



# Assessment of ecosystem diversity and urban boundaries:

Combining surface reflectance and emissivity

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## **Project Collaborators**

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- Professor Goodenough (CFS Victoria, BC)
- Dr. Tony Trofymow (CFB Esquimalt, BC)
- Dr. Elizabeth Middleton (NASA/GSFC)

The project will also provide research experience for graduate and undergraduate students. They will participate through established programs with GSFC, UMBC and FIU, Miami, providing support to process and organize data, and assisting with preparation of reports.

# Background

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

*Hypotheses:* 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 and anthropogenic ecosystem compositions compare with regard to land cover types, diversity, function, and spectral properties?

•How do environmental characteristics, associated with natural factors and effects of urban pressure and UHI, affect vegetation composition and function, and ecosystem health?

•How do species, functional type, and ecosystems biodiversity composition differ spectrally, in relation to gradients of anthropogenic and non-anthropogenic stressors?

•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.
- 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 Olympic National Park, WA and the Great Victoria Watershed (GVWD) test site on Vancouver Island, BC and rural, sub-urban 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 1: Vancouver Island Canada/Hoquiam, WA

> > Data: **MASTER & AVIRIS ASTER EO-1** Hyperion

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Vancouver Nanaimo Abb otsford Victoria Kalispel Spokane Seattle Coeur d'Alene 0.0 Tacoma • Washington Yakima Kennewick Longv • Portland

amloops

• Salem Corvallis

Oregon Eugene

Idaho

Boise

Calgar

Grants Pass Klamath Medford Ealls

Eureka Redding

> Reno Nevada •Chico 7 Oarson City Yuba City .

 Sacramento Santa Rosa . Stockton San Francisco San Jose

California

Fresno • Visalia

Santa Mana • Bakersfield Palmdale Lompoc Santa Clarita

Riverside • Hemet Los Angeles

· Cesanelda

Las Vegas



### **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.

### 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 k m	11 km	7 km
I hermal infrared emission instru	ments and data char	acteristics (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 k m		
ASTER	Number of bands	Spectral bands (discrete)	Spatial resolution
Spectral range 0.5 - 12 µm	14		vary
VNIR channel centers	4	0.54, 0.66, 0.81, 1.65 µm	15 m
SWIR channel centers	5	2.17, 2.21, 2.26, 2.33, 2.40 µm	30 m
TIR channel centers	5	8.30, 8.65, 9.10, 10.60, 11.30 μm	90 m
Swath width	60 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°		



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.







#### VSWIR Images (AVIRIS and Hyperion)

Aggregation to 30, 60 and 90 m

Correction for atmospheric effects to Surface Reflectance (R %)

Spectral <u>R</u> library generation, *R* characteristics and Classification (SAM or other) into dominant land cover & veg.

<u>R</u>Separation of predominantly natural from anthropogenic ecosystems

<u>R</u> bio-indicators of stress or change by dominant vegetation type

#### **TIR Images (MASTER & ASTER)**

Aggregation to 30, 60 and 90 m

Correction for atmospheric effects to Land Surface Temperature (LST), LST library

<u>Combined R & LST</u> *Properties, Classification* of dominant land cover and veg. types

**<u>Combined R & LST</u>** distribution and variation by Ecosystem Type (natural/anthropogenic)

<u>Combined R & LST</u> *bio-indicators of stress* for dominant vegetation type

Assessment of the relationship between **ρ** and **ε** indicators of stress for dominant vegetation and ecosystem types, development of a strategy for the generation of **combined products/applications** 

Evaluation of the health of vegetation types and ecosystems

Evaluation of the extent and effects of UHI





Nater

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.



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.



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).

Bio-physical parameter	Spectral bio-indicators (examples)	References
<b>VSWIR indicators</b>		
Chlorophyll in crops	(R <sub>NIR</sub> / R <sup>-1</sup> <sub>720-730</sub> ) - 1	Gitelson et al., 2005
Canopy greenness	$EVI = G * \frac{R_{NIR} - R_{red}}{R_{NIR} + C_1 R_{red} - C_2 R_{blue} + L}$	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

Bio-physical parameter	<b>Spectral bio-indicators</b> ( <i>examples</i> )	References		
TIR indicators				
Water stress (100% vegetative cover)	CWSI = $(\Delta T canopy - \Delta T nws)/(\Delta T max - \Delta T nws)$ , where T max for non-transpiring and T nws is "non-water-stressed" baseline temperature. 0 = no stress, 1 = maximal stress, $\Delta$ = (T surface - Tair)	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 $\Delta$ = (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 $\mu$ m, 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 <i>a</i> and <i>b</i> 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		

## **Deliverables and Milestones**

- Best estimates of atmospheric properties needed to atmospherically correct the two HyspIRI-like data sets;
- Radiometric calibration coefficients for the sensors used to create the HyspIRI-like data sets;
- Algorithm to spatially convolve the airborne or spaceborne imager data sets to 30, 60, and 90 m spatial resolution;
- Classifications of HyspIRI-like data for a broad range of natural and anthropogenic ecosystems and vegetation functional types;
- Development of HyspIRI-like spectral bio-indicators of ecosystem health; and
- Presentation materials for the HyspIRI workshop and final report delivered one month prior to the completion of the project.





- 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.



### Thank You