

Integrating the past, present, and future of Landsat: An update

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Roy et al. L8 Submission

- Landsat occupies a valuable niche in global satellite remote sensing; the platform collects images at a spatial resolution indicative of human interactions with terrestrial ecosystems yet with an image extent compatible with systematic analyses over large areas. A ground-segment based on reliable science provides free and open access to Level 1 products of known, documented, and traceable quality. The integrity of this production chain and the confidence users have in Landsat data has led to a proliferation of science and applications – the Landsat program provides an example to be emulated.

Context and needs:

- Context:
 - Inventory, monitoring,
 - National and international reporting
 - National Forest Inventory, Carbon Accounting Programs
 - Basic data:
 - Land cover
 - Land cover change, events, types
 - Forest structure
- At the national level for multiple time periods in a scientifically robust, transparent, and repeatable manner**

How can we get there?

- Field plots
 - Detailed measures, limited number and distribution
- Lidar plots, via transects
 - Detailed measures, limited number and distribution, but less so than field plots
- Fine resolution optical satellite data (Landsat)
 - Wall-to-wall and dense time series
 - compositing

Messages

- Image compositing
 - Types
 - Options / implications
- Information needs drive compositing decisions
 - Composites to support land cover, change in cover, and structural attribution
- International options differ from CONUS
- Science and product implications

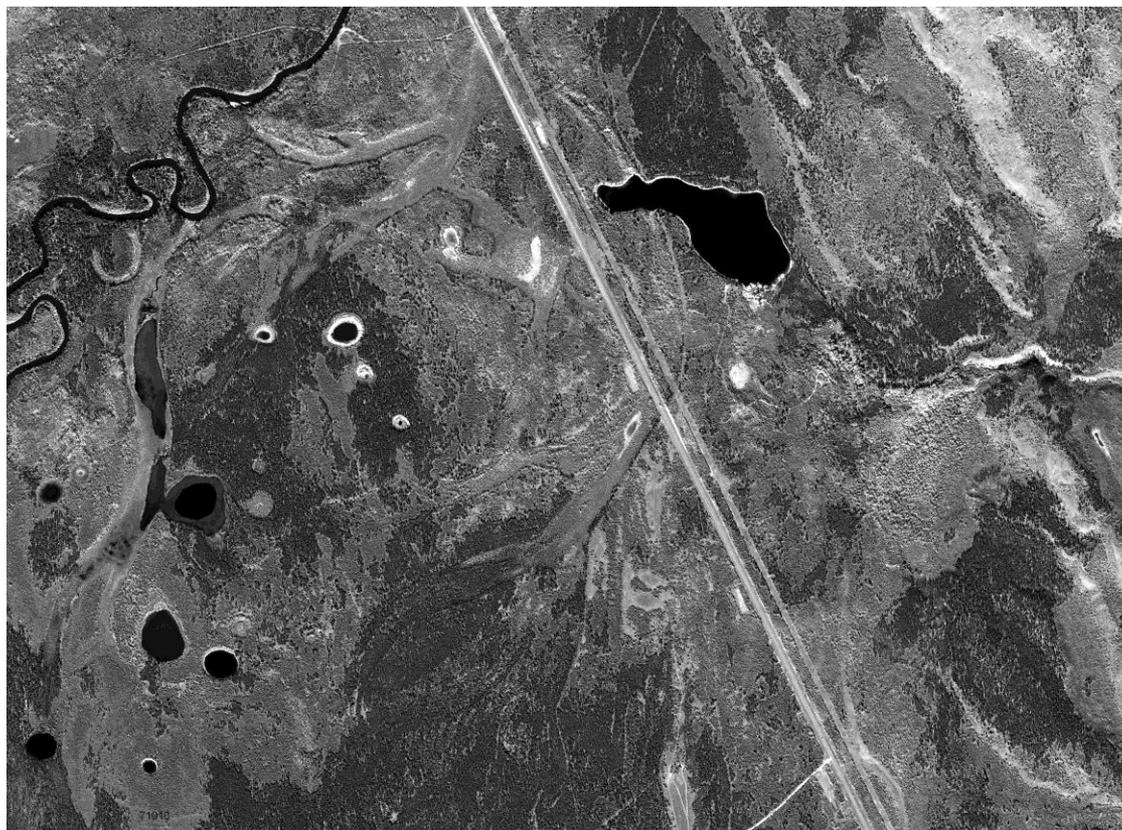
Part 1. Structure

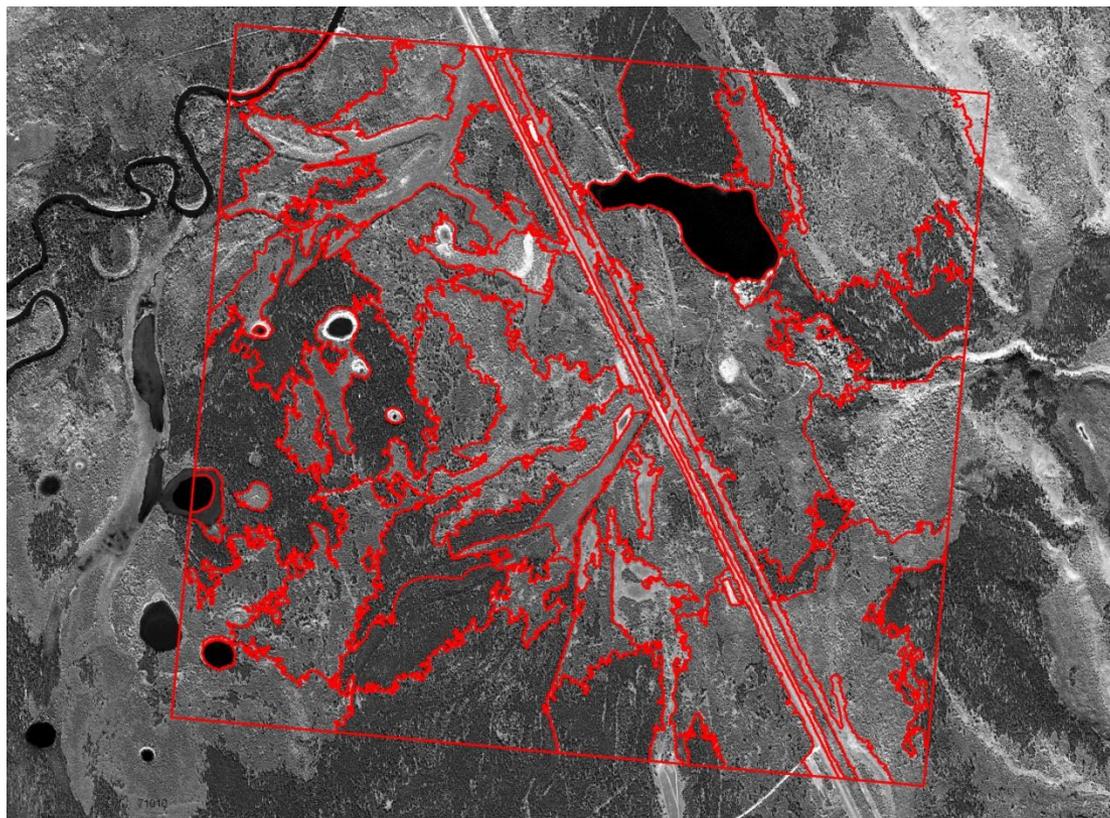
- Or, context for Landsat estimates of forest structure in the boreal

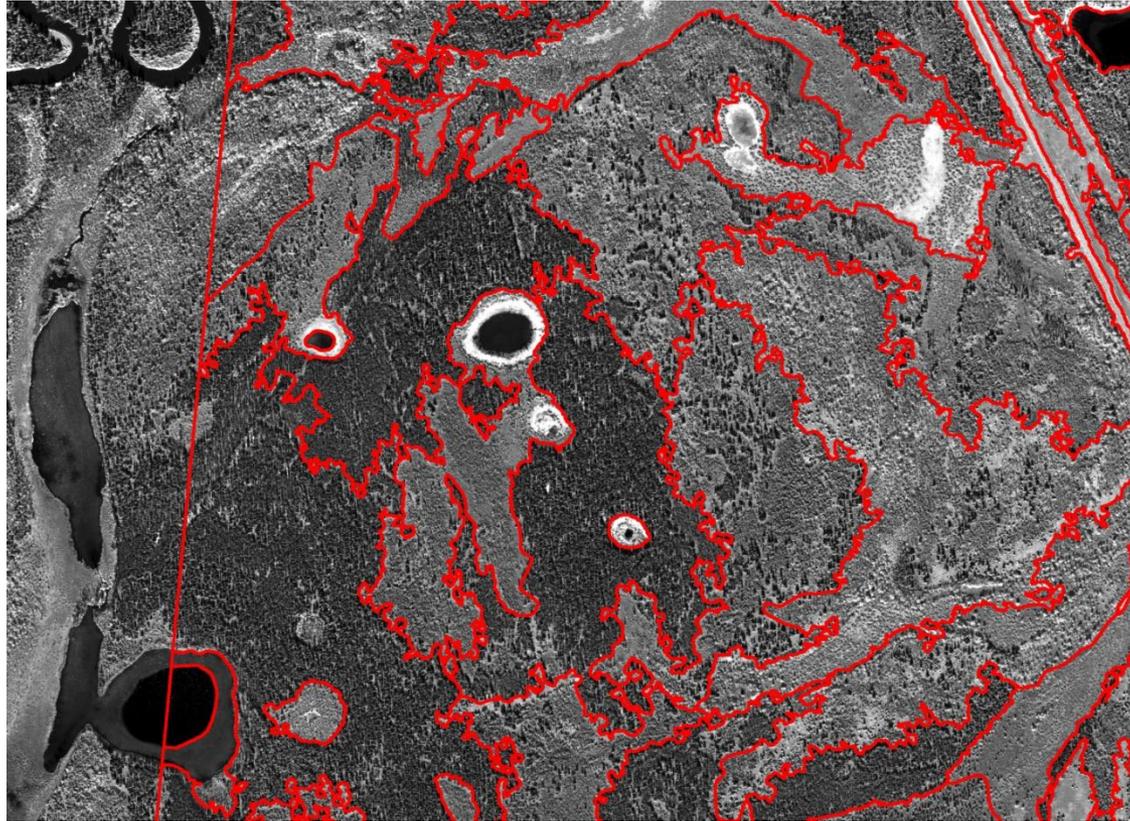
Effort:

