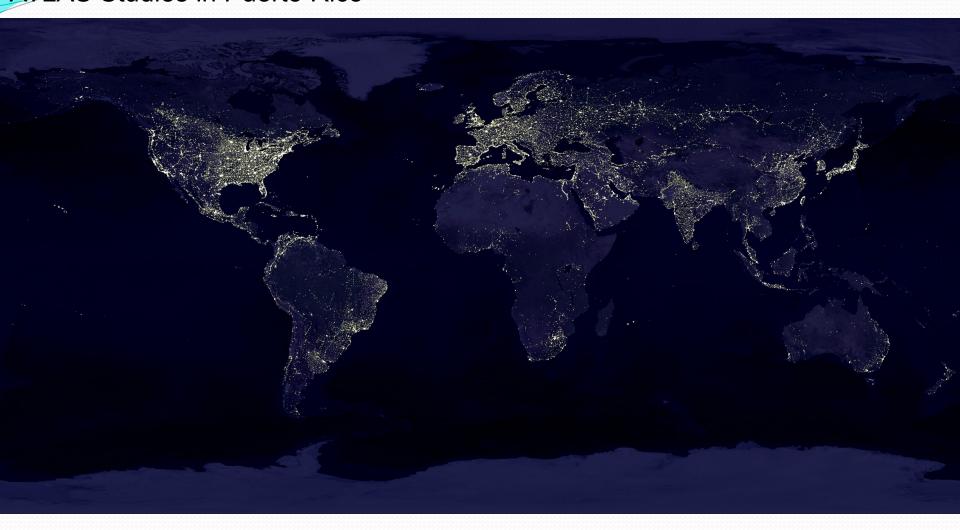
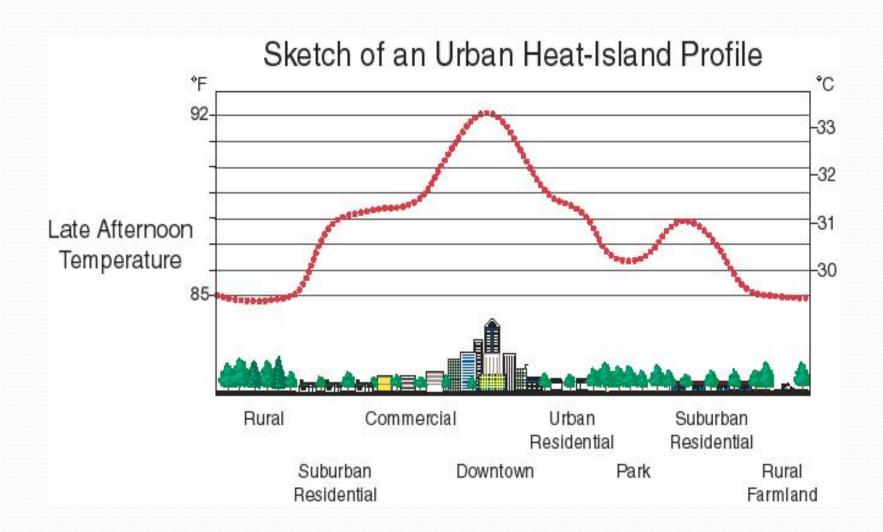
Potential for Improved Urban Land Surface Characterization and Land Surface Temperature Analysis Using HyspIRI Data: Previous ATLAS Studies in Puerto Rico



Jeffrey C. Luvall and Dale A. Quattrochi NASA Marshall Space Flight Center





TRMM Precipitation Radar Study of Urban-Induced Rainfall Featured in August 11th, 2003 Issue of TIME magazine

How Cities Make Their Own Weather

1 In big cities, heat-absorbing roofs,

pavement and

trap the sun's

rays and warm the air

auto exhaust

blacktop

HEN HOUSTON IS HIT BY A SUDden storm, the city may be partly to blame. Increasingly, urban centers don't merely endure bad weather; they help create it. Researchers believe the phenomenon may be more common now than ever before.

Scientists have known for 200 years that the temperature in a city can be higher than that in its environs-something they learned when an amateur weather watcher detected a 1.58°F temperature difference between London and its suburbs. Modern cities, with their cars and heat-trapping buildings, can create an even bigger temperature gap, sometimes as much as 10° F.

Islands of urban heat can do funny things with weather. Hot city air, like hot air anywhere else, rises-even more so because of the turbulence caused by tall buildings. When that air is damp enough and collides with colder layers above it, water can condense out as a sudden burst of rain, especially if there are few frontal systems to disrupt the layers, as in summer. In a spot storm above a city or just downwind of it, it's likely that nature alone isn't behind the downpour.

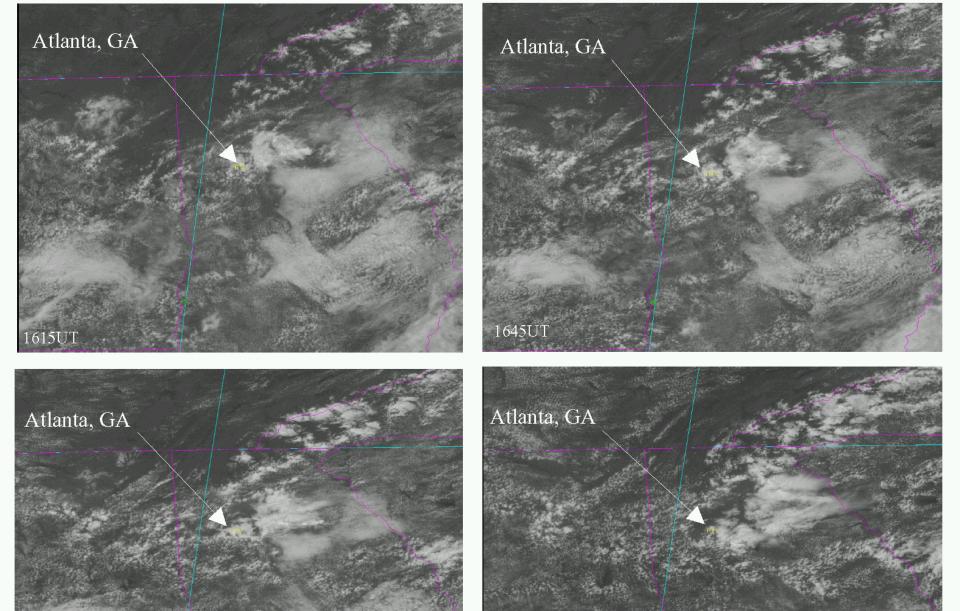
NASA and the University of Arkansas have been using satellite mapping and ground-based temperature readings termine how widespread this phenor is. This spring researchers got a st when they turned their attention to Houston. Because it's near a coast and sea breezes tend to cool and disperse hot air, Houston was thought to be comparatively safe from homemade rain. Now it appears that the opposite may be true. "The sea breeze may exacerbate the rainfall," says research meteorologist Marshall Shepherd of NASA's Goddard Space Flight Center. The warm air and sea air collide, he explains, and "move straight up like the front ends of two cars that hit head on, providing a pump of moist

Hot, waterlogged cities can be cooled off in the usual ways-by limiting auto exhaust, for example. Using light-colored roofing and paving materials in place of black, heat-absorbing tar will also help. As a bonus, the cooler roof will reduce the need for air conditioning. - By Jeffrey Kluger

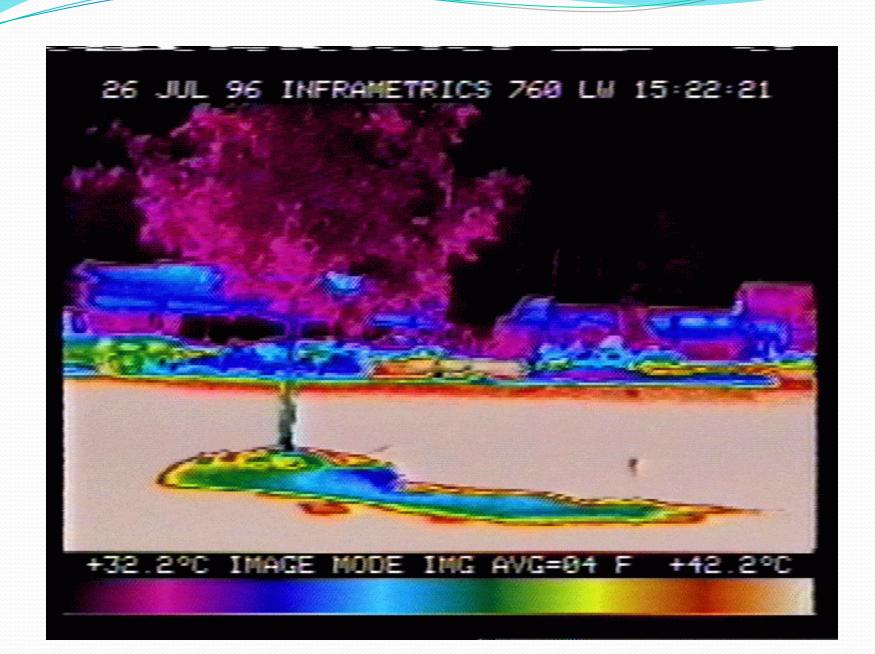
air that helps thunderstorms develop.'

