



Planetary Computer

Dr. Bruno Sánchez-Andrade Nuño
Principal Scientist, AI For Earth

***Computing, Data, and Cyberinfrastructure for a
Systems Approach to Studying the Earth
A Two-Part Virtual Workshop
Part 2: Friday, February 12, 2021, 12:00 - 3:00pm ET***

The National Academies of Sciences, Engineering, and Medicine committee on “*Advancing a Systems Approach to Studying the Earth: A Strategy for the National Science Foundation*” has been tasked to develop a compelling vision for a systems approach to studying the Earth in order to inform approaches to integrated **research at NSF** and to provide guidance as to how NSF can support the research community. The final report will also identify the facilities, infrastructure, coordinating mechanisms, computing, and workforce development needed to support a more **integrated approach** for studying components of the earth system. This two-part workshop seeks feedback on computational, data, and analytic support for Earth systems research, including guidance on harnessing existing, planned, and future NSF-supported cyberinfrastructure. The workshops will focus on and answer the following critical questions:

Investing to build the leading platform for technology solutions to environmental challenges



Environmental Sustainability



CARBON

2030 Carbon Negative



WATER

2030 Water positive



WASTE

2030 Waste free



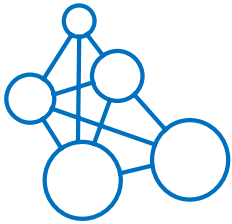
ECOSYSTEMS

Planetary Computer

2025 Land positive

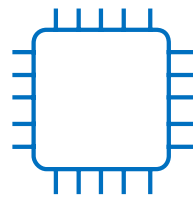
Support and advocate for
public policy initiatives

Pillars



Grants

Increase **access to cloud and AI technologies** through grants
aka.ms/ai4egrants



Technology

Build **ML tools** that support and accelerate the work of environmental scientists
aka.ms/ai4etech