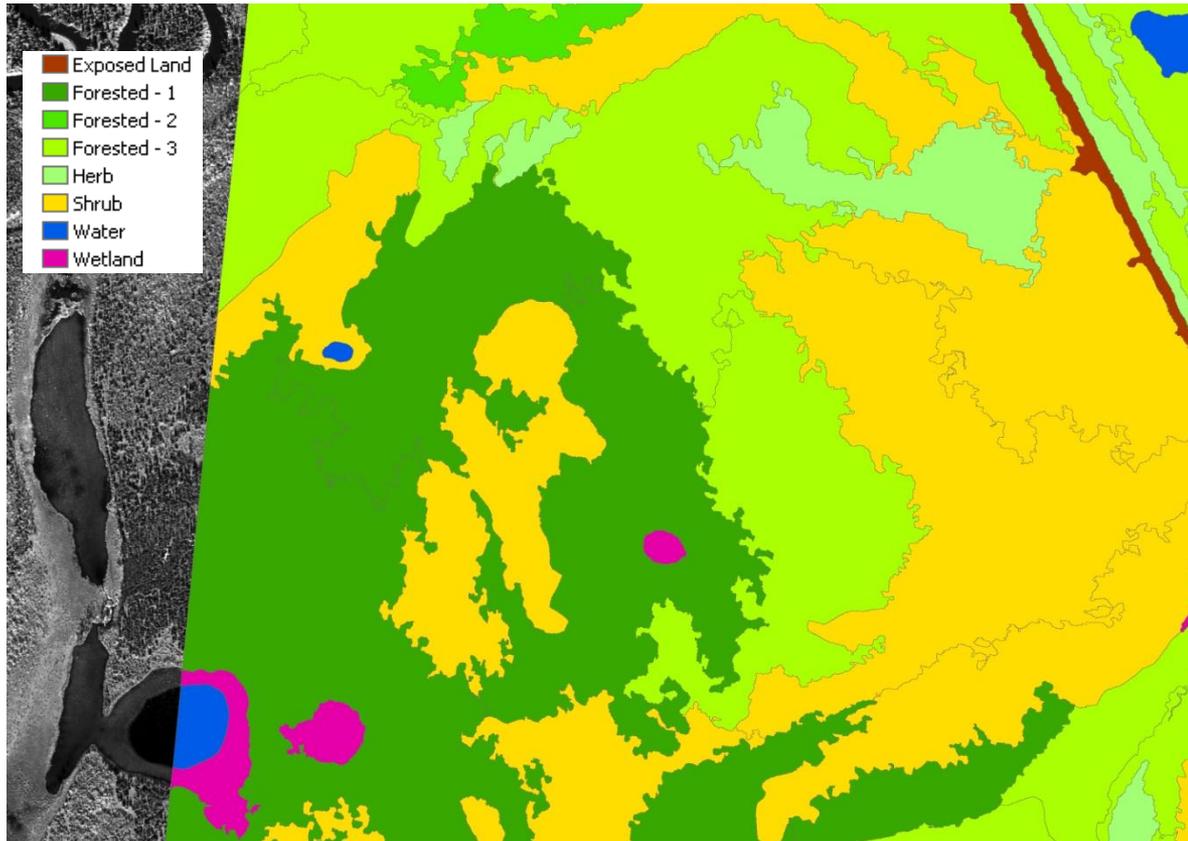
Spatial resolution example

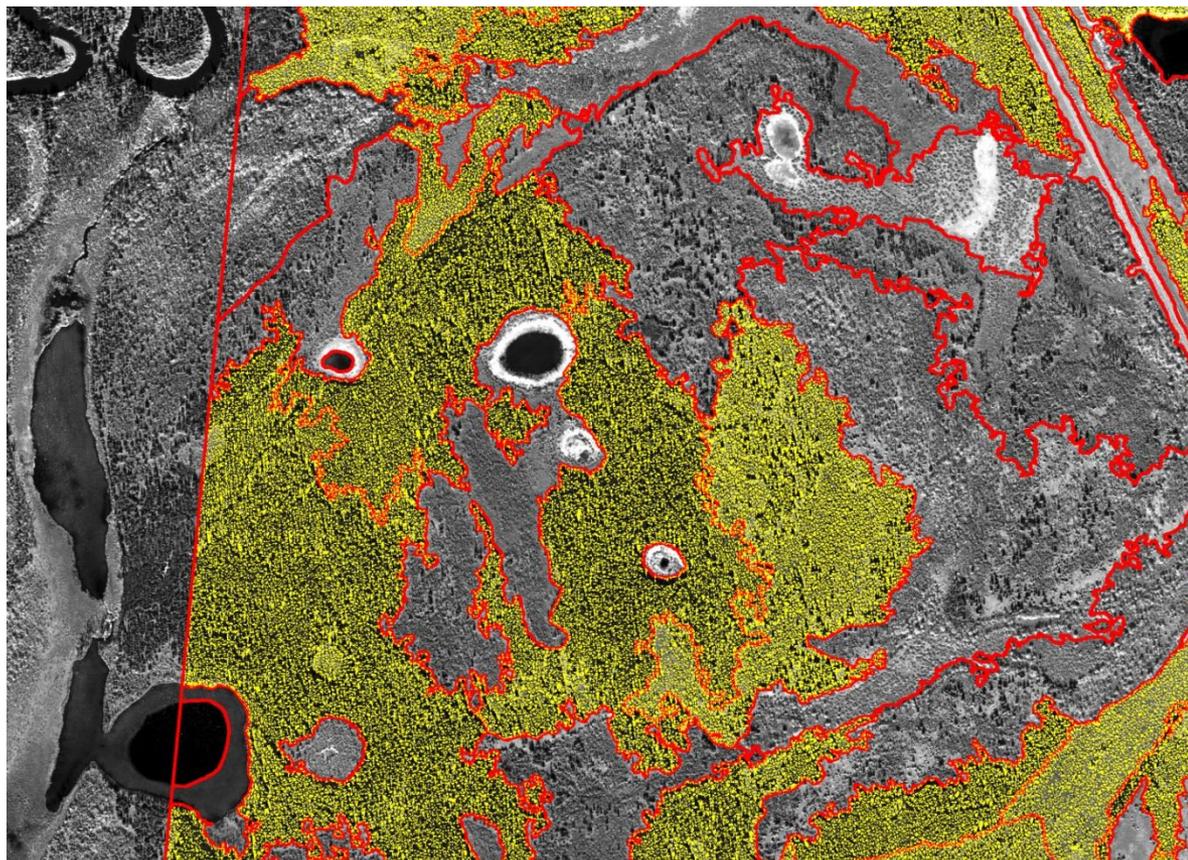
- Goal: stand level forest inventory information
- Support regional and national inventories
- Very high spatial resolution satellite imagery (>1m panchromatic)