3 In cities 2 Late in the day, the accumulated 4 The moist air and prevailing wind blows the near large bodies of water, moist air flows warm city air collide and drive heat starts to be released. The in toward the each other downwind of lighter, warmed air begins to rising urban ai higher, hitting cities get more the cooler layer rain than those of air above and creating cloud COLD AIR WIND DIRECTION

High concentrations of buildings, roads and other artificial structures retain heat, leading to warmer temperatures. evening in August 2000 demonstrates this urban heat-island effect: the hot zone downtown stands in sharp contrast to the cooler suburbs to the south



1845UT



Surface Radiation Budget

$$Q^* = (K_{in} + K_{out}) + (L_{in} + L_{out})$$

 $Q^* = Net Radiation$

K in = Incoming Solar

K_{out} = Reflected Solar

L in = Incoming Longwave

L_{out} = Emitted Longwave

Surface Energy Budget

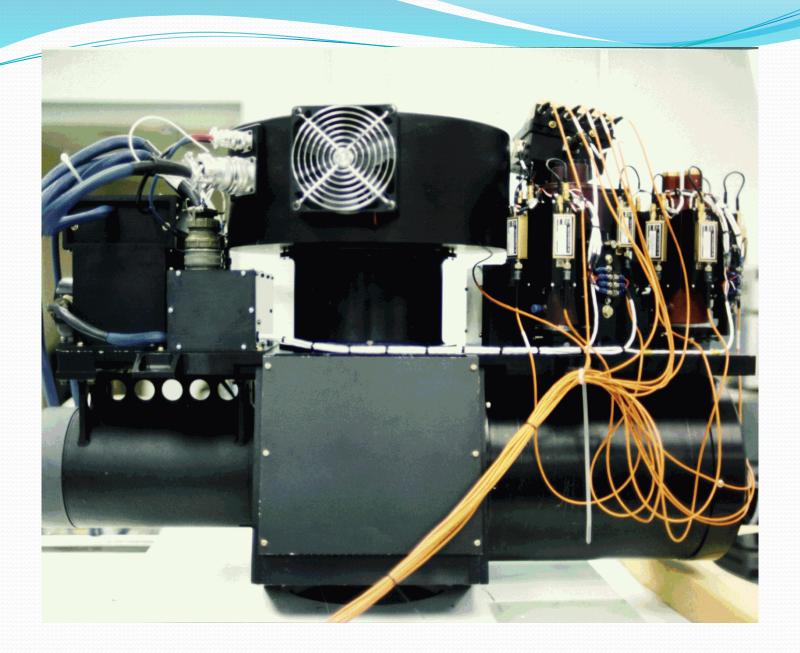
$$Q^* = H + LE + G$$

H = Sensible Heat Flux

LE = Latent Heat Flux

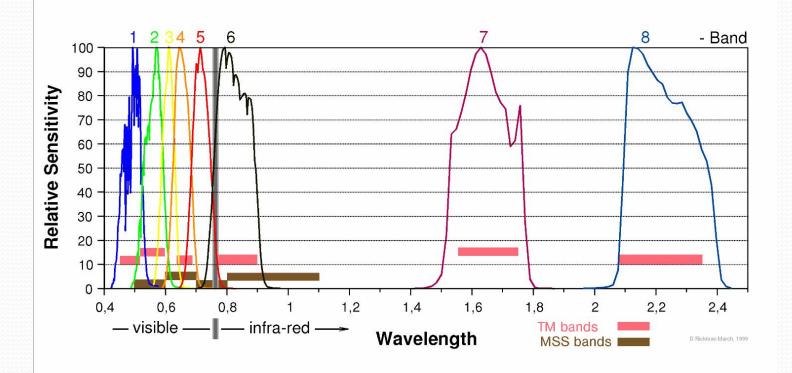
G = Storage (maybe + or -)





ATLAS

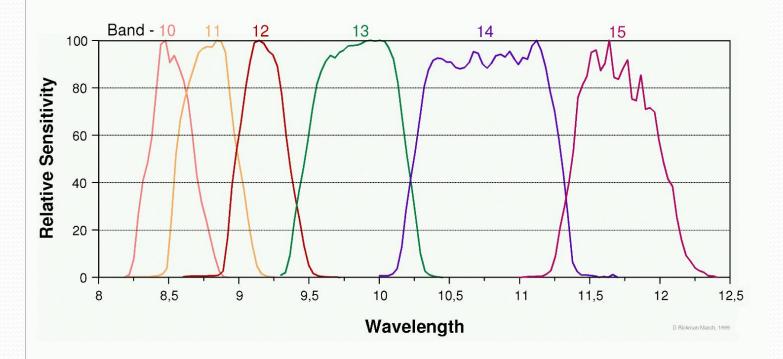
Typical Spectral Response Curves Reflected Bands