Data

Host key **environmental and conservation datasets**
aka.ms/ai4edata

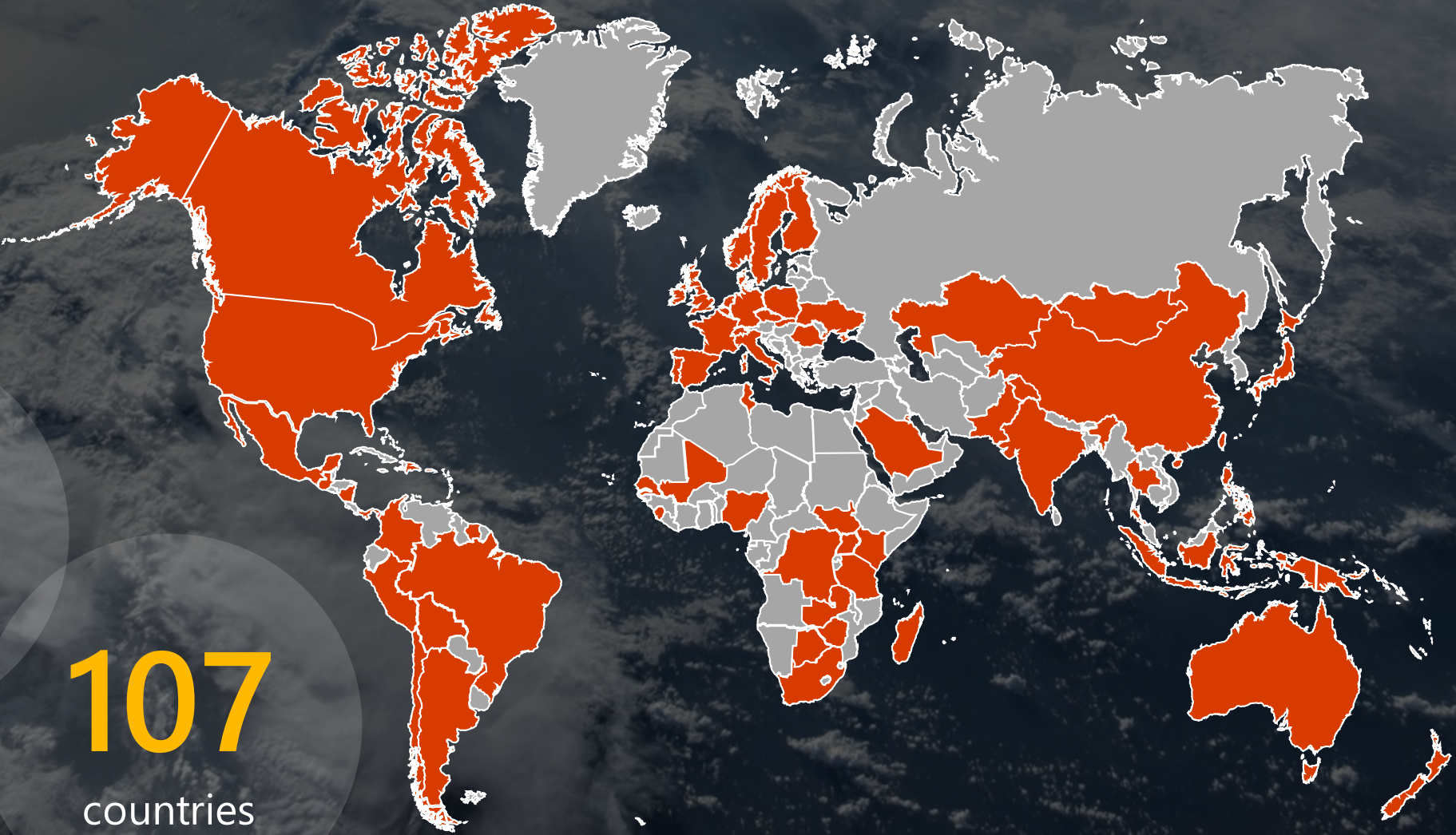
AI for Earth

701

projects

107

countries



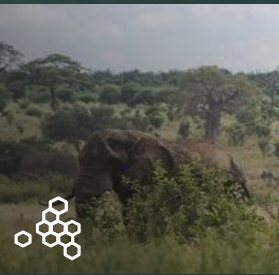
Featured AI for Earth Partners



iNaturalist



Wild Me



Conservation Metrics



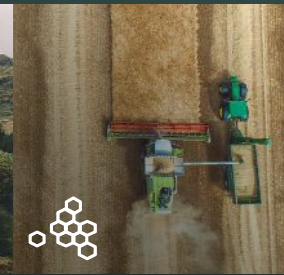
SilviaTerra



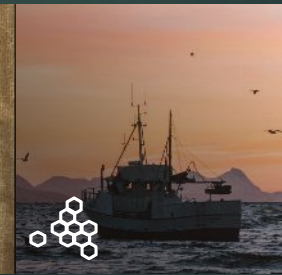
SunCulture



Terrafuse



Ag-Analytics



OceanMind



PAWS

Strategic Partnerships

TECHNOLOGY PARTNERS



SIGNATURE PROJECTS



ACADEMIC INSTITUTIONS

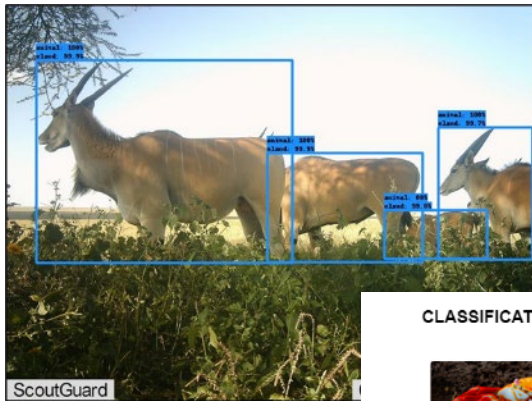


GRANTS & COMPETITIONS



AI for Earth Data Science – some examples

Accelerating camera trap surveys

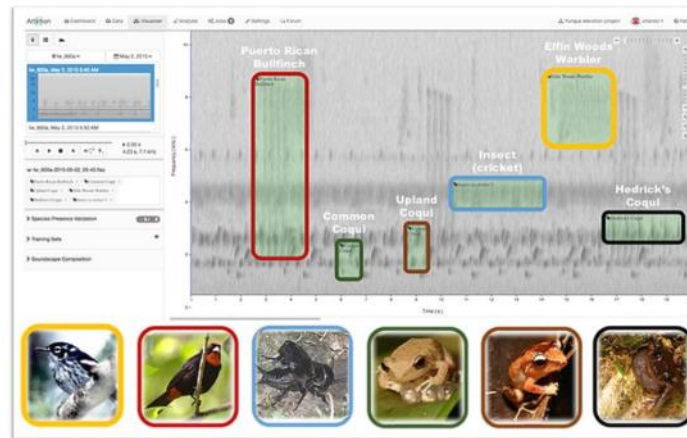


CLASSIFICATION RESULTS

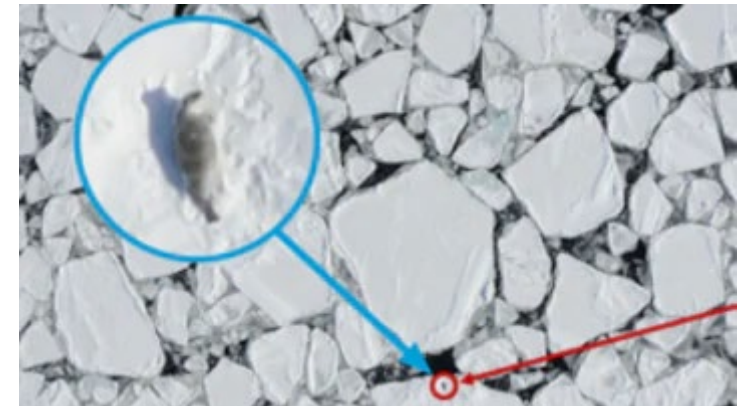


<i>Grapsus grapsus</i> (sally lightfoot crab)	83.24%
<i>Cardisoma guanhumi</i> (blue land crab)	0.91%
<i>Hemigrapsus nudus</i> (purple shore crab)	0.7%
<i>Metacarcinus magister</i> (dungeness crab)	0.68%
<i>Ocypode quadrata</i> (atlantic ghost crab)	0.61%