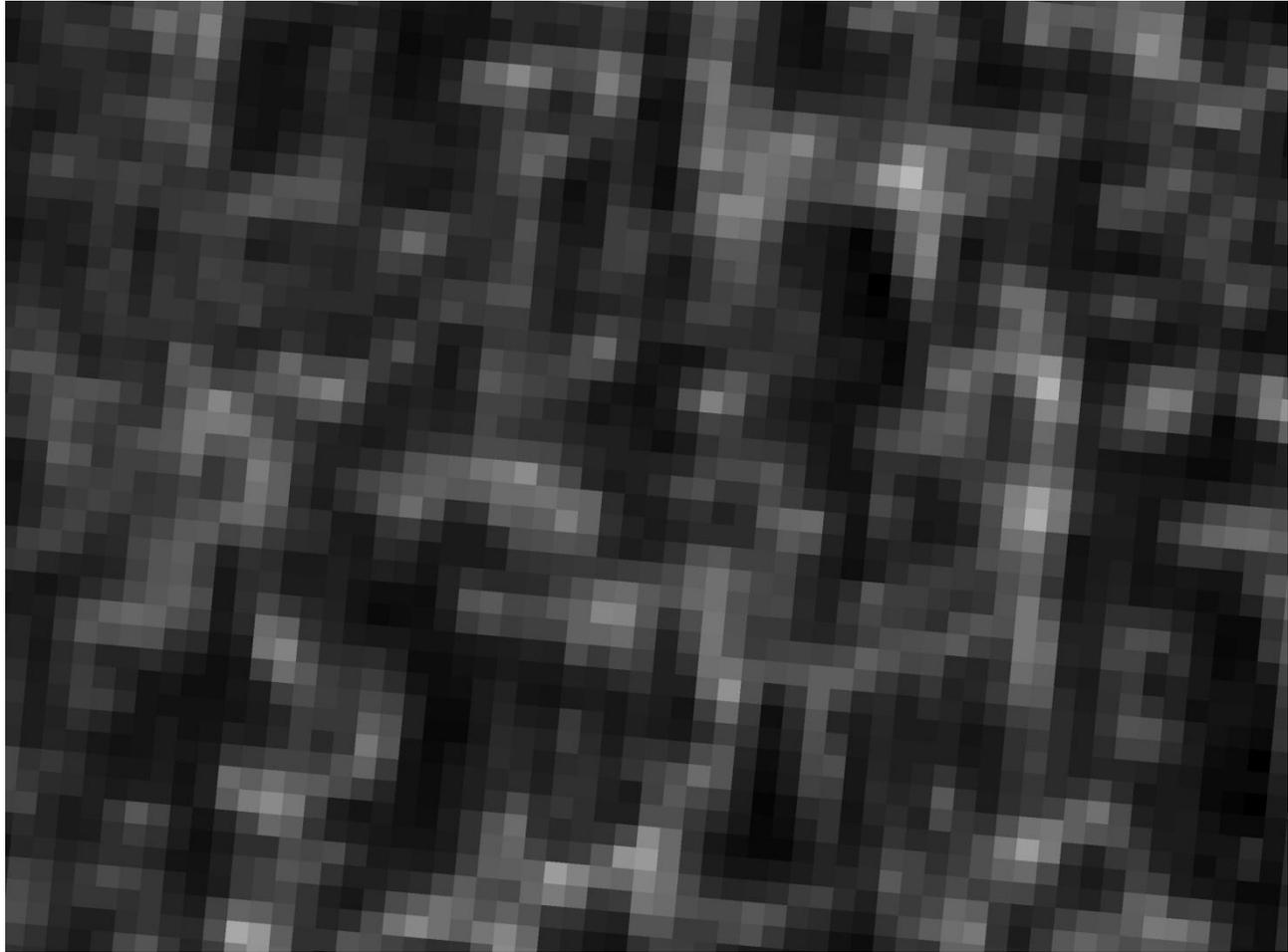


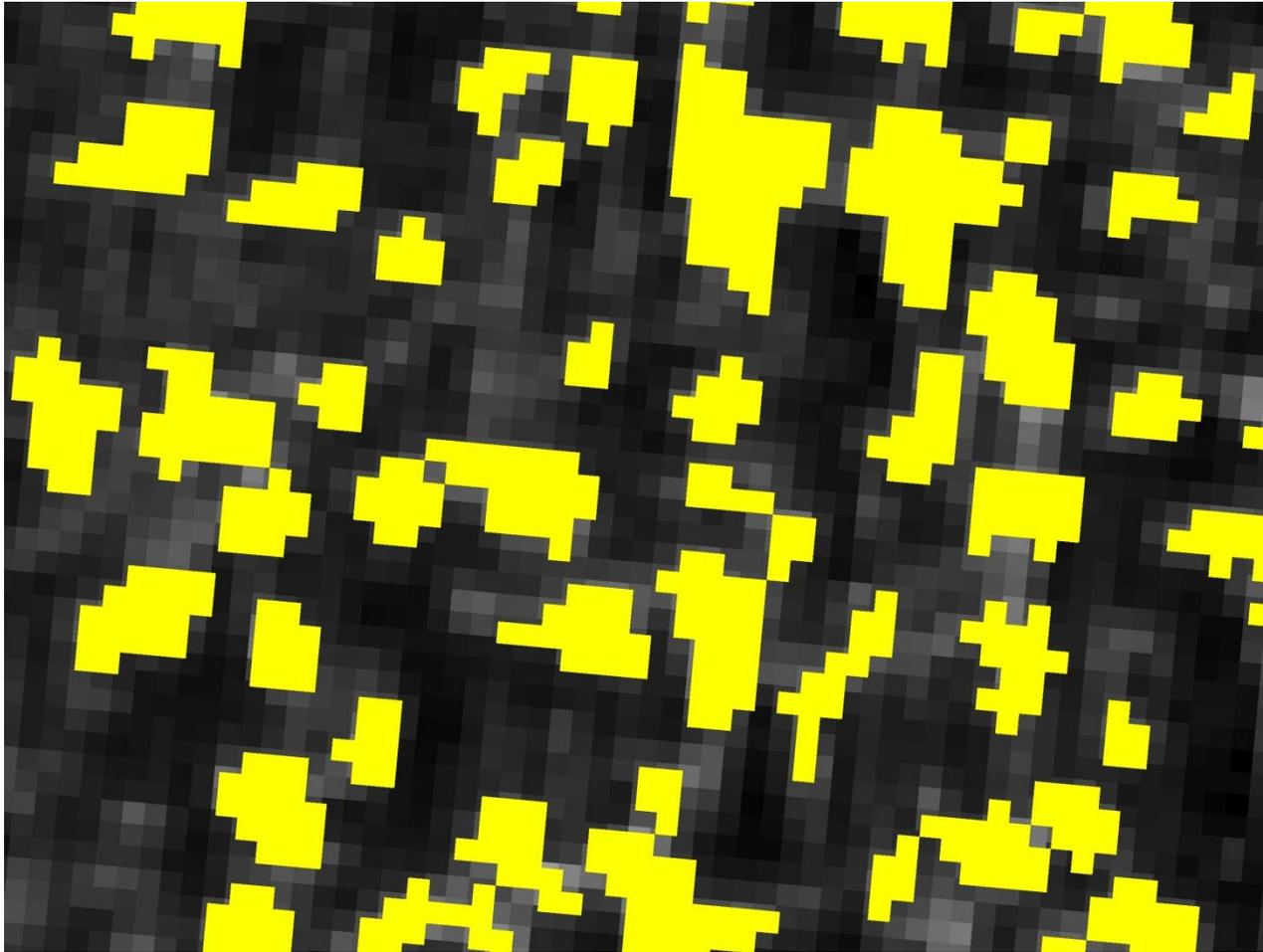












Gougeon, F. (1995). A crown-following approach to the automatic delineation of individual tree crowns in high spatial resolution aerial images. *Canadian Journal of Remote Sensing*. 21(3):274-284.

Stand and crown metrics

Stand segment metrics

Photo plot ID: 4

Stand ID: 94

Area: 14.85 ha

Stand type: Conifer

Leading species: Lodgepole pine

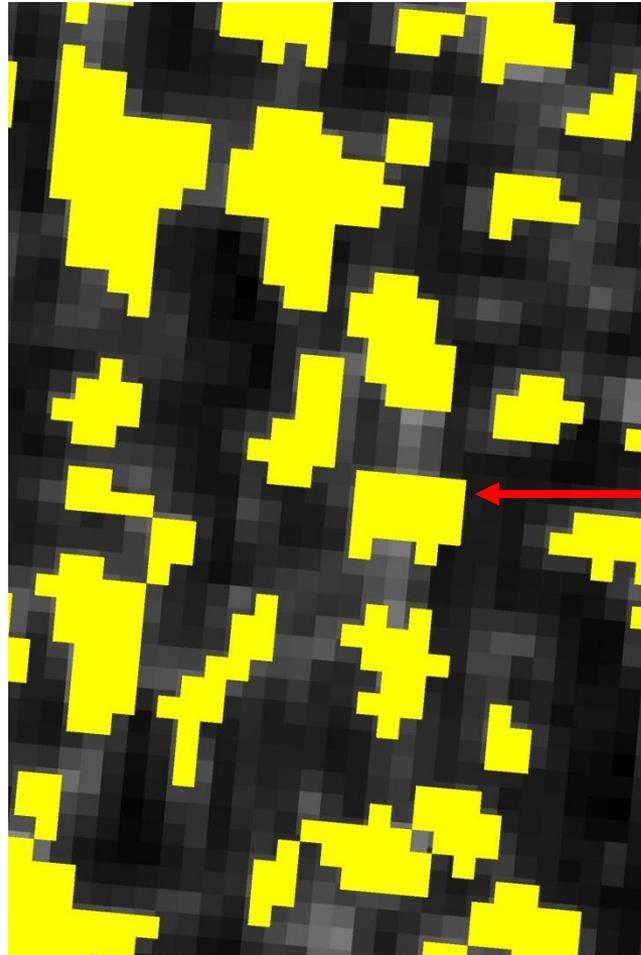
Mean crown area: 17.8 m²

25th percentile: 9 m²

50th percentile: 14 m²

75th percentile: 23 m²

Crown closure: 37%



Result:

Stand height with
RMSE of **2.84 m**

Crown metrics

Crown ID: 1385

Length: 3.5 m

Area: 6.8 m²

40th percentile

Compare to studies using HSR imagery with ground plots or LiDAR samples

- Peuhkurinen et al. (2008) IKONOS and ground plots
 - mean stand height, RMSE = **3.1 m**
- Hilker et al. (2008) QuickBird and LiDAR
 - mean stand height, 2 ha minimum stand size
 - underestimated stand height by an average of **3.5 m**
- Chen et al. (2012) QuickBird and LiDAR
 - mean plot height, plot size = 0.04 ha
 - RMSE for *best* model was **3.37 m**
- Mora et al. (2013). QuickBird and LiDAR
 - mean stand height, RMSE = **2.3 m**

