

Hyperspectral Feature Detection Onboard the Earth Observing One Spacecraft

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3

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1

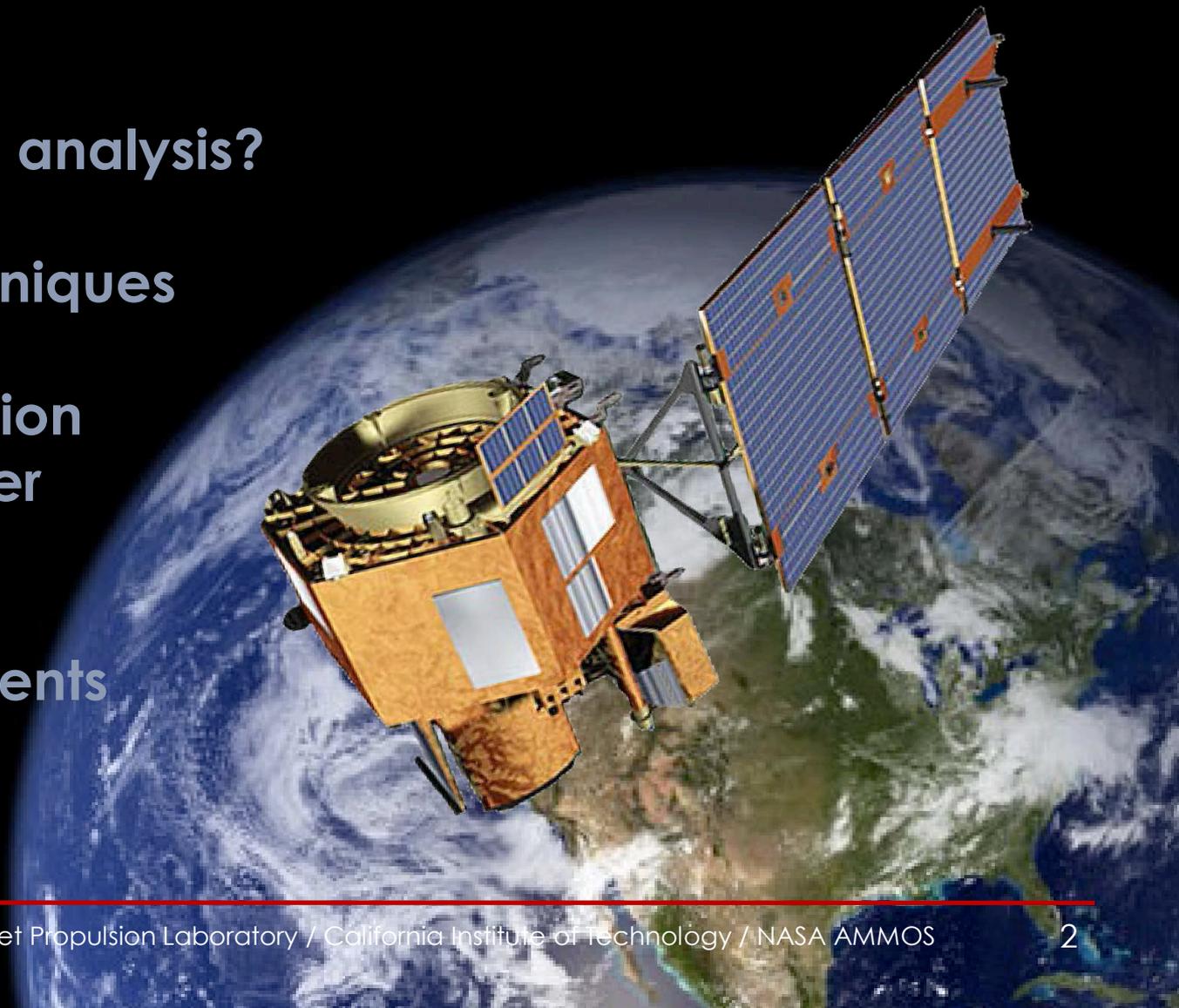
Agenda

Why onboard
hyperspectral analysis?

Enabling techniques

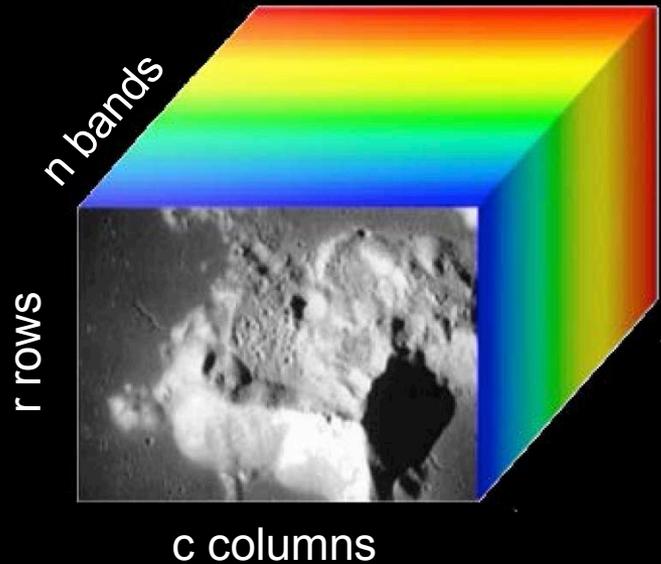
1. Superpixel segmentation
2. Endmember detection

EO-1 experiments



Imaging spectrometers are important tools for exploration

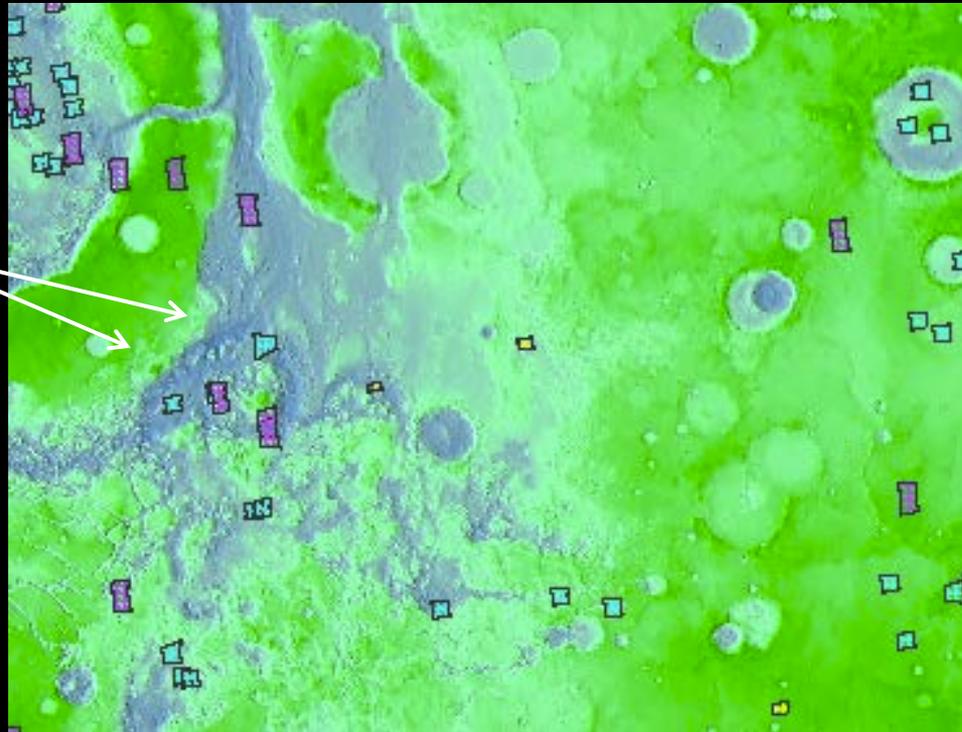
- Moon (M³)
- Mars (OMEGA, CRISM)
- Jupiter (Galileo NIMS)
- Saturn (Cassini VIMS)
- Vesta (Dawn VIR)
- Comets (Rosetta)
- In-situ
- Terrestrial applications



Images courtesy NASA/Brown/JPL and other missions

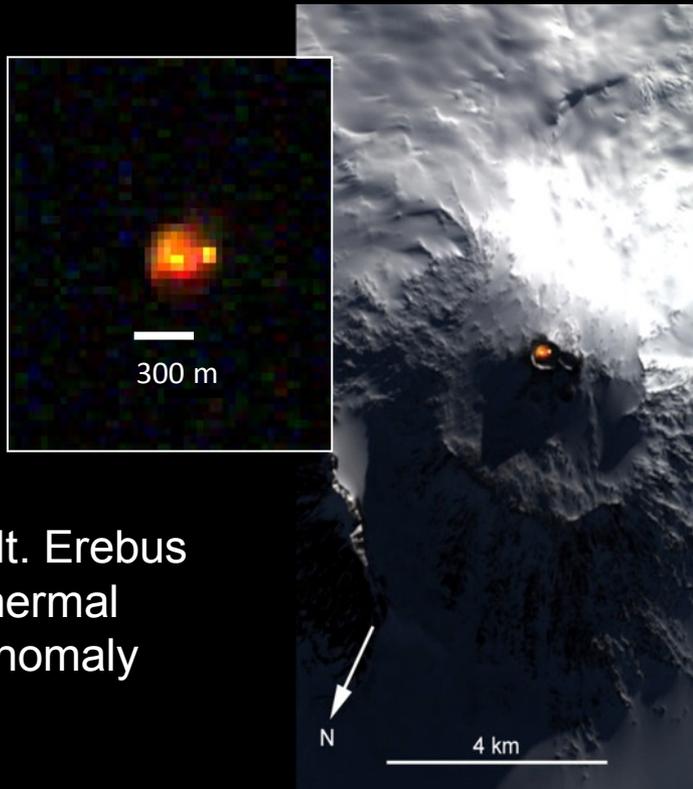
Challenge: bandwidth constraints limit duty cycle

MRO/CRISM
coverage by
full-resolution
images



images courtesy NASA / Brown University. Online map: <http://crism-map.jhuapl.edu/>