Multi-species bioacoustics classification for wildlife surveys

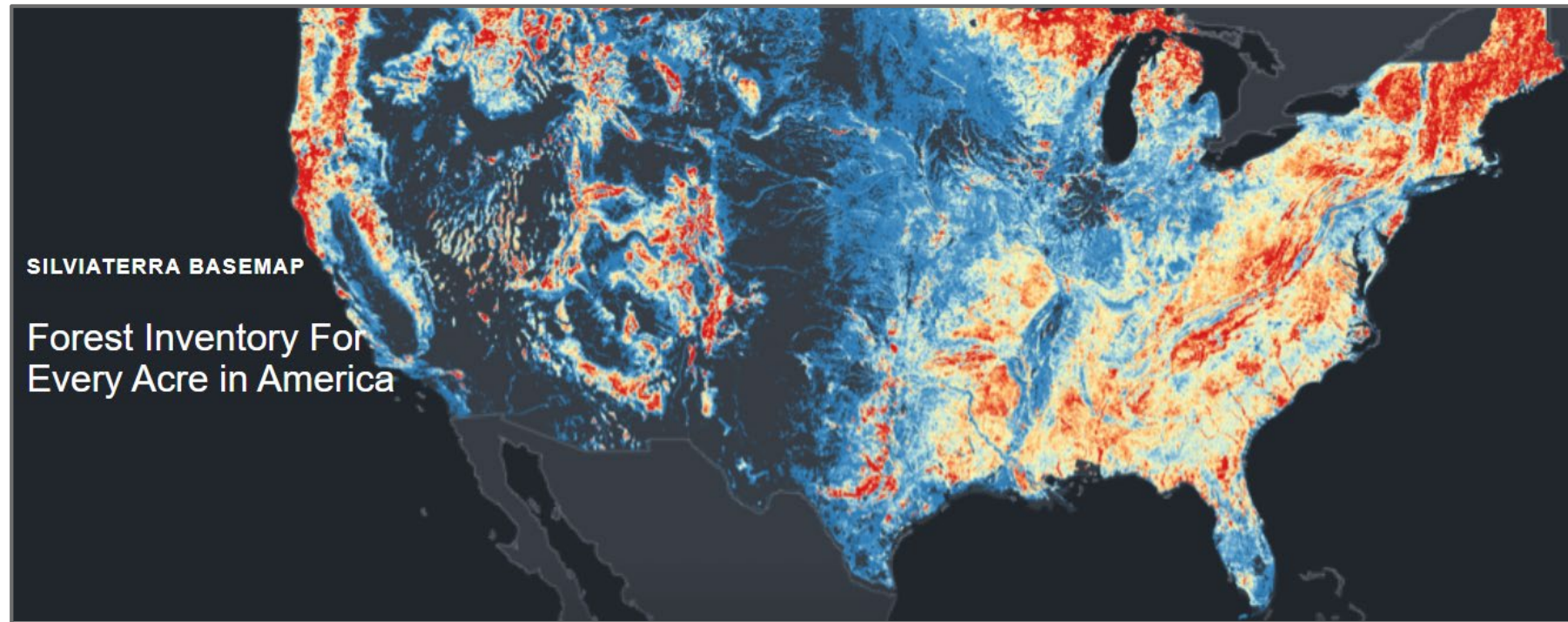