Compare to other studies using Landsat and LiDAR

- Wulder and Seemann (2003)
 - mean stand height, RMSE = **3.3 m**
- Pascual et al. (2011)
 - mean stand height, RMSE = **1.9 – 2.3 m**
- Maselli et al. (2011)
 - mean stand height, RMSE = **3.01 m**
- Varhola and Coops (2013)
 - mean plot height (plot size = 0.25 ha), RMSE = **3.24 m**

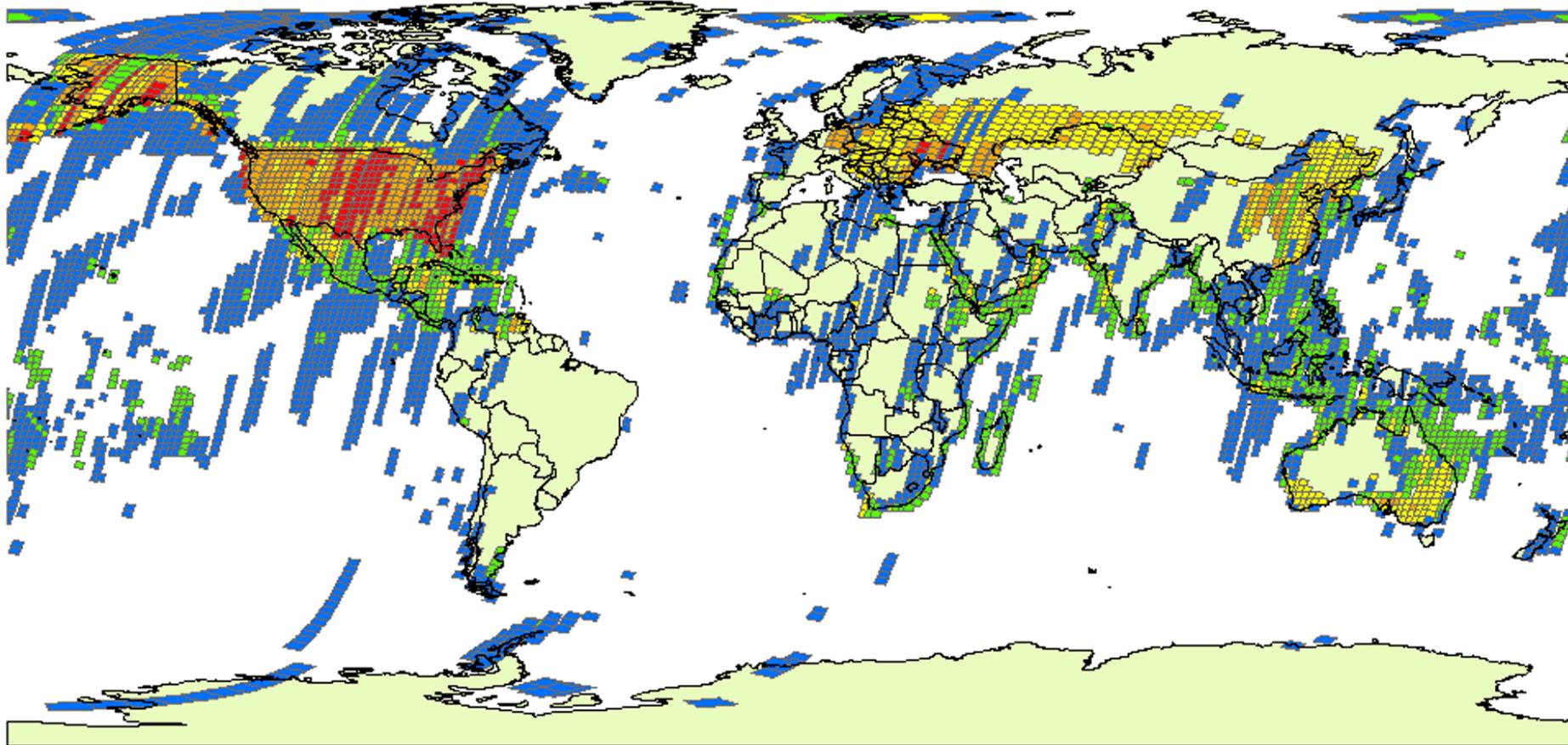
What have we learned?

- Similar RMSEs are found when using either field plots or lidar plots
- Similar RMSEs are found when using very high spatial resolution imagery or Landsat
- Value in using lidar plots to provide a well distributed set of plot-like data to calibrate and validate models of forest structure

Part 2. Compositing

- Or, when the best available pixel is not good enough...

