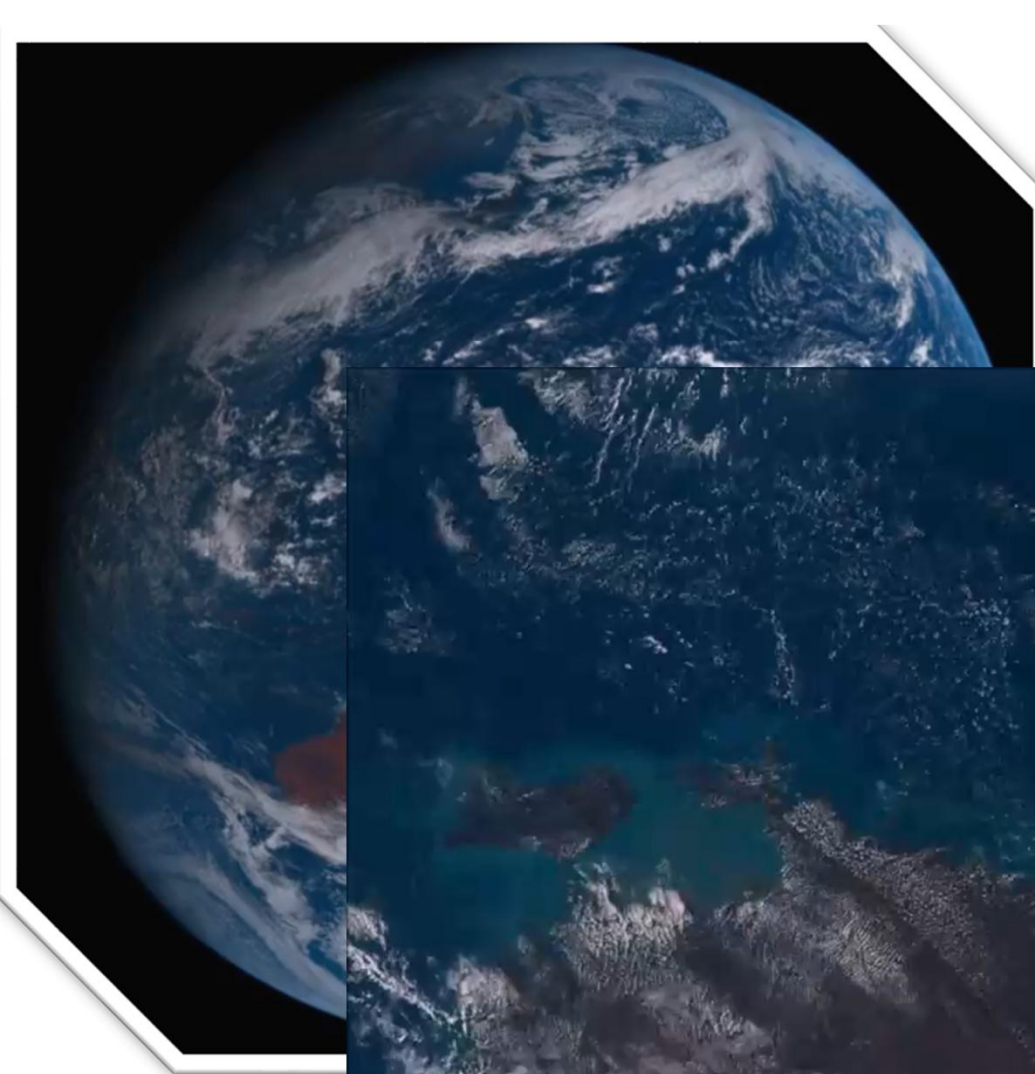
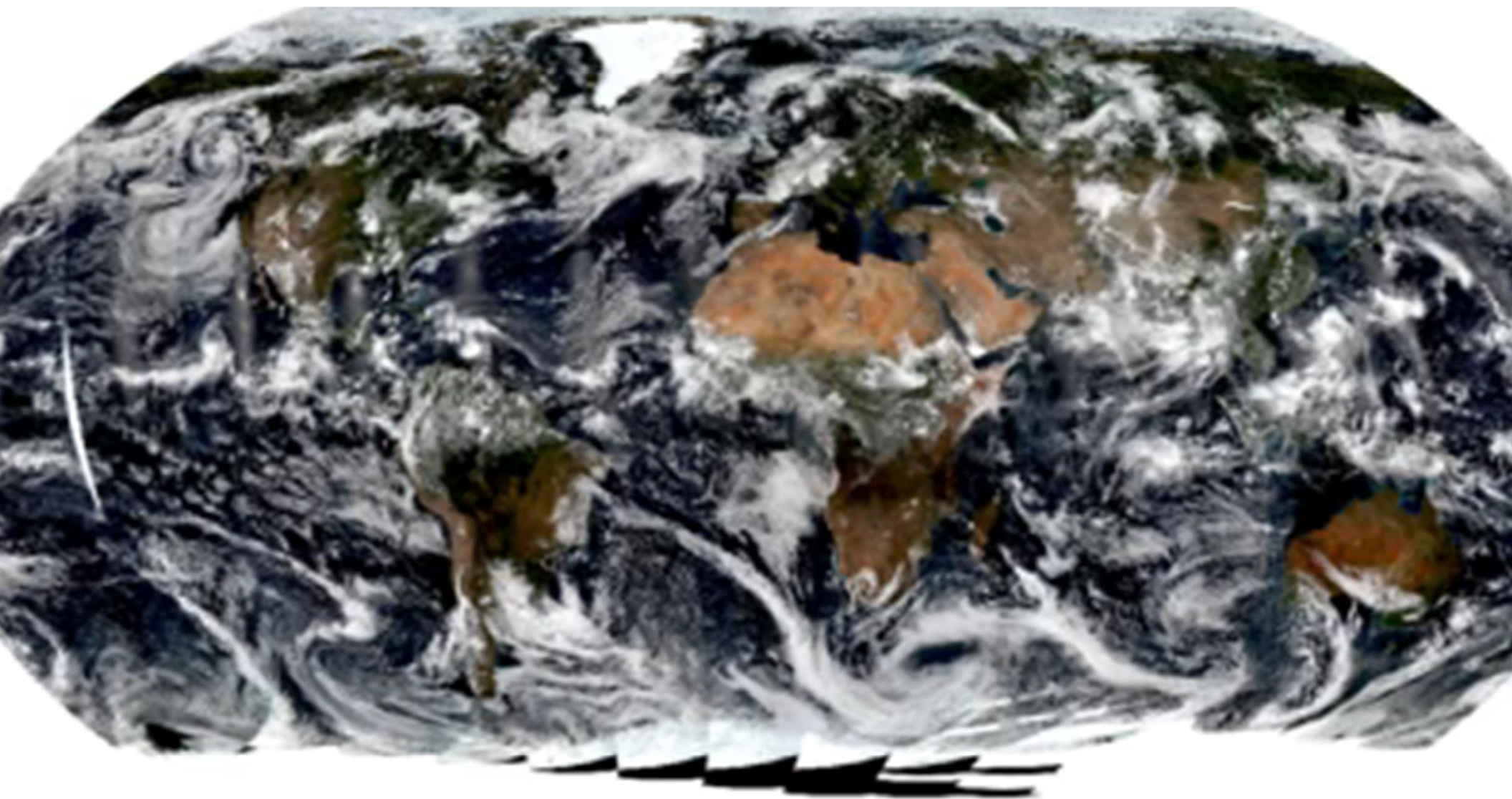


Google Earth Engine Workshop

Gennadii Donchyts







A satellite-style map of the Earth showing continents and oceans. The text is overlaid on the map.

Google

Earth Engine:

**Google's Cloud Platform for
Big Earth Data Analytics**

Google Earth



3-D Globe
Visualization

Earth Engine

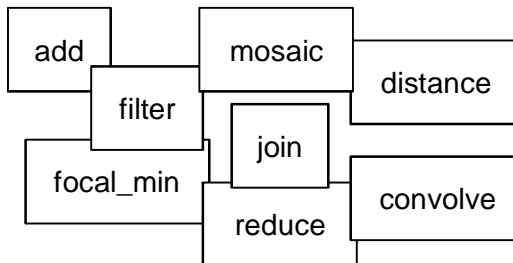
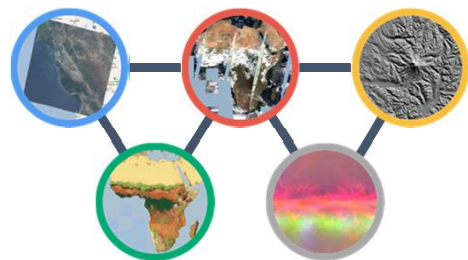


Geospatial
Analysis

Requests

Results

Geospatial
Datasets

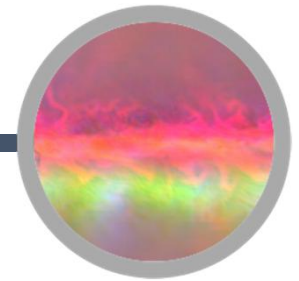
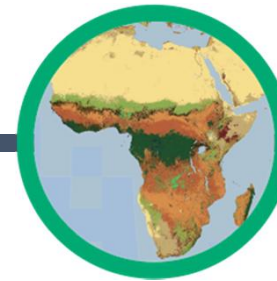


Algorithmic
Primitives



Storage and Compute

The Earth Engine Public Data Catalog



Landsat 4, 5, 7, 8

Raw, TOA, SR, ...

MODIS

Daily, NBAR, LST, ...

Terrain

SRTM, GTOPO, NED, ...

Land Cover

GlobCover, NLCD, ...

Atmospheric

NOAA NCEP, OMI, ...

...
... and many more, updating daily!

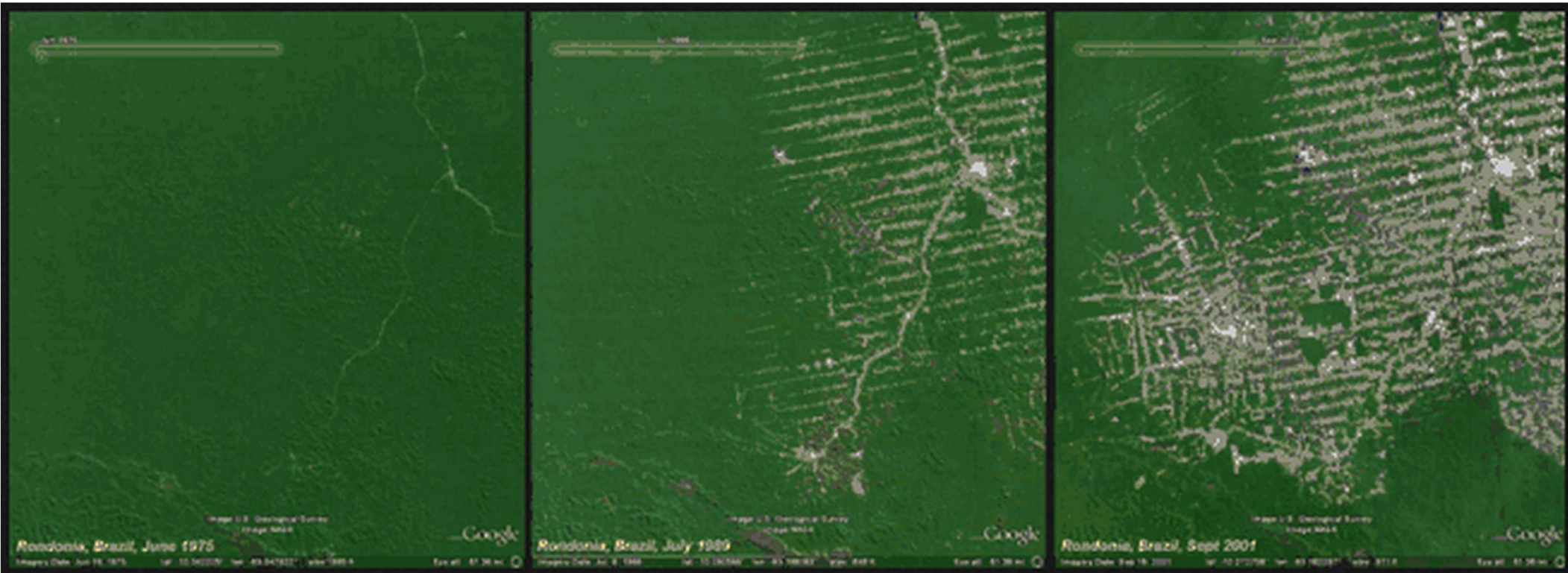
> 200 public datasets

> 5 million images

> 4000 new images every day

> 5 petabytes of data

Earth Engine: Origins



Global Forest Watch

GLOBAL
FOREST
WATCH
2.0

HOME

COUNTRIES

STORIES

MAP

BLOG

DATA

ABOUT

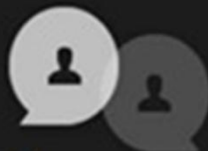
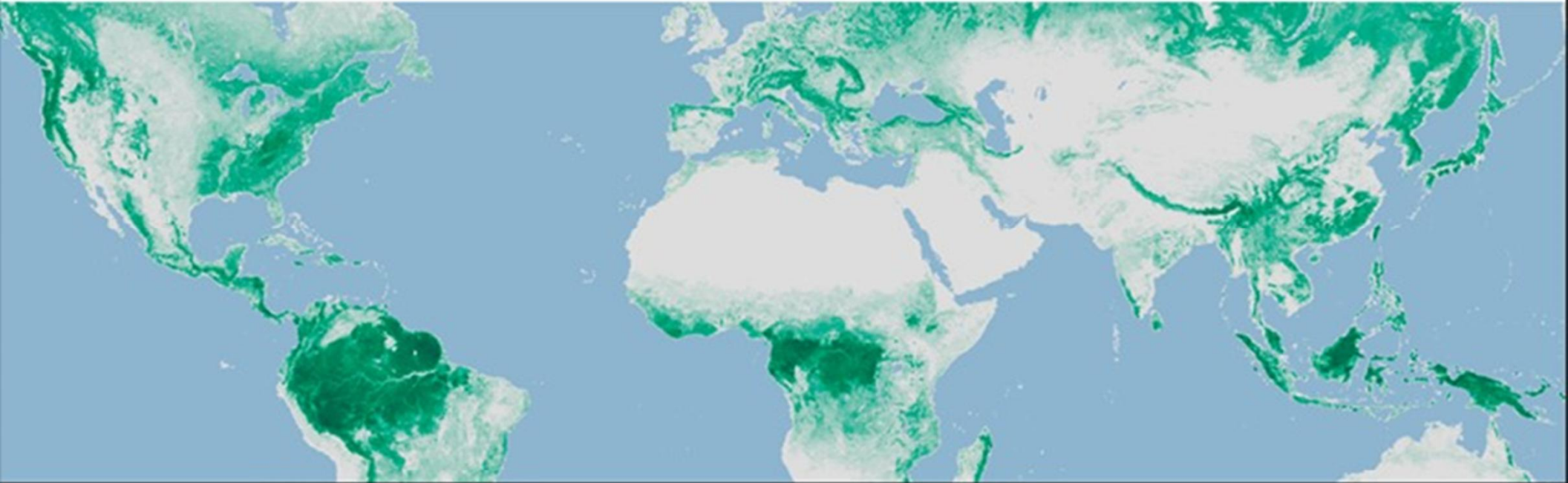
ENGLISH