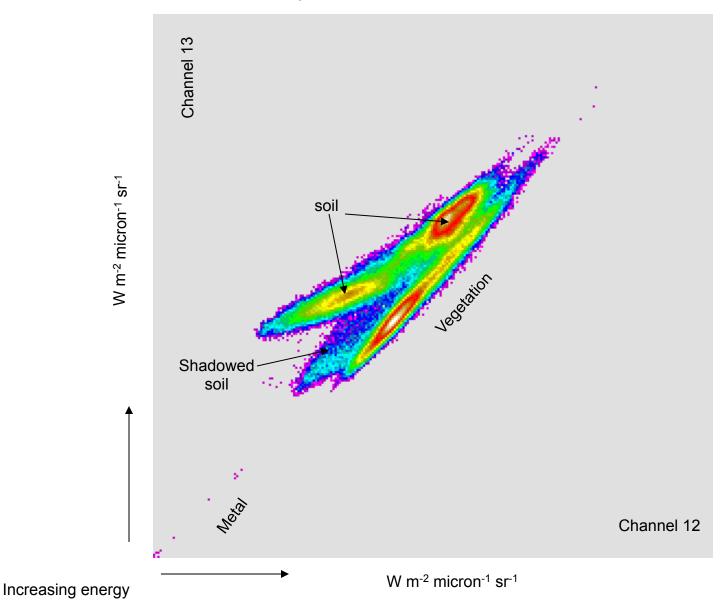
ATLAS

Typical Spectral Response Curves

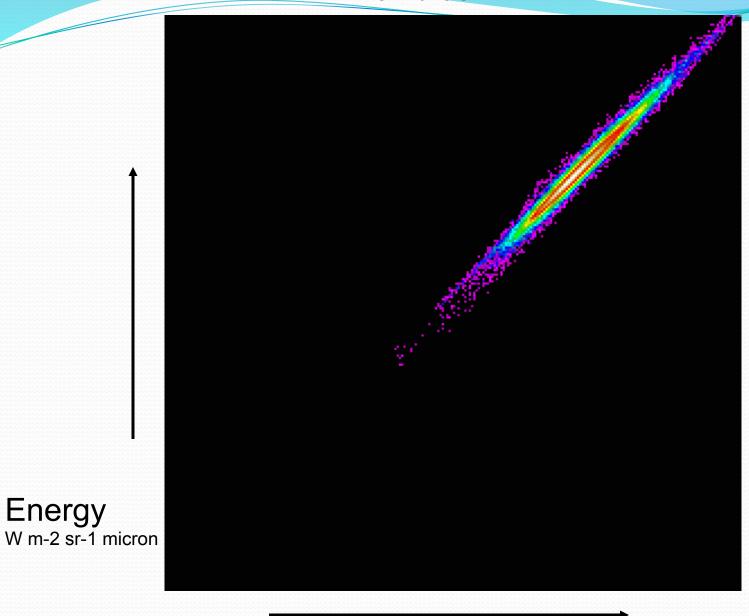
Thermal Bands

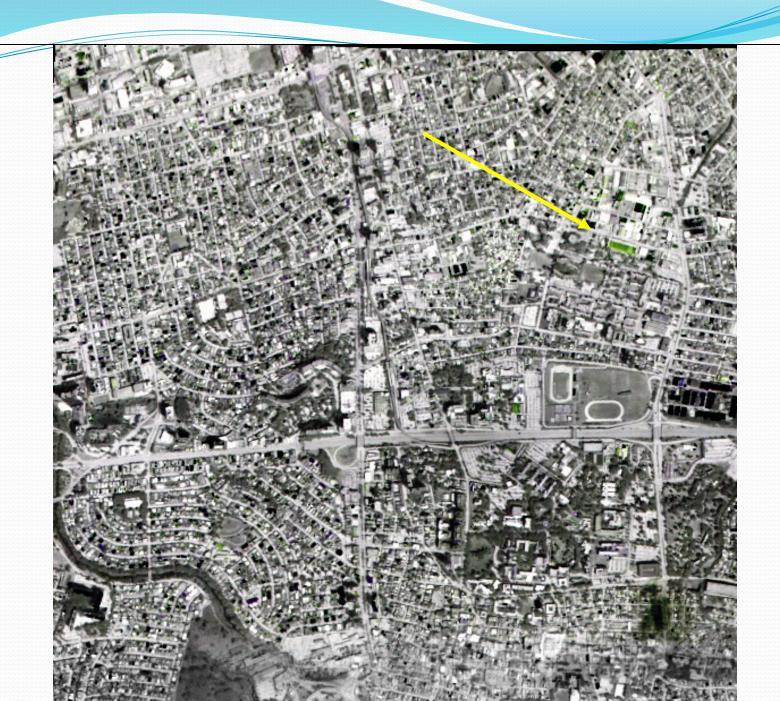


Emissivity Differences in Natural Surfaces



San Juan CH 10 vs CH 11





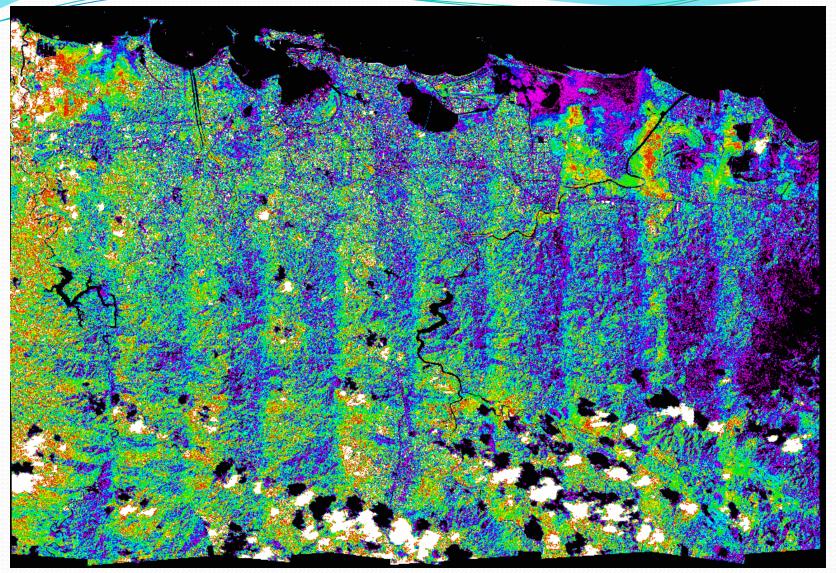




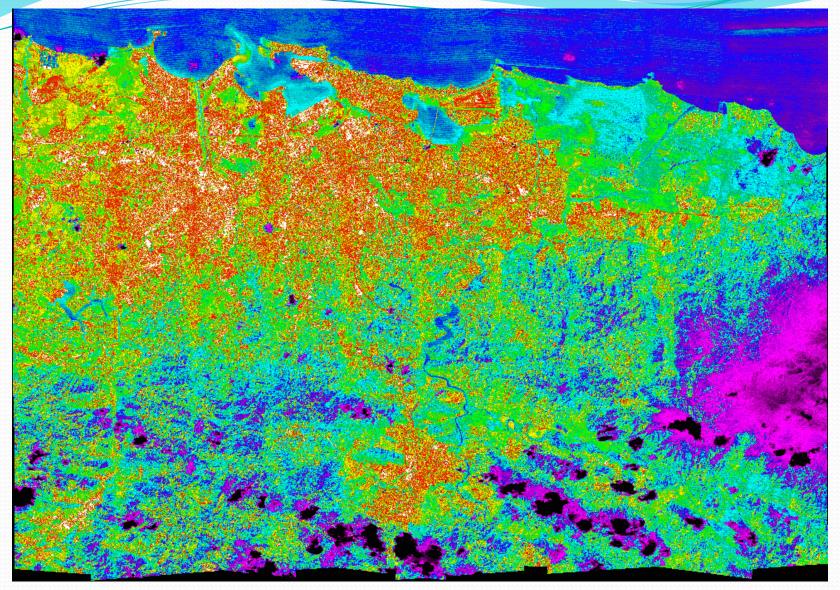
San Juan F5 Mosaic True Color



San Juan F5 Mosaic Albedo

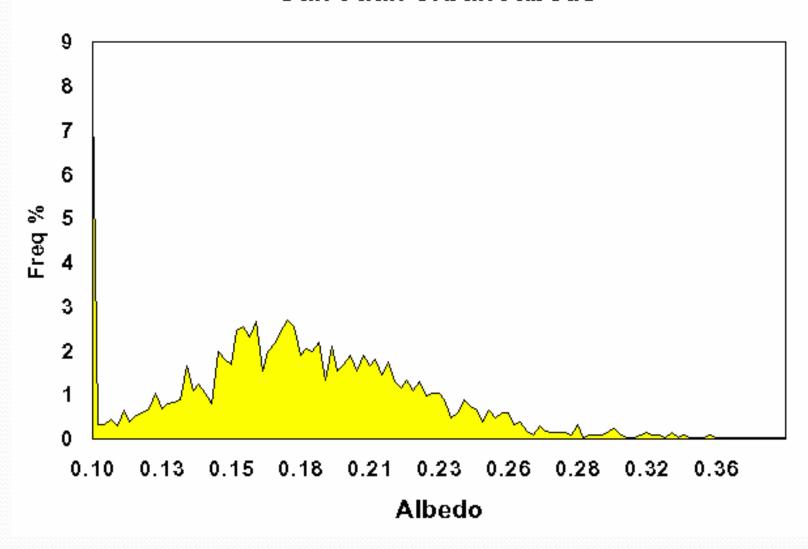


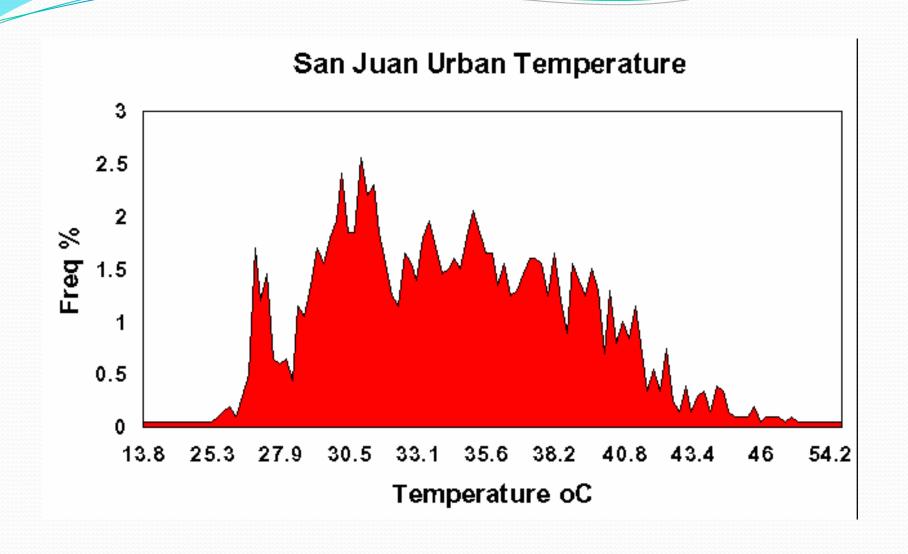