Challenge: respond to targets of opportunity



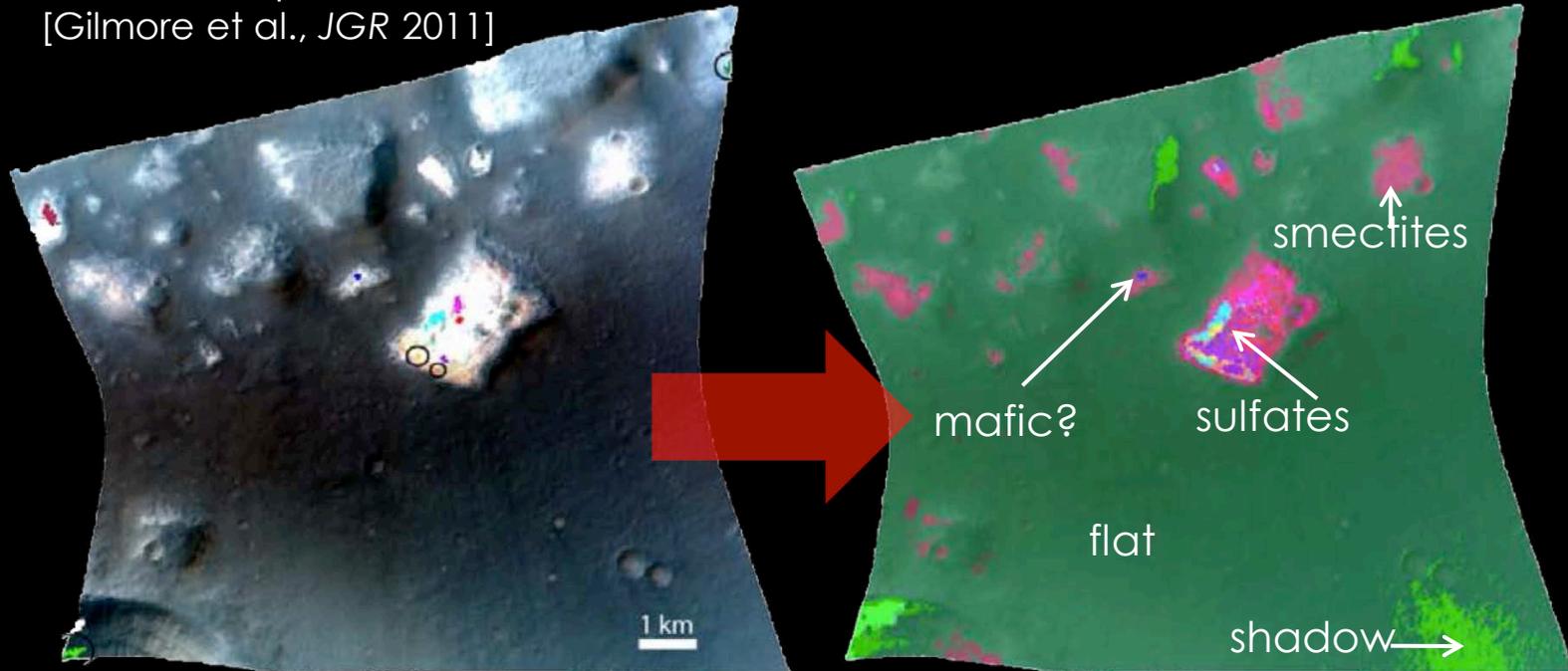
Mt. Erebus
thermal
anomaly

Hartley 2
morphology,
plumes



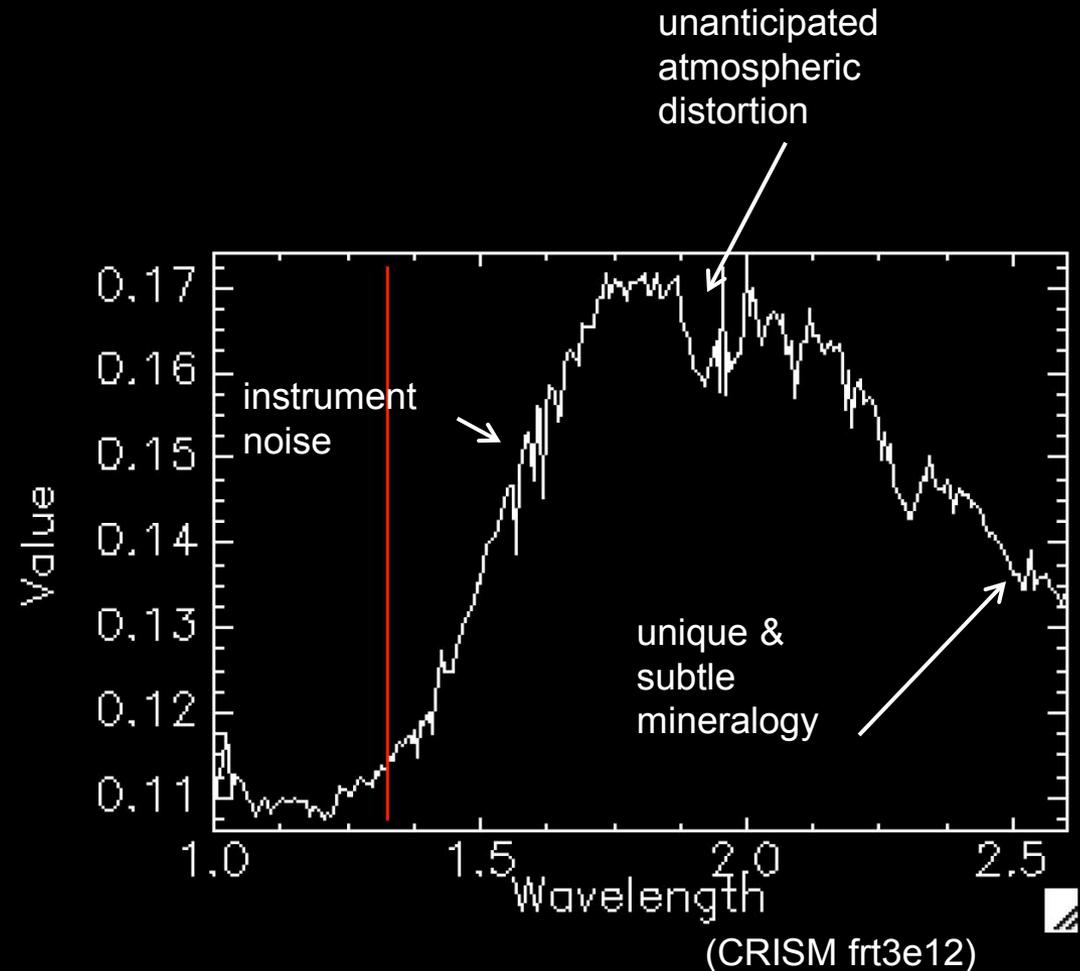
Solution: onboard spectral discovery and mapping

CRISM interpretation
[Gilmore et al., *JGR* 2011]



Onboard analysis is hard

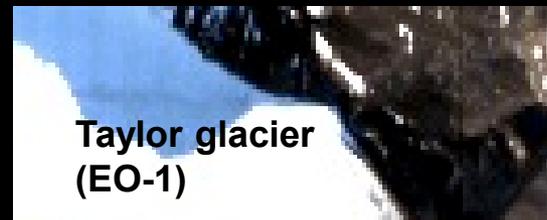
1. Planetary spectral data can be noisy
2. Signals are subtle, difficult to interpret
3. Targets are often unknown in advance
4. Very limited onboard computation



Superpixel representations exploit the *spatial* dimension

- Spatially contiguous, homogeneous regions
- Improves robustness to artifacts
- Reduces noise by \sqrt{n}
- Reduces data set by 75x for later processing

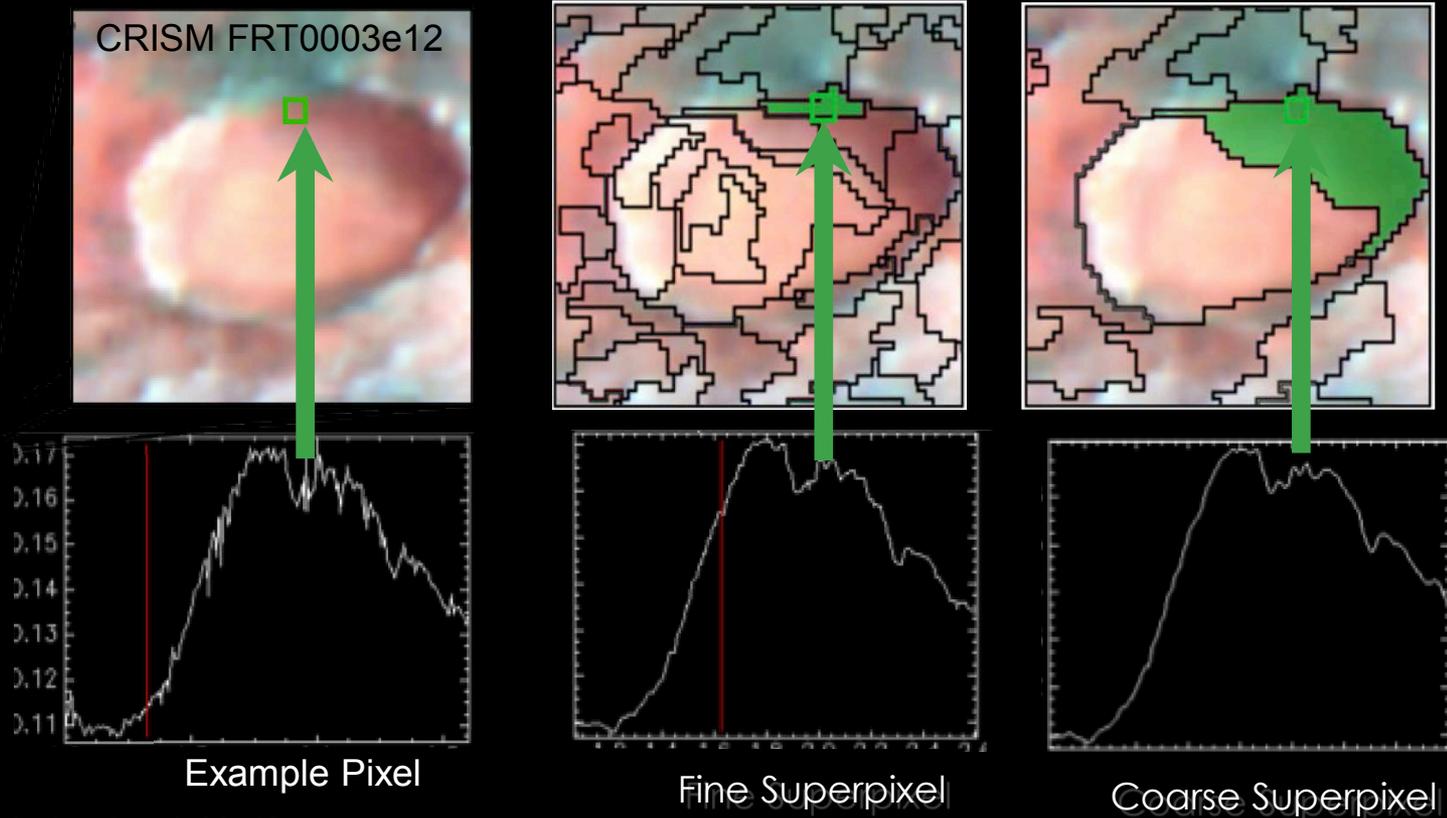
260,000 pixels



<3500
superpixels



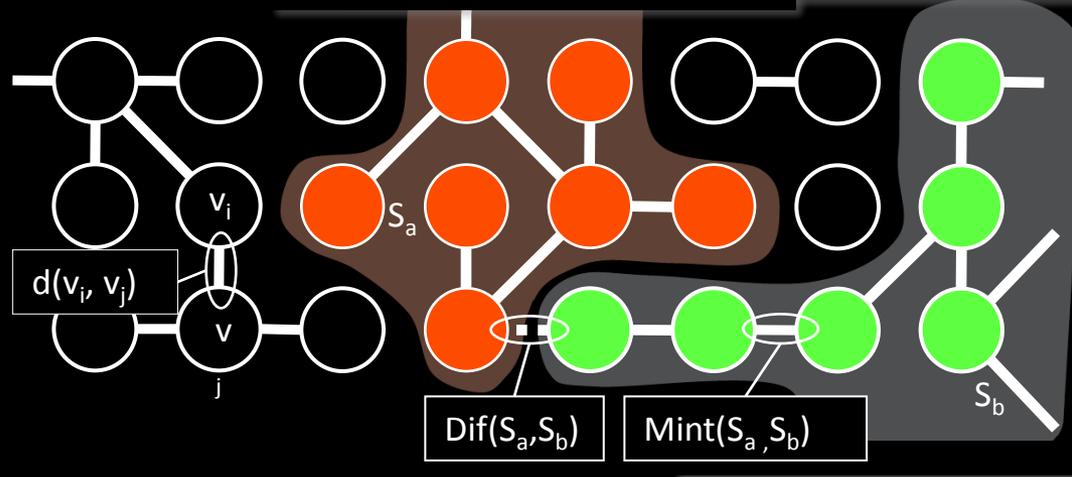
Superpixel representations exploit the *spatial* dimension