+

Find out what is happening
in forests right now

44,479
ALERTS IN THE PAST
YEAR

3
NEW FOREST
STORIES



Join the community



Analysis tool



Stay updated

JavaScript & Python API

```
// Make a median composite from two years of Landsat 7.  
// Load the image collection.  
var collection = ee.ImageCollection('LANDSAT/LE7_L1T');  
// Filter it down to 2011 and 2012.  
var filtered = collection.filterDate('2011-01-01', '2012-12-31');  
// For each pixel, for each band, calculate the median and make an image  
// of the result. The median tends to remove clouds, shadows, data gaps.  
var medianImage = filtered.median();  
// Add the image to a map.  
Map.addLayer(medianImage, {bands:['B3', 'B2', 'B1'], min:35, max:170});
```

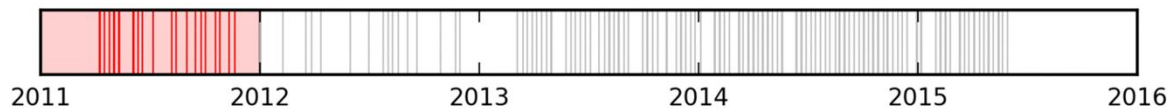
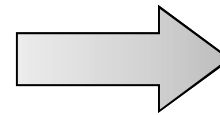
Code Editor link:

<https://code.earthengine.google.com/d7e0877fe7ad644bd6b50e6843150469>

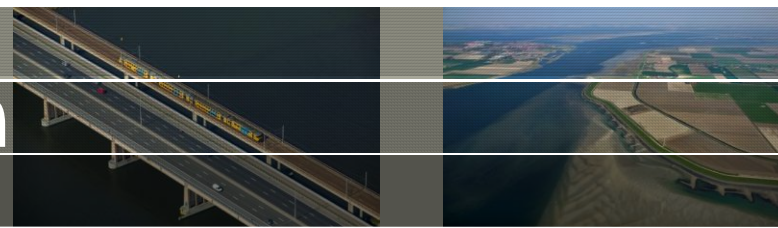
Coastline Extraction – Zandmotor



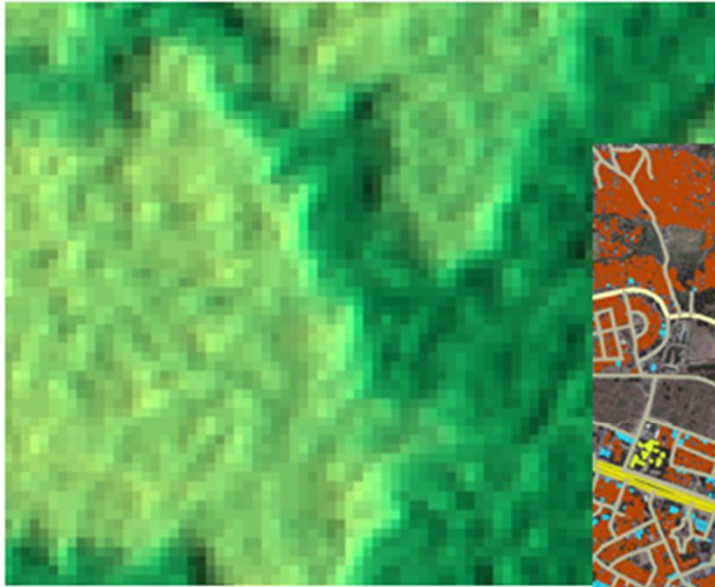
Zandmotor [2011/01/01 to 2012/01/01]



Participatory DEM generation



SRTM 30m



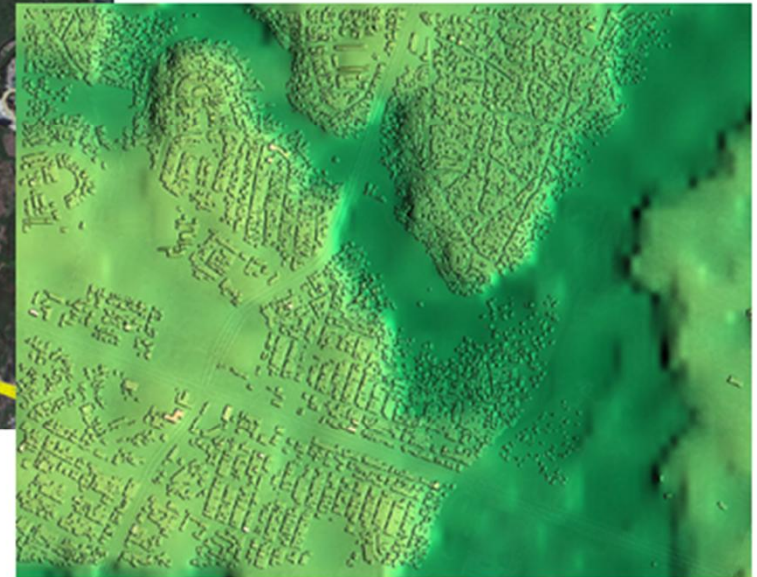
OSM data



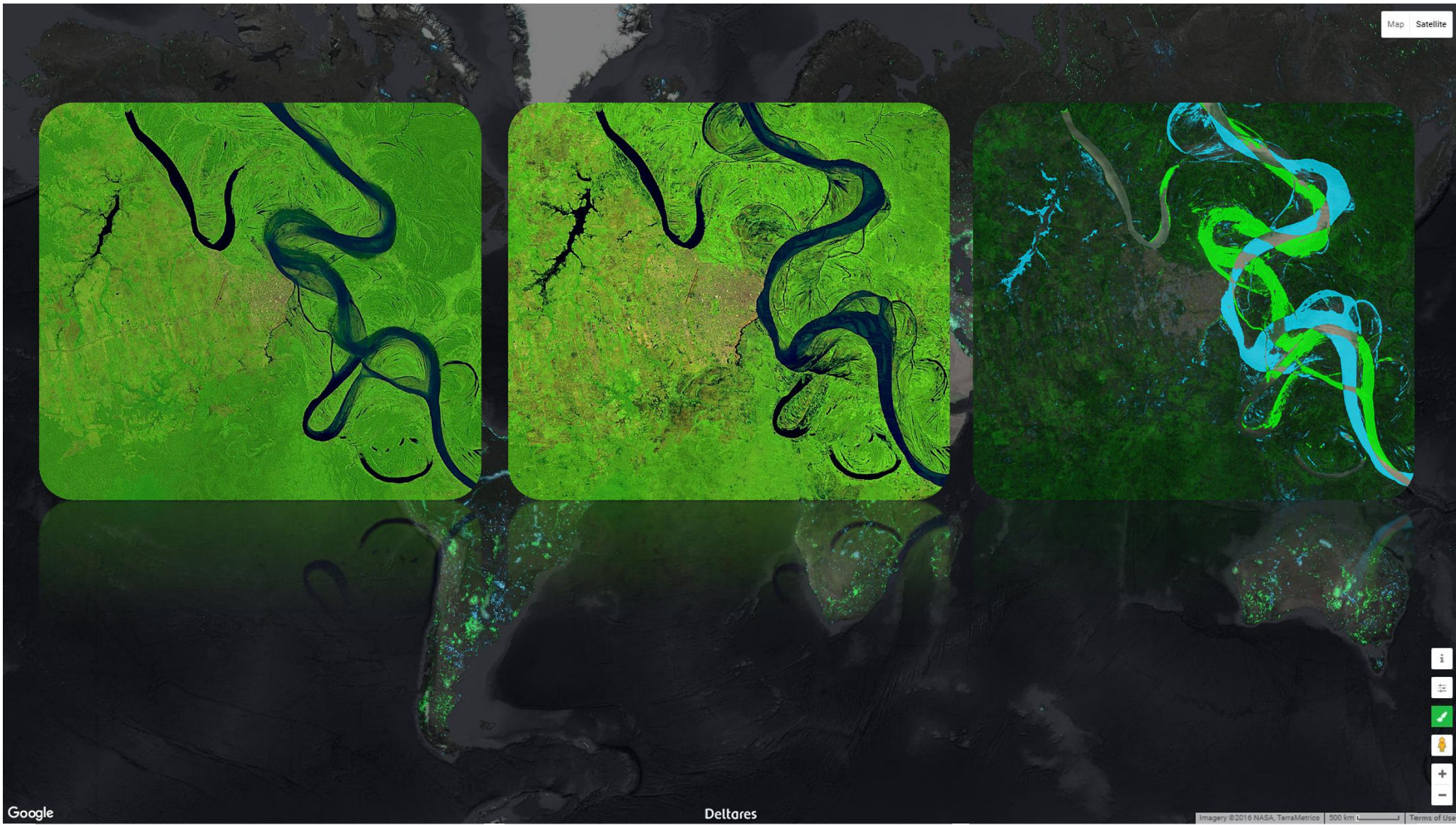
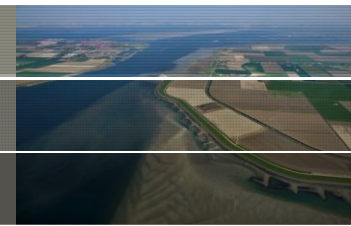
SRTM + OSM fusion

=

HighRes Digital Surface Model



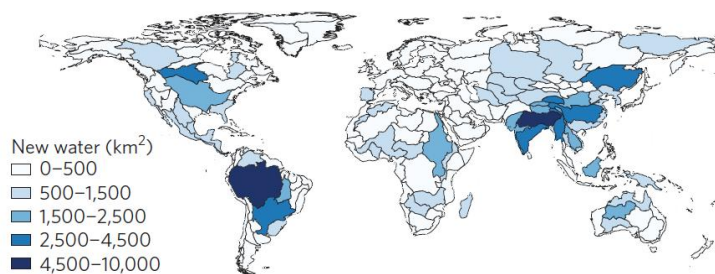
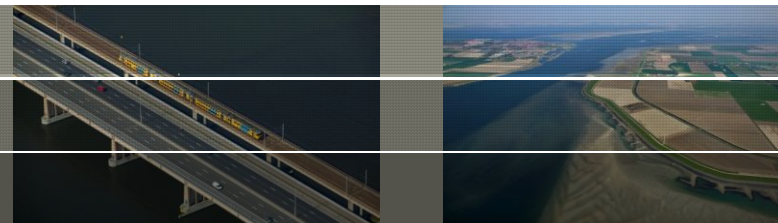
Aqua Monitor



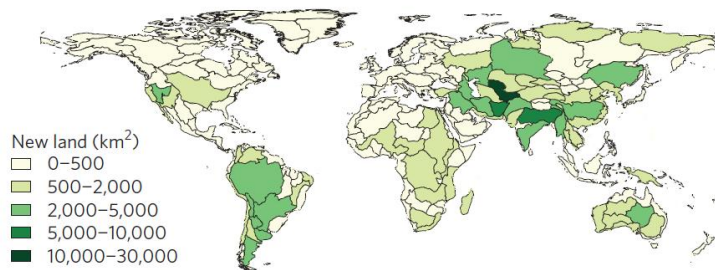
<http://aqua-monitor.appspot.com>

Deltares

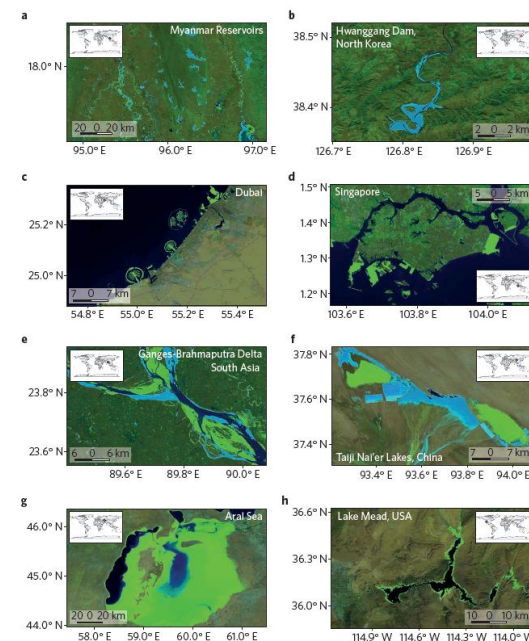
Aqua Monitor

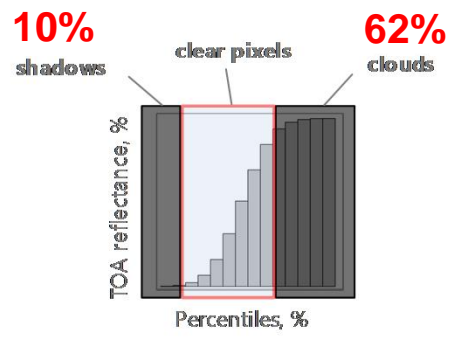


Name	Area (km ²)
Tibetan Plateau	7,661
Amazon River	7,058
Ganges-Brahmaputra Rivers	5,495
Rio de la Plata	4,410
River Nelson	4,101
India	3,677
River Amur	3,494
Yangtze River	3,238
Tigris and Euphrates Rivers	2,636
Thanlyin; Sittang; Ayeyarwady	2,592
Indus River	2,312
Mississippi River	2,231
Mekong River	2,074



