Assessment of ecosystem diversity and urban boundaries:

Preliminary findings, exploring surface reflectance and emissivity

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Background

<u>Problem</u>: With the increase in the population density, the effect of ever expanding conversion of land from rural to urban has become a problem of critical importance. With increasing urbanization, land cover type and land surface temperature (LST) in urban and rural areas display significant differences, such as higher LST and lower moisture content.

<u>Hypotheses</u>: The distribution, and spectral and spatial characteristics of optical and thermal data co-vary, in significantly different way in natural and anthropogenic environments; The combination of high spectral resolution optical and thermal infrared imagery will provide a powerful capability for more precise land cover type discrimination and ecosystem monitoring than possible using current satellite systems.

Science Questions

- How do natural environmental characteristics change under the effects of urbanization, affecting vegetation composition and function, and ecosystem health?
- How do natural vs. anthropogenic ecosystem compositions compare (similar environmental conditions) with regard to land cover types, diversity, function, and spectral properties?
- How do natural ecosystems respond to impinging environmental changes, particularly to urban growth and land cover change and the associated impacts of urbanization?

Goals

- 1. Using together VSWIR and TIR measurements, to assess the differences in natural and anthropogenic ecosystem composition and their vegetation bio-physical parameters.
- 2. Provide data that can elucidate how urbanization impacts the environment. Seeking common spectral trends associated with vegetation function, induced by natural and anthropogenic factors such as effects of urbanization and UHI, we would contribute toward improving the current capabilities for vegetation assessments.
- 3. Generate HyspIRI-like datasets and tools for work with the data.

Study Sites

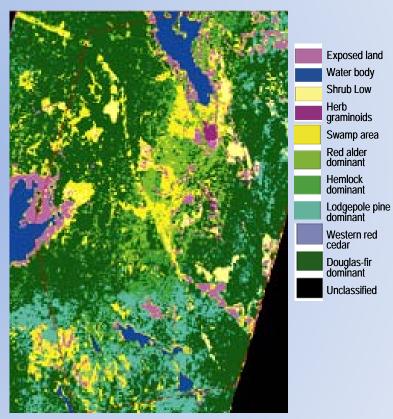
The study includes two independent locations with different regional climate and ecosystem types. Data aggregated to higher and lower spatial scales will be compared for the same locations.

Vancouver Island, Canada/Hoquiam, WA:

includes portions of unique natural ecosystems such as the Great Victoria Watershed (GVWD) test site on Vancouver Island, BC and rural, suburban and urban environment associated with the city of Victoria, BC.

Jasper Ridge Biological Preserve (JRBP), CA:

provides Mediterranean-type climate, with five major vegetation types: evergreen forest, deciduous forest, chaparral shrublands, herbaceous perennial wetlands, and annual grasslands



Hyperion land cover classification of the Greater Victoria Watershed (GVWD) test site on Vancouver Island (September 10 2001, Goodenough et al. 2003).



Area 2: Jasper Ridge Biological Preserve (JRBP), CA



Data:MASTER & AVIRIS Hyperion

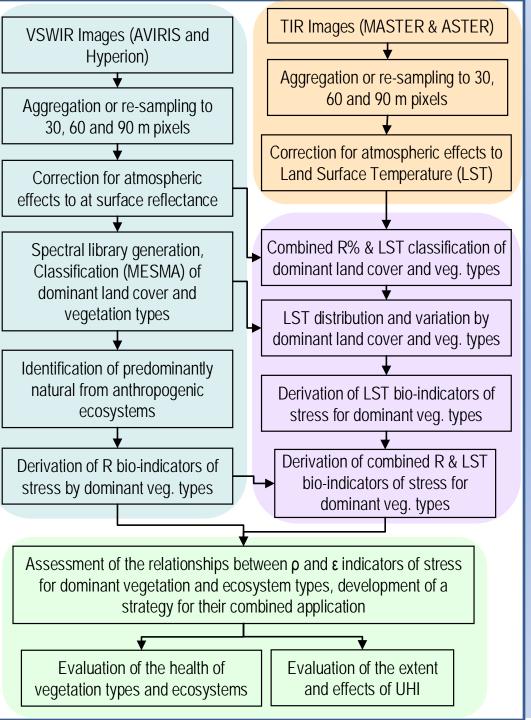
Calibration of TIR and VISWIR data

Lake Tahoe is a highaltitude, large-sized lake on the California-Nevada border near Reno, Nevada.

The <u>Ivanpah Playa</u> test site is approximately 3 km by 7 km in size with excellent spatial uniformity, hence its use for the vicarious calibration of reflective bands.



Sample MASTER imagery from the Ivanpah test site, reflectance (left), vegetation in green; thermal (right) side.



Capability of HyspIRI data for assessments of vegetation type and function

The analysis will be conducted using the native data resolution (~ 30 m), and aggregated to 60 and 90 m images, comparing the results.