San Juan F5 Mosaic Temperature





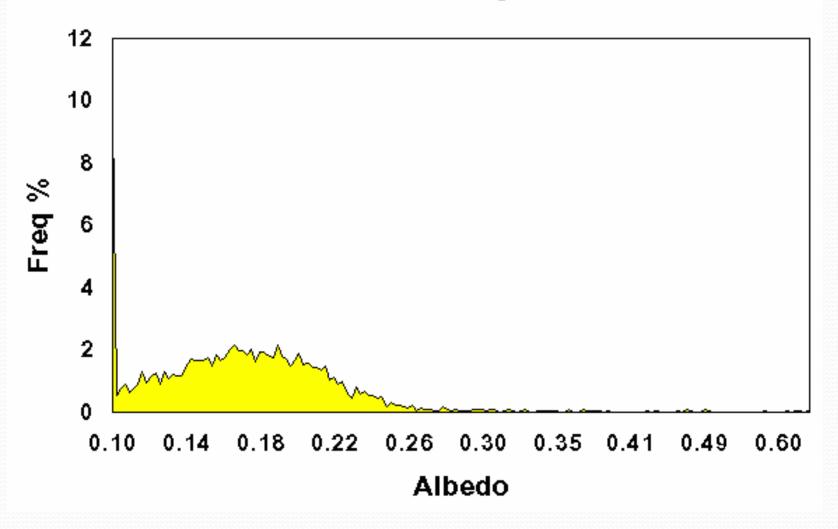
San Juan Urban Albedo



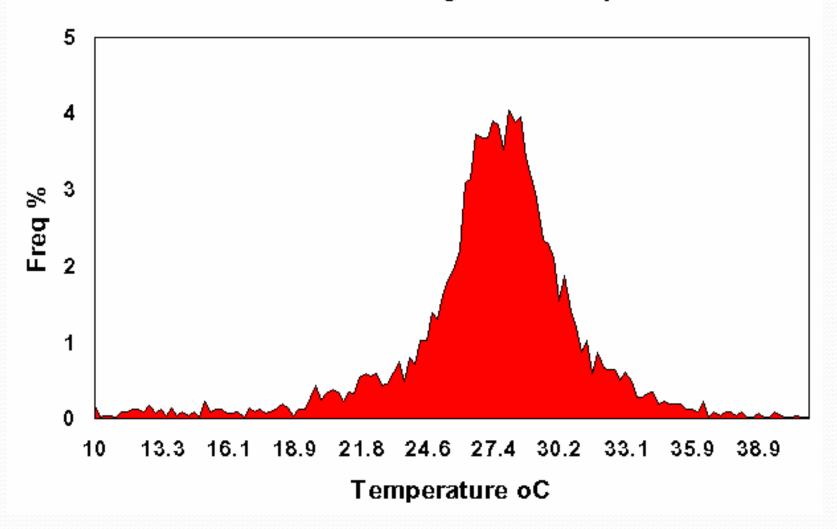




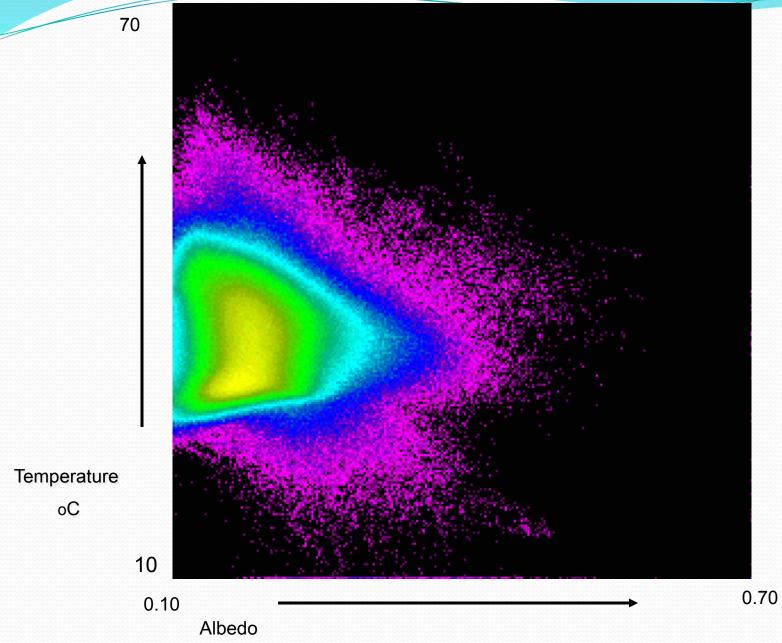
San Juan Urbanizing Area Albedo



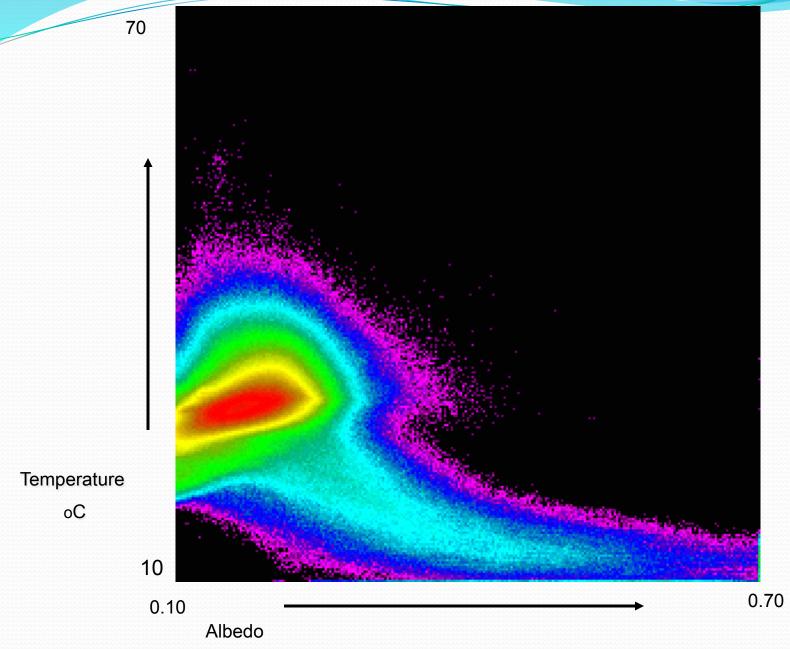
San Juan Urbanizing Area Temperature



San Juan Puerto Rico Albedo vs Temperature



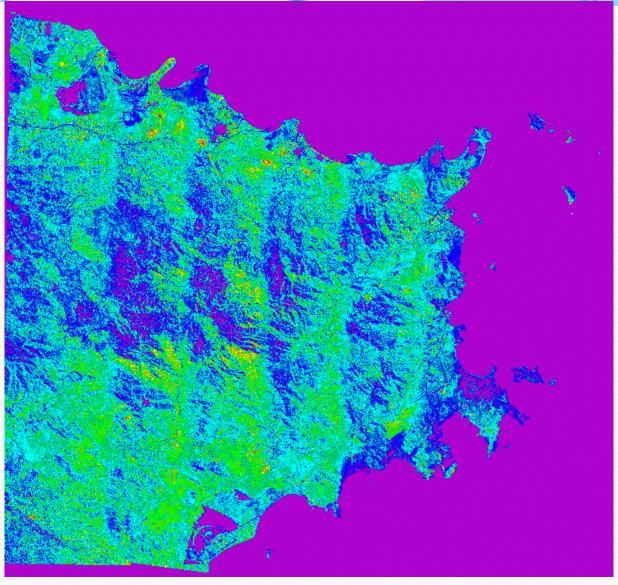
San Juan Puerto Rico Urbanizing Area Albedo vs Temperature