Name	Area (km ²)
Aral Sea	27,841
Meghna River	19,580
Hamun Lake	6,044
Tigris and Euphrates Rivers	4,897
Amazon River	4,888
Lake Eyre	4,672
Rio de la Plata	4,020
India	3,851
Lake Urmia, Caspian Sea	3,759
Great Salt Lake; Malheur Lake	3,752
Ural River	3,455
Amur River	3,344
Ob River	3,068

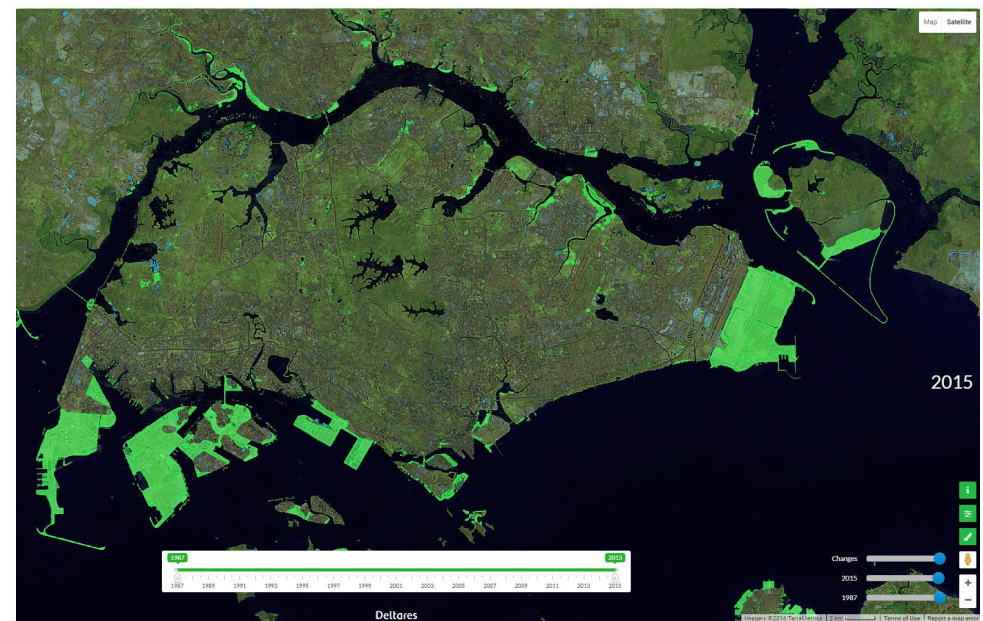
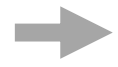
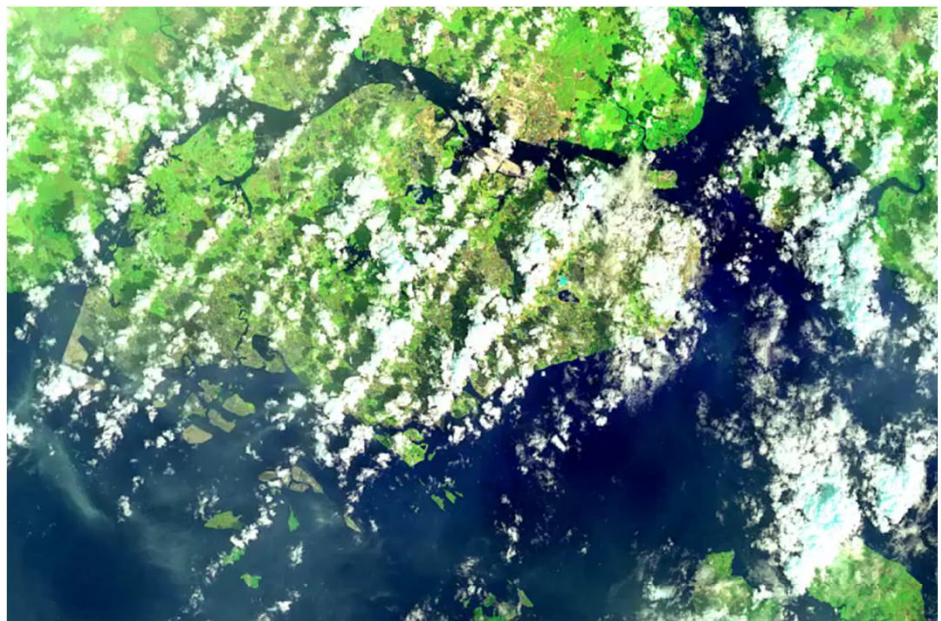




$$F(\rho) = \int_{-\infty}^{\rho^i} P(\rho) d\rho$$

$$I = \frac{\rho_{green}^i - \rho_{nir}^i}{\rho_{green}^i + \rho_{nir}^i}$$

$$I = \beta_0 + \beta_1 T + \epsilon$$



Singapore

Surface water changes (1985-2016)

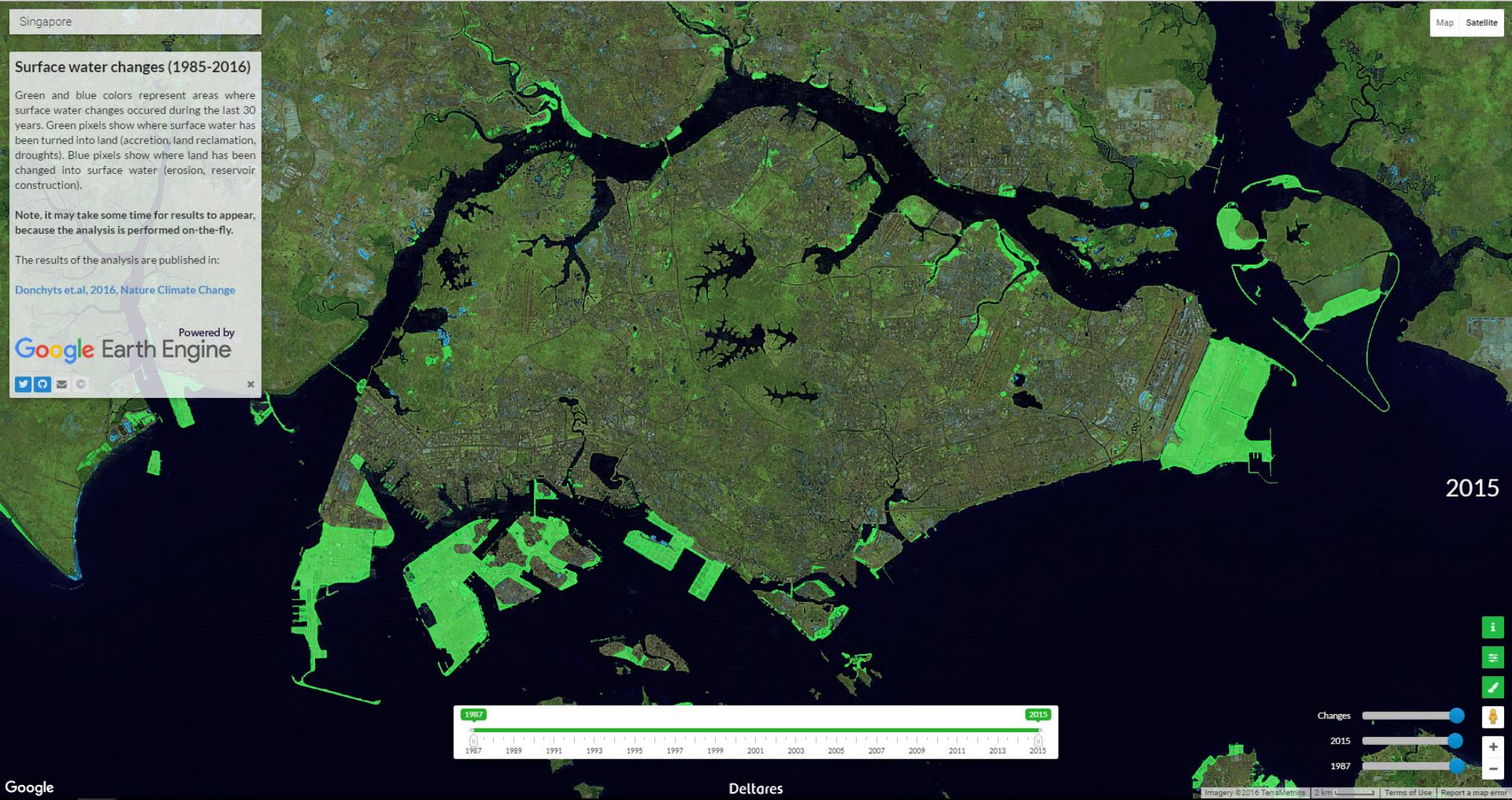
Green and blue colors represent areas where surface water changes occurred during the last 30 years. Green pixels show where surface water has been turned into land (accretion, land reclamation, droughts). Blue pixels show where land has been changed into surface water (erosion, reservoir construction).

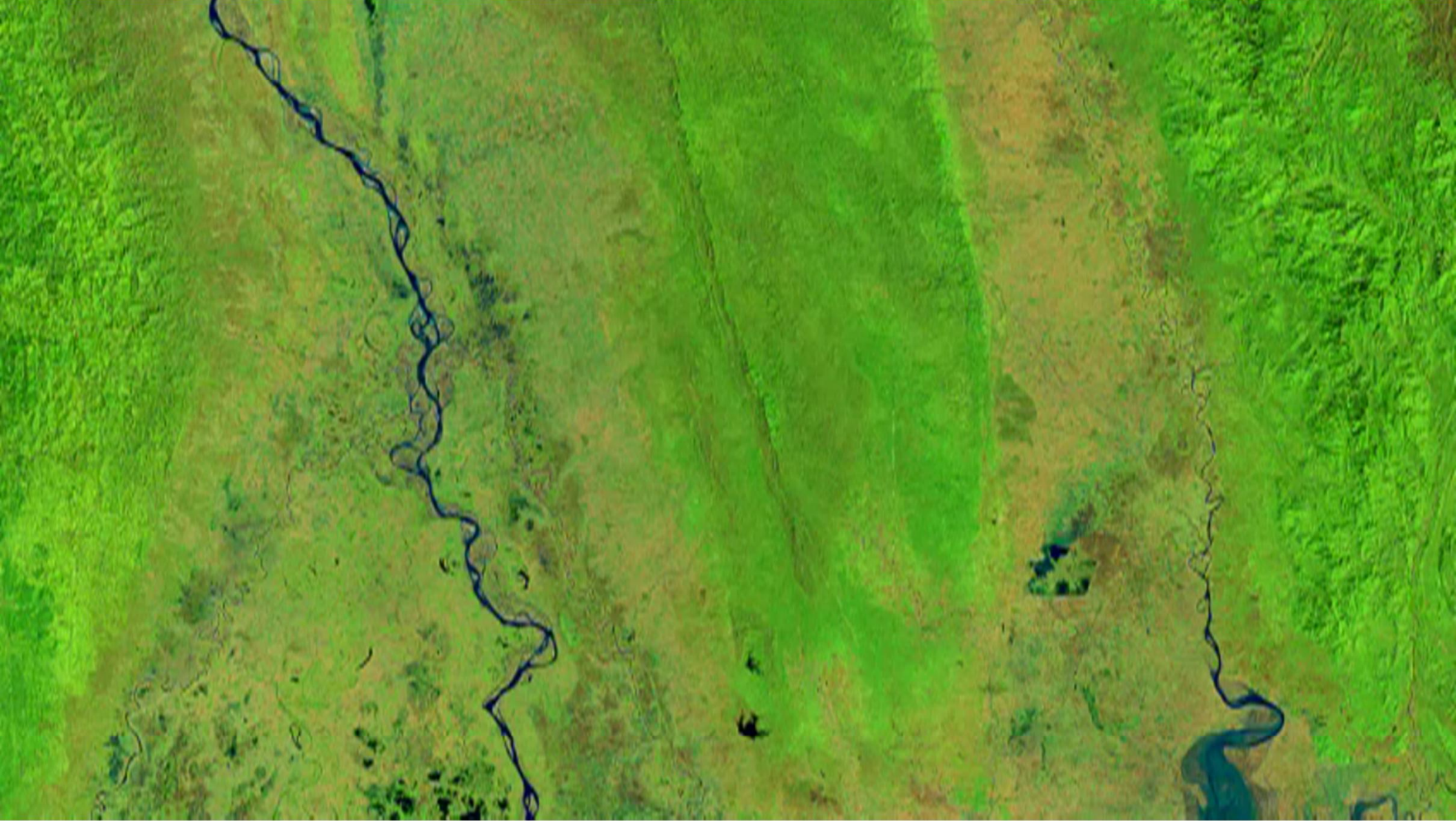
Note, it may take some time for results to appear, because the analysis is performed on-the-fly.