[Thompson et al., TGARS 2010]

Superpixel representations exploit the *spatial* dimension

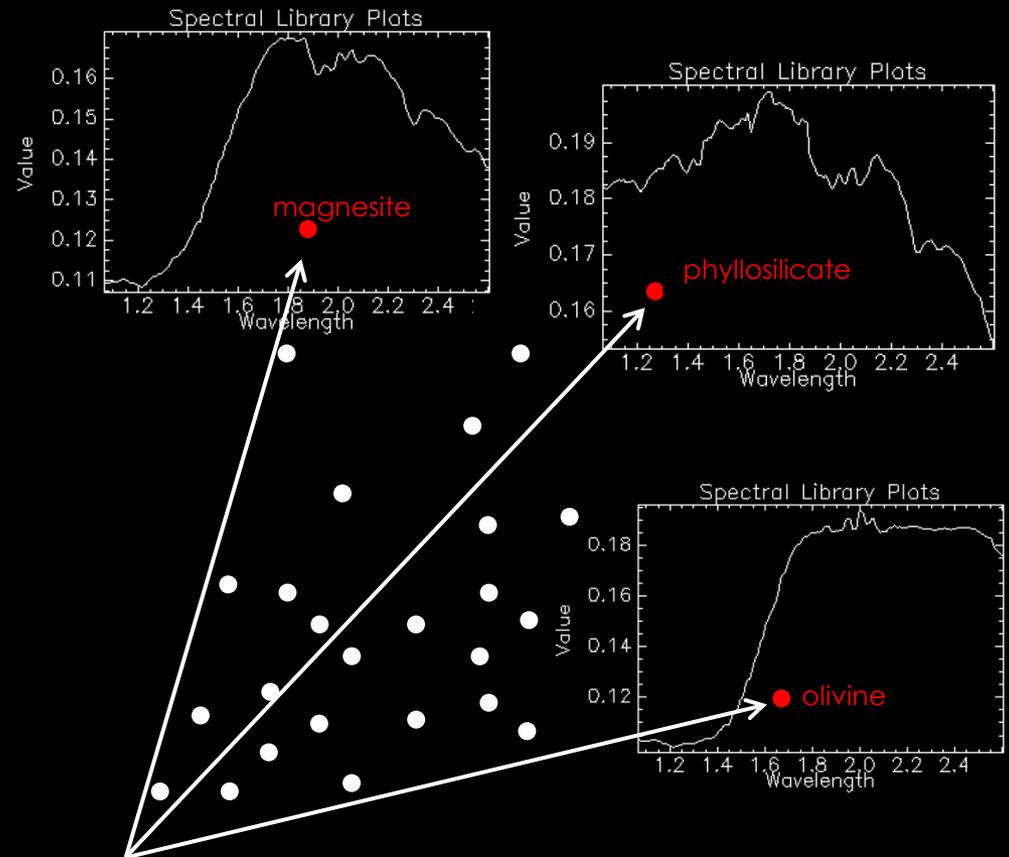
- Felzenszwalb graph partitioning algorithm [2004]
- Compute spectral distances between neighbors
- Agglomerative clustering grows minimum spanning trees



Endmember detection finds spectrally distinct superpixels

Linear mixing model: spectra are convex combinations of a small number of endmembers

Endmembers often correspond to unique materials



Endmember detection finds spectrally distinct superpixels

Linear mixing model: spectra are convex combinations of a small number of endmembers

Endmembers often correspond to unique materials

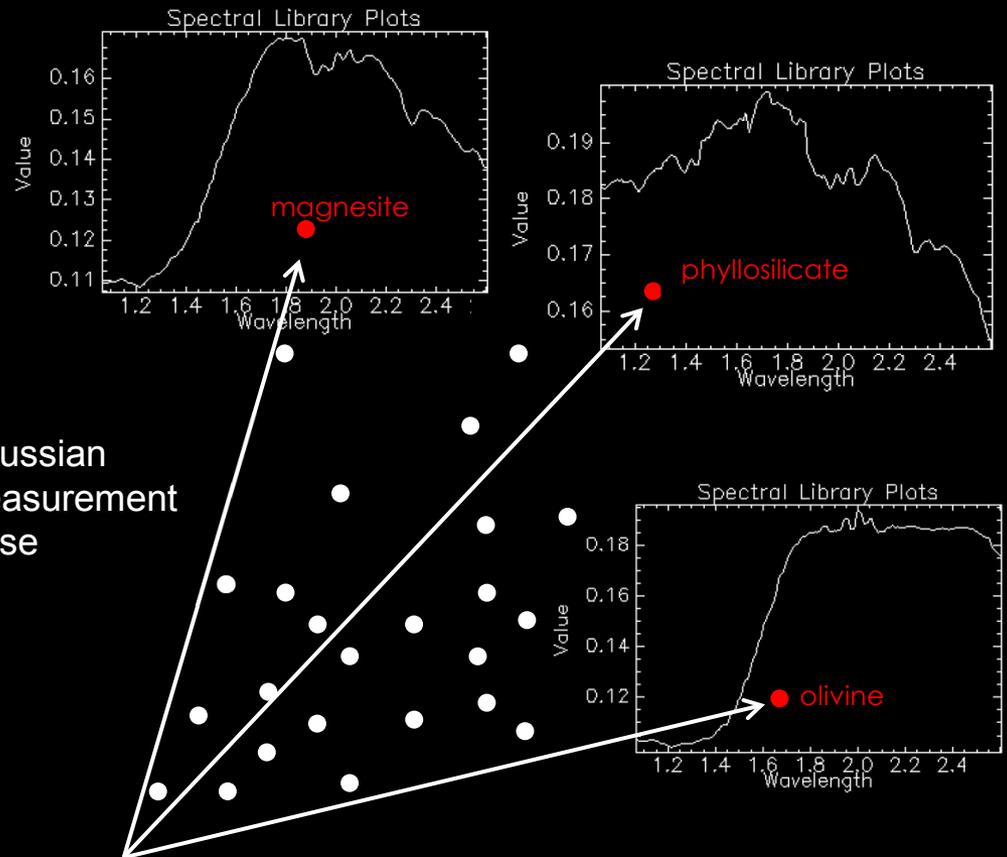
$m \times n$ matrix of endmembers

measurement at superpixel j

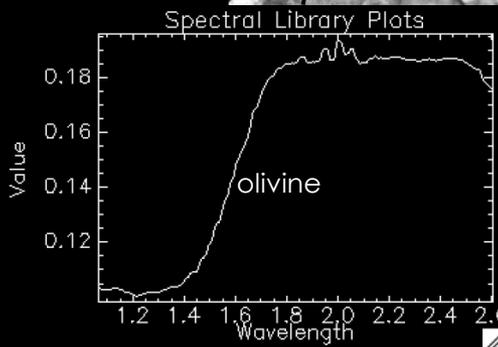
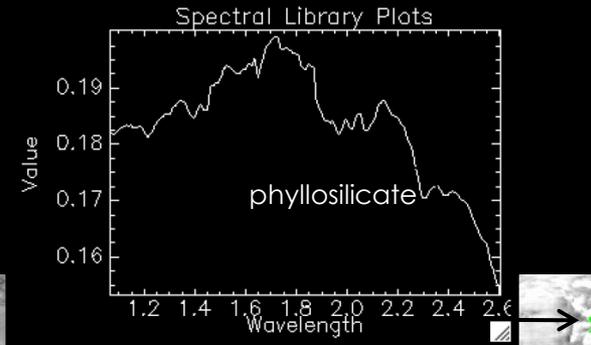
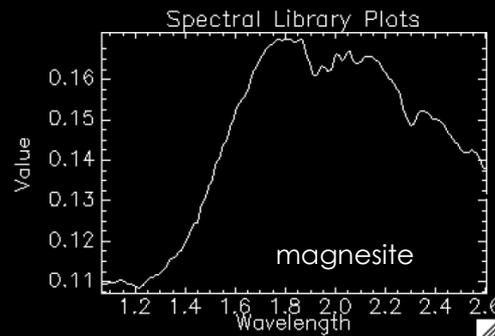
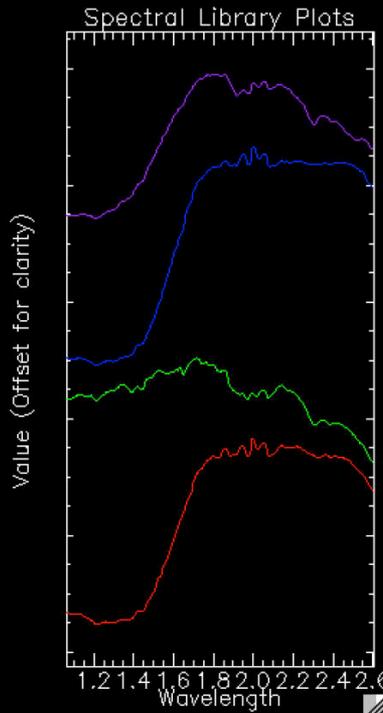
$$\rho_j = \mathbf{Q} \mathbf{a}_j + \mathcal{N}(0, \sigma^2)$$