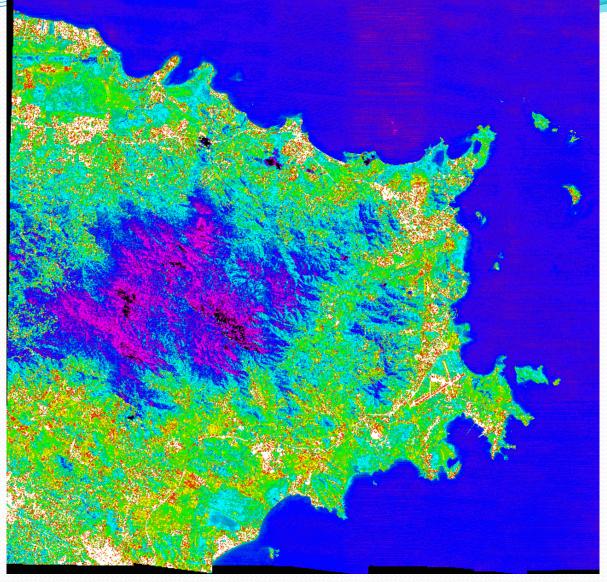
El Yunque F4 Mosaic True Color

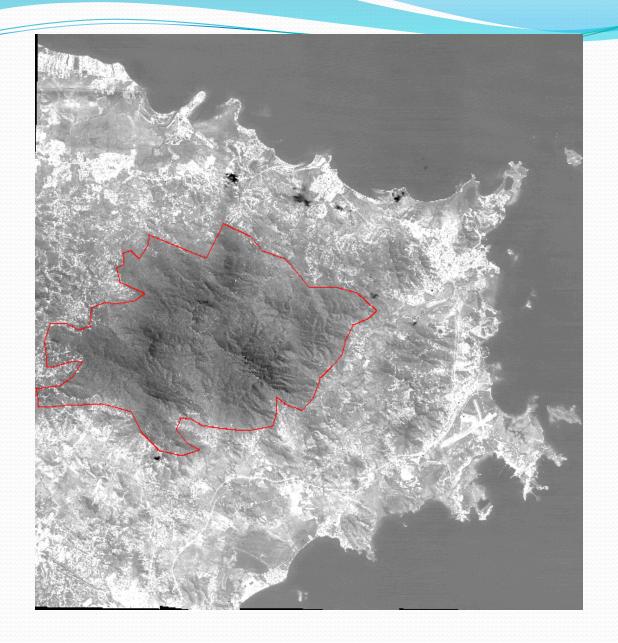


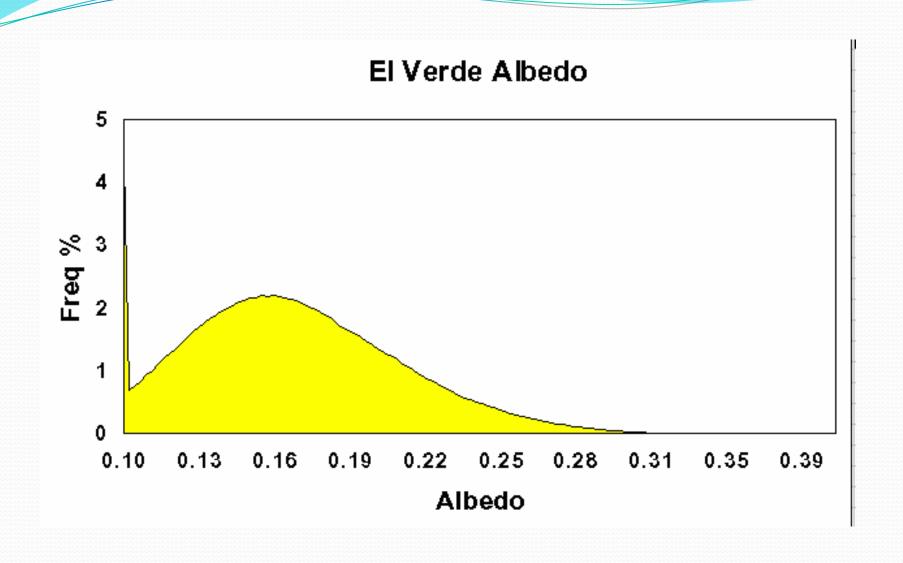
El Yunque F4 Mosaic Albedo

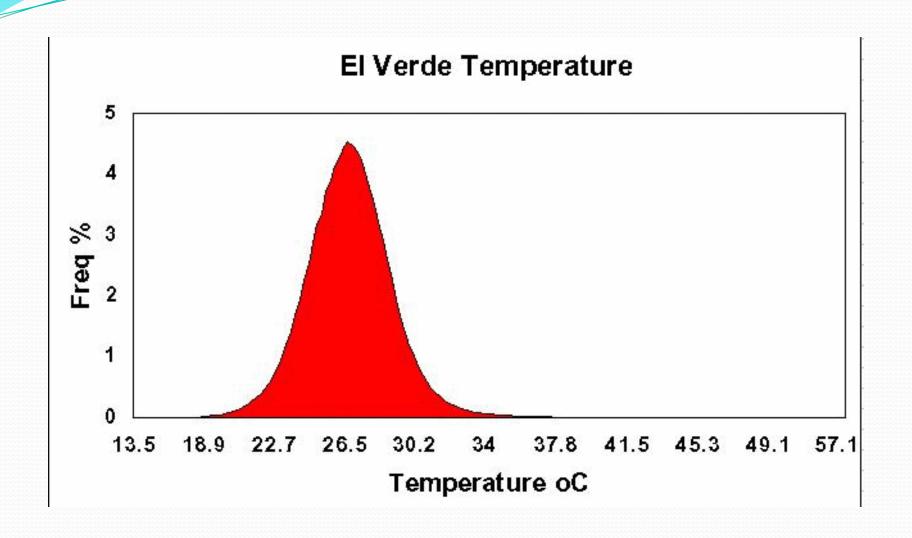


El Yunque F4 Mosaic Temperature

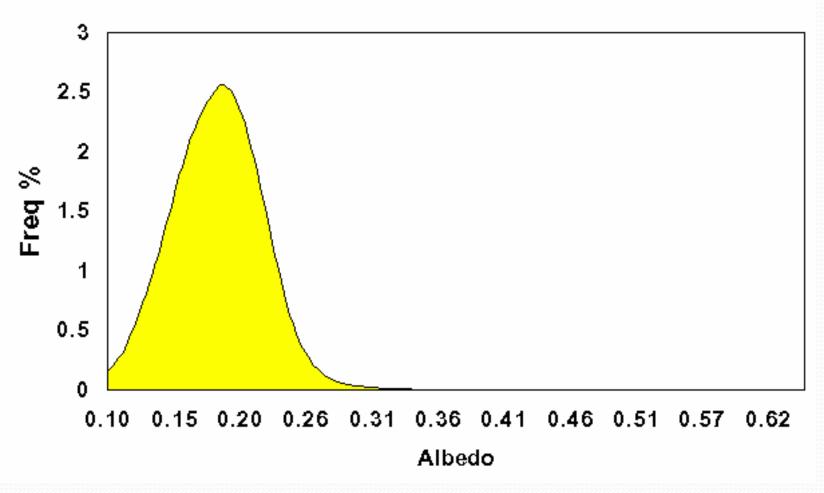


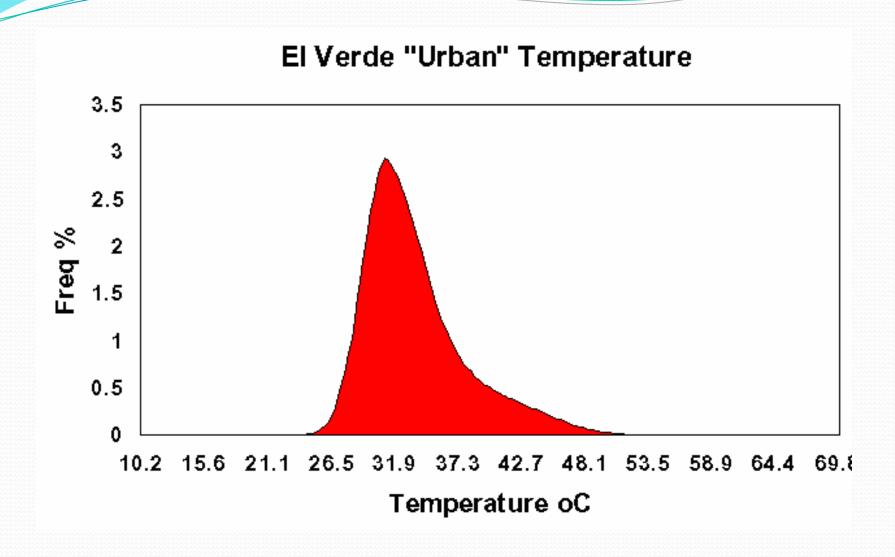


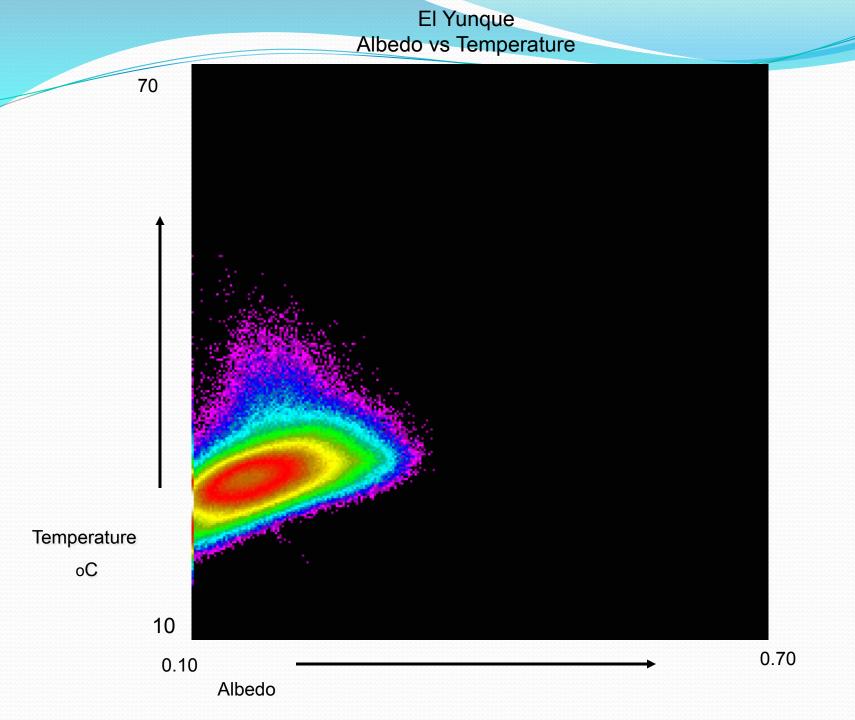












Urban Heat Island Mitigation Pilot Project

EPA/NASA Marshall Space Flight Center

Jeffrey C. Luyall Dale Quattrochi Maury Estes Doug Rickman

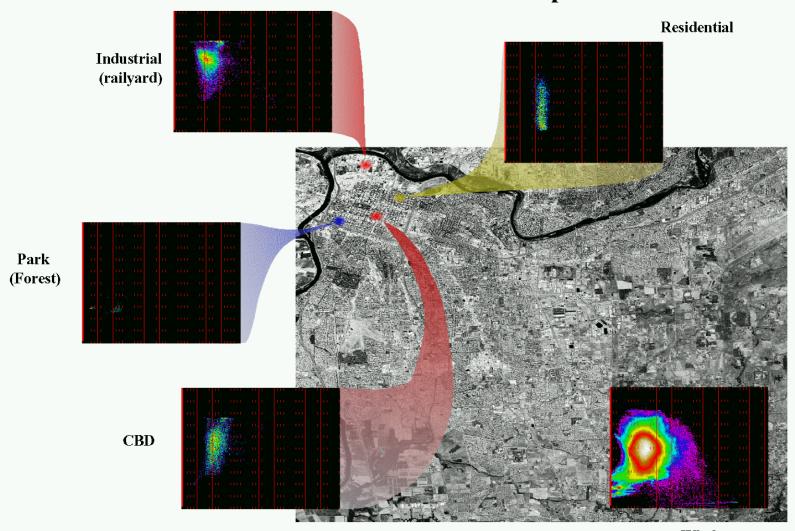


Urban Heat Island Mitigation Strategies

- ▲ Albedo Modification
 - Lighter colored roofs and pavements
 - New materials/coatings
- ▲ Plant trees and increase green space
 - Shade buildings, rooftops, parking lots and roads
 - Cool the air through transpiration
- ▲ Rooftop gardens
 - Keep roofs cool by shading and/or transpiration
 - storm water reduction



Sacramento Skattergrams Albedo vs Temperature



Whole Mosaic

Climate Impacts of Land Cover and Land Use Changes in Tropical Islands under Conditions of Global Climate Change. Daniel E. Comarazamy, J. E. González, J.C. Luvall, D. L. Rickman, & R. D. Bornstein. *In Press*

A Land-Atmospheric Interaction Study in the Coastal Tropical City of San Juan, Puerto Rico

Daniel E. Comarazamy, J. E. González, J. C. Luvall, D. L. Rickman, and P. J. Mulero

Earth Interactions

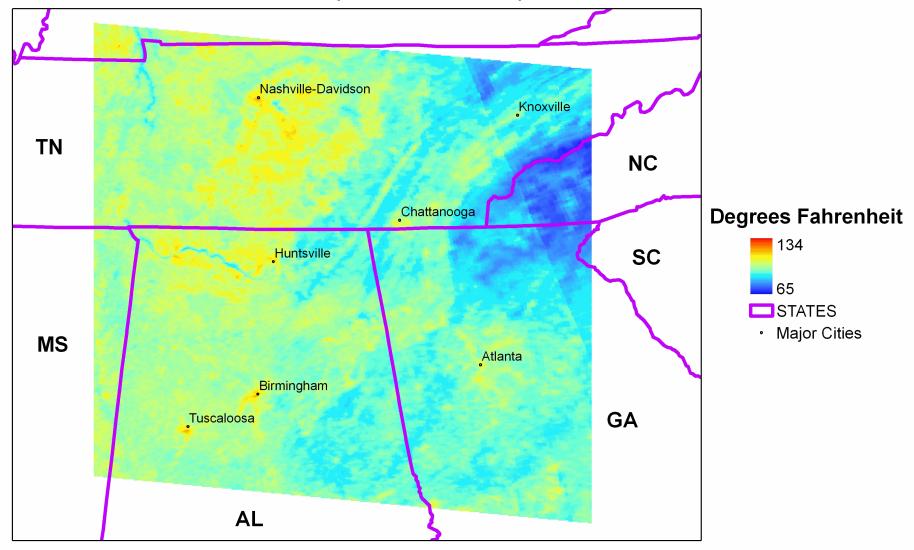
Volume 14, Issue 16 (November 2010) pp. 1-24

What can HyspIRI data provide for urban heat island research and applications?