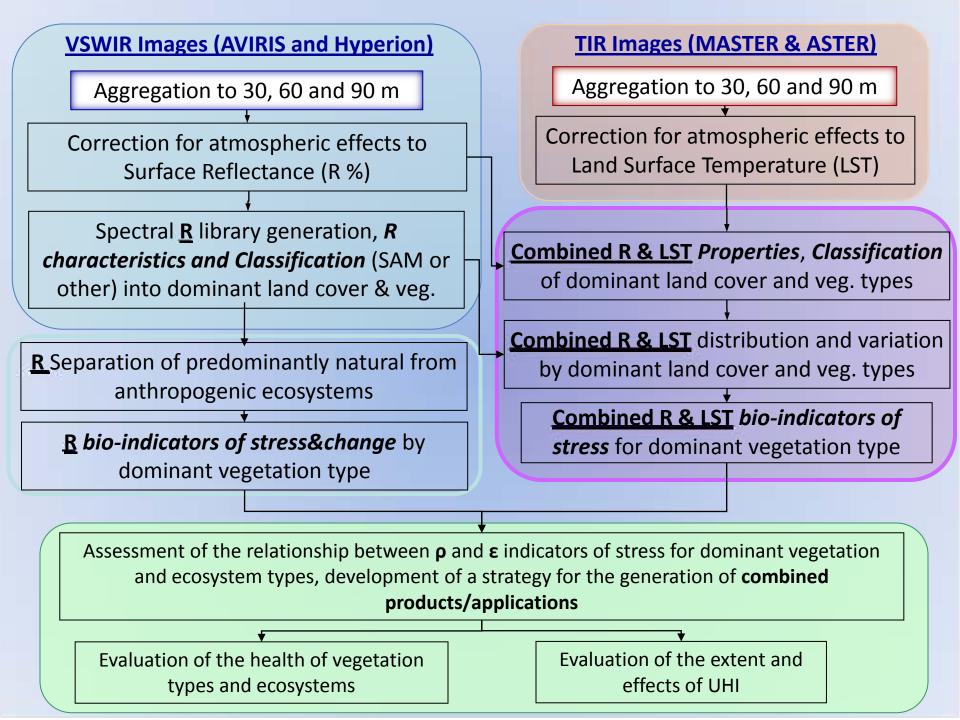
Data

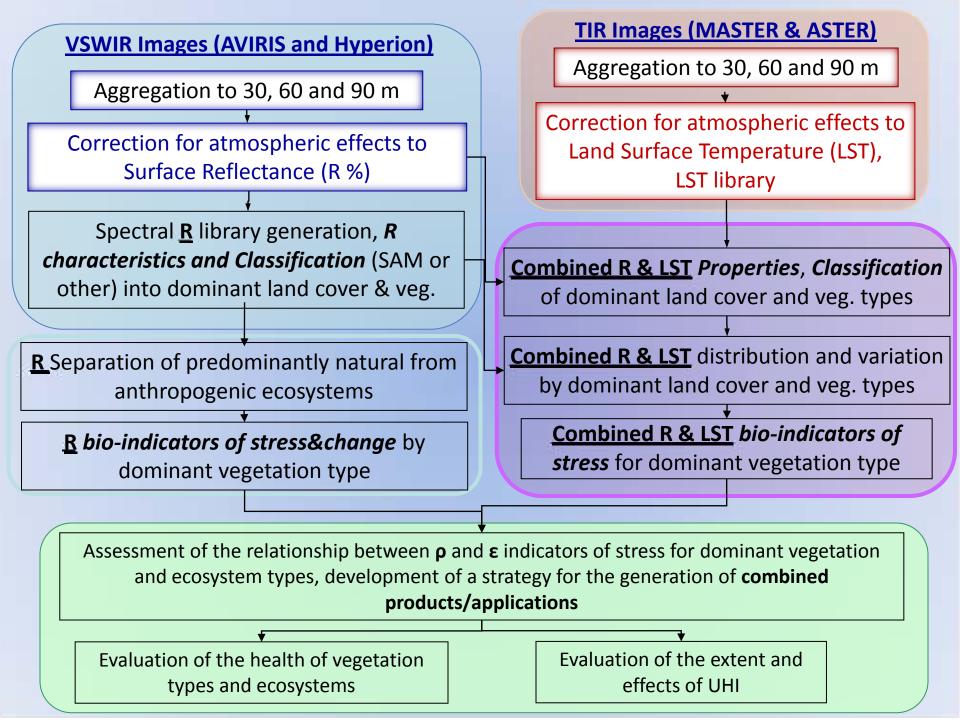
Parameters	HyspIRI VSWIR	AVIRIS	EO-1 Hyperion
Spectral range	0.38 - 2.5 µm	0.4 - 2.5 μm	0.4 - 2.5 μm
Band width	10 nm	10 nm	10 nm
number of bands	~ 220	224	220 (196 calibrated)
Spectral Coverage	contiguous	contiguous	contiguous
Spatial resolution	60 m	20 m	30 m
Swath width	145 km	11 km	7 km

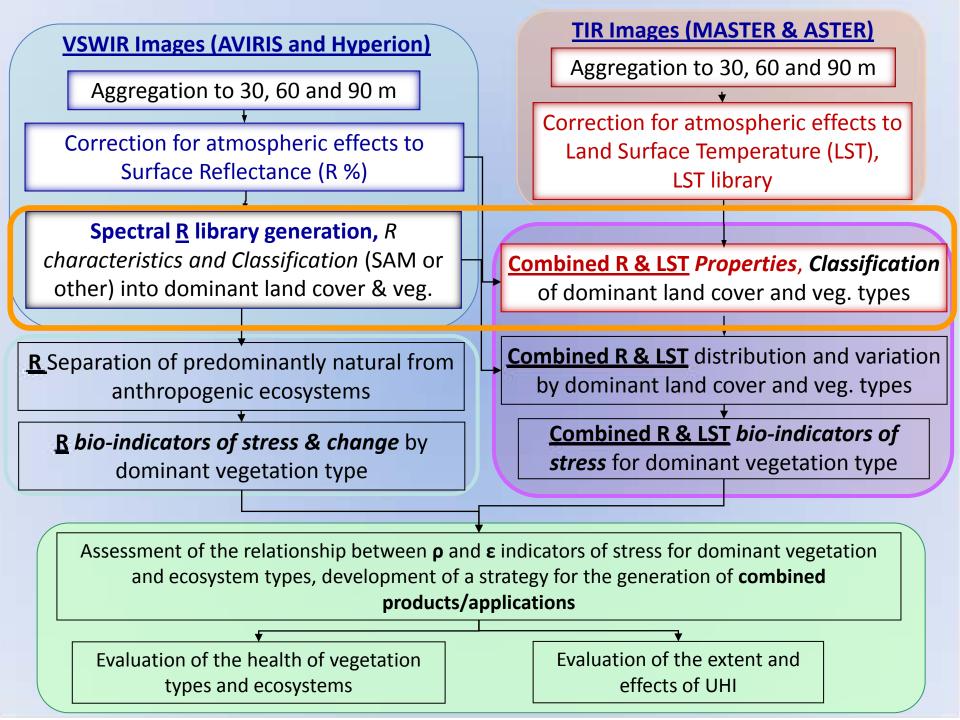
Thermal infrared emission instruments and data characteristics (summary/overview)

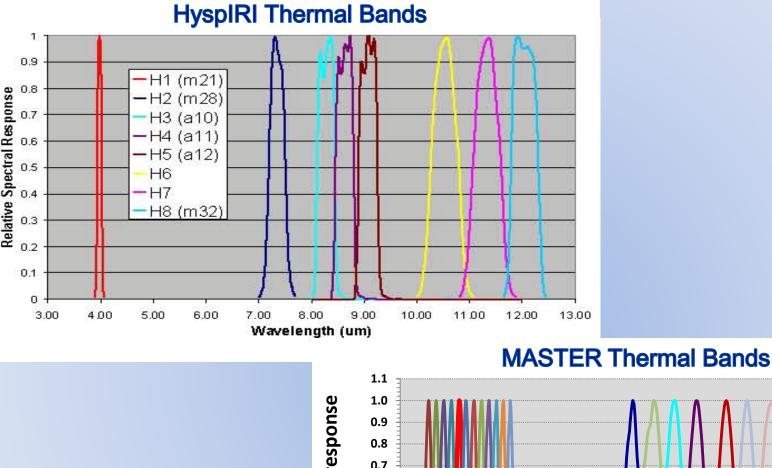
HyspiRi TIR	Number of bands	Spectral bands (discrete)	Spatial resolution
TIR channel centers	8	3.98, 7.35, 8.28, 8.63, 9.07, 10.53, 11.33, 12.05 µm	60 m
Swath width	600 km		