MSS L3 1978



MSS L3 1978

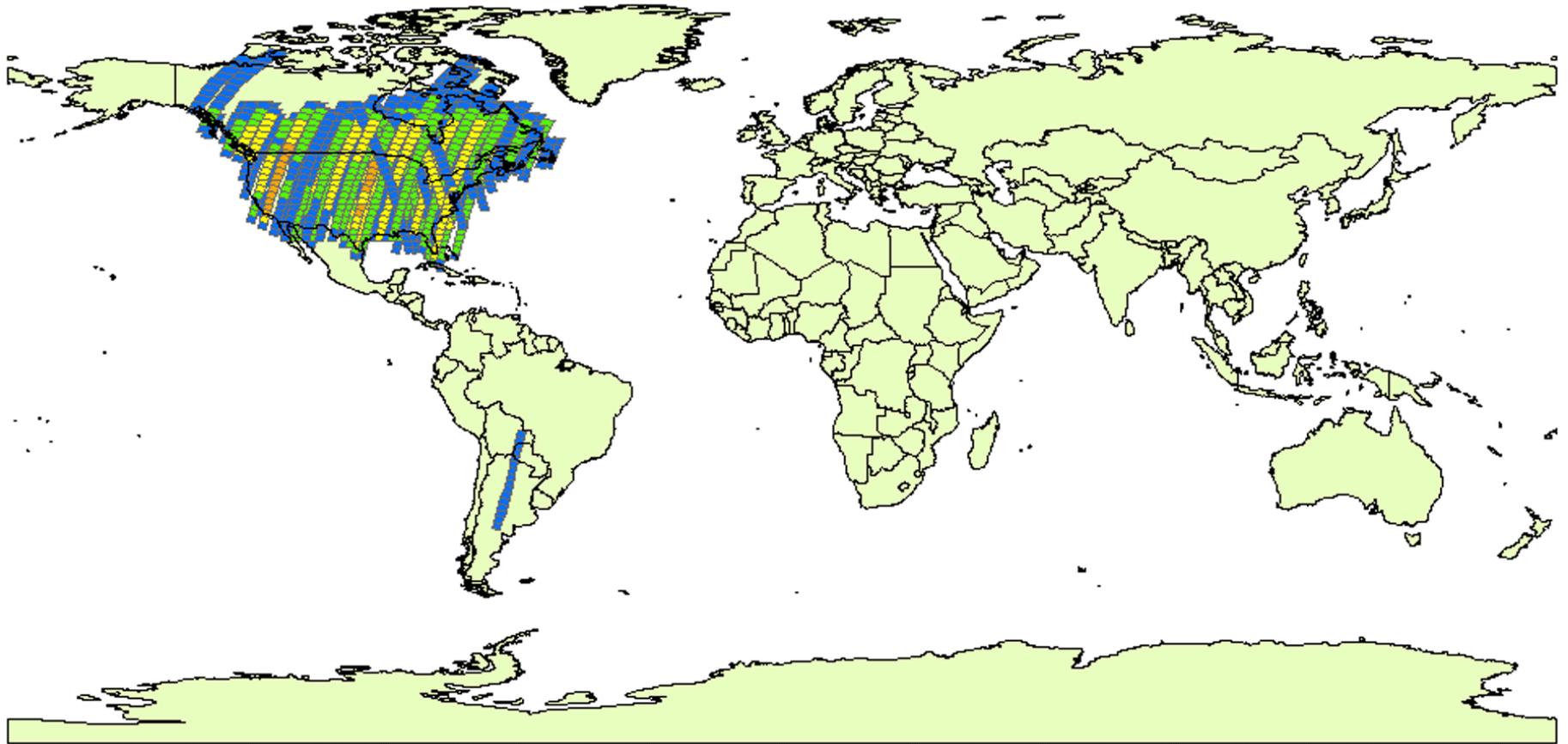
Total Marketable Scenes: 22,315

Unique Marketable Locations: 5,819

1-2 3-5 6-9 10-12 13-16

http://landsat.usgs.gov/documents/StateOfTheArchive_web.ppt

TM L4 1983



TM L4 1983

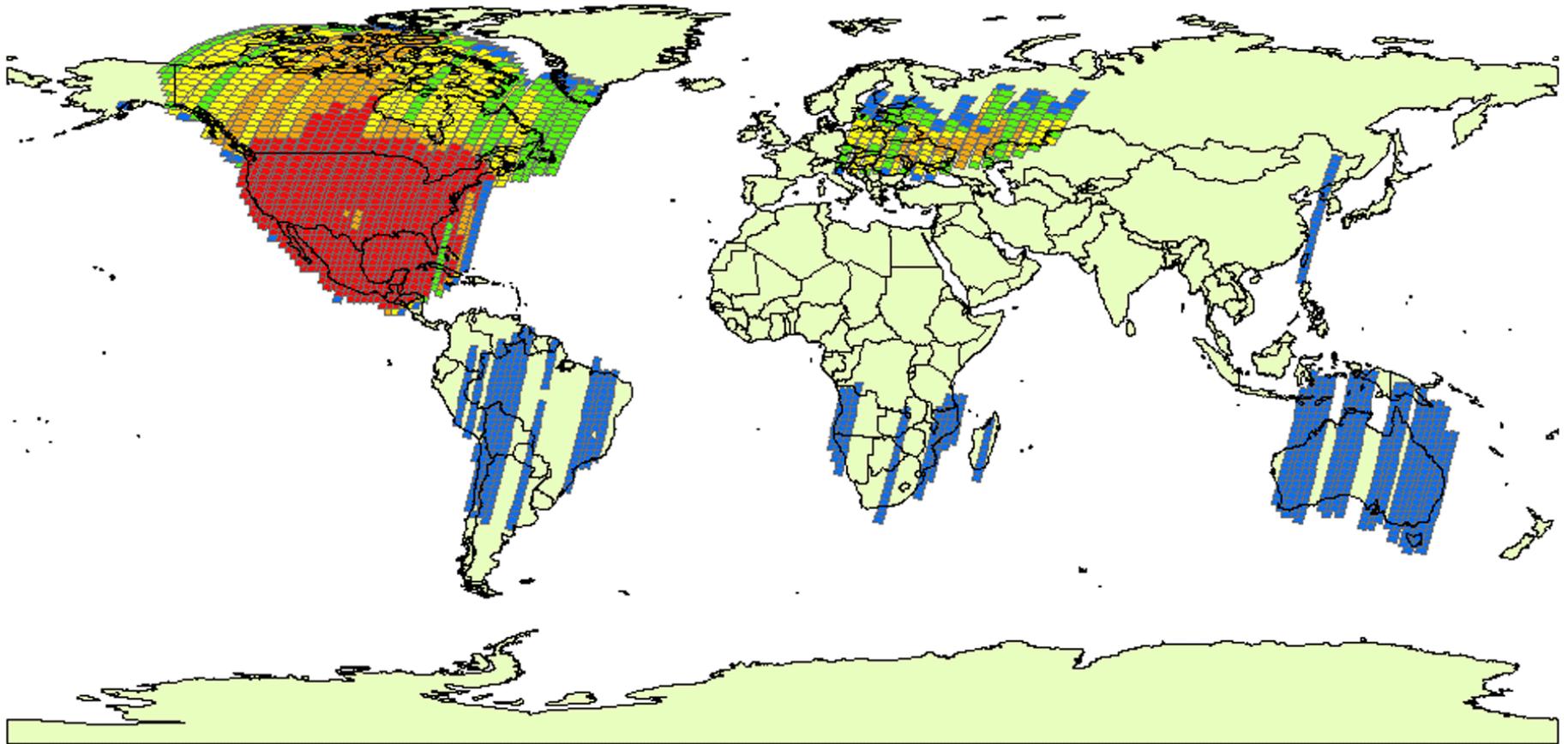
Total Marketable Scenes: 2,126

Unique Marketable Locations: 1,247



http://landsat.usgs.gov/documents/StateOfTheArchive_web.ppt

TM L5 1993



TM L5 1993

