from IS was the result of better constraints on the data due to higher dimensionality
- Improved sensitivity to burned surfaces from MTIR bands remained relatively small compared to VSWIR alone Important questions when developing L3 burned area and severity products: what are the trade-offs?
- So far most post-fire studies, including this one, have remained limited to post-fire data only...

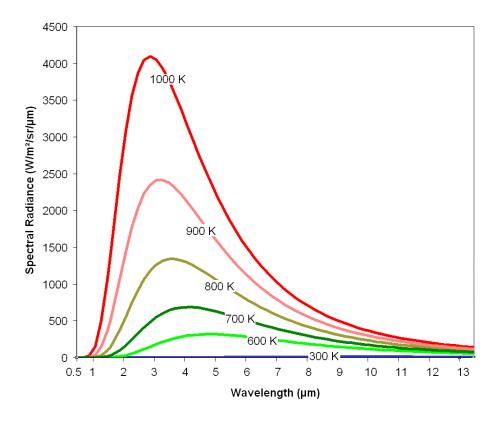
  But, the recent 2013 Rim and 2014 King fires enable unseen opportunities by providing before and after data With HyspIRI this data will routinely be available
- Airborne LiDAR data acquired over Rim fire also allows exploration of synergies with structural data as GEDI precursor

(contributions of Stavros and Tane in this workshop)

## Fire intensity using the 4 µm band

HyspIRI will provide unique 60 m observations in the MTIR range The 4  $\mu m$  band is optimal for active fire detection and characterization





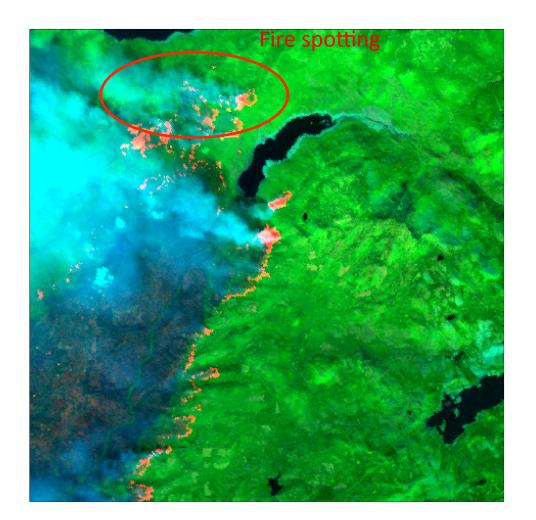
Active fire fronts cover only a fraction of a 1 km MODIS pixel Partially burned pixels may remain under the detection threshold

HyspIRI will be able to detect smaller fires

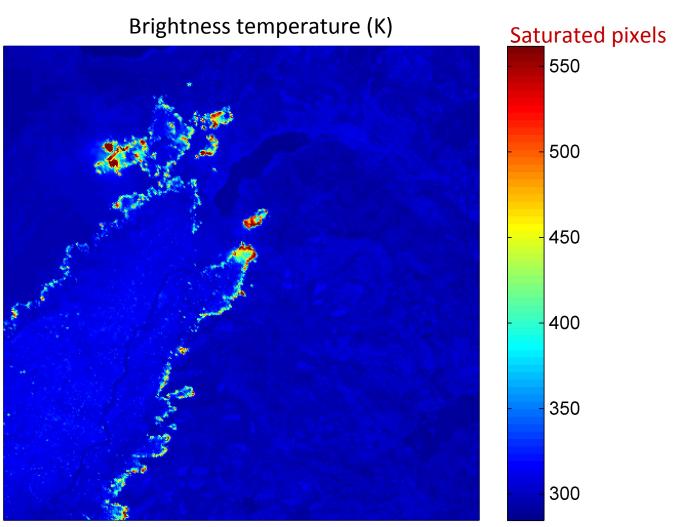
ASTER image overlaid with MODIS grid Yellow crosses show detected active fire pixels

## Fire intensity: 2014 King fire

MASTER's 4  $\mu m$  band saturates around 600 K HyspIRI's 4  $\mu m$  band will only saturate around 1200 K

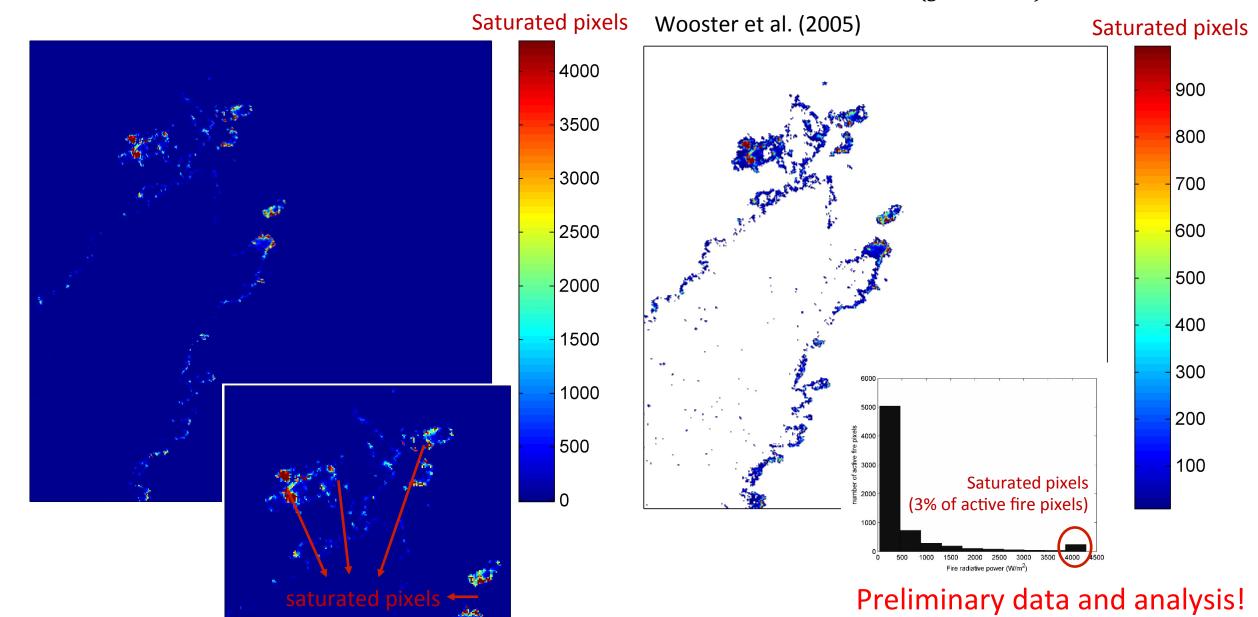






## Fire intensity: 2014 King fire

Fire radiative power  $(Wm\uparrow-2)=a\times (T\downarrow b, fire\uparrow 8-T\downarrow b, background Redus emissions (g C m\uparrow-2)=a\times FRP$ 



## Conclusions on fire intensity

HyspIRI's  $4~\mu m$  band will allow fire intensity characterization with far more spatial detail

This will allow to further disentangle relationships between topography, fuels and weather in fire behavior studies

It allows direct quantitative measurements of instantaneous emissions