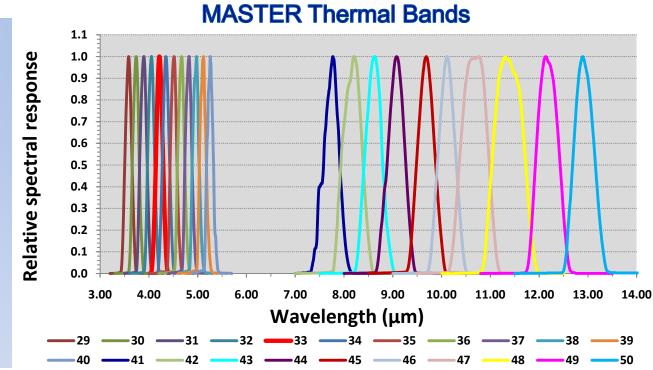
MASTER	Number of bands	Spectral bands (discrete)	Spatial resolution
Spectral range 0.4 - 13 µm	50		
0.45-2.39 μm	25	0.05 μm a part	5-25 m (B200)
3.15-5.27 µm	15	0.15 μm a part	10-30 m (DC-8)
7.75-12.87 µm	10	0.40-0.80 µm a part	50 m (ER-2)
Total field of view	85.92°		88 38



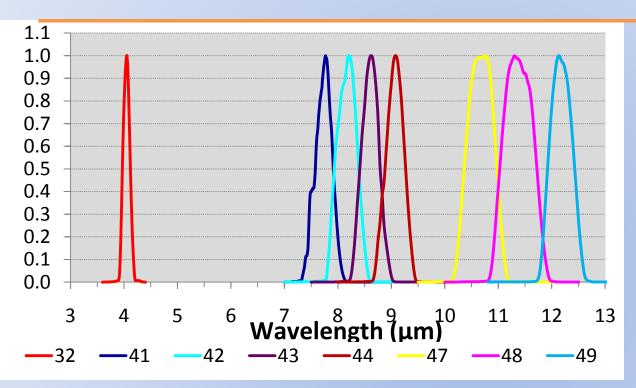


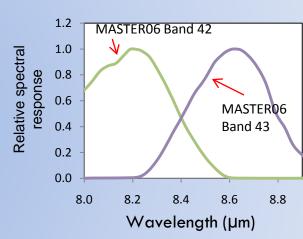


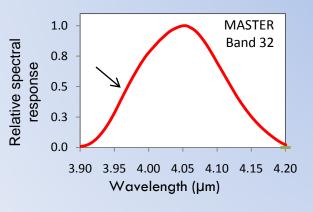


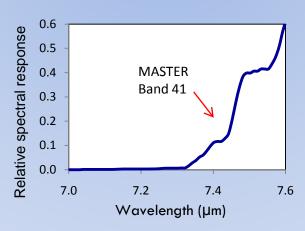


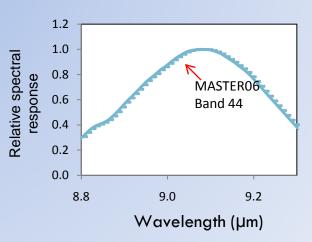
MASTER Thermal Bands Relative to HyspIRI











Spatial Aggregation of HyspIRI-like Thermal Bands



Original image pixel size varies (49.1 m, bands 32, 41, 42, 43, 44, 47, 48 and 49)

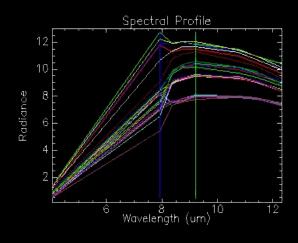


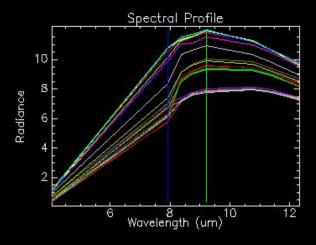
Aggregation to 60 m (MASTER, band 44)

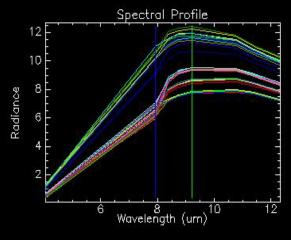


Aggregation to 90 m (MASTER, band 44)

Atmospheric Correction of Thermal Data



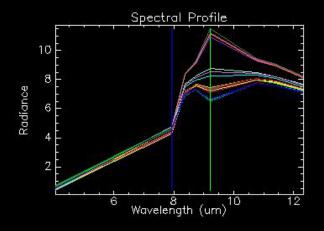


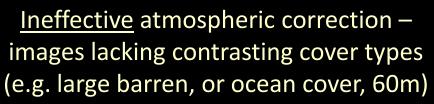


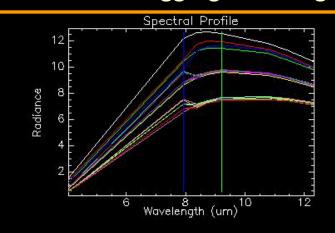
Max hit-normalized regression, 49.1 m resolution

Max hit-top of bins, 60 m resolution

All-top of bins, 90 m aggregated image



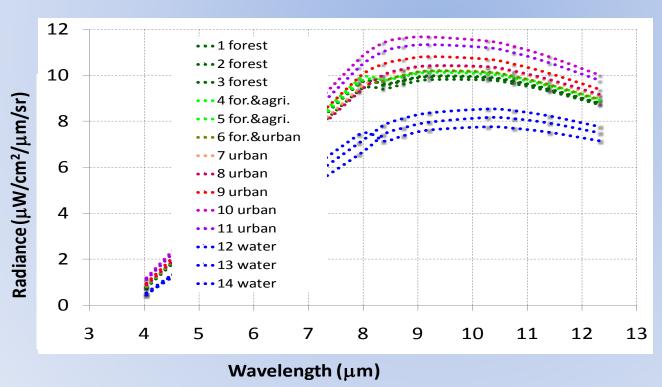




Effective atmospheric correction (60 m resolution, max hitnormalized regression)

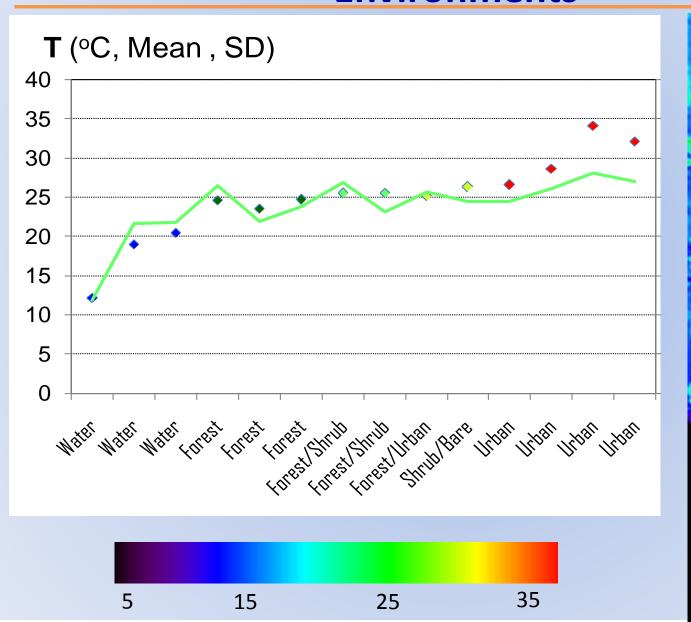
Thermal Radiance of Rural Urban and Environments