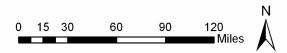
- Scale consistent with complex urban surfaces
- •Well calibrated data allows quantifiable radiative & energy budgets:
 - Analysis of albedo
 - Analysis of surface temperature
 - Analysis of emissivity
- Scale consistent to plan alteration of the urban fabric to mitigate the urban heat island
- Swath width scale for urban climatology studies urbanization of RAMS.
- Global cities.
- Only mutispectral thermal at 60 m
- ■5 day repeat for thermal for short & long term trend analysis
- Allow for functional classification of urban surfaces

Thanks

NASA/MODIS Land Surface Temperature at 1:30 PM (June 27, 2012)

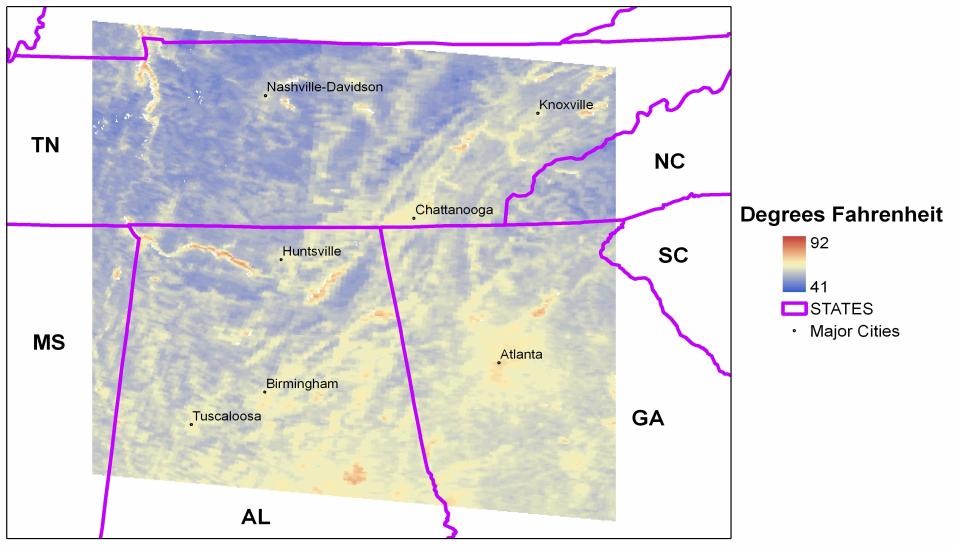


Prepared by: Dr. Mohammad Al-Hamdan USRA at NASA/MSFC mohammad.alhamdan@nasa.gov July 17, 2012

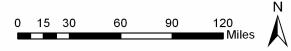




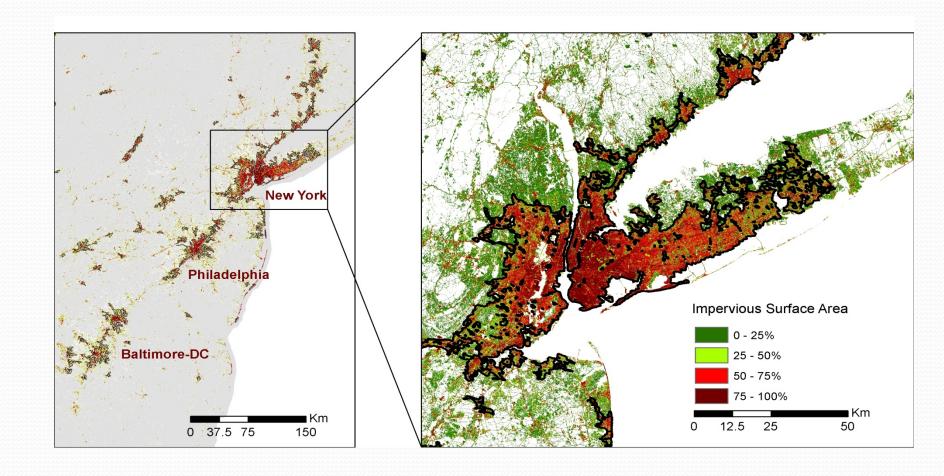
NASA/MODIS Land Surface Temperature at 1:30 AM (June 27, 2012)



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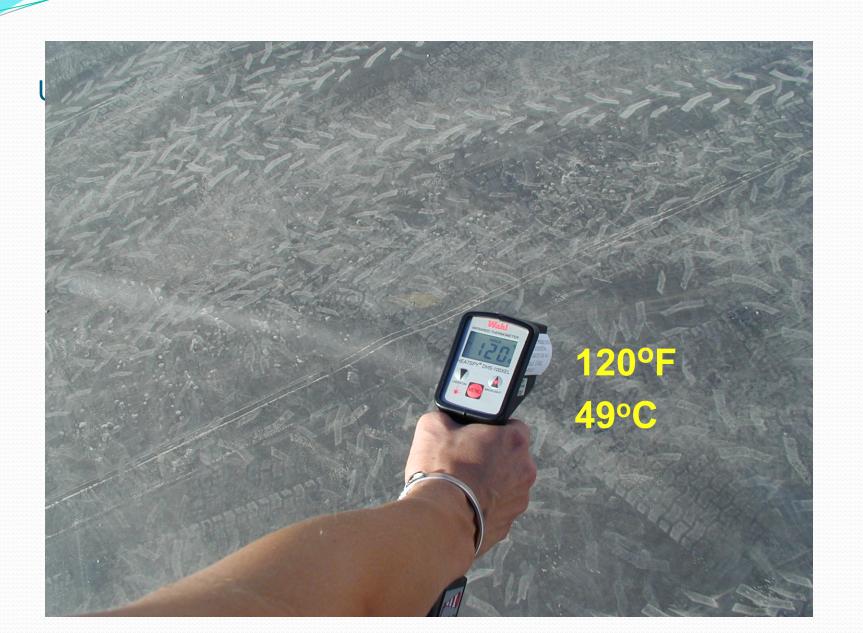




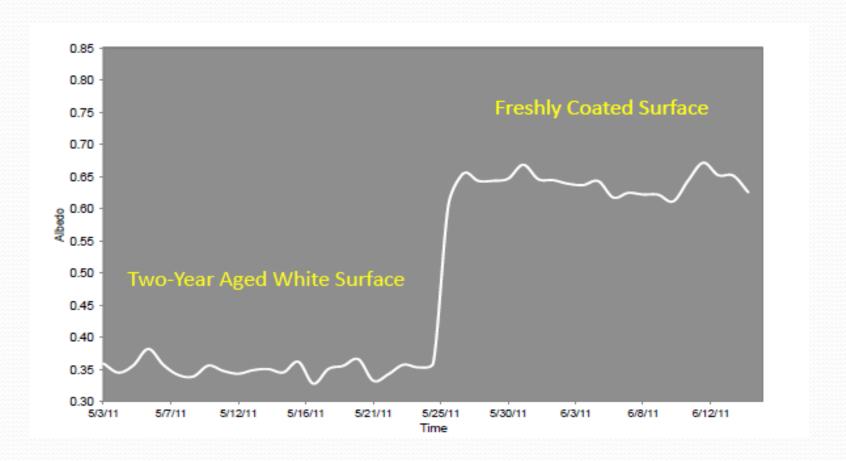
Urban Thermal Remote Sensing Impervious Surfaces