Accelerating aerial wildlife surveys in Alaska

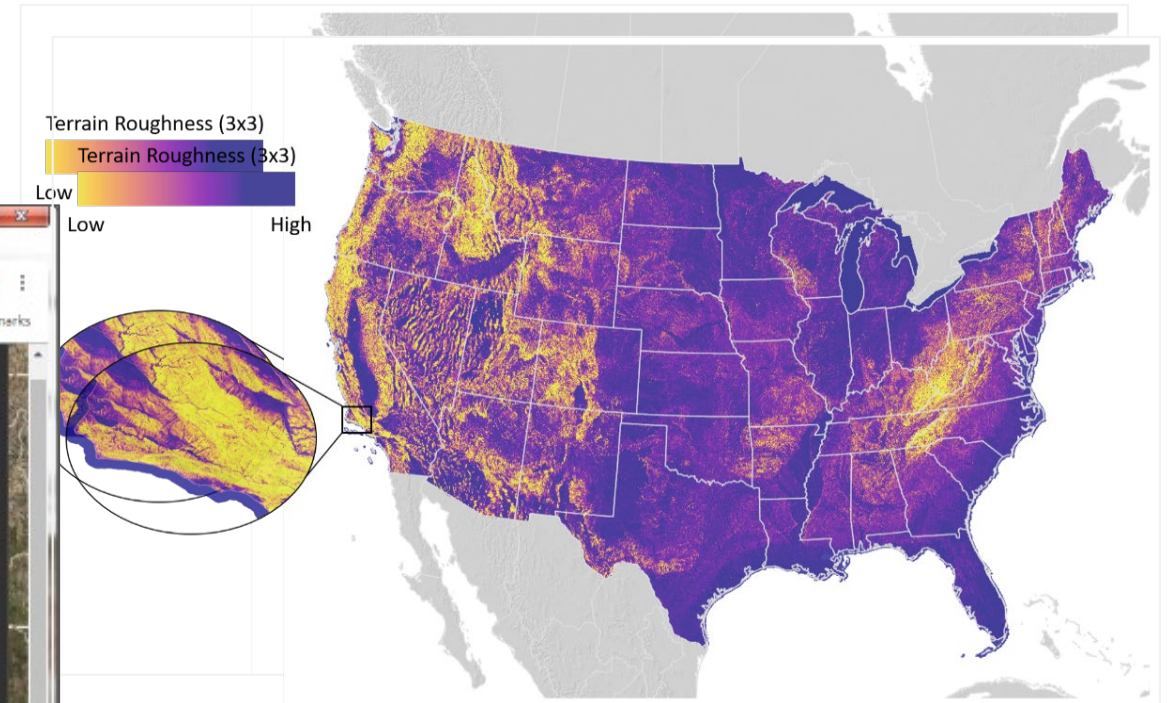
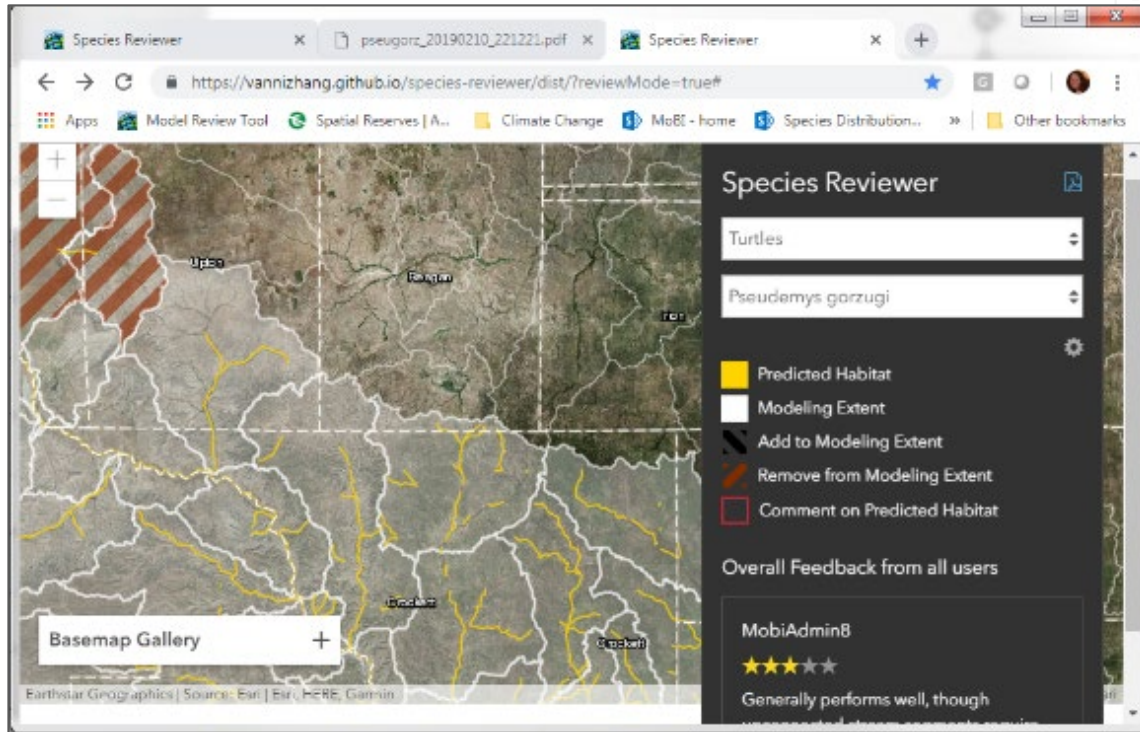