- Assess the Variations in Thermal Radiance for ecosystems ranging from natural to urban
- Convert Radiance to Emissivity and Temperature to quantify the temperature and emissivity differences in ecosystems

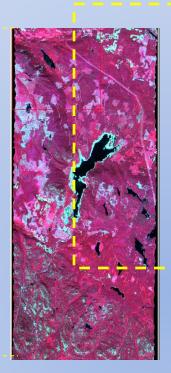


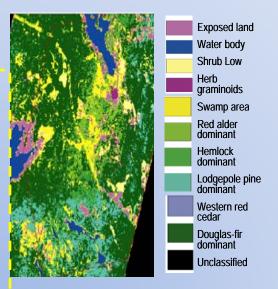


Temperature Gradient of Rural Urban and Environments



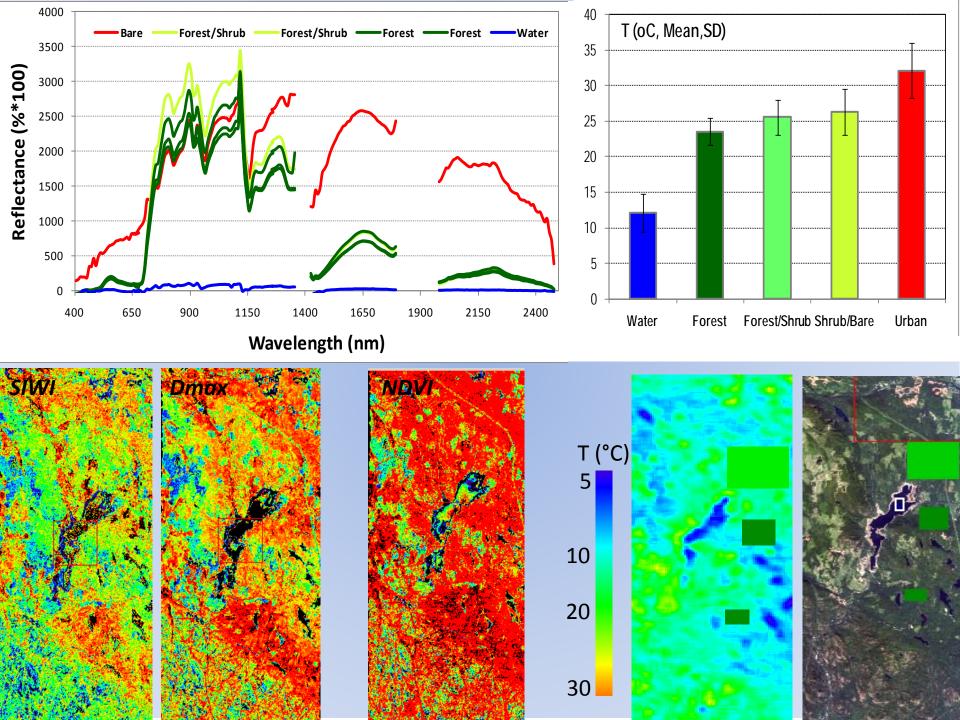


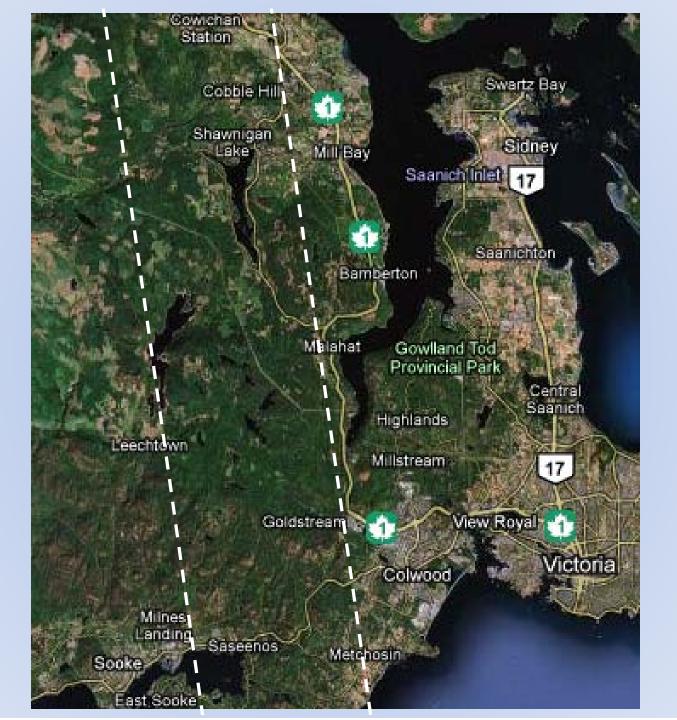


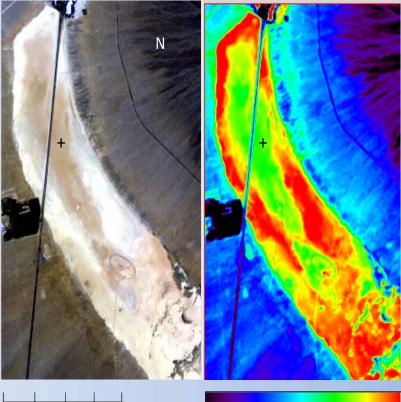










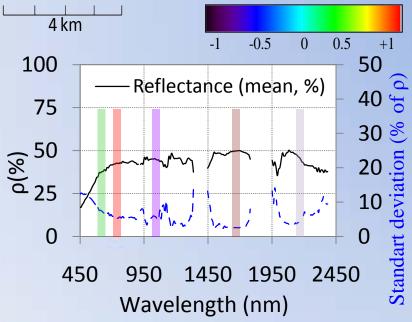


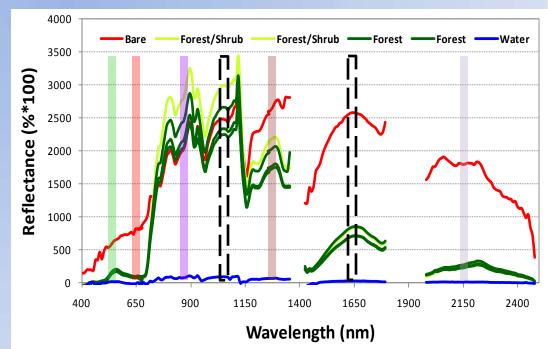
Site Spatial Uniformity / Spectral Variability