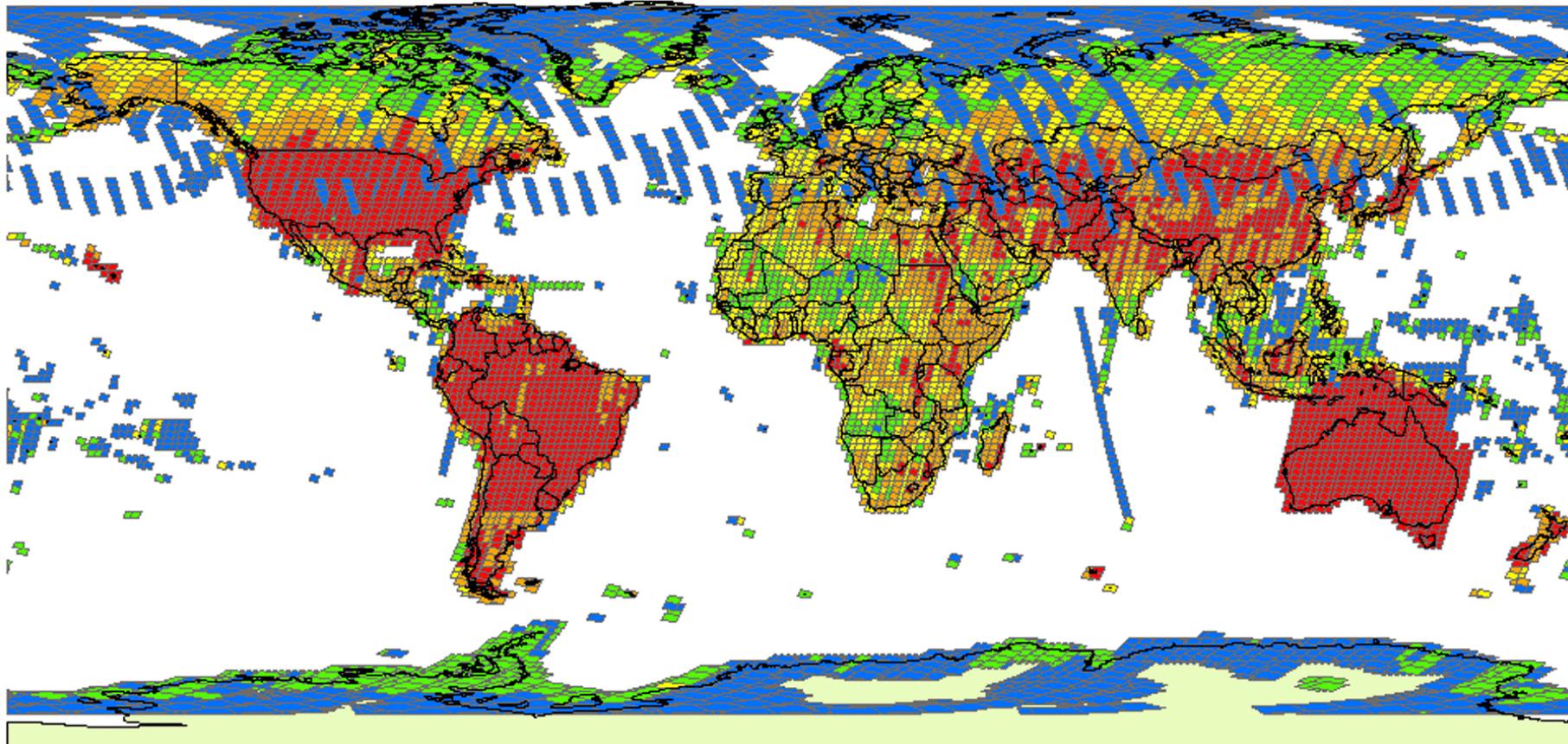
Total Marketable Scenes: 32,479

Unique Marketable Locations: 2,987

1 - 4 5 - 9 10 - 12 13 - 17 18 - 23

http://landsat.usgs.gov/documents/StateOfTheArchive_web.ppt

ETM+ L7 2012



ETM+ L7 2012

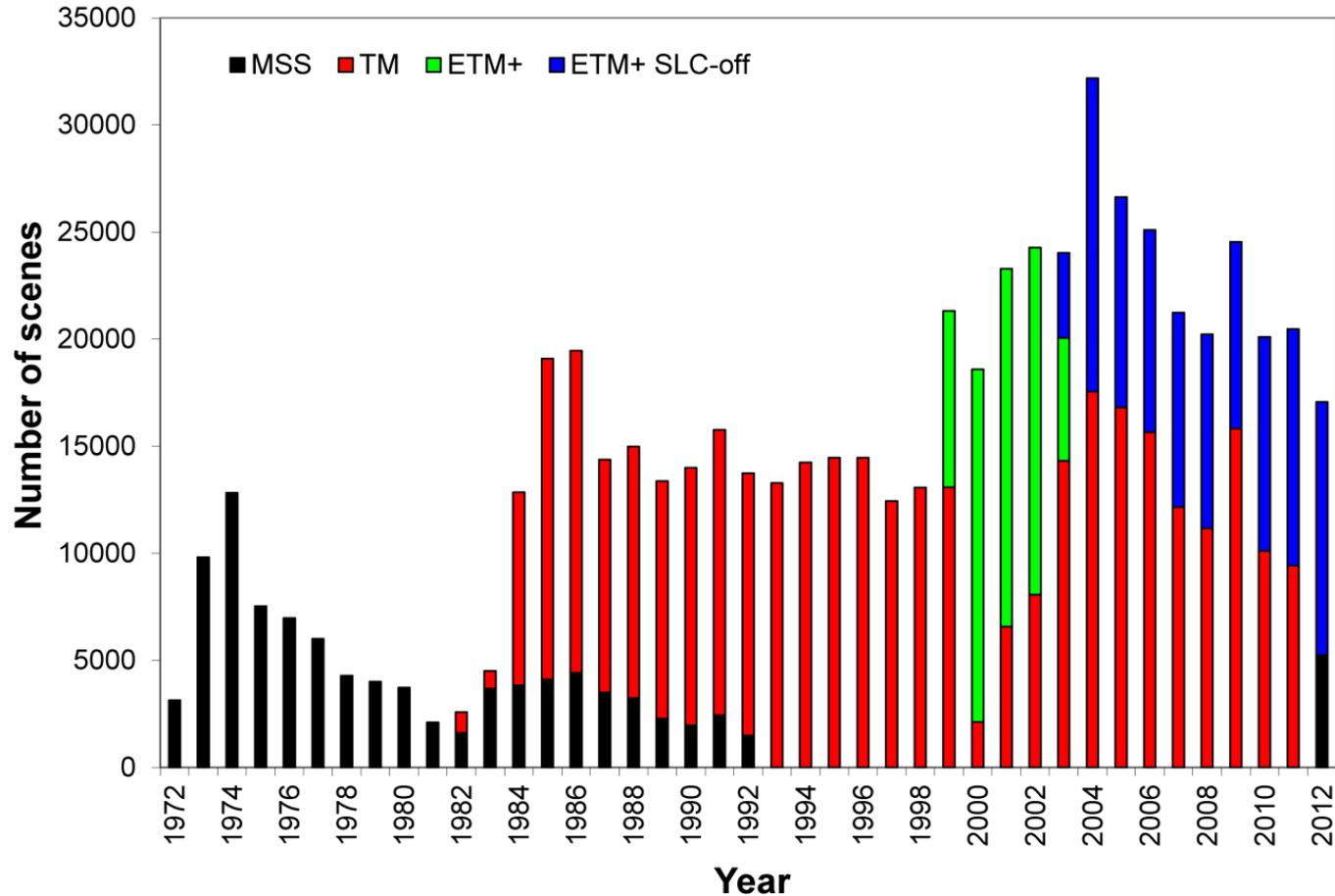
Total Marketable Scenes: 137,630

Unique Marketable Locations: 13,296

1-4 5-9 10-13 14-18 19-23

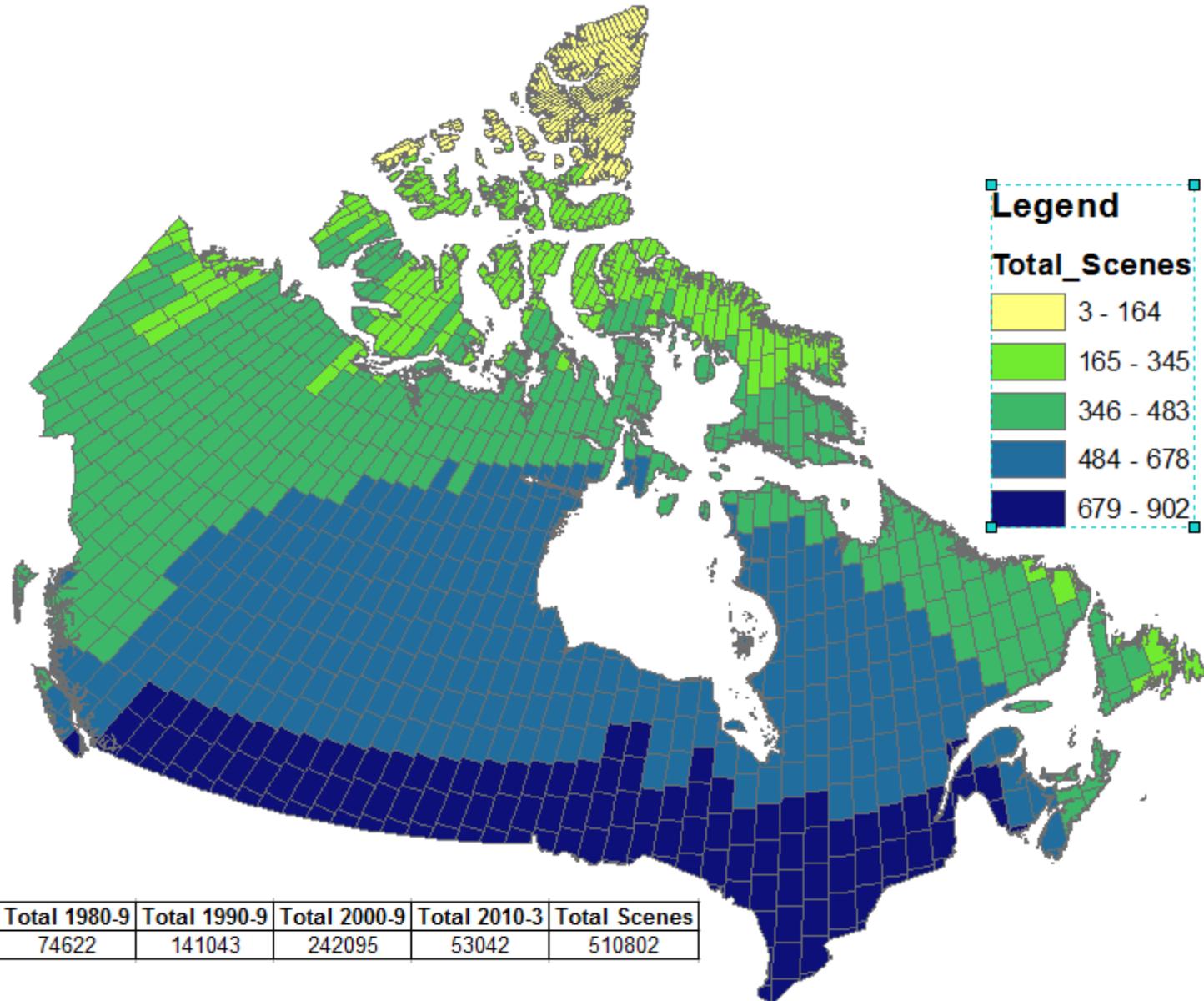
http://landsat.usgs.gov/documents/StateOfTheArchive_web.ppt