n -vector of nonnegative mixing coefficients

Gaussian measurement noise



Superpixel endmember detection

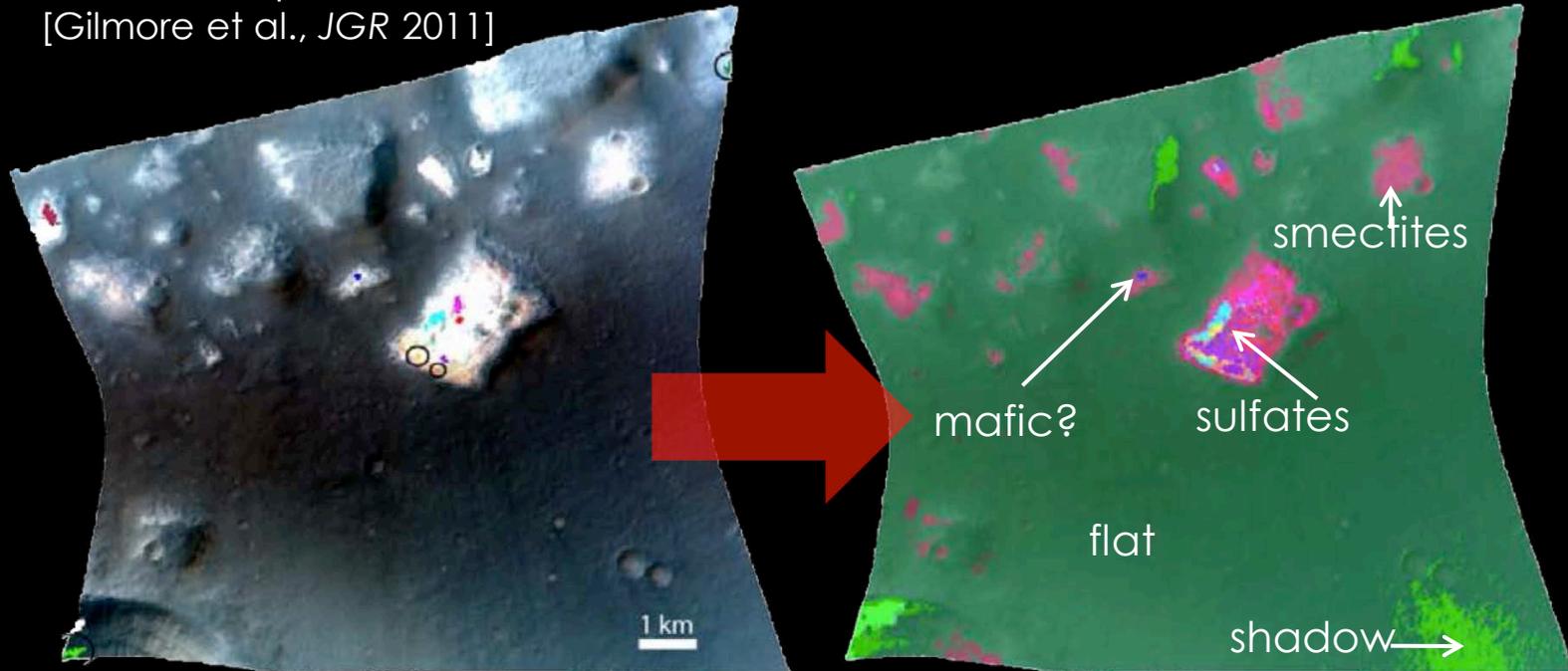


SMACC endmember
detection in CRISM 3e12
[Thompson et al., TGARS 2011]

Summary Classification

Spectral angle map for fast, fully automatic scene classification

CRISM interpretation
[Gilmore et al., *JGR* 2011]



Agenda

Why onboard
hyperspectral analysis?

Two tricks to make it
work

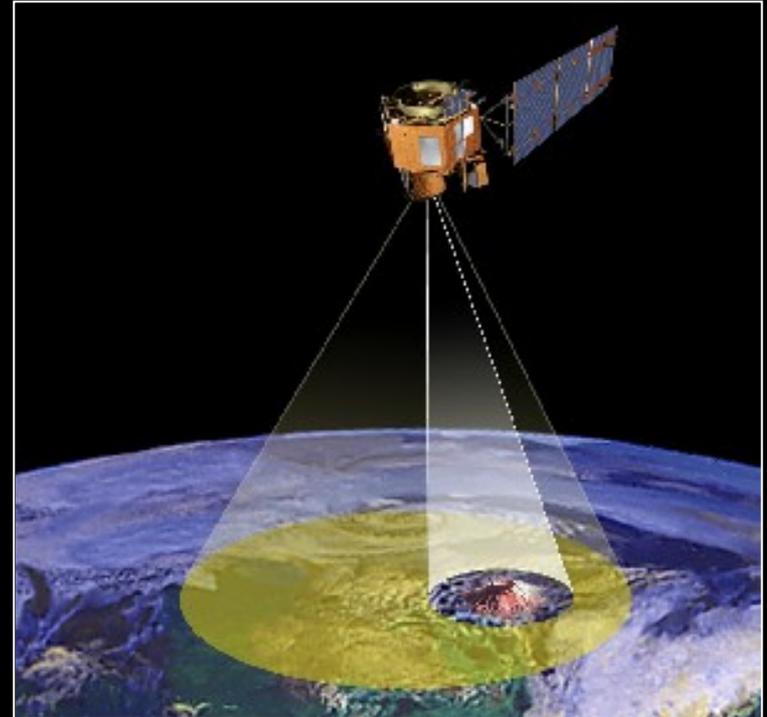
1. Superpixel
segmentation
2. Endmember
detection

EO-1 experiments



The EO-1 Spacecraft

- Currently in an “extended mission” phase
- Used in sensorweb and autonomous science operations since 2004 [Chien et al., 2005]
- Detects transient events such as floods and volcanoes
- Mongoose-V 32-bit microprocessor for onboard data analysis
 - 12MHz clock speed
 - No hardware floating-point arithmetic
 - Limited memory (16 MB application max)



EO-1 Selective downlink of volcanic activity “hot spot” in thermal imagery [Davies et al. 2005]

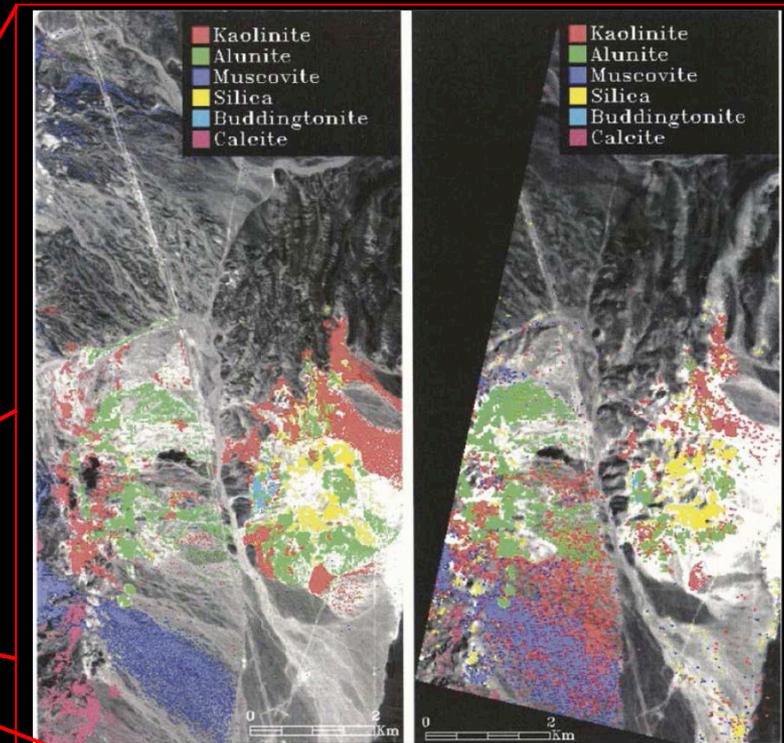
The Hyperion imaging spectrometer

- High resolution hyperspectral imager
- 220 spectral bands from 0.4 to 2.5 μm
- 30 meter spatial resolution, provides 7.5 x 100 km land area per image
- A reflectance product is available for onboard use
 - 12 bands selected in advance (once per observation)
 - 256x1024 pixel subframe