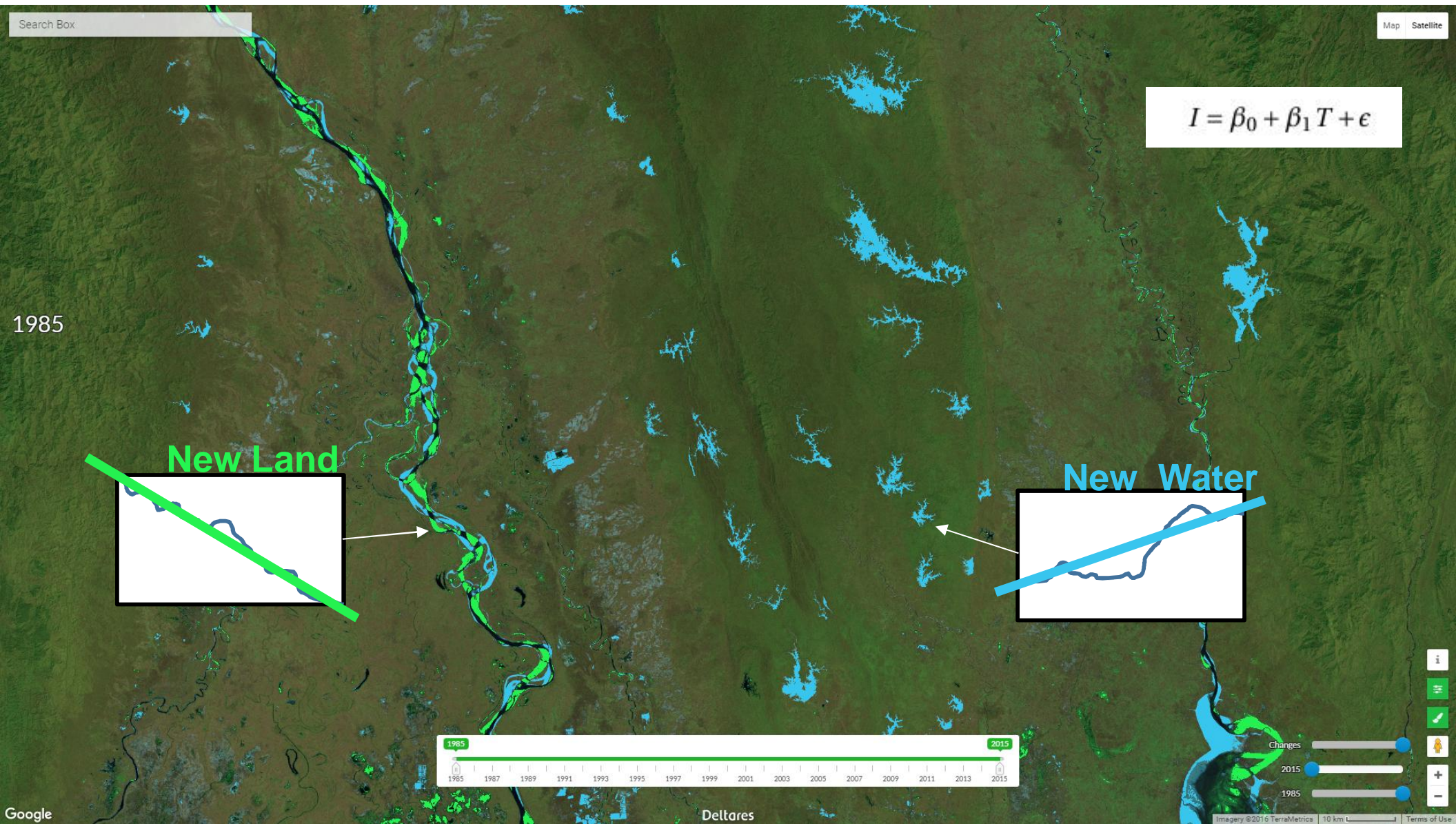
The results of the analysis are published in:

[Donchyts et.al, 2016, Nature Climate Change](#)

Powered by
Google Earth Engine

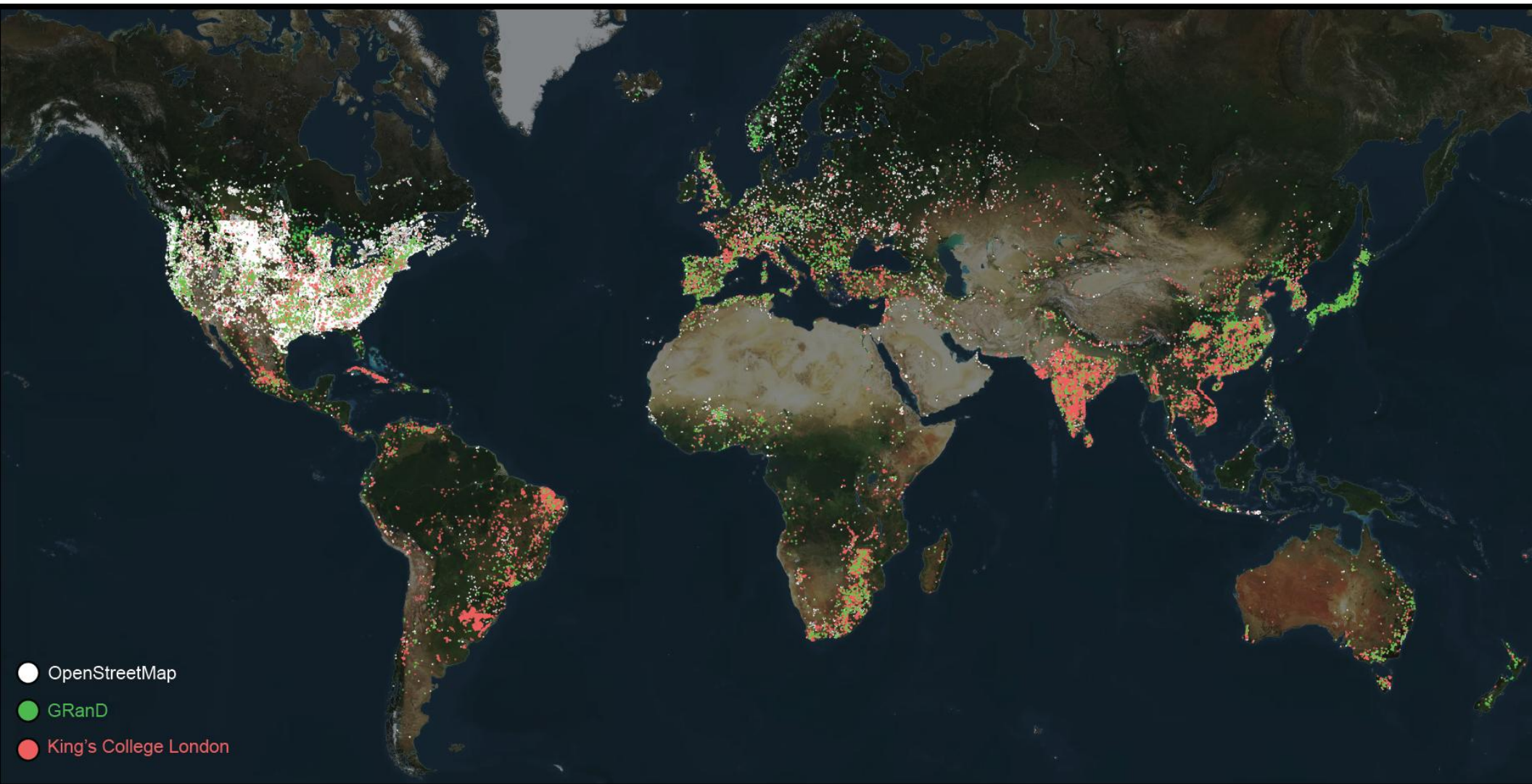






<http://aqua-monitor.appspot.com>

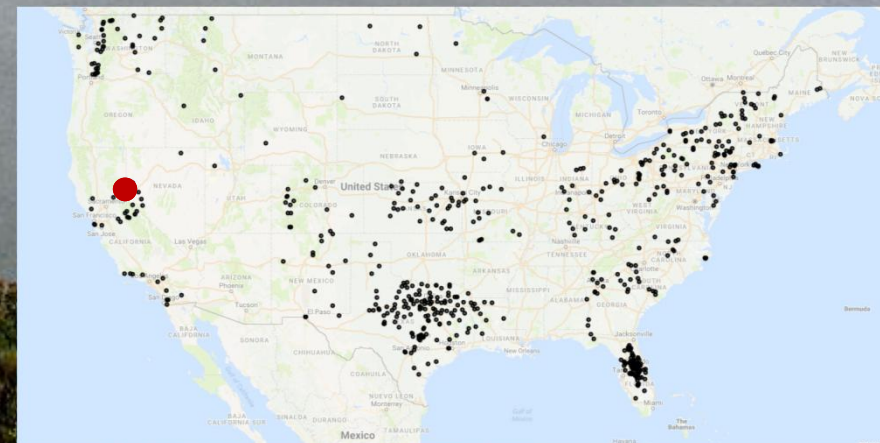
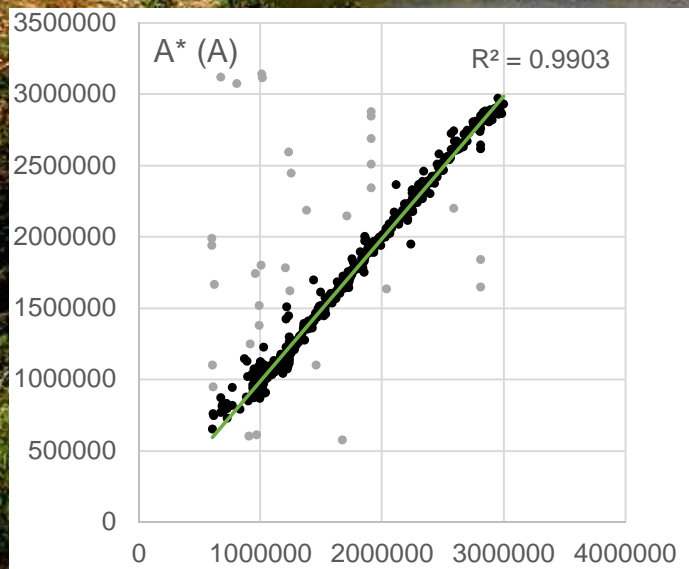
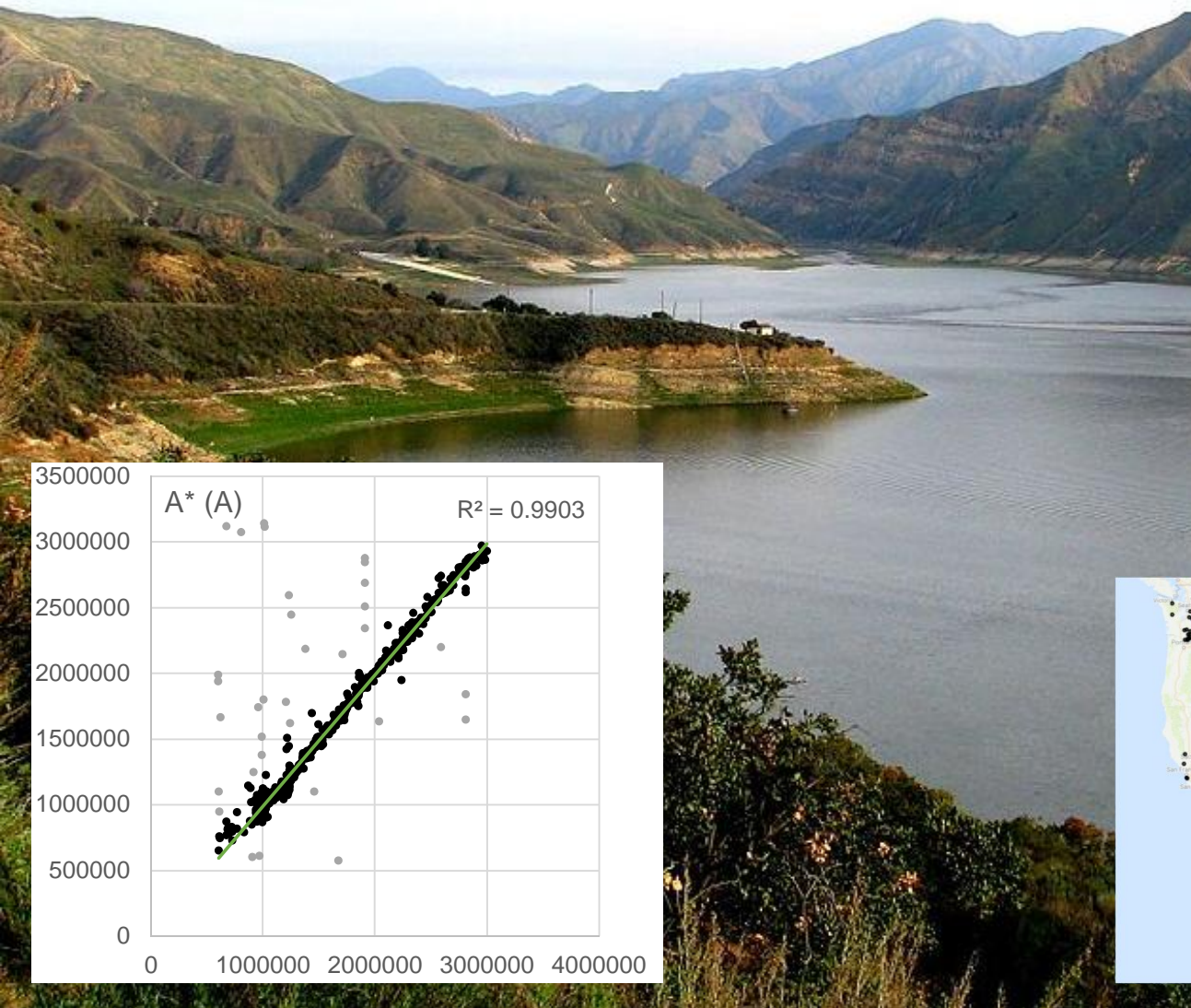
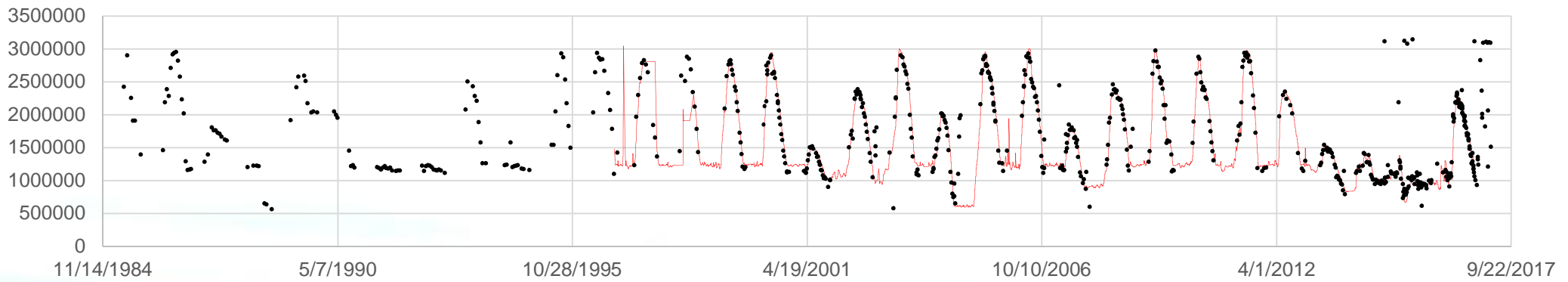
About 150 000 reservoirs mapped today (GRanD \approx 6000)