Carbon Markets: SilviaTerra

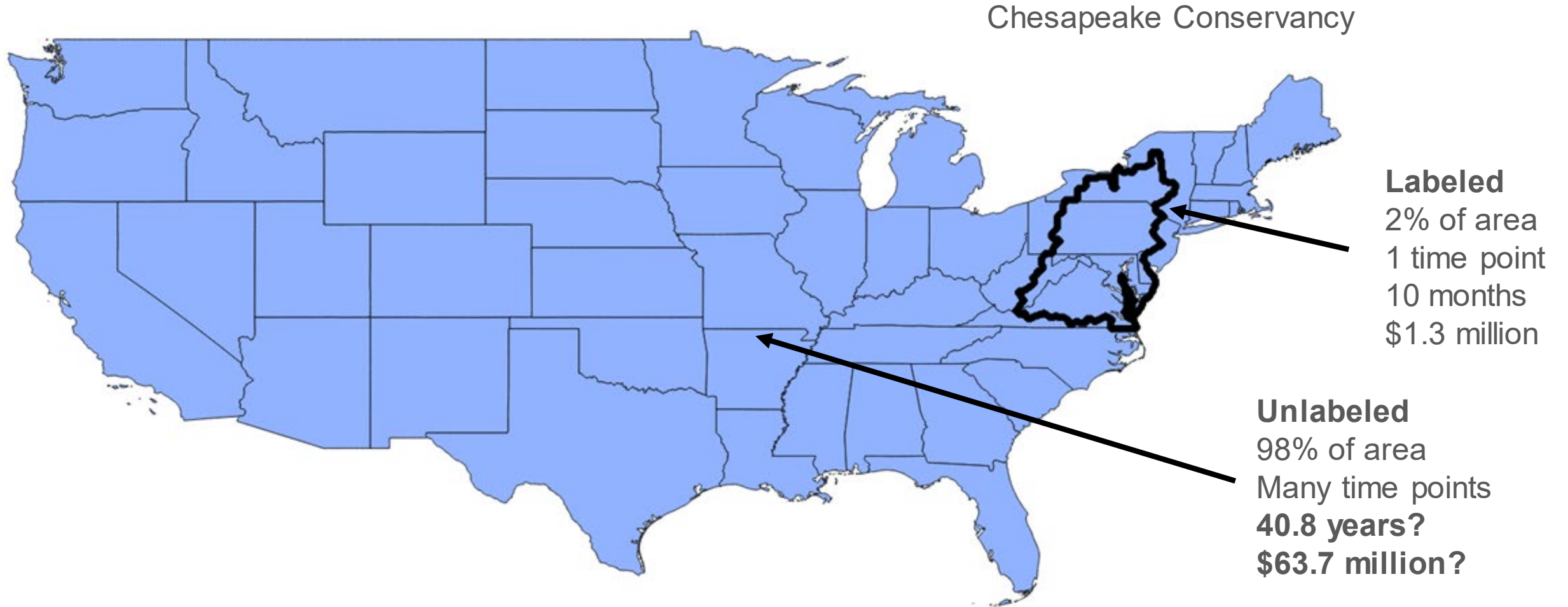
92B trees mapped on 537M acres

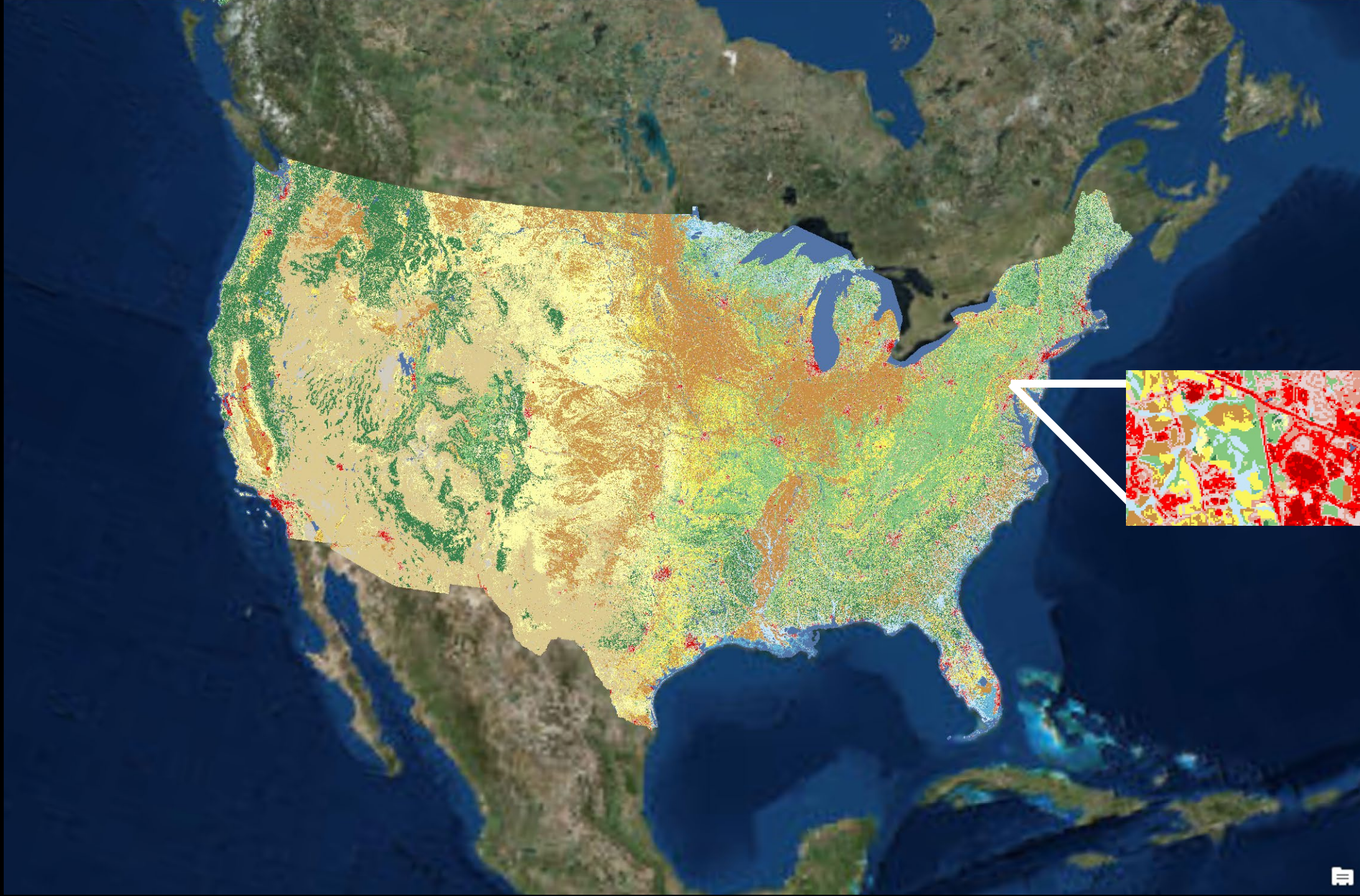


NatureServe Map of Biodiversity Importance

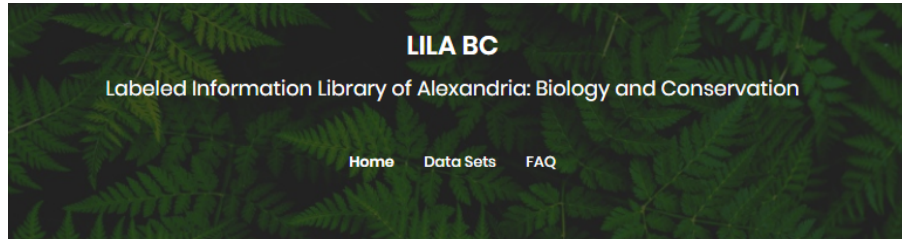


Need ML to scale human effort





LILA



Home

LILA BC is a repository for data sets related to biology and conservation, intended as a resource for both machine learning (ML) researchers and those that want to harness ML for biology and conservation.

Machine learning depends on labeled data, but accessing such data in biology and conservation is a challenge. Consequently, everyone