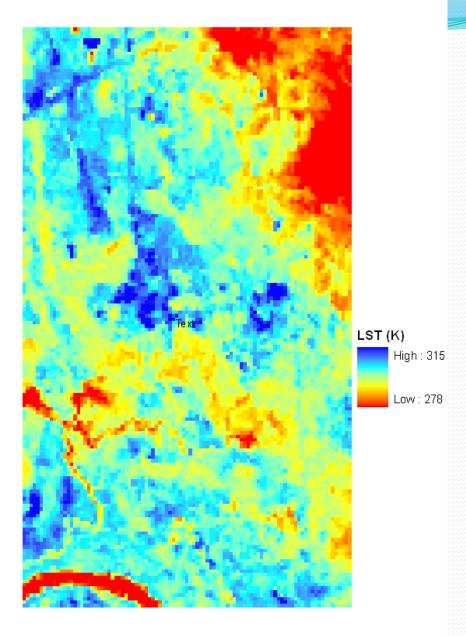




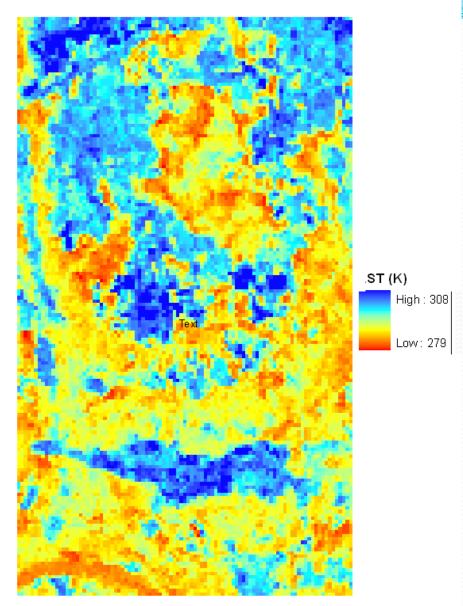
Urba 83°F 28°C



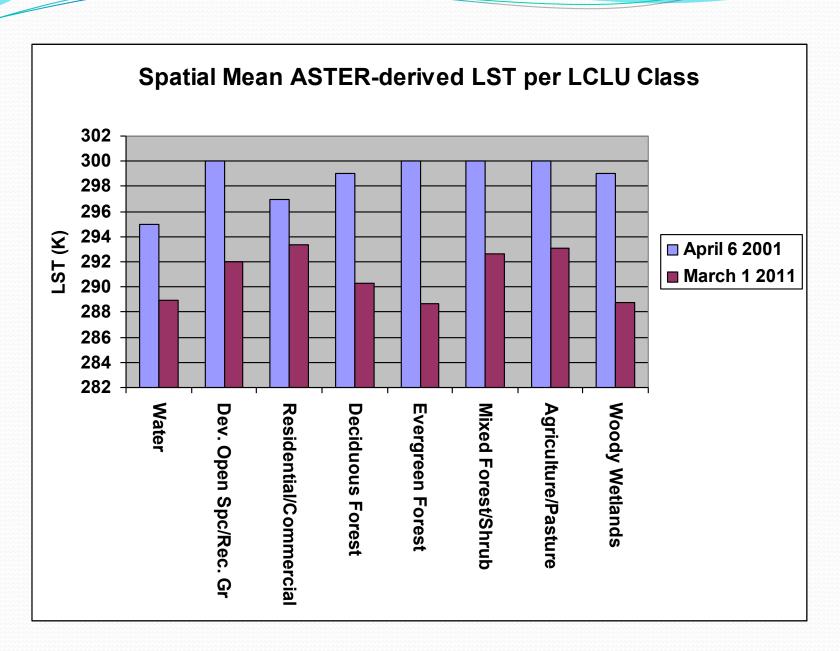
ASTER LST (K) (APR 6, 2001)



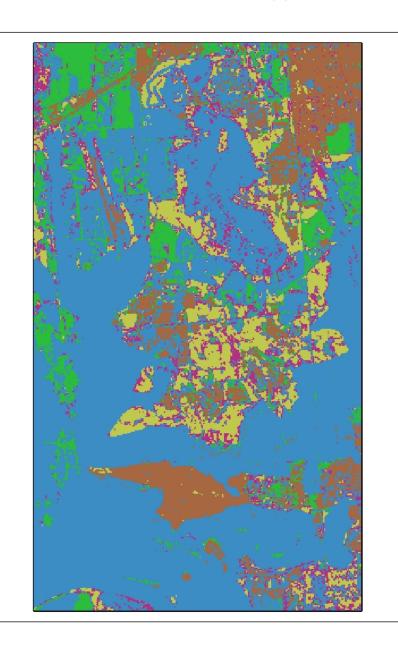
ASTER LST (K) (MAR 1, 2011)



Land Cover and Climate Change Impacts



Emissivity 1992

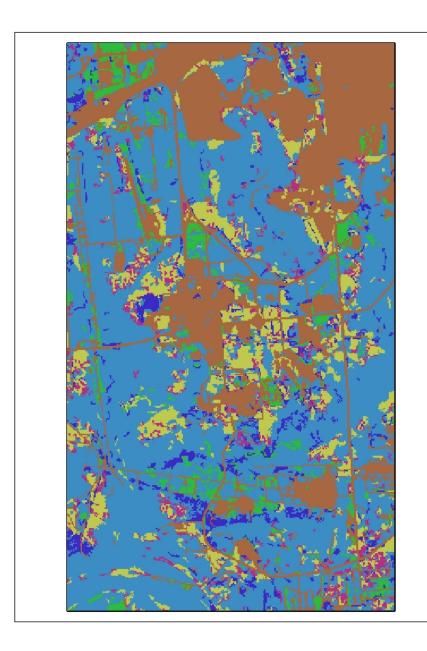


Emissivity

- 0.969 (Used for Bare Soil; and Developed Pixels)
- 0.974 (Used for Shrub Pixels)
- 0.980 (Used for Crops Pixels)
- 0.989 (Used for Deciduous Forests (assuming they're mostly Broadleaf); Wetlands, and Water Pixels
- 0.9895 (Used for Mixed Forests Pixels)
- 0.990 (Used for Evergreen Pixels (assuming they're mostly Needle)

Based on a look-up table in Snyder et al. 1998 and given that our analysis is for a period when the vegetation is green.

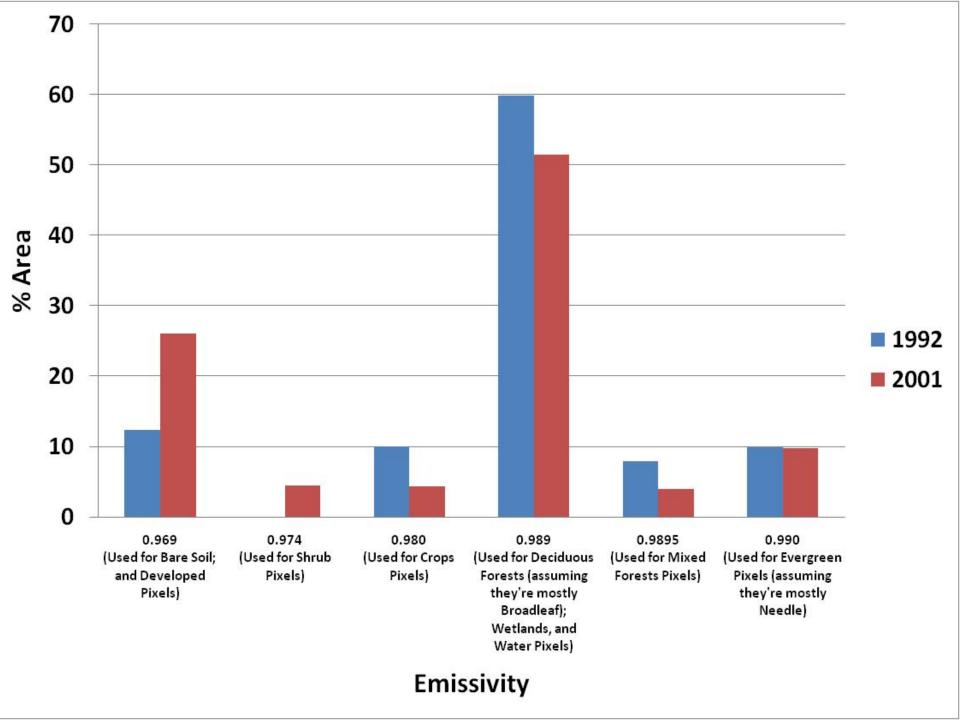
Emissivity 2001



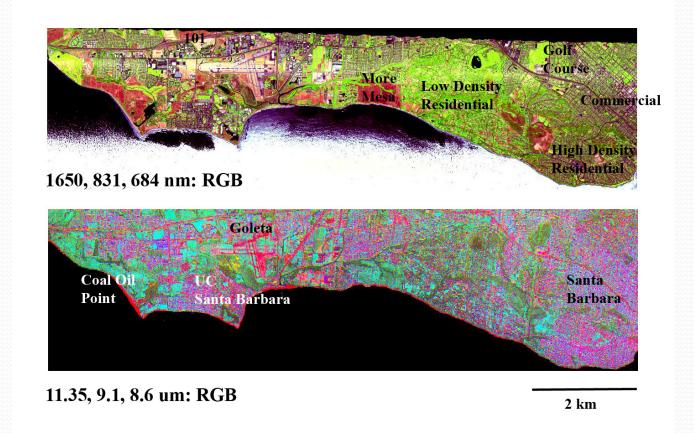
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Based on a look-up table in Snyder et al. 1998 and given that our analysis is for a period when the vegetation is green.



Study Site: Santa Barbara



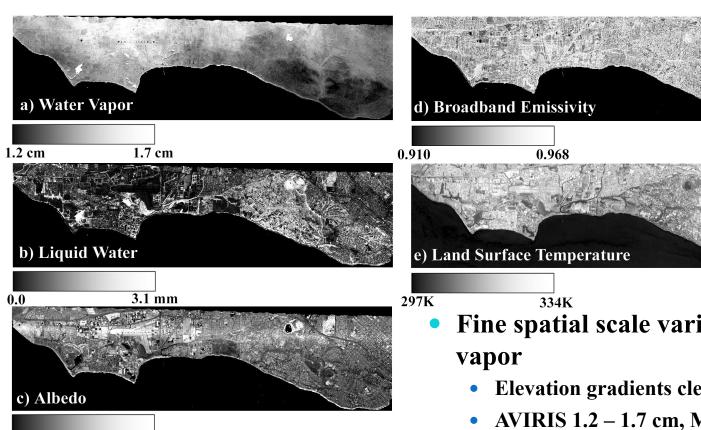
- Mixed urban-natural systems, ~ 150,000 people
- AVIRIS-MASTER pair, June 19, 2008
 - 7.5 m AVIRIS, 15 m MASTER
 - Spatial degradation, 15 m AVIRIS, 60 m AVIRIS/MASTER