USGS archive holdings, Canada

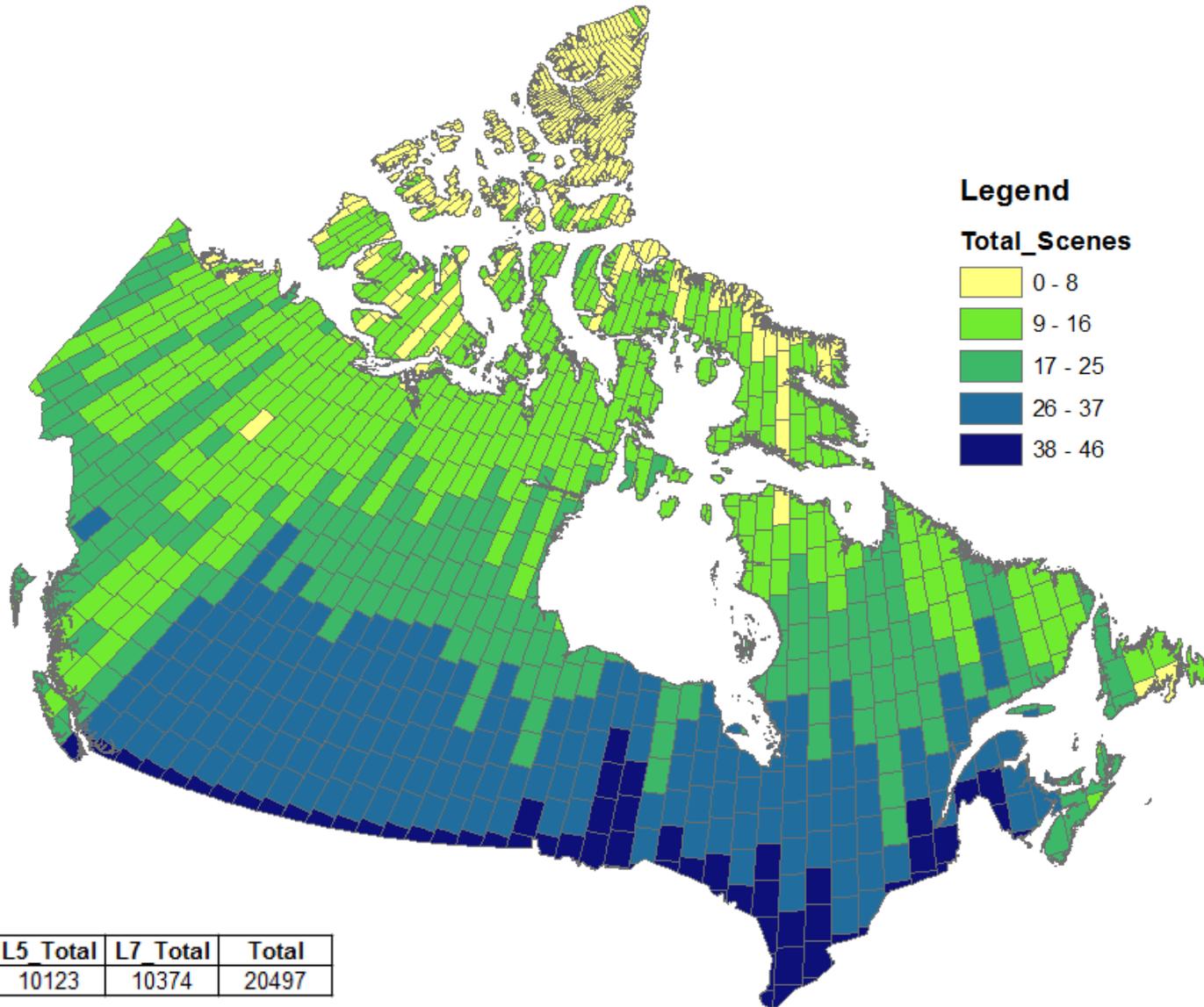


- > 605,000 images in archive

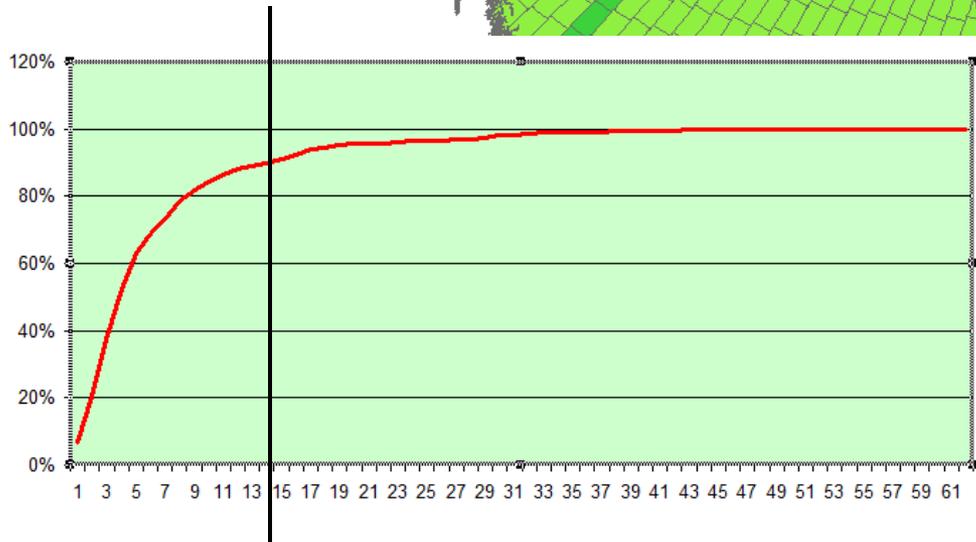
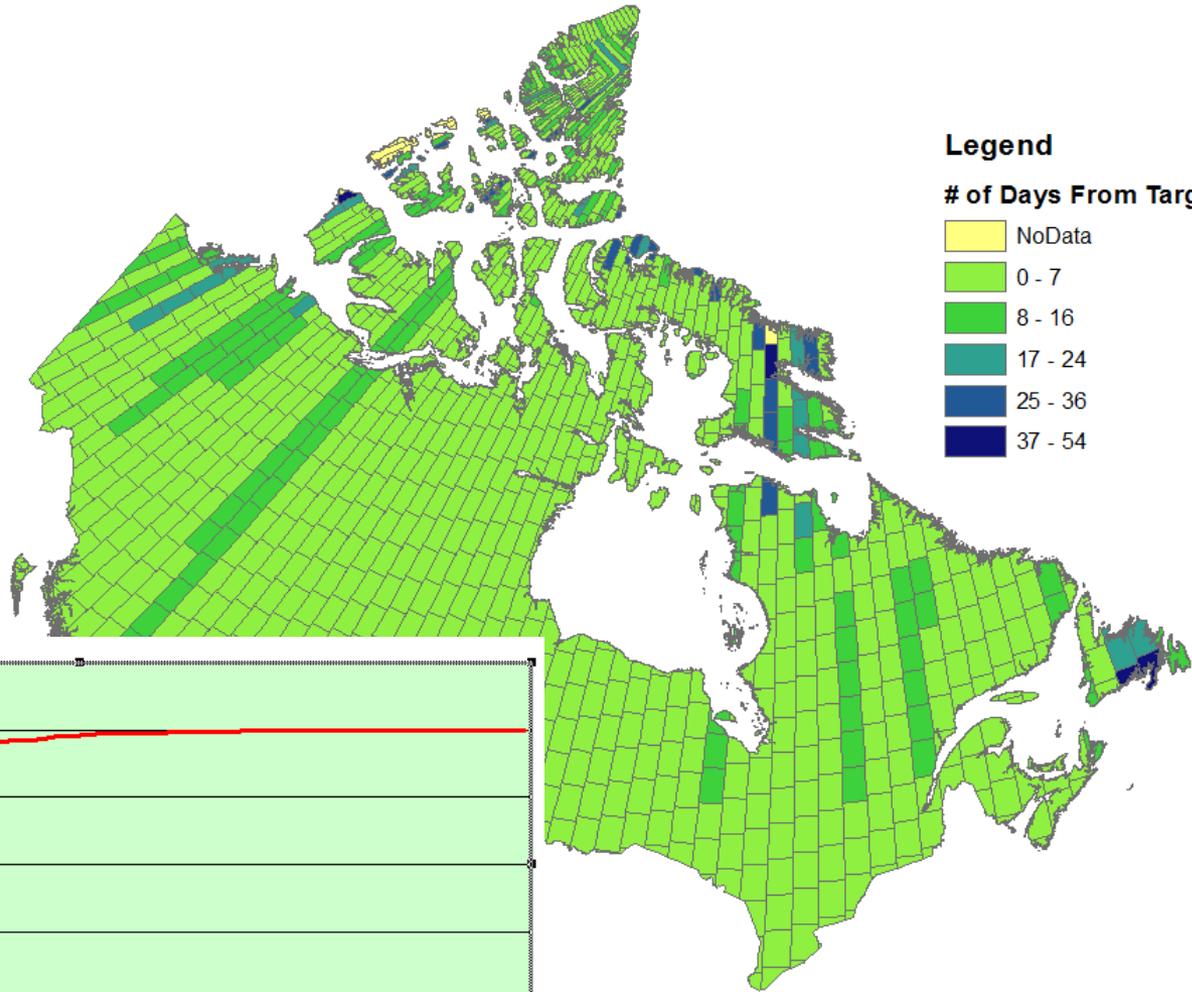
Available Images for 1286 Path Row Scenes Over Terrestrial Canada



Available 2010 Images for 1286 Path Row Scenes Over Terrestrial Canada



Distance off Target Date for Best Image between June 1 and Sept 30, 2010



Cloud / Shadow...

Best Available Pixel Processing

- Develop pixel based processing, not scene based
- Create composites from many images
- Cloud and shadow screening critical

Hansen, M. C. and Loveland, T. R. 2012. A review of large area monitoring of land cover change using Landsat data. *Remote Sensing of Environment*. Vol. 122, pp. 66-74.

Roy, D. P.; Ju, J.; Kline, K.; Scaramuzza, P. L.; Kovalskyy, Y.; Hansen, M.; Loveland, T. R.; Vermote, E., and Zhang, C. Web-enabled Landsat Data (WELD): Landsat ETM+ composited mosaics of the conterminous United States. *Remote Sensing of Environment*. 2010; 114:35-49.