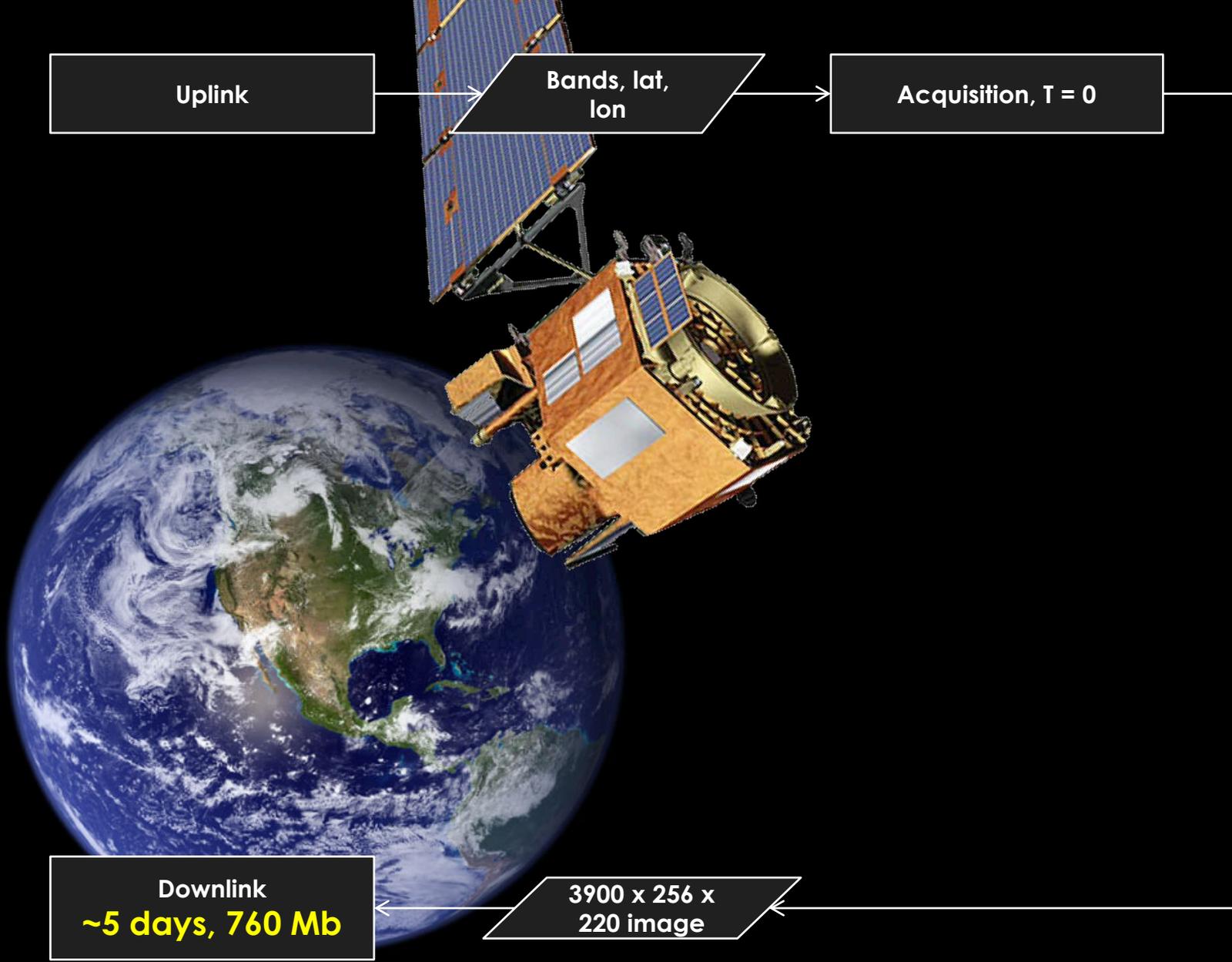
Hyperion view of Cuprite, NV

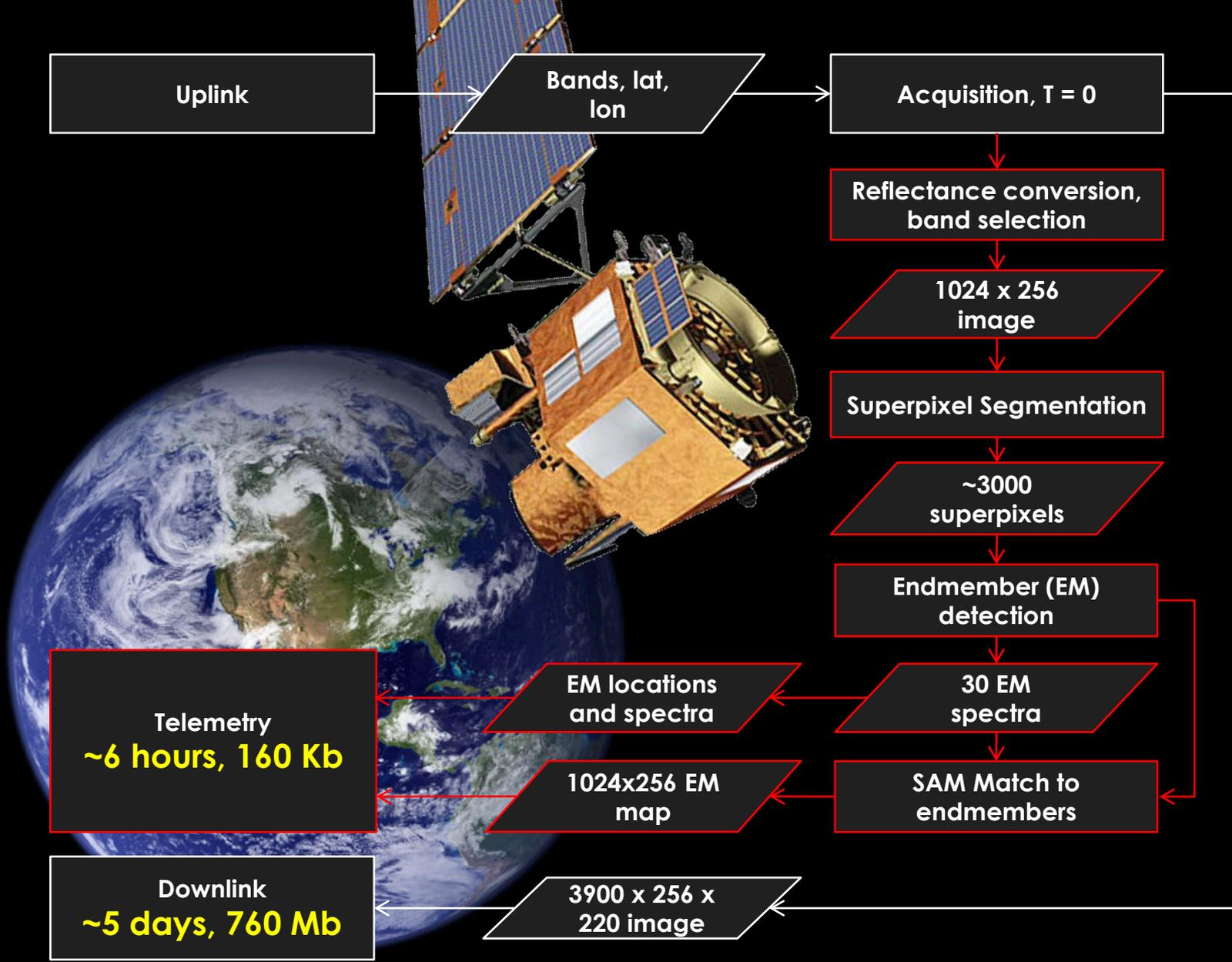
Gold Standard Cuprite maps
[Kruse et al., IEEE TGARS 2003]



AVIRIS

Hyperion





Cuprite NV, USA – Sept. 2011

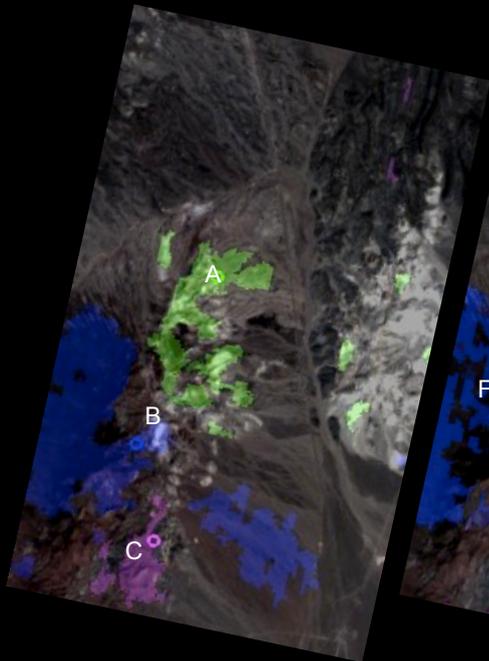


Credit: nvghosttowns.com

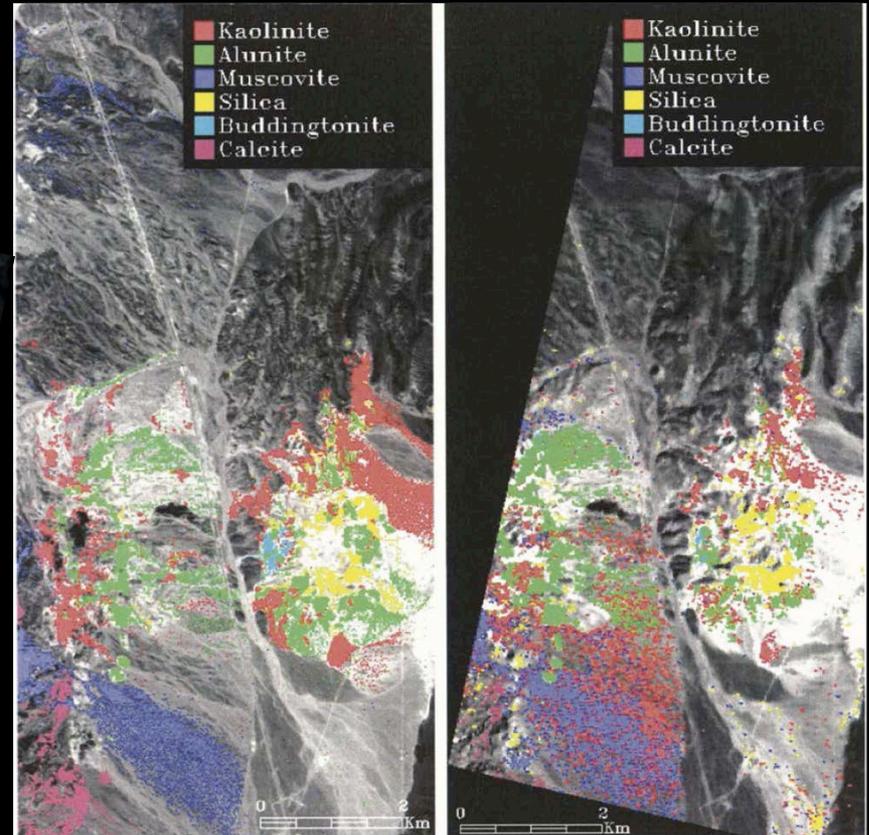
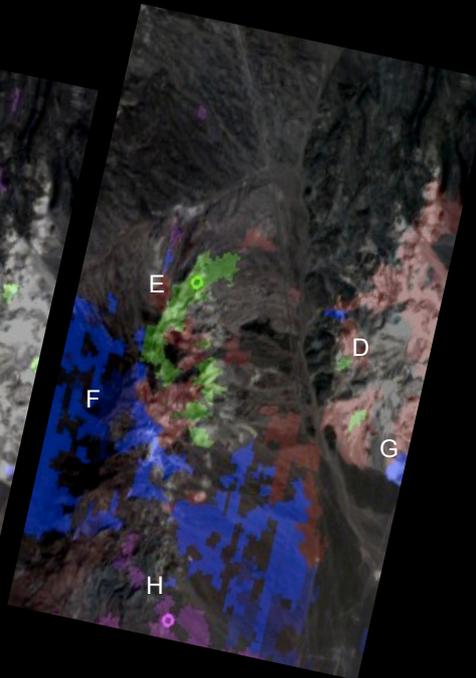
Cuprite NV, USA – Sept. 2011

Gold Standard Cuprite maps
[Kruse et al., IEEE TGARS 2003]

Sept. 19, 2011
Overflight



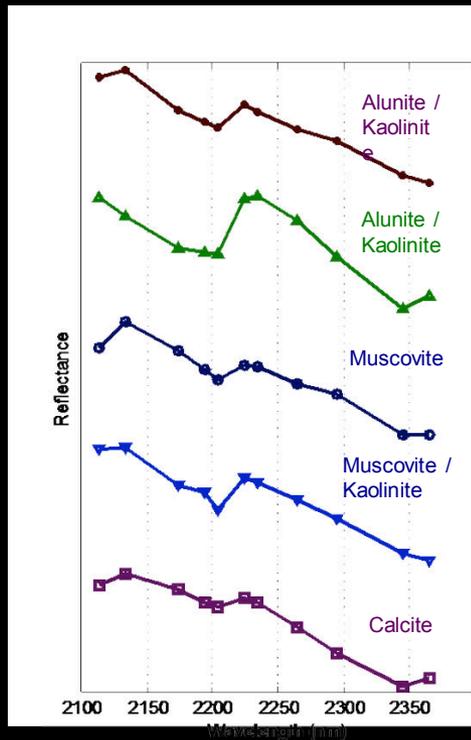
Sept. 27, 2011
Overflight



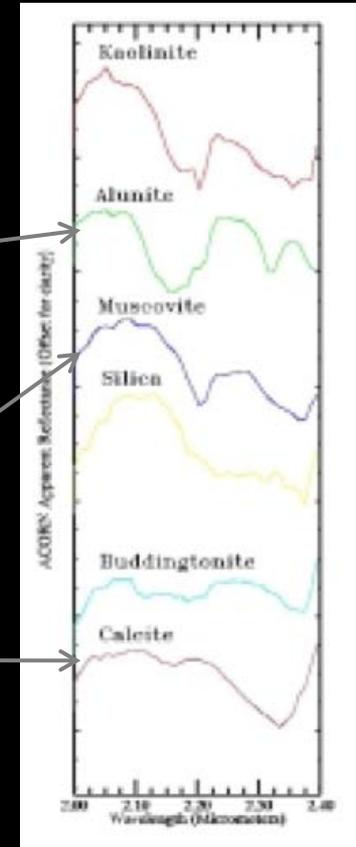
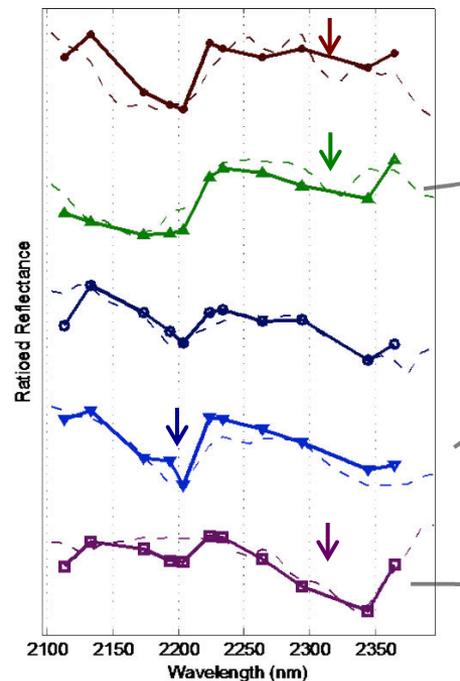
Cuprite NV, USA – Sept. 2011