Results

0.04

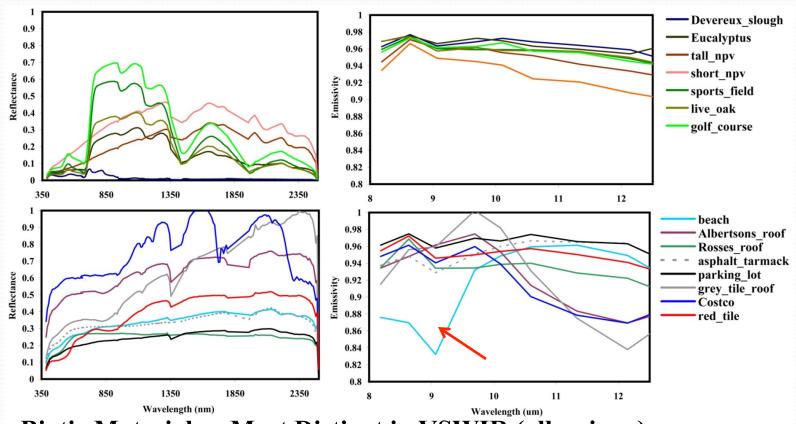
0.45

AVIRIS-MASTER Products



- Fine spatial scale variability in water
 - **Elevation gradients clear**
 - **AVIRIS 1.2 1.7 cm, MASTER 0.78 cm**
- Liquid water and emissivity positively correlated
 - Asphalt also high emissiivty
- Albedo and LST poorly correlated

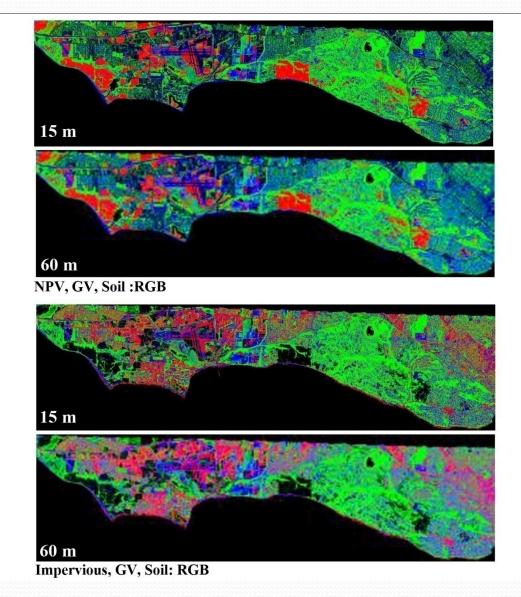
Reflectance and Emissivity Spectra



- Biotic Materials Most Distinct in VSWIR (all unique)
 - NPV low emissivity in TIR (Differs by stature)
- Abiotic Materials Varies
 - AVIRIS: Painted roofs, red tile
 - Soils and some road surfaces are not distinct*
 - MASTER: Quartz beach sands, various roof types
 - Asphalt surfaces are near black bodies

VSWIR-TIR
improves
discrimination of

VIS-NPV Fractions: 15 and 60 m



- GV and NPV fractions scaled well between all spatial resolutions
 - 7.5, 15 and 60 m
- Soil tended to be overmapped at the expense of Impervious at coarser scales

Two error sources
Asphalt – soil
Red Tile Roof (so variable requires 3 ems)

Fraction Validation: 60 m a) GV b) NPV y = 0.998x - 0.0947 $R^2 = 0.8085$ 0.8 0.8 NAIP Fractions Phenology 0.6 0.6 0.4 0.4 0.2 0.2 y = 1.05x + 0.041 $R^2 = 0.829$ **GV-NPV** 0.6 0.8 0.2 0.4 0.6 0.8 0 0 **Fractions** c) Impervious d) Soil good at HyspIRI Under-1 y = 0.690x - 0.068Mapped NAIP Fraction 0.6 0.4 0.2 scales $R^2 = 0.409$ 0.8 0.6 0.4 Soil Overmapped y = 1.443x + 0.110.2 0.2 (red tile/Asphalt) $R^2 = 0.7184$

0.2

0

0.6

HyspIRI Fractions

0.8

0.2

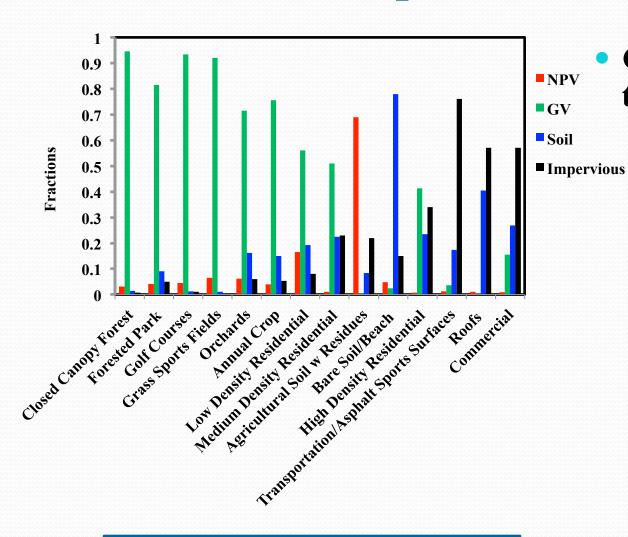
0.4

HyspIRI Fractions

0.6

0.8

Land-cover Composition

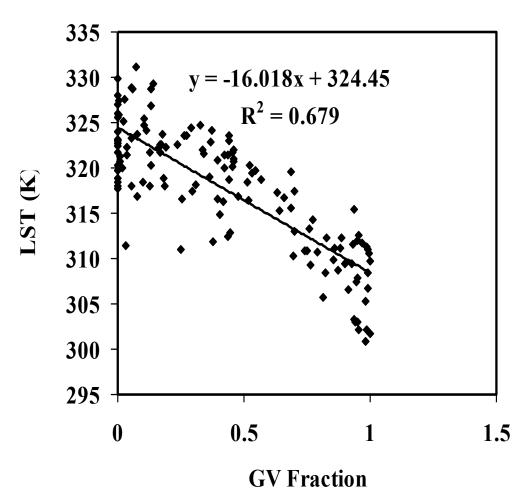


- Cover fractions follow the expected pattern
 - High GV, low Temperature
 - High Impervious, high Temperature
 - Residential
 - Low density, high GV, Low T, low Imp
 - High density, low GV, High T, High Imp

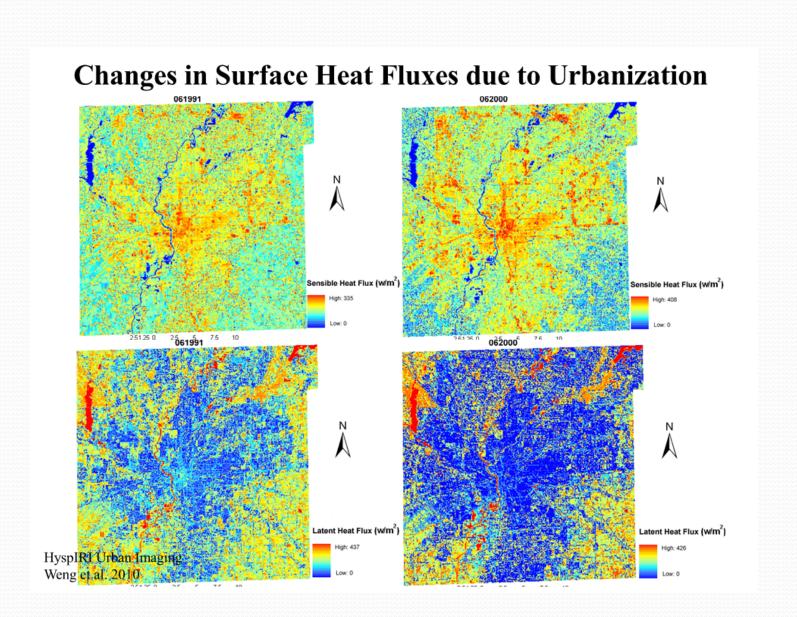
306K

325K

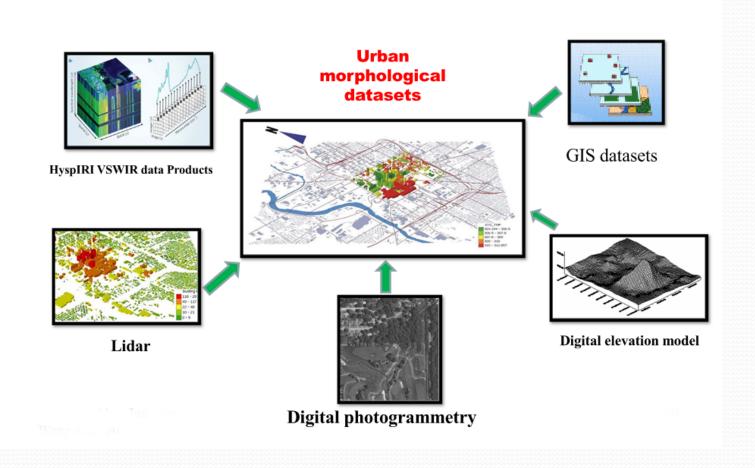
Green Cover and LST



- Standard inverse relationship between GV Cover and LST
- Considerable scatter
 - The scatter is likely to be the most interesting part
 - Moist soils evapotranspiring
 - Closed canopies with variable ET



HyspIRI Combined with other RS/GIS data to generate Urban Morphological Datasets











Images showing some of the unique characteristics that define individual cities (clockwise from upper left: Boise, ID; Annapolis, MD; St Louis, MO, and Tucson, AZ).