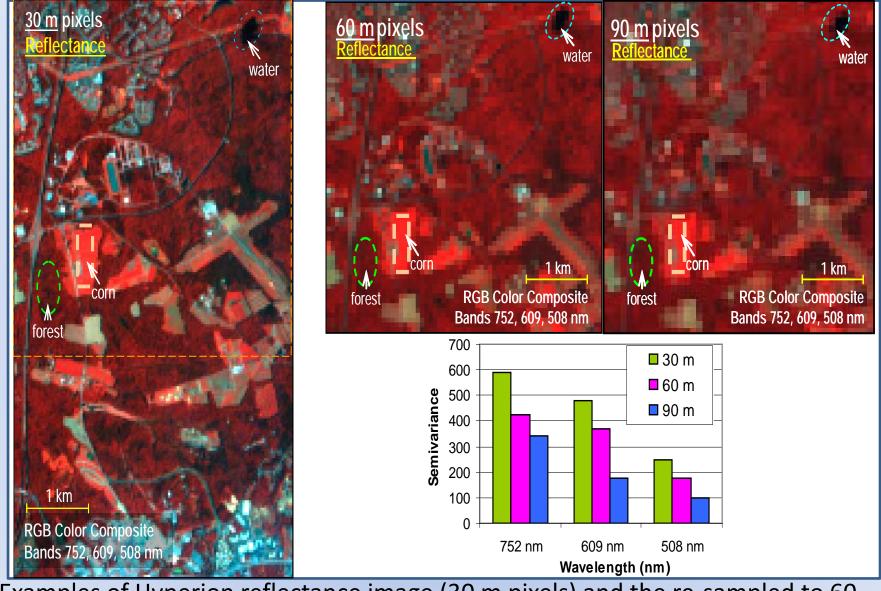
Example from Ivanpah Playa, Hyperion

- A. Natural color composite (RGB: 651, 549, 447)
- B. B. Getis Gi* statistics (band 549) calculated for 197 Jul day image

The Gi statistics calculated for G, R, NIR, SWIR bands produced similar results.







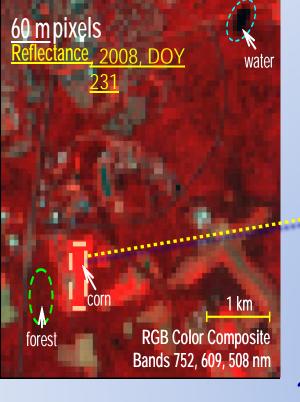
Examples of Hyperion reflectance image (30 m pixels) and the re-sampled to 60 and 90 m subset, demonstrate the spatial differences between HyspIRI-like data (60 m) and 30 and 90 m data.

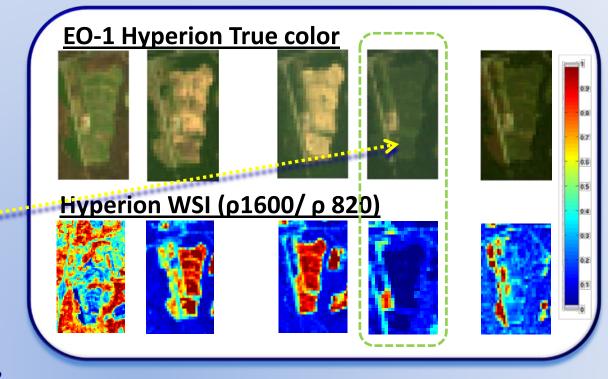
30 m pixels

60 m pixels

Pixel	xel Vegetation Indices:						Albedo		
size	V1	PRI	REIP	Dmax	NDWI	NDVI	water	corn	forest
30 m	1.81	-0.14	721	0.749	0.14	0.81	0.03	0.20	0.14
60 m	1.88	-0.15	721	0.748	0.15	0.82	0.04	0.20	0.13

While there were significant differences in the spatial variability between the original 30 m and the aggregated to 60 and 90 m data, the spectral properties of the major land cover types did not significantly differ.





2008, 108 DOY **Spring**

195 231 **Summer**

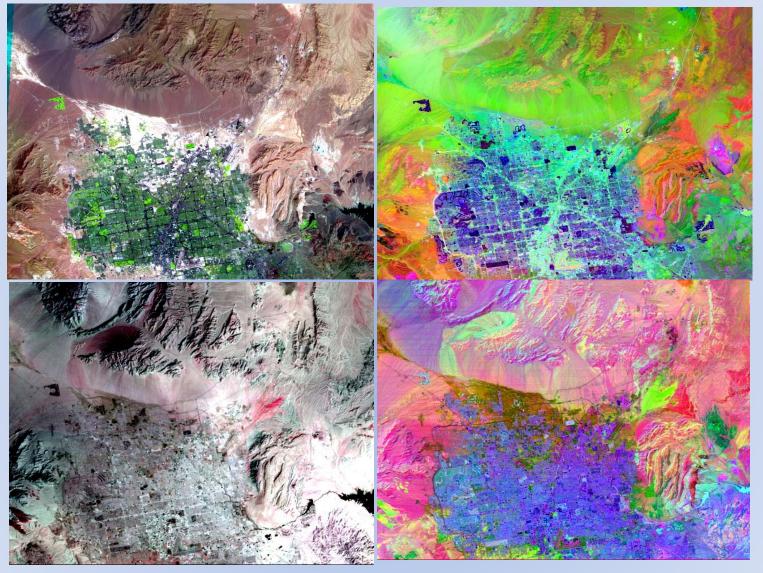
Fall

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Cover Type	Hyperion, 2008	V1	PRI	REIP	Dmax	WBI	Albedo
Corn	13-Jun	1.03	-0.04	712	0.36	0.96	0.461
	18-Aug	1.81	-0.06	722	0.75	1.09	0.197
	3-Oct	1.15	0.04	721	0.51	0.98	0.155
Forest	13-Jun	1.12	-0.06	712	0.89	1.00	0.257
	18-Aug	1.56	-0.03	722	0.51	1.01	0.140
	3-Oct	1.61	-0.10	712	0.42	0.94	0.127
Water	13-Jun	0.15	0.01	712	0.16	1.23	0.058
	18-Aug	0.52	0.02	712	0.10	1.46	0.031
	3-Oct	0.62	-0.07	712	0.08	0.93	0.036

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Bio-physical parameter	Spectral bio-indicators (examples)	References
TIR indicators		
Water stress (100% vegetative cover)	CWSI = $(\Delta T canopy - \Delta T nws)/(\Delta T max - \Delta T nws)$, where Tmax for non-transpiring and Tnws is "non-water-stressed" baseline temperature. $0 = no stress$, $1 = maximal stress$, $\Delta = (T surface - T air)$	Idso et al. 1981, Jackson, 1981
Irrigated vs. non- irrigated (100% veg.)	IG = (Tmax – Tcanopy)/(Tcanopy – Tnws),	Jones 1999
Increase in sensible heat due to UHI	Rn = G + LE + H, $Rn + A = G + LE + H$; where Rn is net radiation, G ground heat, LE latent heat, H sensible heat, A anthropogenic heat	Kato and Yamaguchi, 2005
Combined VSWIR	& TIR indicators	
Relative water status (Veg. cover < 100%)	WDI trapezoid, defined by the fractional vegetation cover and Δ = (surface - air temperature); 0.0 well-watered (latent heat flux is limited only by atmospheric demand), 1.0 no available water	Moran et al. 1994
Water supply	WSVI=NDVI/Ts, where Ts is the brightness temperature at ~11 µm, the smaller this index is, the more severe the drought is	Hatfield et al. 2008
Temperature— vegetation dryness index	TVDI = $(Ts - Ts_min)/(a + bNDVI - Ts_min)$, where Ts_min is the minimum temperature in the triangle; and a and b are the interception and slope of the dry edge, respectively	Wand et al. 2004
After WSVI and TDVI	Like WSVI and TVDI, but using VSWIR parameters sensitive to water stress, such as NDWI or SIWI.	New, to test



ASTER imagery, Las Vegas area: VSWIR aggregated to 90 m (upper left, vegetation in green) and TIR bands (lower left), principle components of reflective bands (upper right) and all VSWIR and TIR (lower right).