benefits when labeled data is made available. Biologists and conservation scientists benefit by having data to train on, and free hosting allows teams to multiply the impact of their data (we suggest listing this benefit in grant proposals that fund data collection). ML researchers benefit by having data to experiment with.

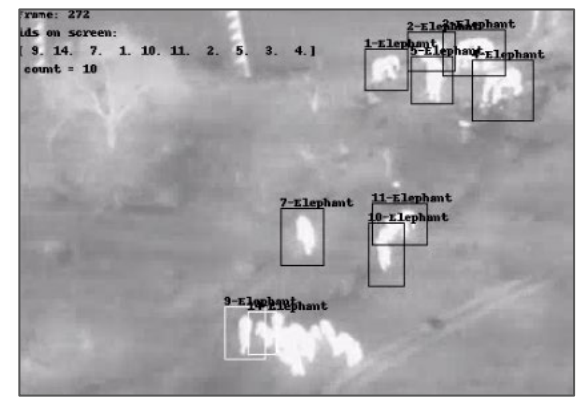
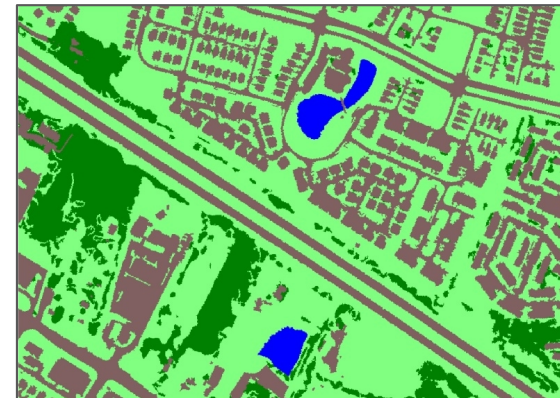
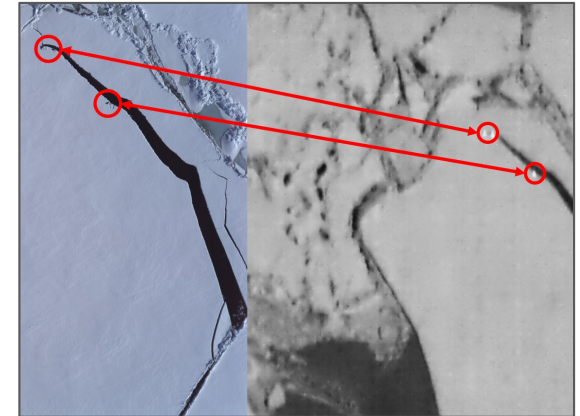
LILA BC is intended to host data from a variety of modalities, but emphasis is placed on labeled *images*; we currently host over ten million labeled images.