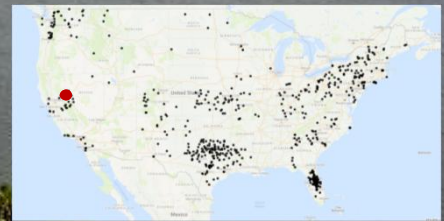
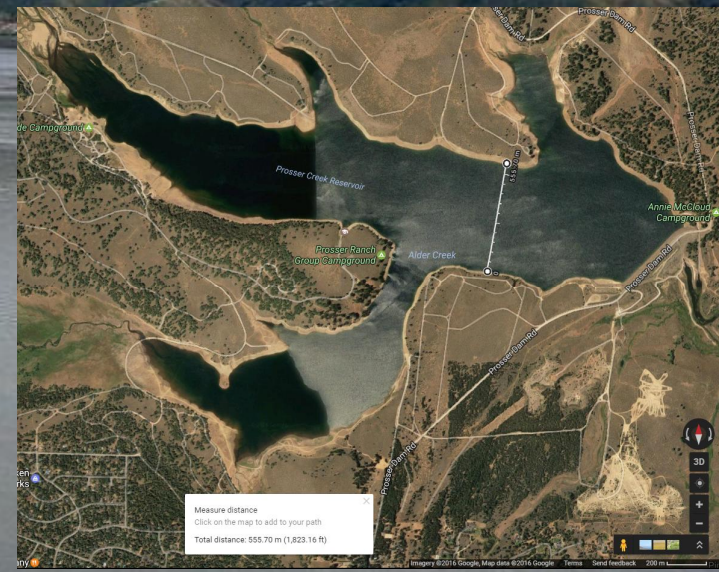
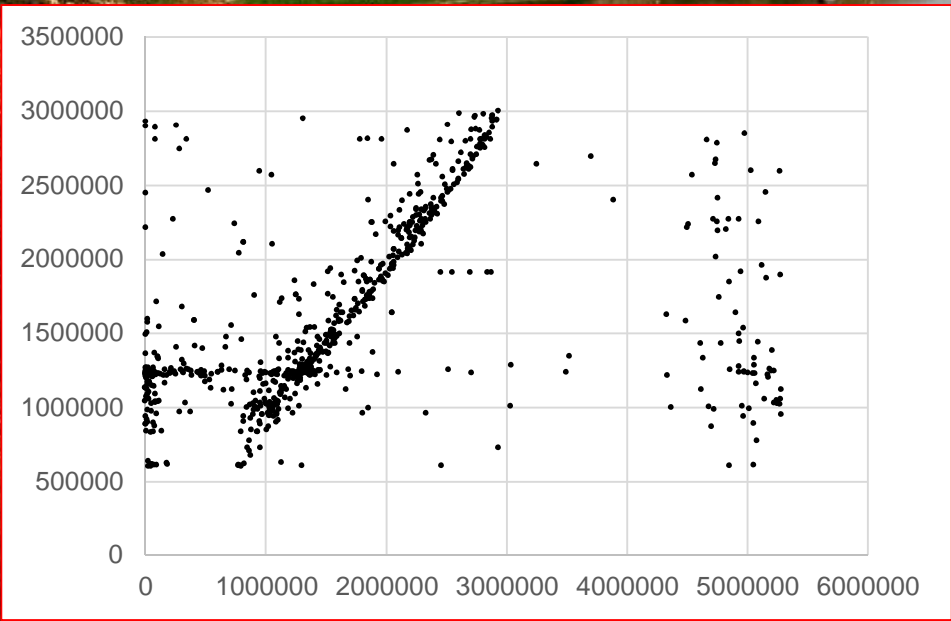
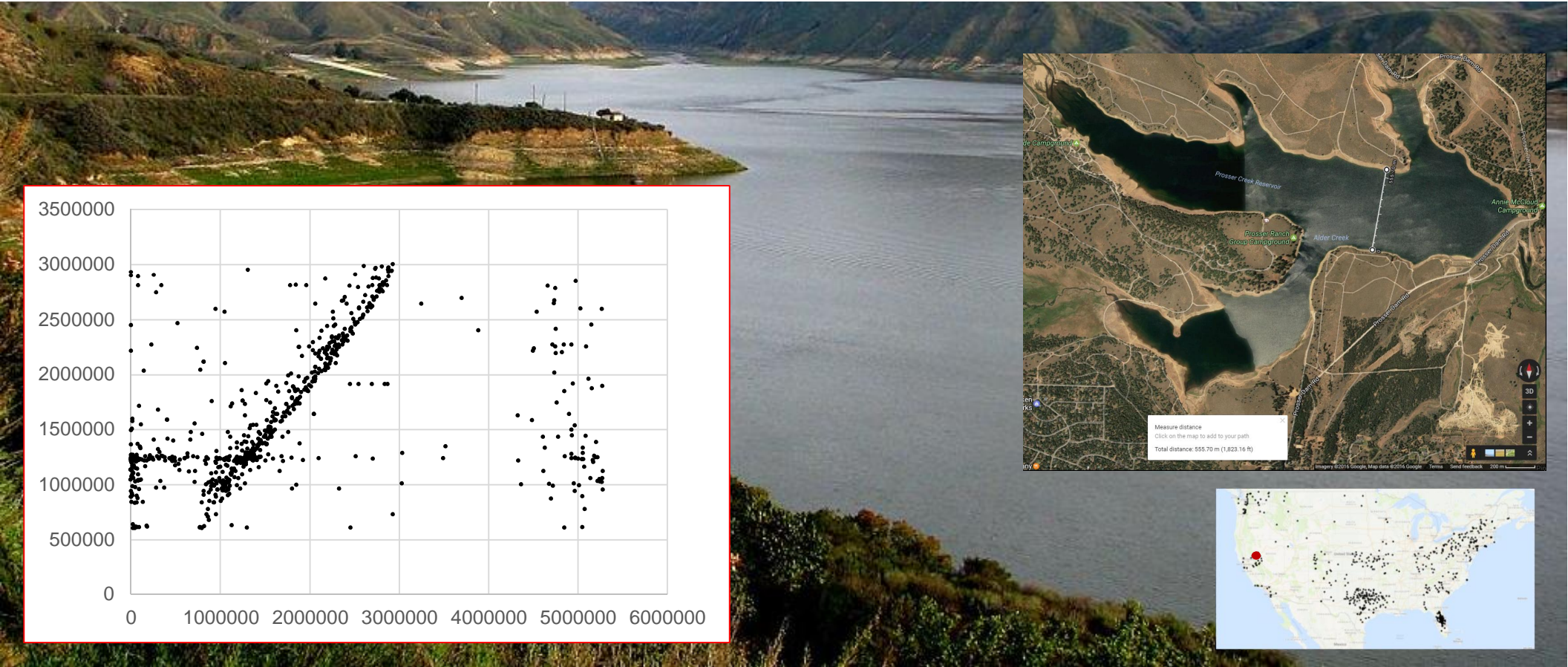
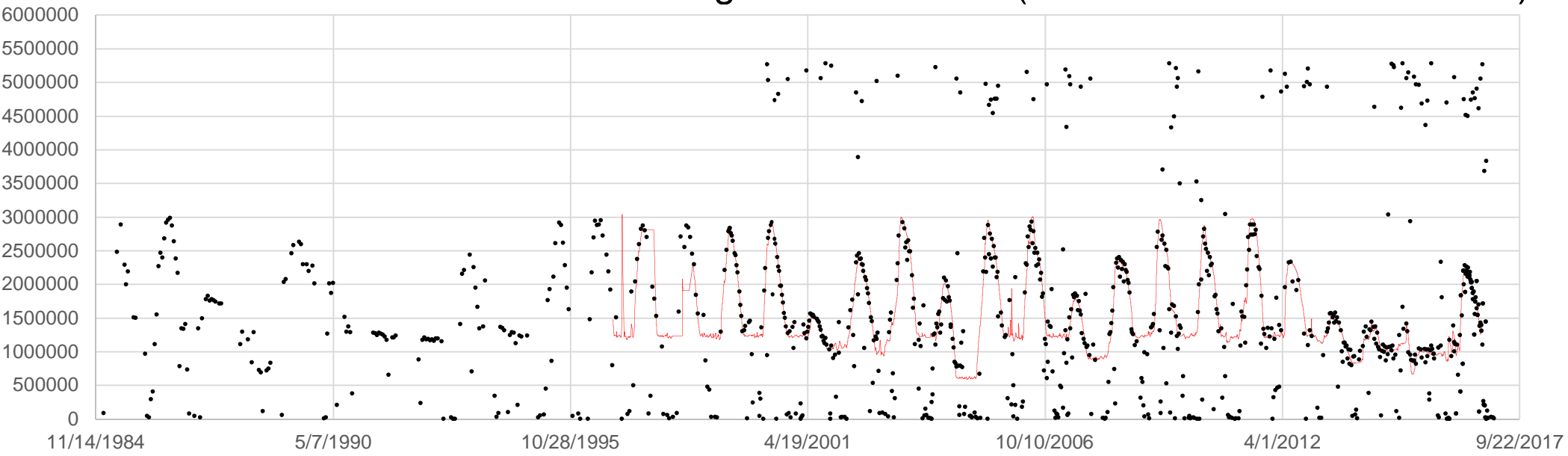
http://bit.ly/global_reservoirs_map_2016

http://bit.ly/agu16_reservoirs

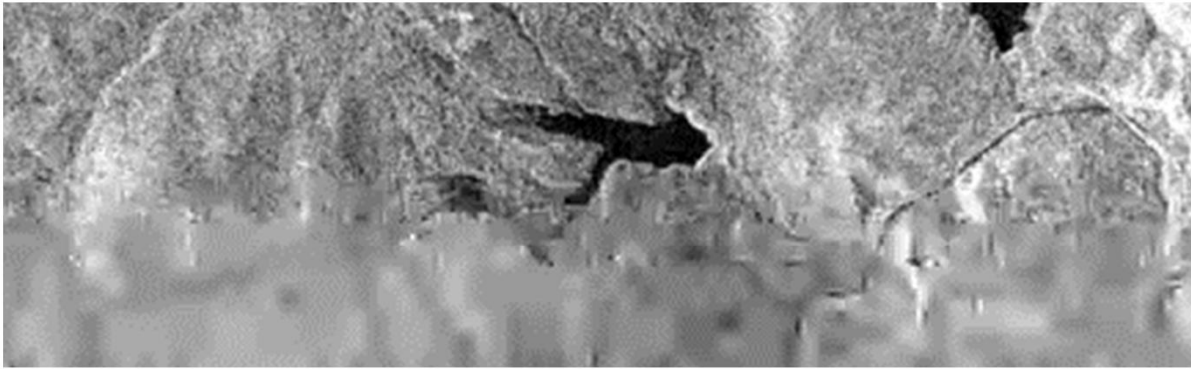
Surface water area estimated from satellite and from local water level station



Surface water area estimates using naive methods (no cloud/snow and NDWI=0)