Deliverables and Milestones

- Best estimates of atmospheric properties needed to atmospherically correct the two HyspIRI-like data sets;
- Algorithm to spatially convolve the airborne or spaceborne imager data sets to 30, 60, and 90 m spatial resolution;
- Development of HyspIRI-like datasets;
- Development of HyspIRI-like spectral bio-indicators of ecosystem health;
- Classifications of HyspIRI-like data for a broad range of natural and anthropogenic ecosystems and vegetation functional types;
- Presentations at HyspIRI meetings, and final report.



Expected Results



- Through the combined use of reflective and emissive data, the proposed work will evaluate the suitability of HyspIRI for delineating ecosystem functional types and tracking the conversion of land from rural to urban.
- The research will produce two HyspIRI-like data sets combining reflectance and thermal data, which will be used to assess the potential of HyspIRI data for delineating land cover and vegetation types, discriminating natural versus urban ecosystems, and assessing ecosystems diversity and health.
- In addition, the work will produce methods and tools for work with the HyspIRI data.

Preliminary Findings

- MASTER's spectral bands 32 and 41 may not provide ideal simulation for HyspIRI
- Atmospheric correction of thermal images at 60m worked as well and better, than at the original resolution
- Thermal atmospheric correction was effective when playa, vegetation and water regions are present and well-delimited
- The 60m resolution can estimate differences in emissivity for urban and rural regions, and demonstrated higher temperatures in urban areas

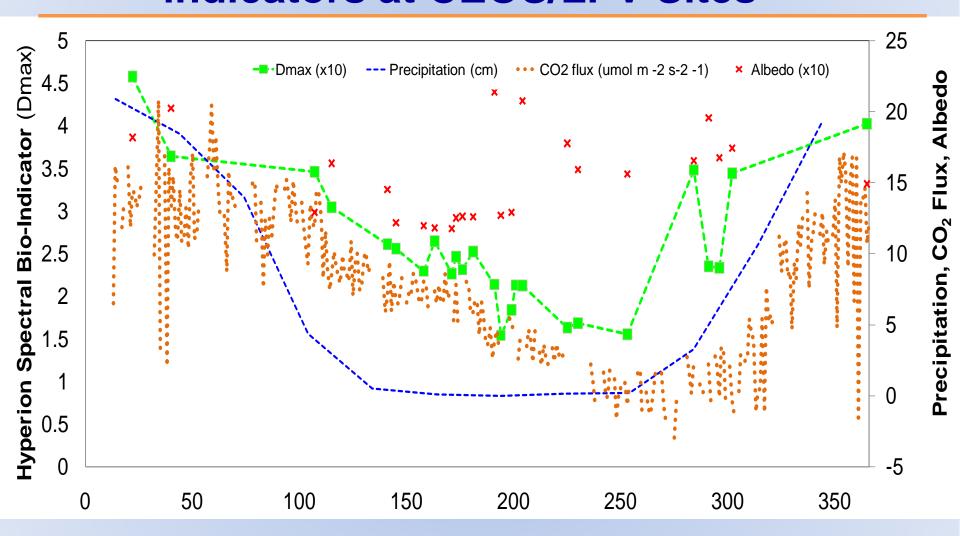
Project Collaborators

- Prof. David Goodenough (CFS Victoria, BC)
- Dr. Nicholas Coops (UBC, Vancouver, CA)
- Dr. Nona R. Chiariello (Stanford University, CA)
- Dr. Tony Trofymow (CFB Esquimalt, BC)
- Dr. Elizabeth Middleton (NASA/GSFC)

Acknowledgments

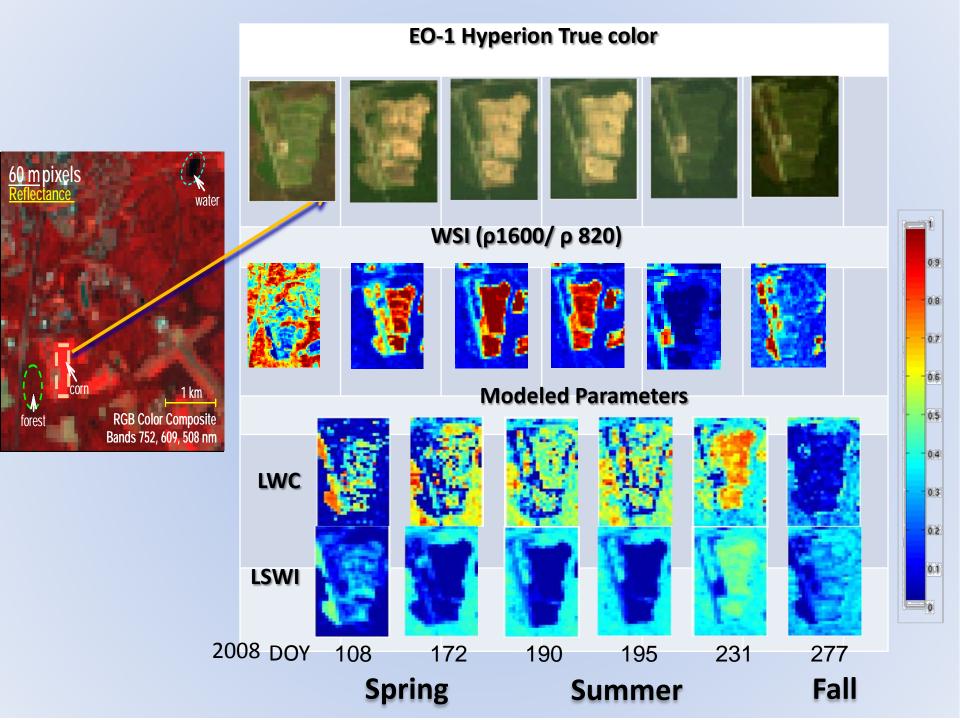
The project provides research experience for graduate and undergraduate students, which participate through established programs, assisting with data processing, analysis and preparation of reports.