Kruse et al. manual analysis (Hyperion)

Endmember spectra



Full spectrum



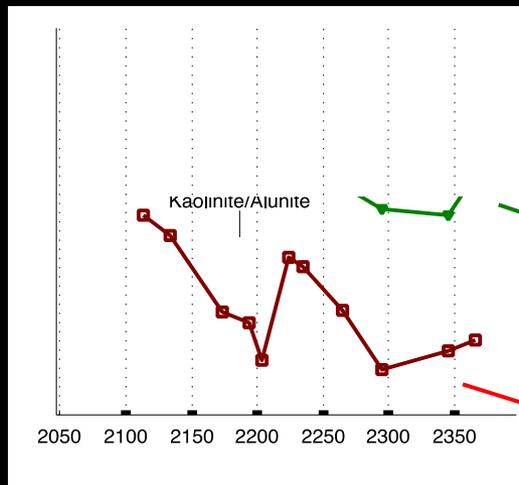
Steamboat Springs, USA – 3 Oct 2012



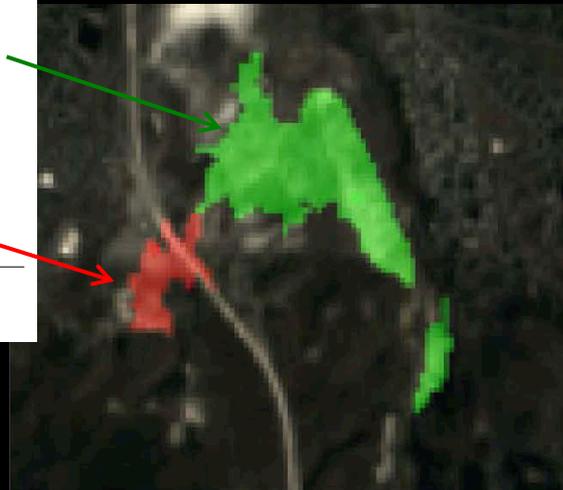
Credit: NREL

Steamboat Springs, USA – 3 Oct 2011

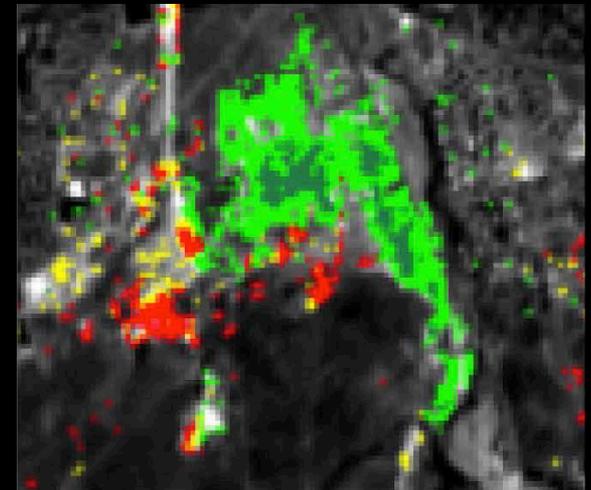
Endmember spectra



Compositional map generated onboard



Authoritative version of [Kruse et al., 2003]



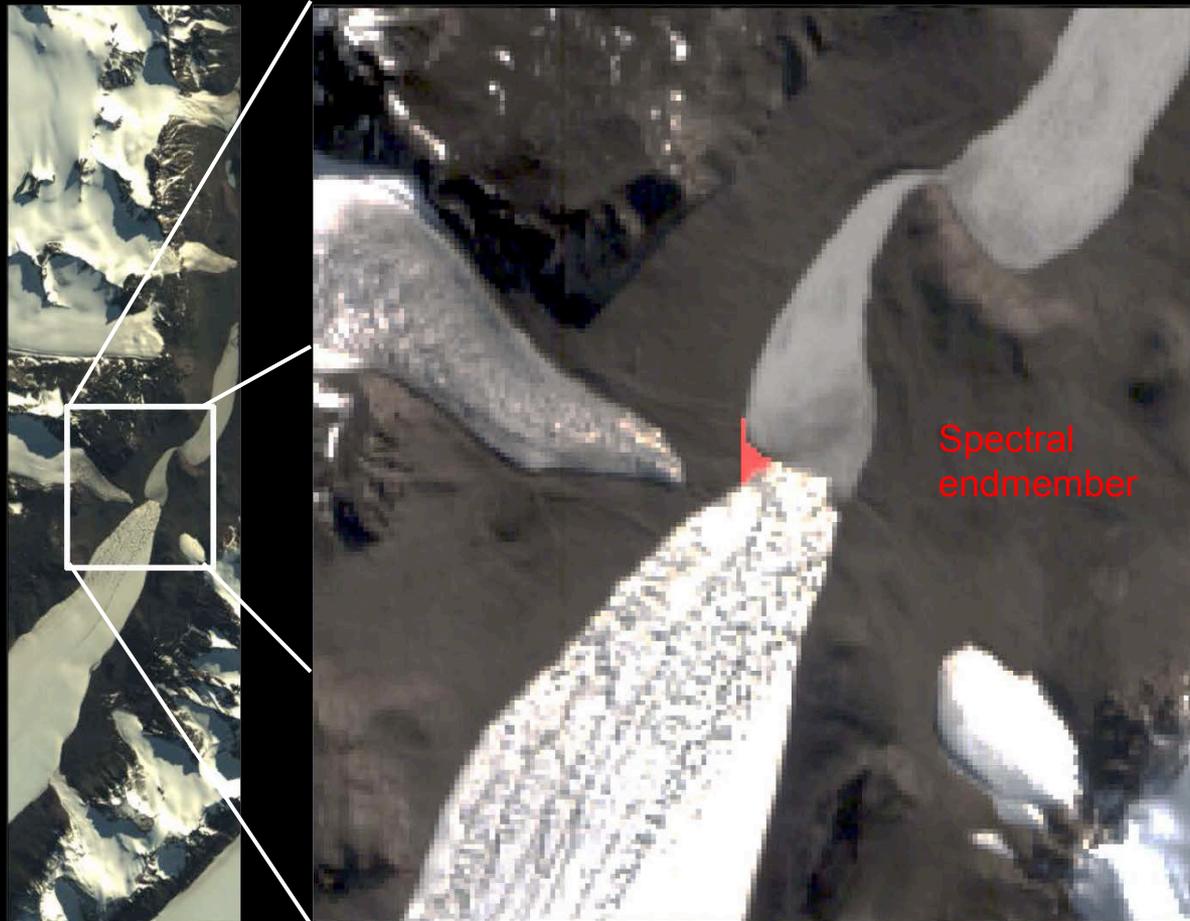
Blood Falls, Taylor Glacier – 7 Feb 2012