We ask that if you use a data set hosted on LILA BC, you give credit to the data set owner in the manner listed on the data set's landing page.

For more information, or to inquire about adding a data set, email info@lila.science.

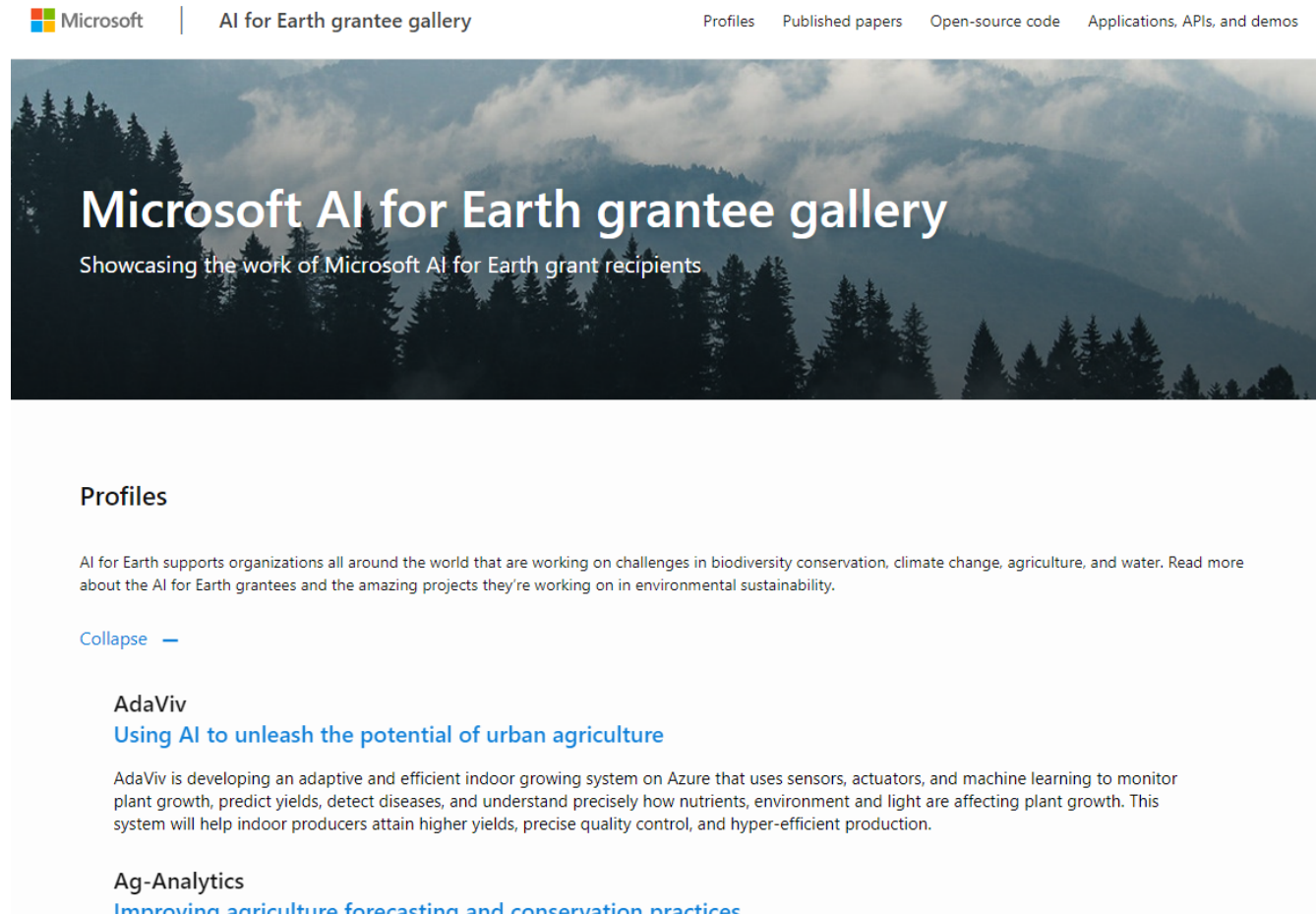
We also maintain a list of other labeled data sets related to conservation.

LILA BC is maintained by a working group that includes representatives from Zooniverse, the Evolving AI Lab at the University of Wyoming, the University of Minnesota Lion Center, Snapshot Safari, and Microsoft AI for Earth. Hosting on Microsoft Azure is provided by Microsoft AI for Earth.



<http://lila.science>

AI for Earth: grantee stories



aka.ms/ai4egrantees

AI for Earth data science:

Using ML to help conservation scientists spend *less time clicking stuff*, and *more time doing conservation*.