0. Input images from multiple sensors

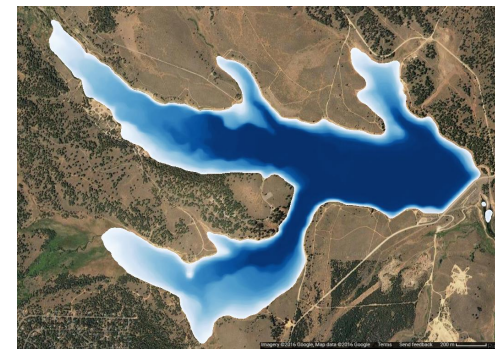


~1200 images, <30m

1. Detect water for cloud-free images

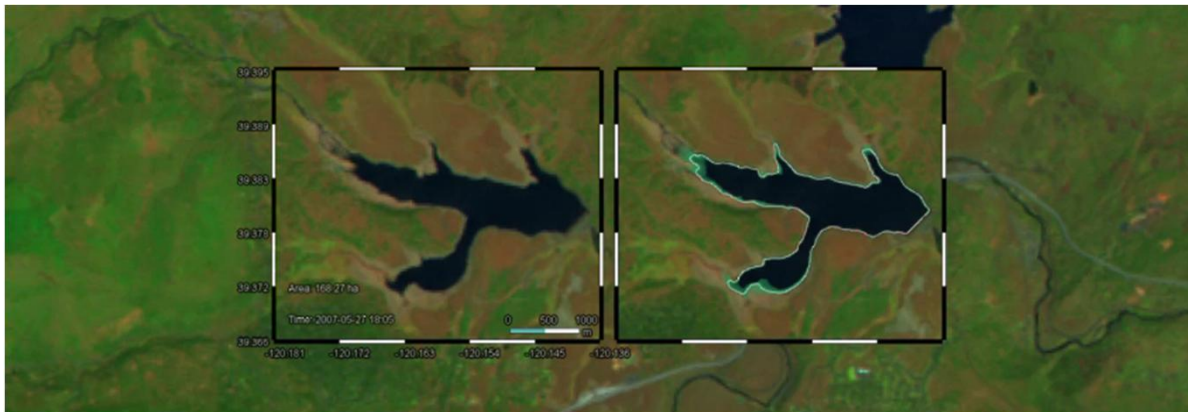


2. Estimate probability density (prior)

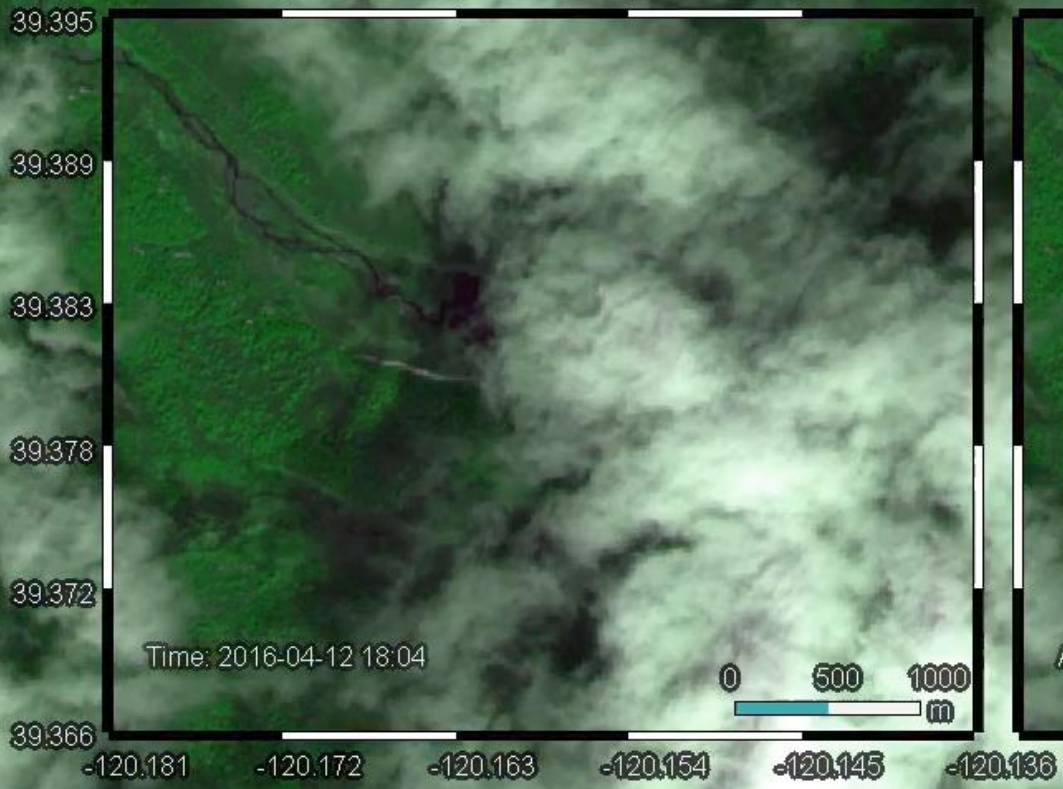


$$P(\text{water} | x, y, \text{cloud}=0)$$

3. Estimate surface area using generative-discriminative algorithm



$$P(W|W^t) = \frac{P(W|W^t) \cdot P(W)}{P(W^t)}$$





39.395

39.389

39.383

39.378

39.372

Time: 2016-06-30 13:06

39.366

-120.181

-120.172

-120.163

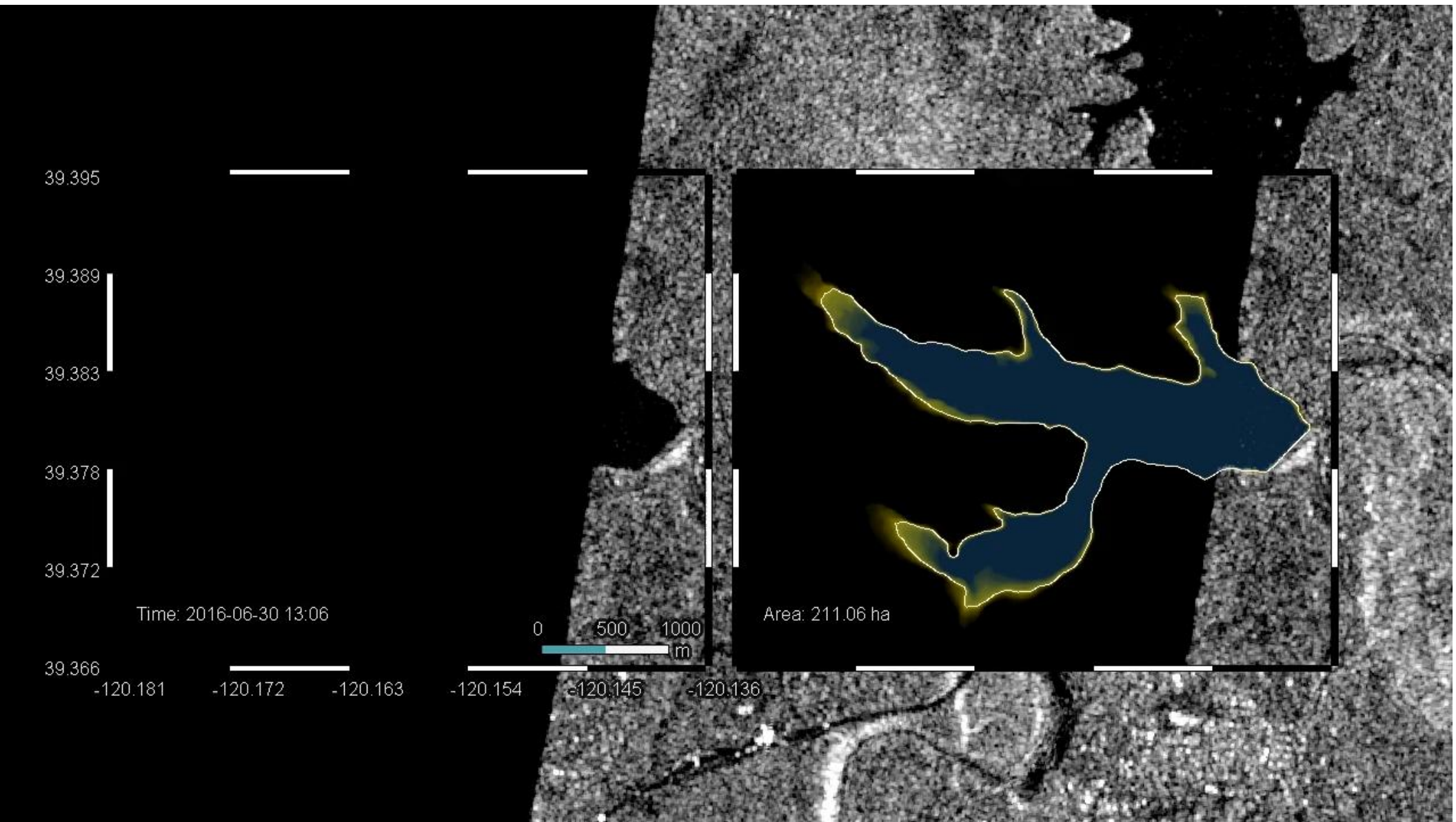
-120.154

-120.145

-120.136

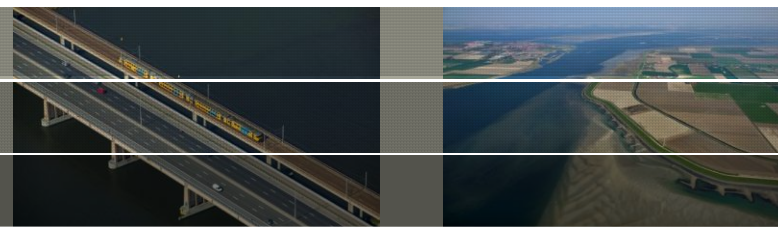


Area: 211.06 ha





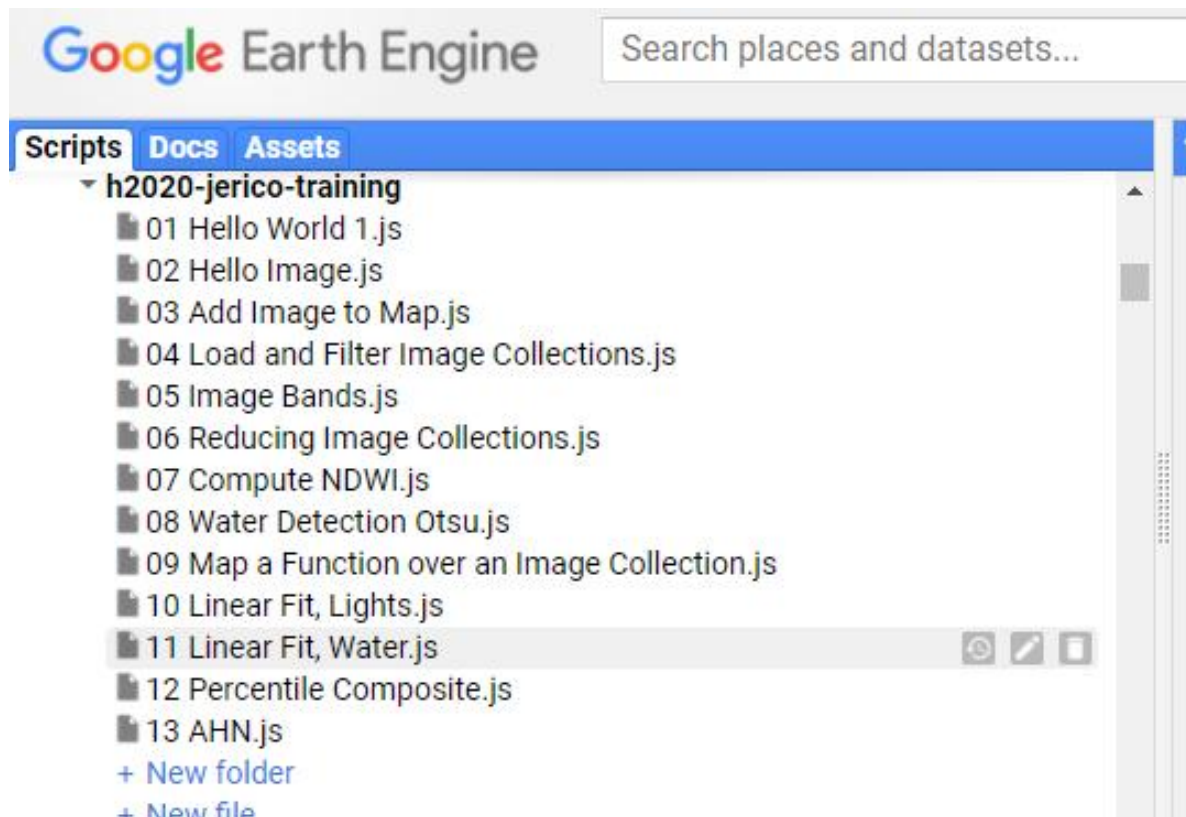
Exercices



Join Group: <https://groups.google.com/forum/#!groupsettings/jerico-next-summerscholar-2017>

Scripts repository: https://code.earthengine.google.com/?accept_repo=h2020-jerico-training

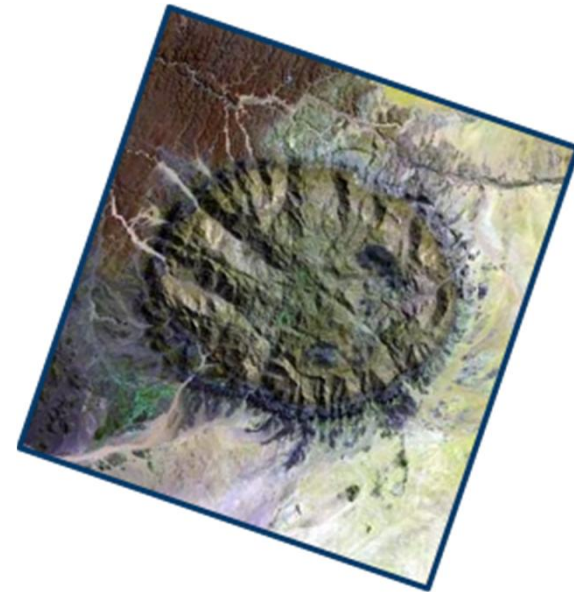
Slides: <http://bit.ly/jerico-next-gee-slides> ~120Mb



Google Earth Engine
Introduction to
planetary-scale
geospatial
analysis



Get an Image



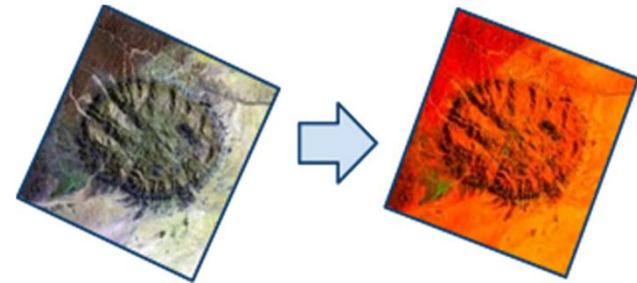
Pick your: projection, resolution, bands, bounding-box, visualization
Access as: Web map, KML, GeoTIFF

Global-scale Algorithm Processing



Get an Image

Apply an algorithm to an image



Library functions or script your own.