Credit: USAP

Blood Falls, Taylor Glacier – 7 Feb 2012

Hyperspectral image

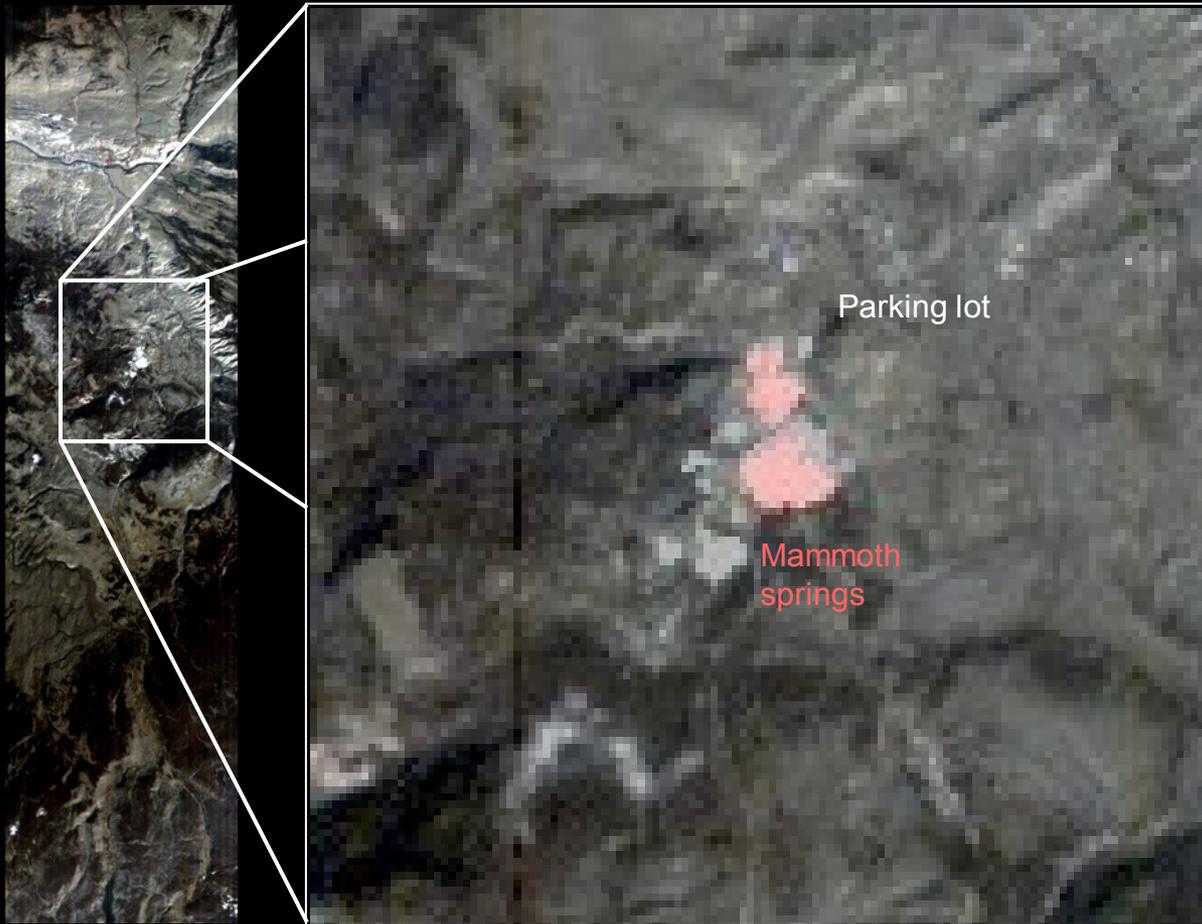


Mammoth Springs, MT - 20 Oct 2011



Mammoth Springs, MT - 20 Oct 2011

Hyperspectral image



Rio Tinto, Spain- 25 May 2012



Credit: Carol Stoker (AMES) / Wikicommons

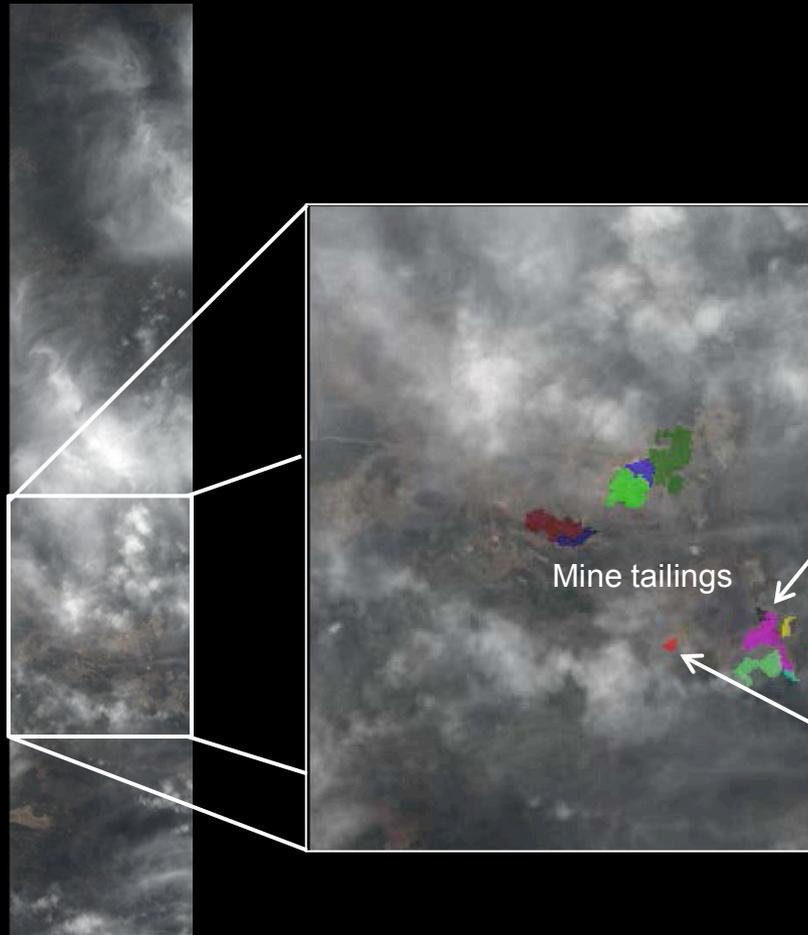
Rio Tinto, Spain- 25 May 2012



Clouded over?

Credit: Carol Stoker (AMES) / Wikicommons

Rio Tinto, Spain- 25 May 2012



Credit: Jesús Municio

Summary

- Automatic onboard mapping summarizes a 1024 x 256 x 220 scene using 20kB
- Runs onboard, using a fraction of a 12MHz processor
- Requires <16MB of volatile memory
- Identifies pure features and returns exemplar spectra
- Operating regularly on EO-1

Future directions

Autonomy can be enabling for hyperspectral imagers

More sophisticated detection algorithms are possible

- Follow up on signatures in a library
- Follow up on signatures **not in** a library
- Linear unmixing

Multi-core architectures and FPGAs can provide faster, full-spectrum analyses

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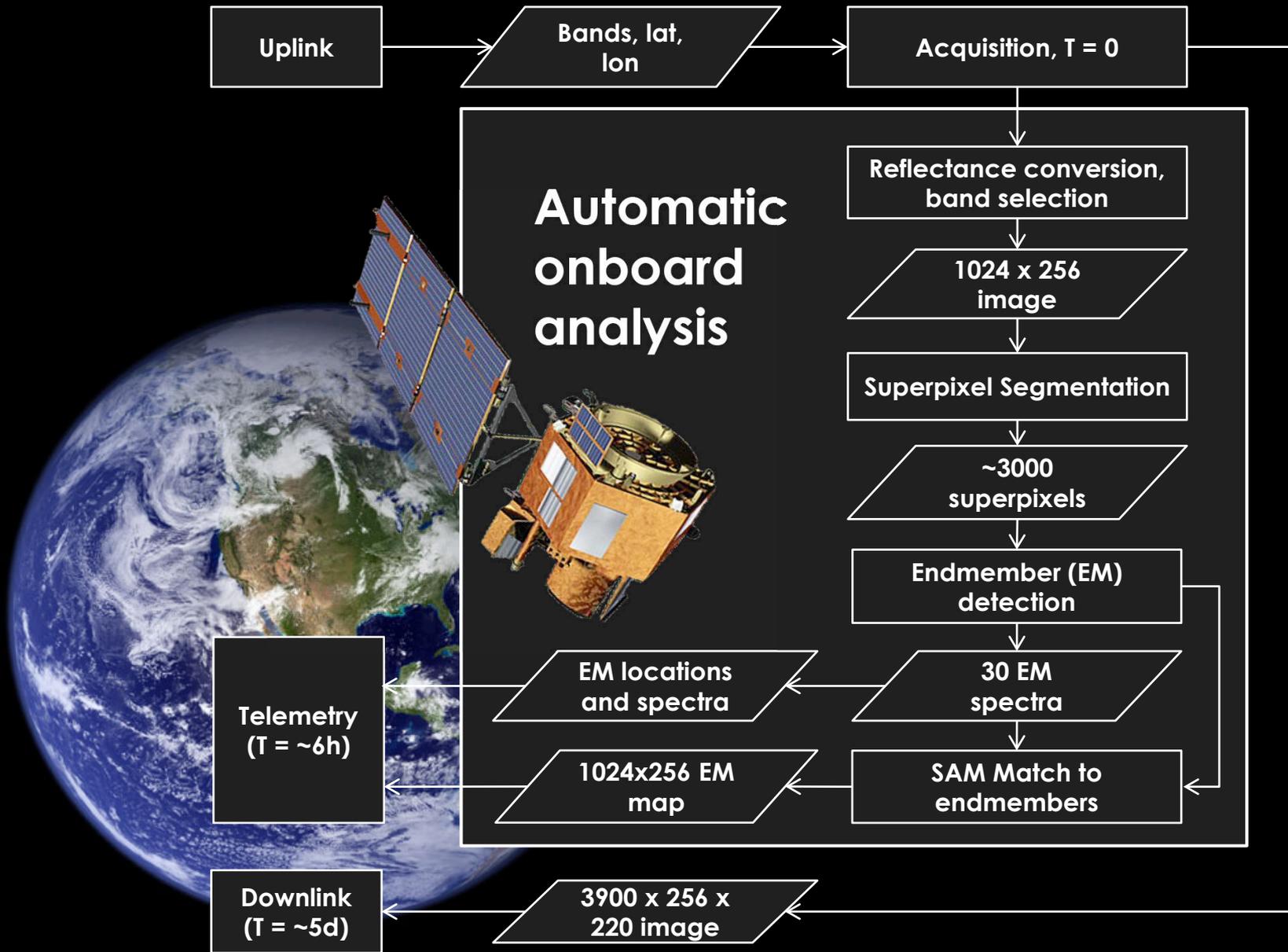
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Jesús Municio por el uso de sus fotos de Río Tinto

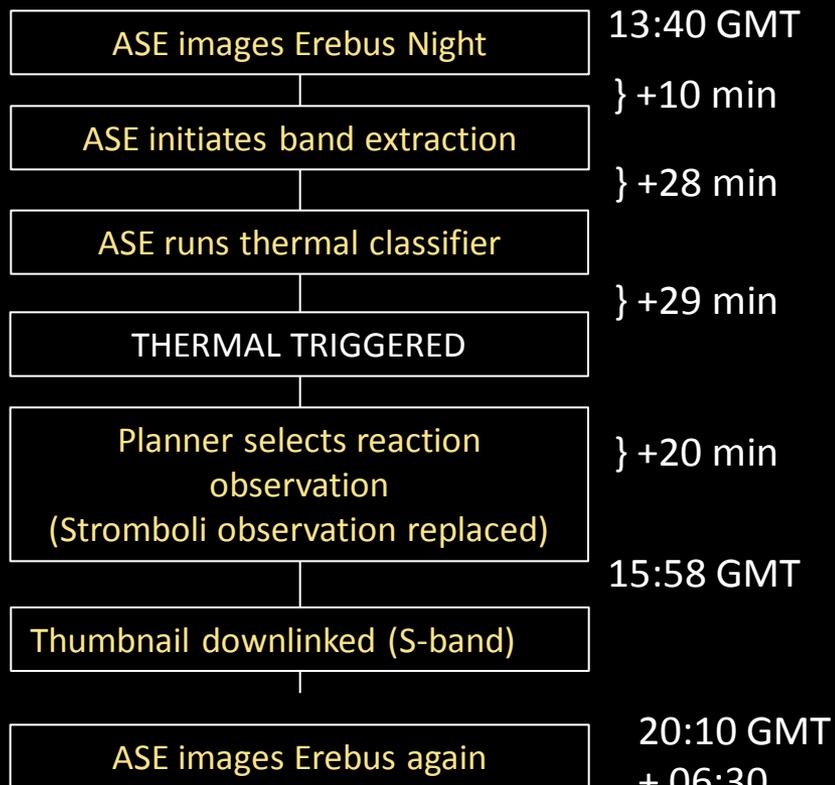
The research was carried out at the Jet Propulsion Laboratory, California Institute of Technology, under a contract with the National Aeronautics and Space Administration (NASA). Copyright 2012. All Rights Reserved. U.S. government support acknowledged.