Temporal profile of Hyperion's Spectral Bioindicators at CEOS/LPV Sites



The temporal profile in the derivative spectral bio-indicator Dmax captures the dynamics in vegetation phenology at Mongu, Zambia

Thank You





1.2

1.0

8.0

0.6

0.4

0.2

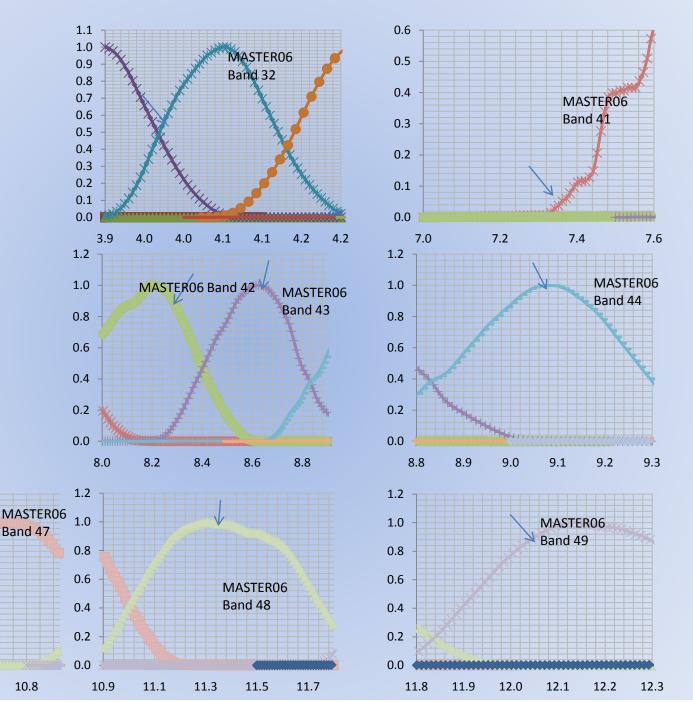
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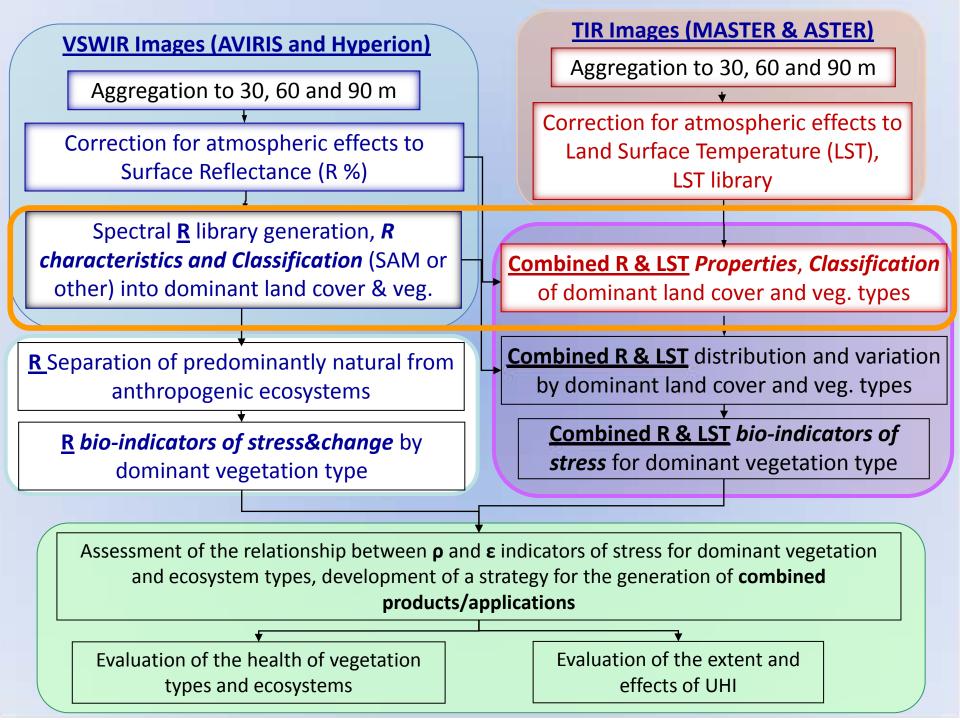
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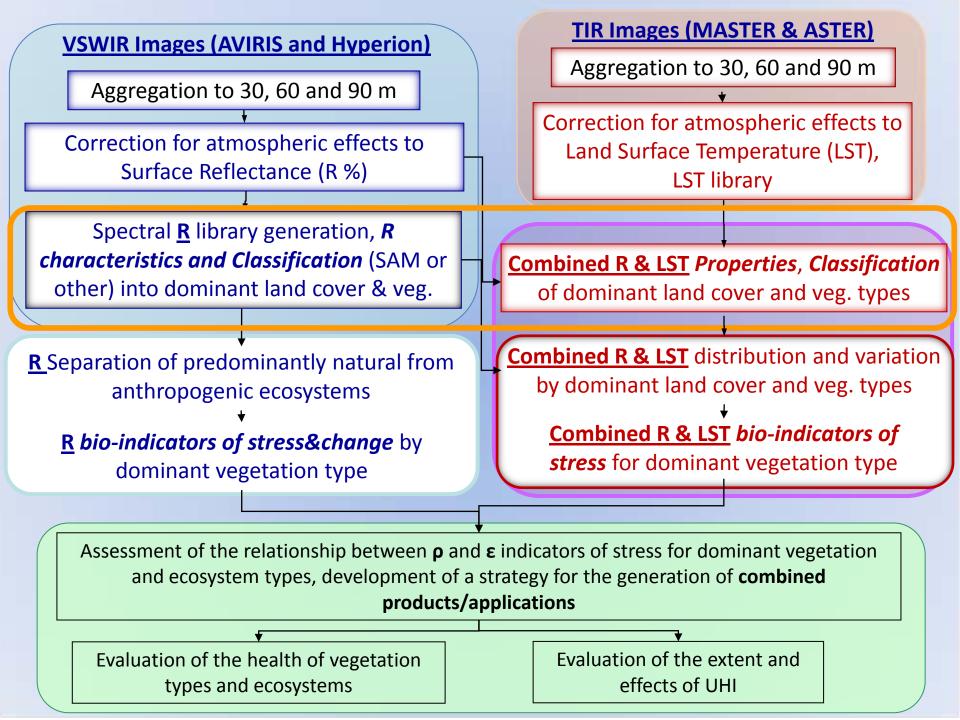
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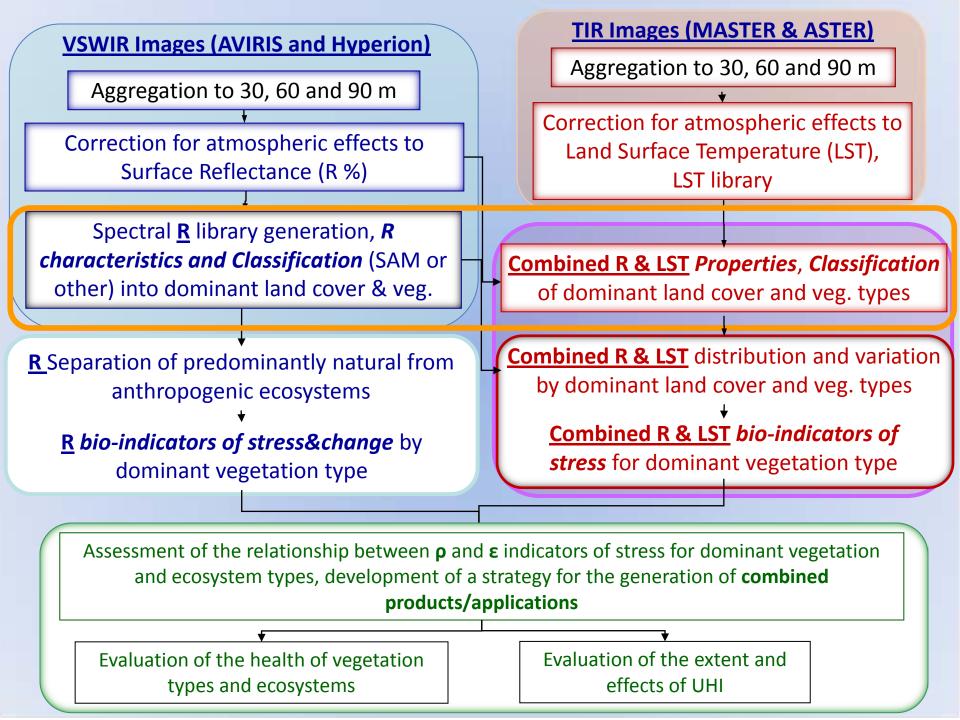
10.6