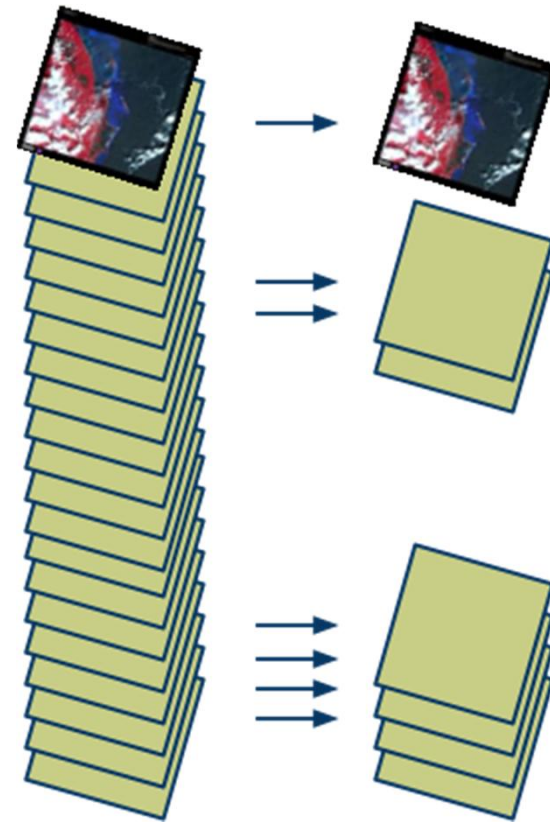
Global-scale Algorithm Processing



Get an Image

Apply an algorithm to an image

Filter a collection



Time, Space & Metadata search

Global-scale Algorithm Processing

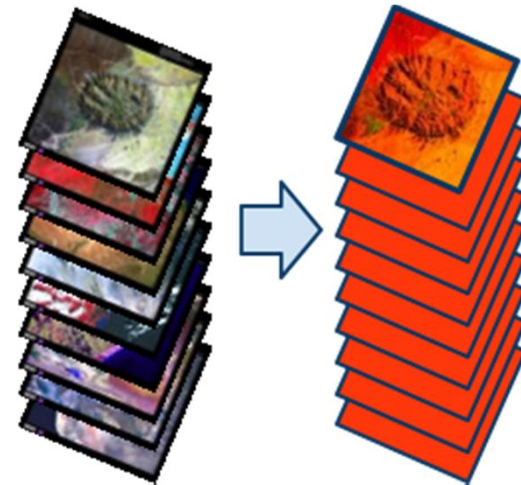


Get an Image

Apply an algorithm to an image

Filter a collection

Map an algorithm over a collection



$N \rightarrow N$

Global-scale Algorithm Processing



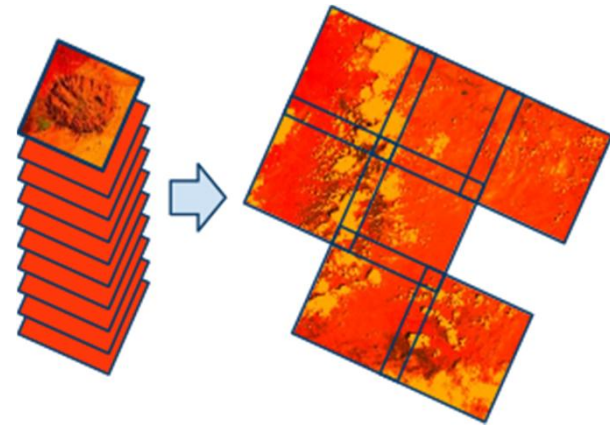
Get an Image

Apply an algorithm to an image

Filter a collection

Map an algorithm over a collection

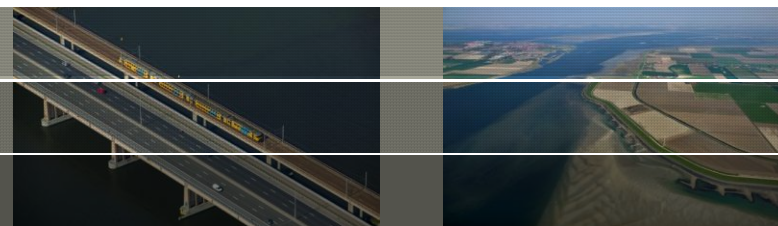
Reduce a collection



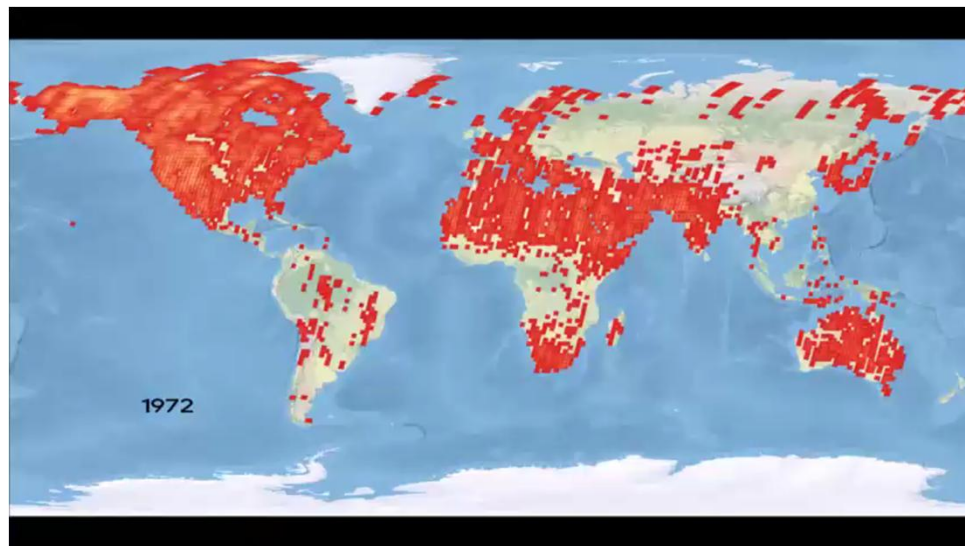
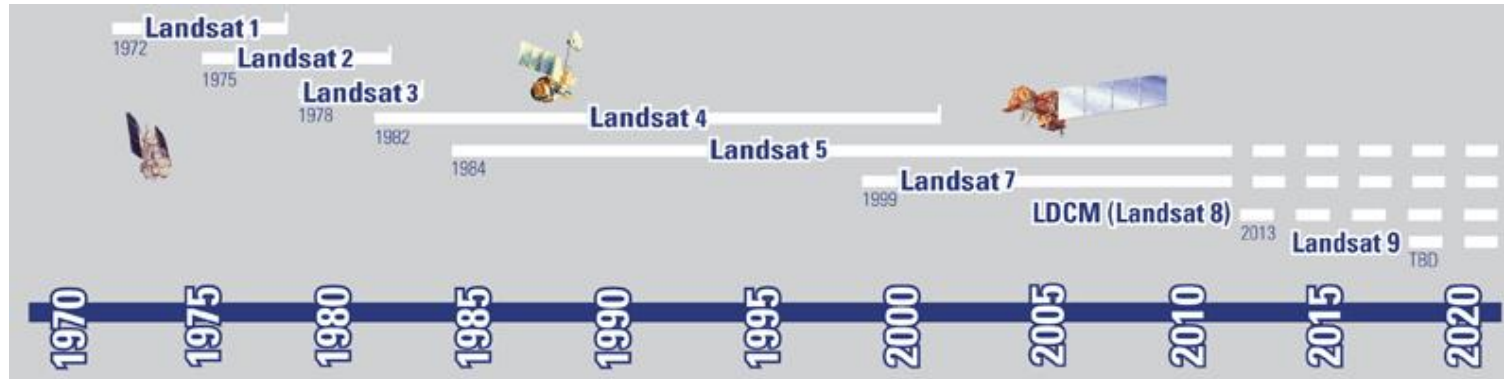
$N \rightarrow 1$ or $N \rightarrow M$



Optical Satellite Image Analysis

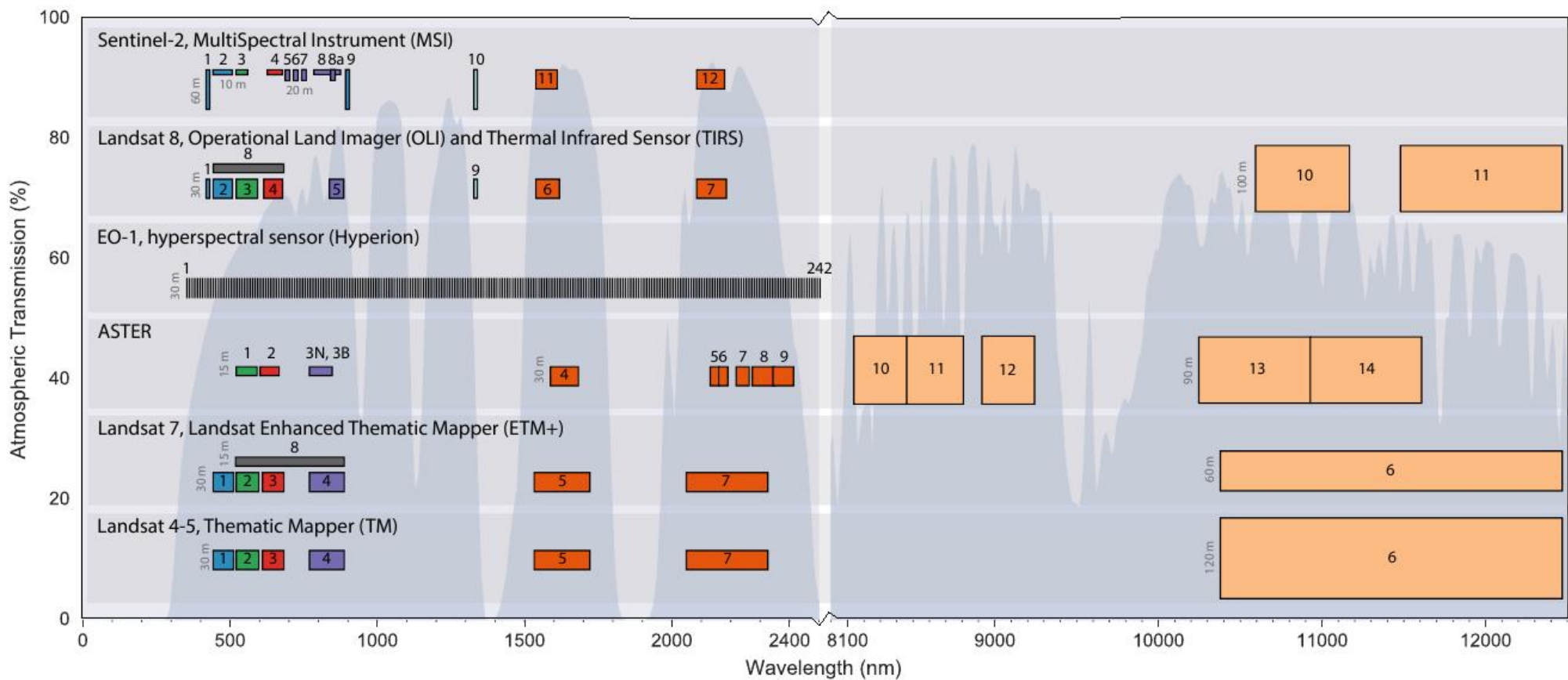


LANDSAT mission

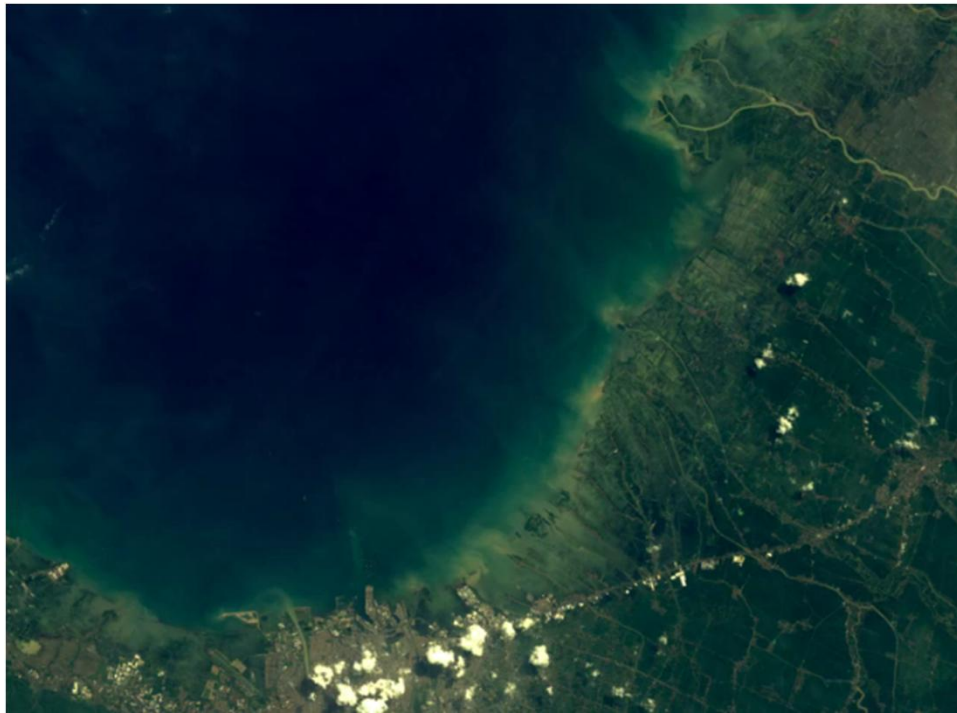
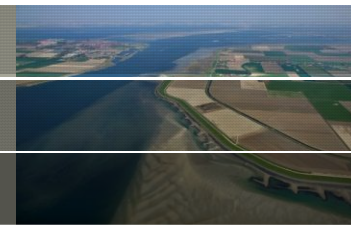


https://youtu.be/YP0et8I_bvY - true/false color composites

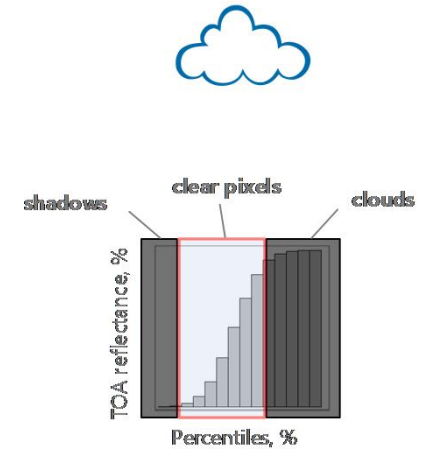
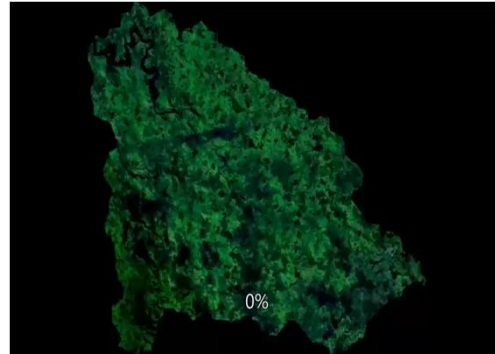
https://youtu.be/YP0et8I_bvY - landsat orbits