Extra slides

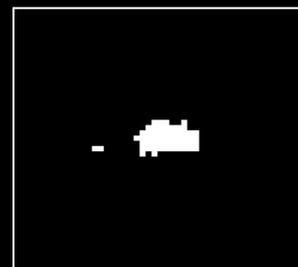


EO-1 Thermal signature detection

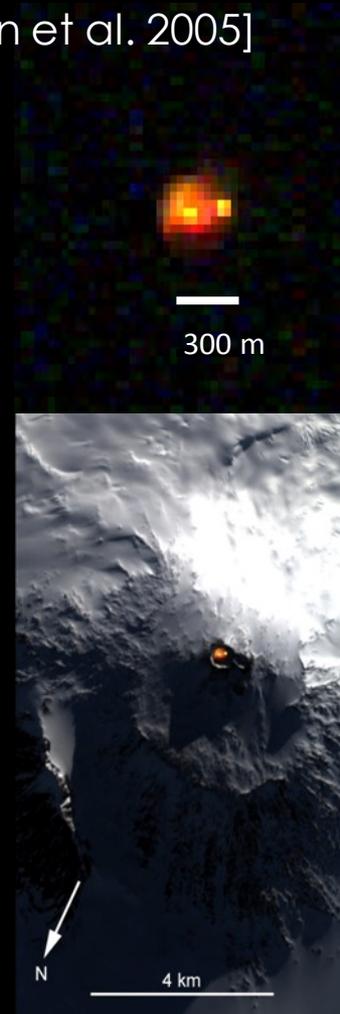
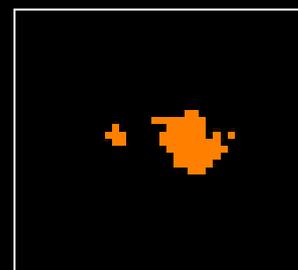
Black Body model used to trigger a second observation [Chien et al. 2005]



ASE Onboard Thermal Classifier Thumbnail



ASE Onboard Thermal Classifier

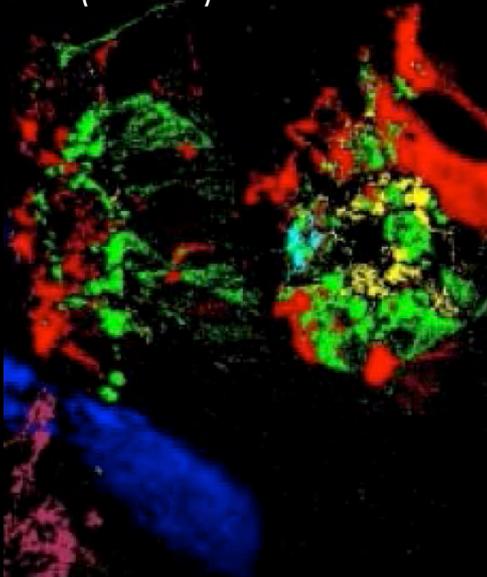


Cuprite NV

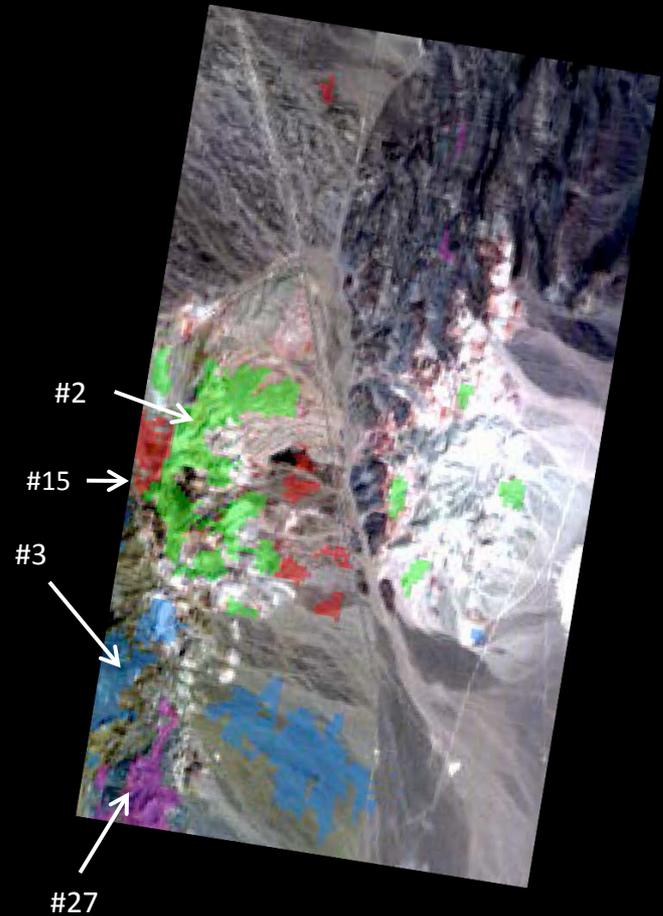
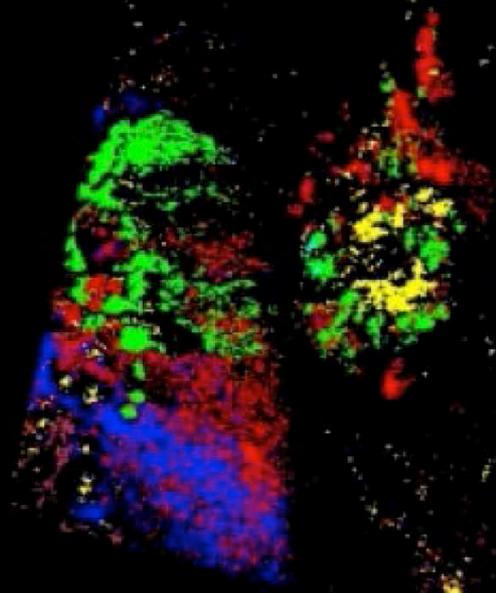
- Kaolinite
- Alunite
- Muscovite
- Silica
- Buddingtonite
- Cacite

EO-1 Onboard

Kruse et al.
(AVIRIS)

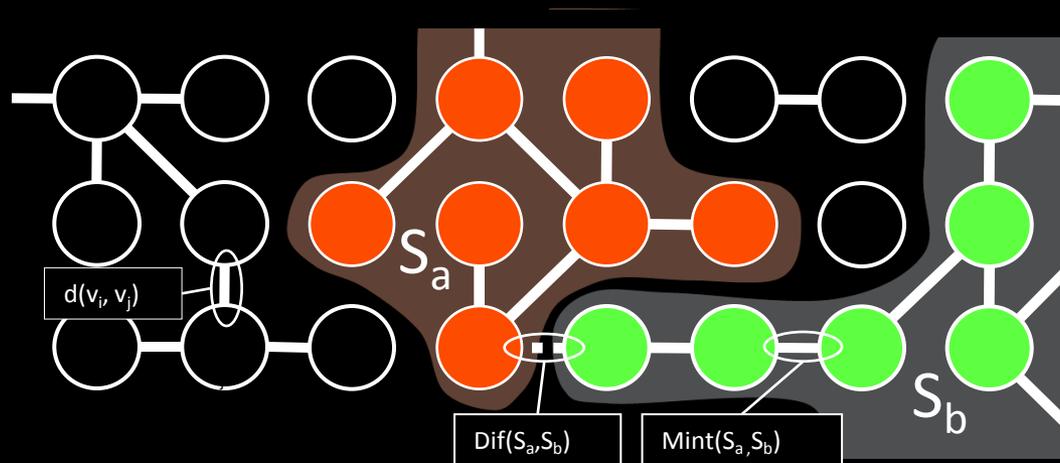


Kruse et al. (Hyperion)



Superpixels via graph partitioning

- Posit spectral distances $d(v_i, v_j)$ between neighbors
- Agglomerate minimum spanning trees [Felzenswalb 2004]
- Merge based on largest internal distance

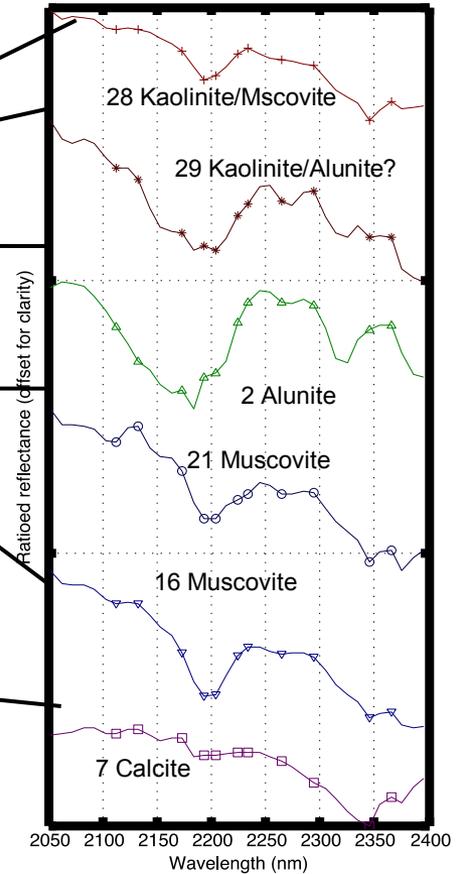
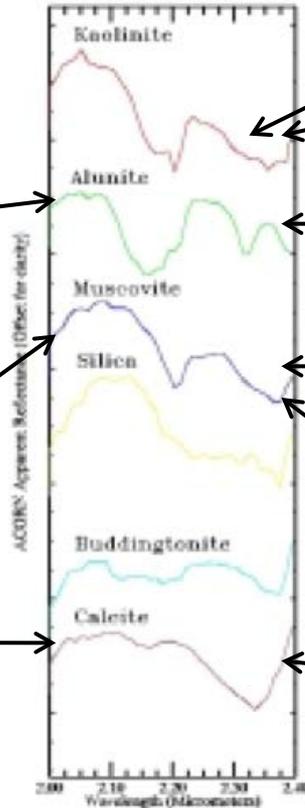
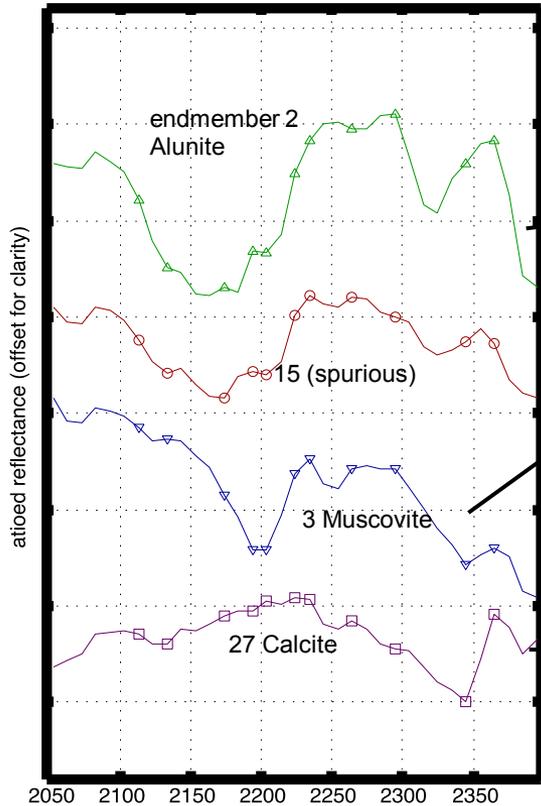


$$Dif(S_a, S_b) > MInt(S_a, S_b) = \min \left(\text{Int}(S_a) + \frac{k}{|S_a|}, \text{Int}(S_b) + \frac{k}{|S_b|} \right)$$

EO-1 Onboard
Sept. 21, 2011

Kruse et al. manual
analysis (Hyperion)

EO-1 Onboard
Sept. 27, 2011



Kruse et al.