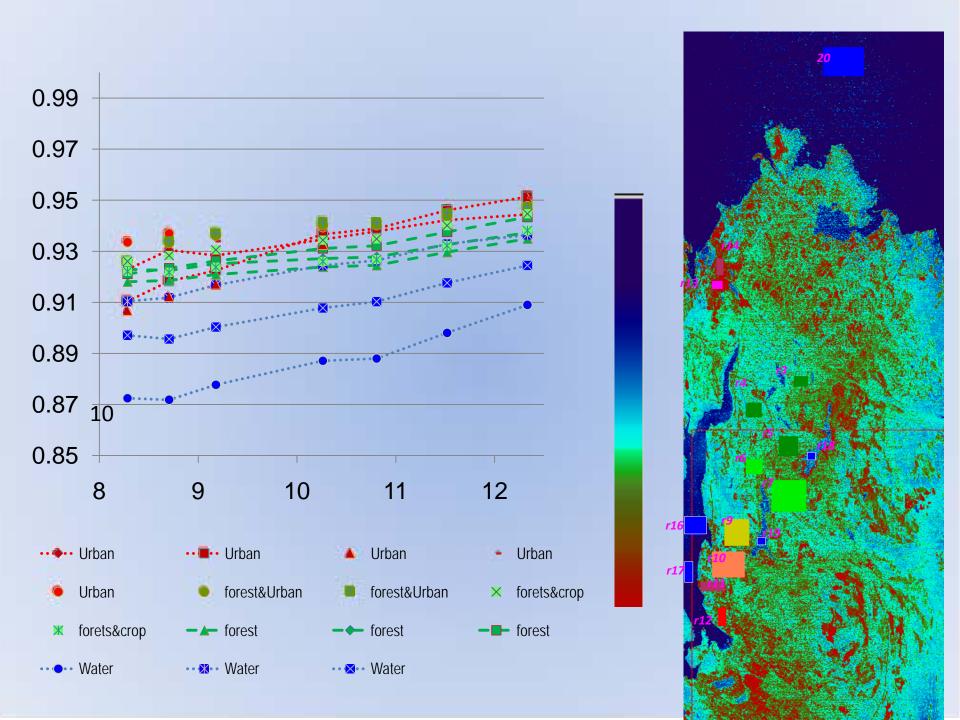
10.8



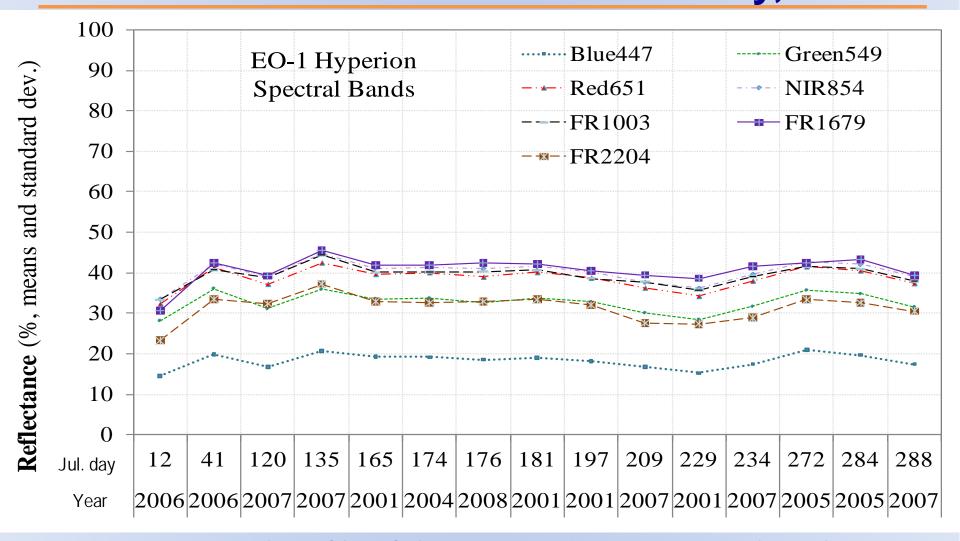








Temporal profile of Hyperion at the CEOS Calibration site at Railroad Valley, NV



Linear temporal profile of the spectra demonstrates the relative stability of the spectral characteristics, as indicated for selected bands

Bio-physical parameter	Spectral bio-indicators (examples)	References	
VSWIR indicators			
Chlorophyll in crops	$(R_{NIR}/R^{-1}_{720-730}) - 1$	Gitelson et al., 2005	
Canopy greenness	$EVI = G*rac{R_{NIR}-R_{red}}{R_{NIR}+C_{1}R_{red}-C_{2}R_{blue}+L}$	Huete et al. 2002	
Chlorophyll concentration estimation	$TCARI/OSAVI = \frac{3*[(R_{700} - R_{670}) - 0.2*(R_{700} - R_{550})*(R_{700}/R_{670})]}{(1+0.16)*(R_{800} - R_{700})/(R_{800} + R_{670} + 0.16)}$	Haboudane et al. 2002, Zarco-Tejada et al. 2004	
Efficiency of photosynthesis	PRI=(R570-R531)/(R570+R531)	Gamon et al., 1992, Suarez et al. 2008	
Canopy foliar water content	$NDWI = rac{R_{NIR} - R_{1240}}{R_{NIR} + R_{1240}}$	Gao, 1996	
Canopy foliar water content	$SIWI = rac{R_{NIR} - R_{1640}}{R_{NIR} + R_{1640}}$	Fensholt and Sandholt, 2003	
Vegetation stress	Average (R675 R705)	Vogelman, 1993	
Veg. structure and foliar biomass	NDVI = (R800-R670)/(R800+R670)	Deering, 1978	
Stress/Chlorophyll	D714/D705, where D is product of first derivative transformation of R	Entcheva, 2004	
Stress/ Chlorophyll	Dmax/D705, where Dmax is the maximum of D in the 670-730nm region	Entcheva, 2004	