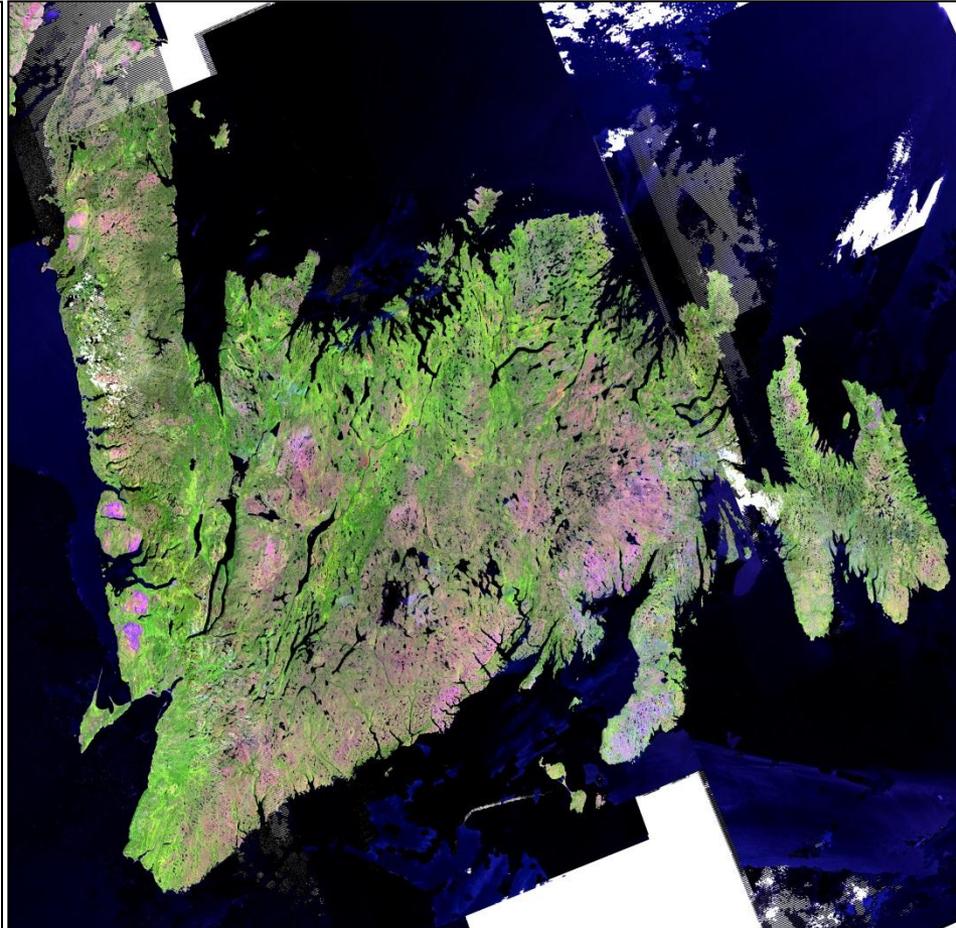
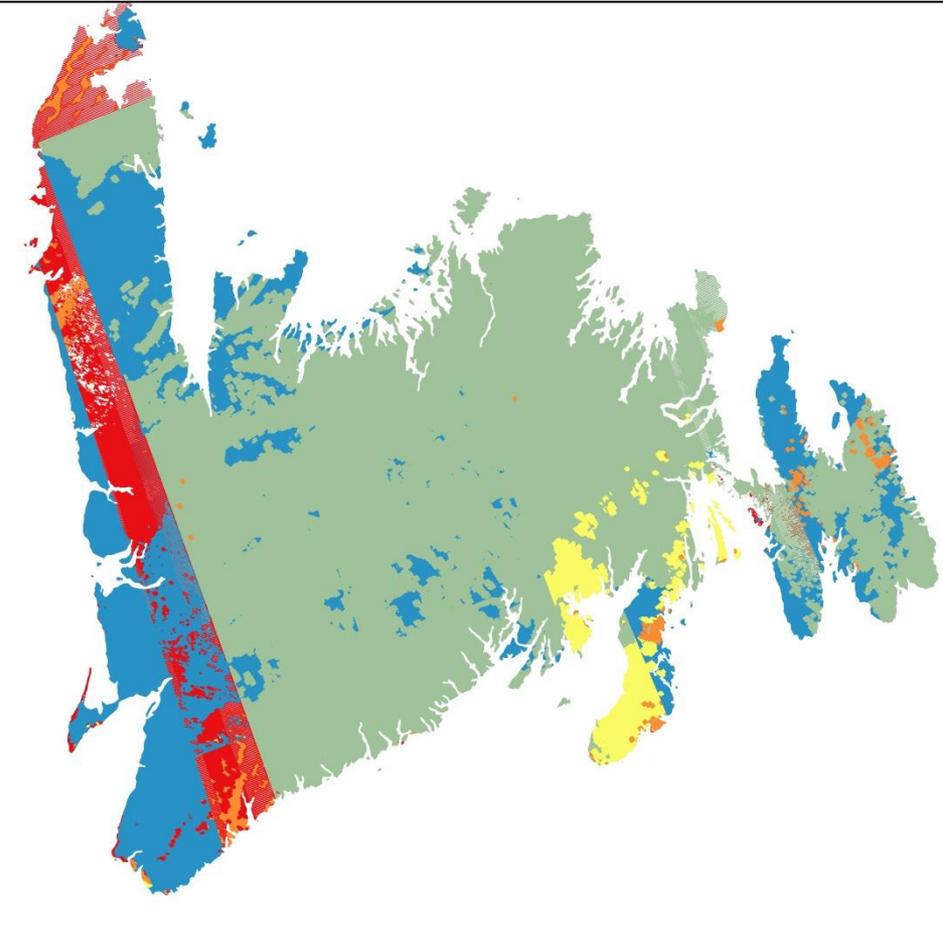
Griffiths, P. et al. 2013. A pixel-based Landsat compositing algorithm for large area land cover mapping. *J. Sel. Topics App. Ear. Obs. Rem. Sens.* Doi: 10.1109/JSTARS.2012.2228167

Pixel Assessment

- A score calculated for each pixel
- “Best” pixel then selected, based on:
 - 1) Acquisition year
 - 2) Acquisition Day of Year (target August 1, day 213)
 - 3) Distance to cloud/shadow
 - cloud / shadow masking from FMASK
 - Zhu et al. 2012. RSE.
 - 4) Reflectance and opacity from LEDAPS (Masek et al. 2006)
 - Use opacity (haze) to aid in pixel selection

Composites - Lexicon

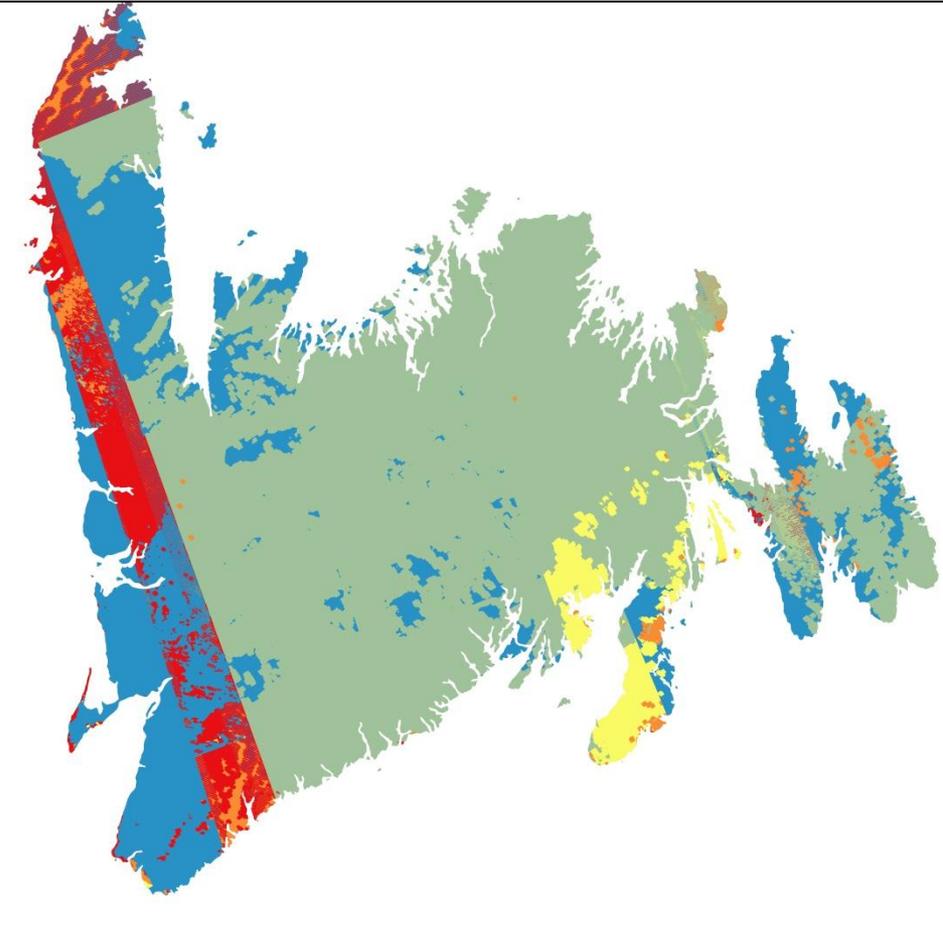
- Rule-based BAP composite
 - determined by DOY, year, distance to cloud, sensor (L5 or L7), and opacity
 - BAP may come from desired date range in previous or subsequent years
 - goal is consistent phenology with minimal “no data” pixels
- Annual BAP composite
 - determined by DOY, distance to cloud, sensor (L5 or L7), and opacity
 - BAP may come from desired date range *for a given year*
 - “no data” pixels are much more likely to occur
- Proxy annual BAP composite
 - use annual BAP composites as source
 - fill in areas of “no data” by averaging (or some other approach) pixel values from previous/subsequent years
 - some intelligence is envisioned whereby a long-term trajectory of pixel values can be used to identify whether or not the pixel is changed
 - contextual rule-base



D2TD (days)

- 0 - 10
- 11 - 20
- 21 - 30
- 31 - 45
- 46 - 62

BAP annual composite (2003)



D2TD (days)



BAP multi-year composite (2003)



BAP proxy composite (2003)

Historic reconstruction:

• Features: Proxy annual BAP composite

- Annual
- Compensates for haze, image availability
- Automated, allows for data reduction
- Allows for applications: cover, change in cover, and surface reflectance for structural attribution
- Change over time compensates for limitations of compositing (haze, phenology, ...). Illogical transitions can be removed. Change capture comes first.
- Note best pixel approaches will differ outside of CONUS and active IC catchments

- Proxy development:
 - Annual BAP layers (limited by date, haze)
 - First pass filter
 - Figuring out how to deal with disturbance
- Land cover change
 - Spectral change; LandTrendr-lite. Disturbance year, pre-dist trend; post-dist trend, etc
- Land cover
 - Stability, logical transitions, spectral info
- Structure
 - Lidar plots to cal / val estimates of ht, biomass, volume, canopy cover, etc.
- Continuity has made possible

Thank you!

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Canada 

Publications:



Natural Resources
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Ressources naturelles
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<http://cfs.nrcan.gc.ca/publications/authors/read/11091>