Visible and false color



The actual images vs. percentile composites



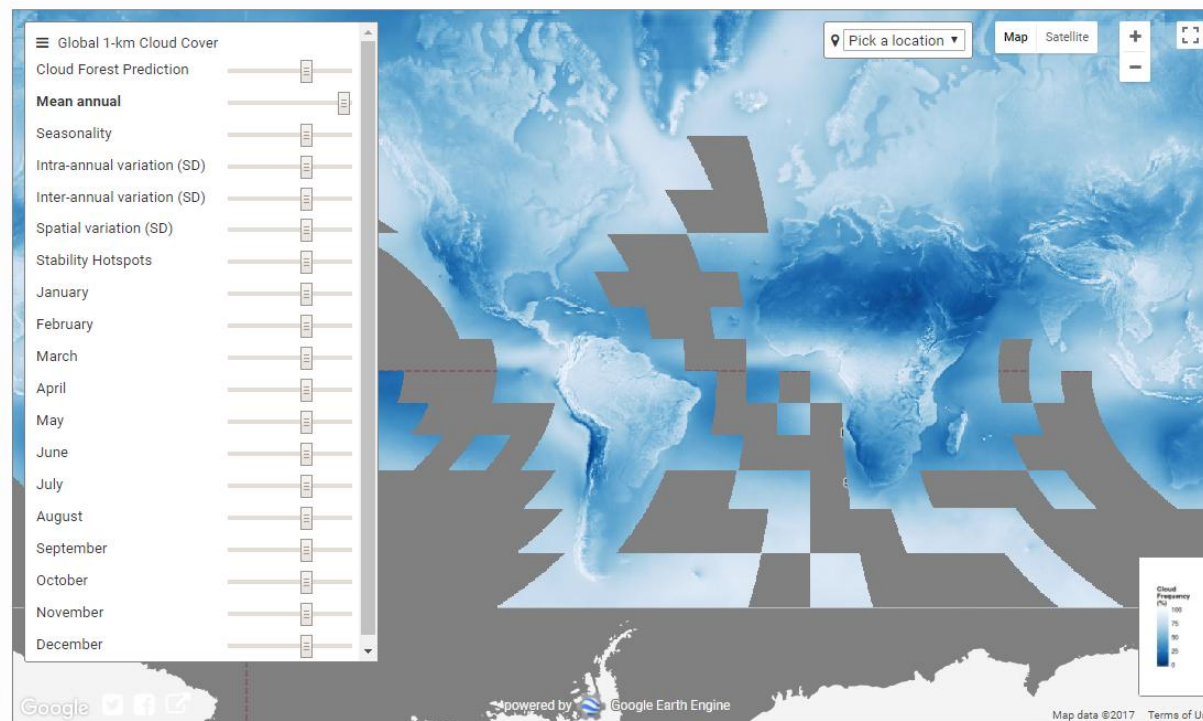
Global 1-km Cloud Cover

The datasets integrate 15 years of twice-daily remote sensing-derived cloud observations at 1-km resolution. For additional information about the integration approach and the evaluations of the datasets, please see the associated journal article:

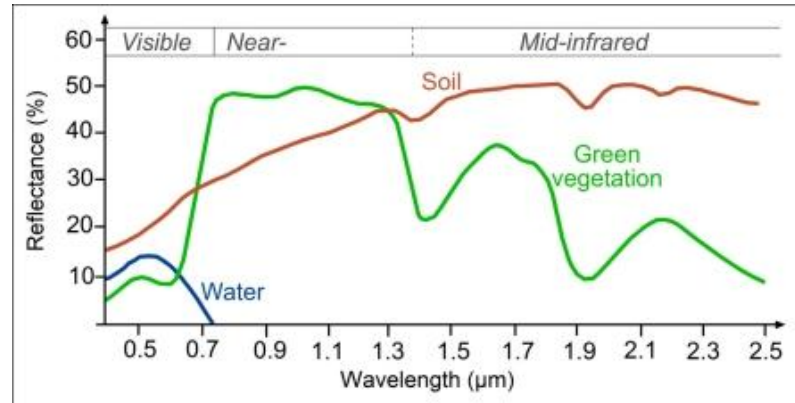
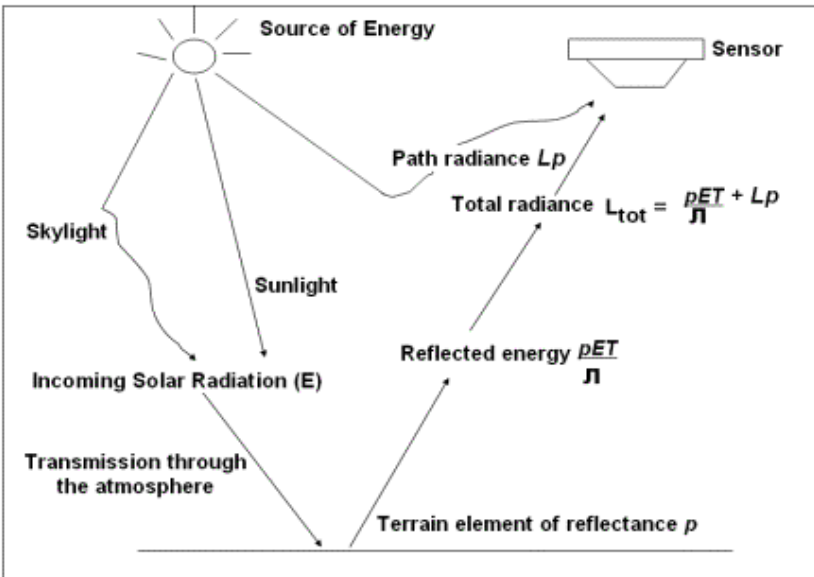
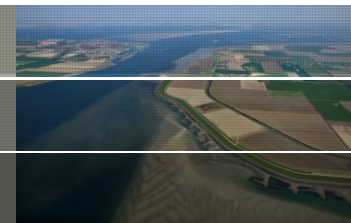
Wilson AM, Jetz W (2016) Remotely Sensed High-Resolution Global Cloud Dynamics for Predicting Ecosystem and Biodiversity Distributions. *PLoS Biol* 14(3): e1002415. doi:10.1371/journal.pbio.1002415

Dataset Details

Cloud cover can influence numerous important ecological processes including reproduction, growth, survival, and behavior, yet our assessment of its importance at the appropriate spatial scales has remained remarkably limited. If captured over large extent yet at sufficiently fine spatial grain cloud cover dynamics may provide key information for delineating a variety of habitat types and predicting species distributions. Here we develop new near-global, fine-grain (~1km) monthly cloud frequencies from 15 years of twice-daily MODIS satellite images that expose spatio-temporal cloud cover dynamics of previously undocumented global complexity. We demonstrate that cloud cover varies strongly in its geographic heterogeneity and that the direct, observation-based nature of cloud-derived metrics can improve predictions of habitats, ecosystem, and species distributions with reduced spatial autocorrelation compared to commonly used



<http://www.earthenv.org/cloud>

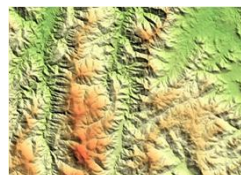


$$L_{\lambda} = \alpha_{\lambda} \times Q_{\lambda} + \beta_{\lambda}$$

$$\rho_{\lambda} = \frac{\pi \times L_{\lambda} \times d^2}{ESUN_{\lambda} \times \cos \theta_s}$$



Cloud
Snow / ice
Topographic correction
Shadows (clouds, hills)



Spatial resolution
Temporal resolution
Spectral resolution
Radiometric resolution

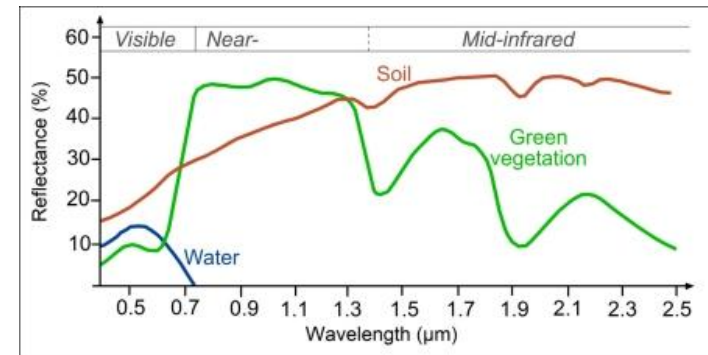
Water detection

$$NDWI_{McFeeters} = \frac{\rho_{green} - \rho_{nir}}{\rho_{green} + \rho_{nir}}$$

$$NDWI_{Xu} = \frac{\rho_{green} - \rho_{swir1}}{\rho_{green} + \rho_{swir1}}$$



Threshold?



Water detection



$$G = \sqrt{G_x^2 + G_y^2}$$
$$\Theta = \text{atan2}(G_y, G_x)$$

Canny Edge Detection

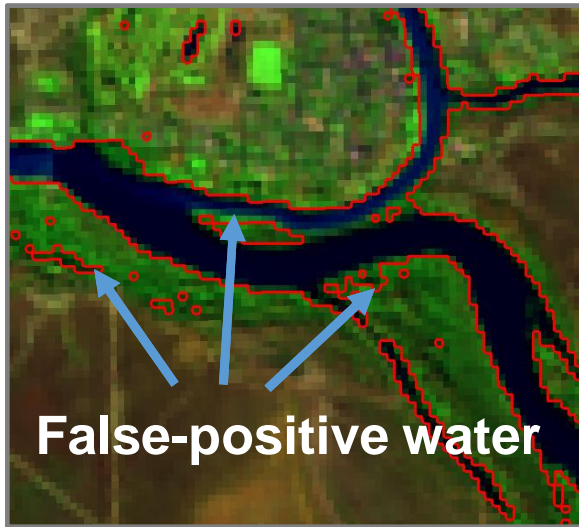


Otsu Thresholding

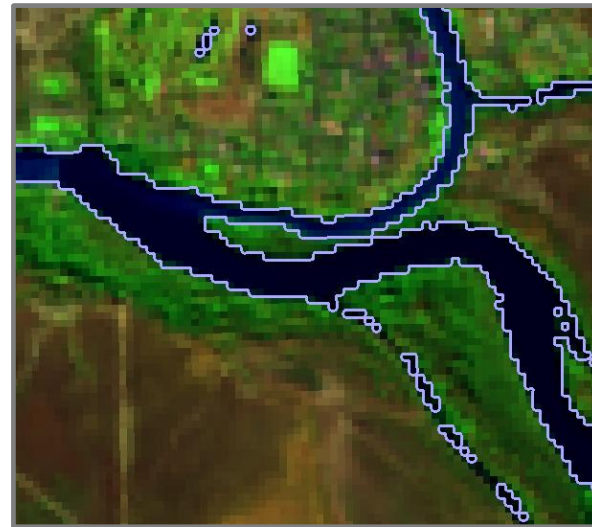


$$\sigma_w^2(t) = \omega_1(t)\sigma_1^2(t) + \omega_2(t)\sigma_2^2(t)$$

Water detection

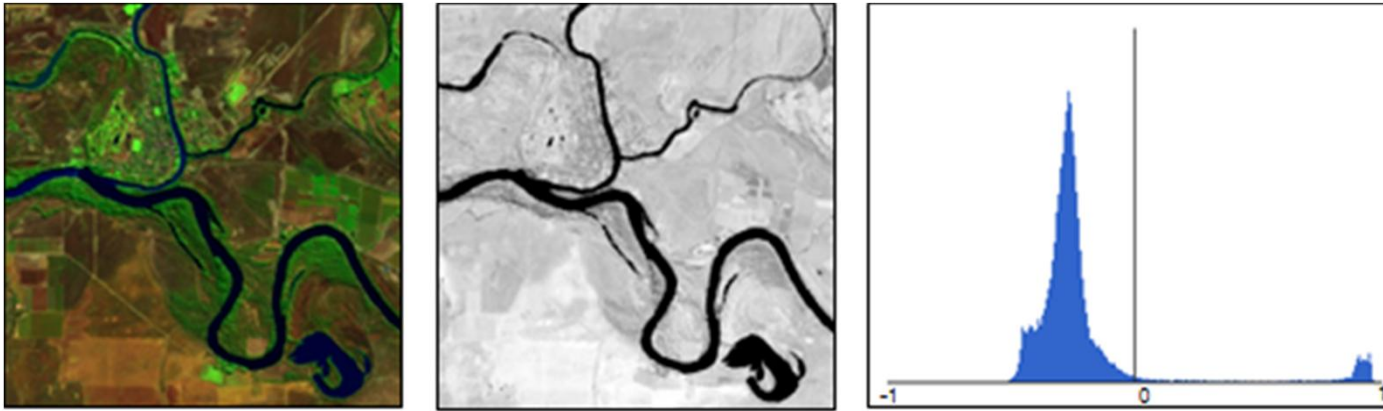


MNDWI=0



MNDWI=0.34

Water detection



Compute Index → Detect Edges → Buffer → Compute Threshold → Threshold

$$I = \frac{\rho_{green} - \rho_{swir1}}{\rho_{green} + \rho_{swir1}}$$

$$C = Canny(I, \sigma, th)$$

$$I_c = \{I | I \in C \oplus S\}$$

$$T = Otsu(I_c)$$

$$Water = \{I | I < T\}$$



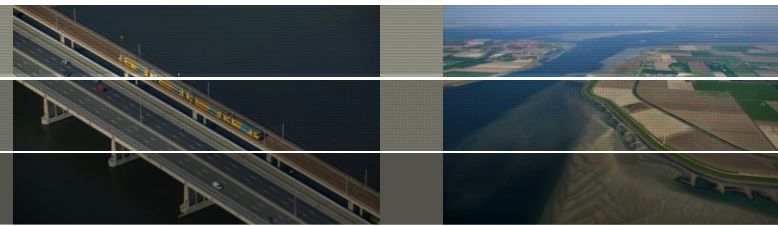
Source: <https://code.earthengine.google.com/65493c1cae67bb86c038e5bef3b4ab2d>

Python: https://github.com/gena/gena.github.io/blob/master/experiments/figure_water_detection.ipynb

Links

1. [Homepage](#) - official Google Earth Engine homepage
2. [Code Editor \(Playground\)](#) - main EE GUI (JavaScript)
3. [Developers Forum](#) - most EE discussions occur here
4. [Issue Tracker](#) - issue tracker
5. [Timelapse](#) - global timelapse video (Landsat)
6. [Private Git Sources](#) - browse your private EE git repositories
7. [Documentation](#) - Users Guide
8. [Earthengine Api](#)
9. [Selected Vector Datasets](#)
10. [Earth Engine resources for higher education](#)
11. [User Summit 2016](#)
12. [gsutil tool](#)
13. [Google Cloud Storage Documentation](#)

Selected Scripts



[Multisensor Chart](#)

[S1 ascending / descending](#)

[S1 speckle filters](#)

[OSM & SRTM](#)

[SLIC unsupervised classification](#)

[LoG sharpening](#)

[Regular Grid](#)

[Elevation Profile](#)

[PCA](#)

$$\frac{\partial I}{\partial t} = \frac{\partial^2 (c \cdot I)}{\partial x \partial y}$$
$$c = \frac{1}{1 + (G/K)^2}$$