Accelerating *camera trap* surveys

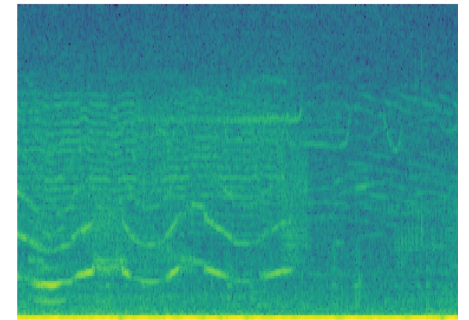
github.com/microsoft/cameratrap



Accelerating *acoustic* wildlife surveys

github.com/microsoft/belugasounds

github.com/microsoft/multi_species_bioacoustic_classification



Accelerating *aerial* wildlife surveys

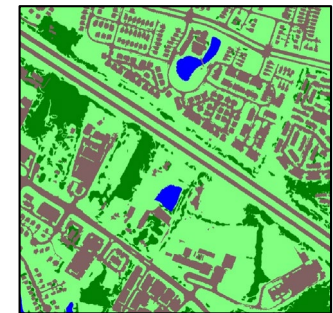
github.com/microsoft/aerial_wildlife_detection

github.com/microsoft/arcticseals



Accelerating *land cover* surveys

aka.ms/landcovermapping



All the fun animal-related work in one place

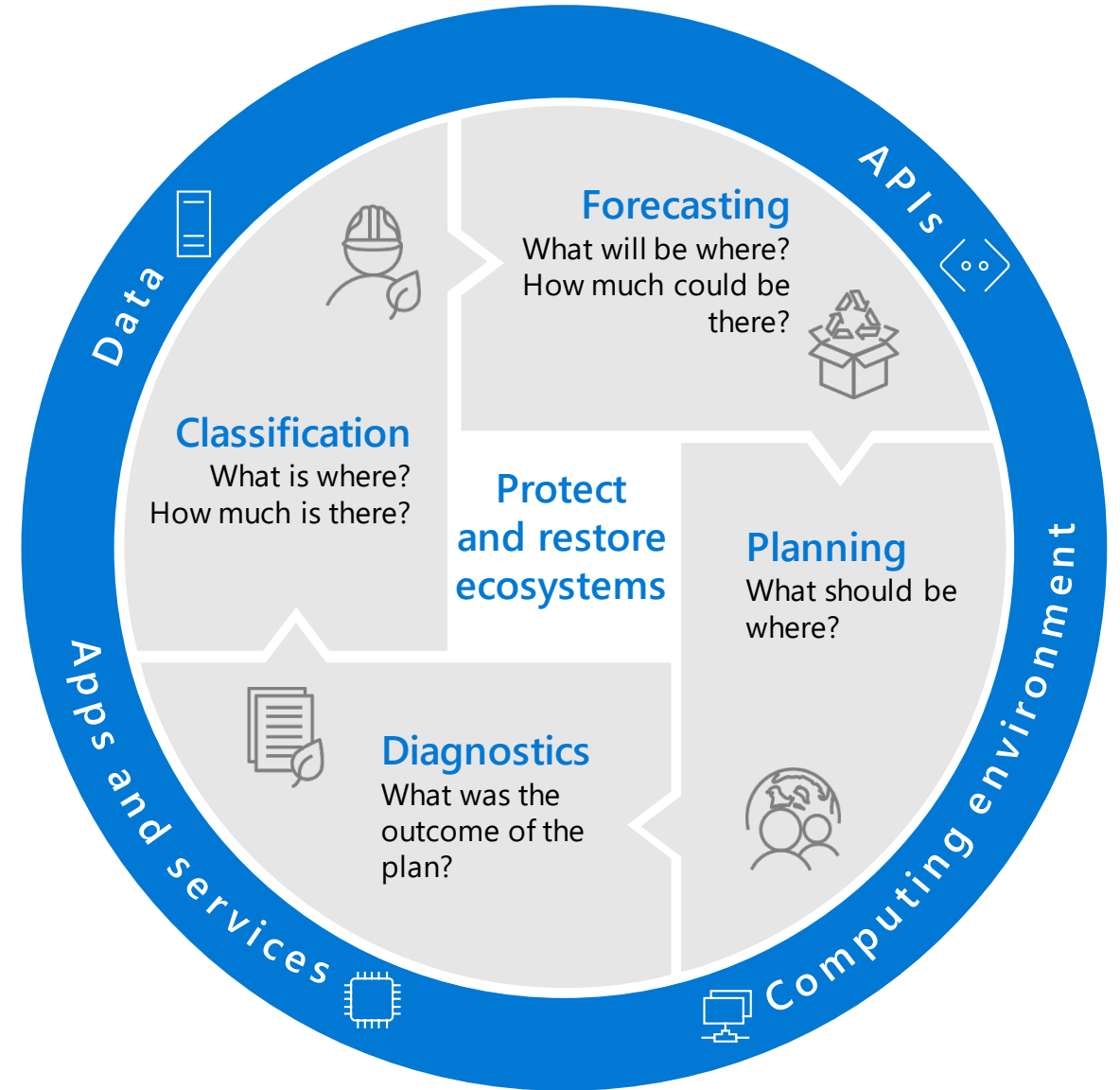
aka.ms/biodiversitysurveys

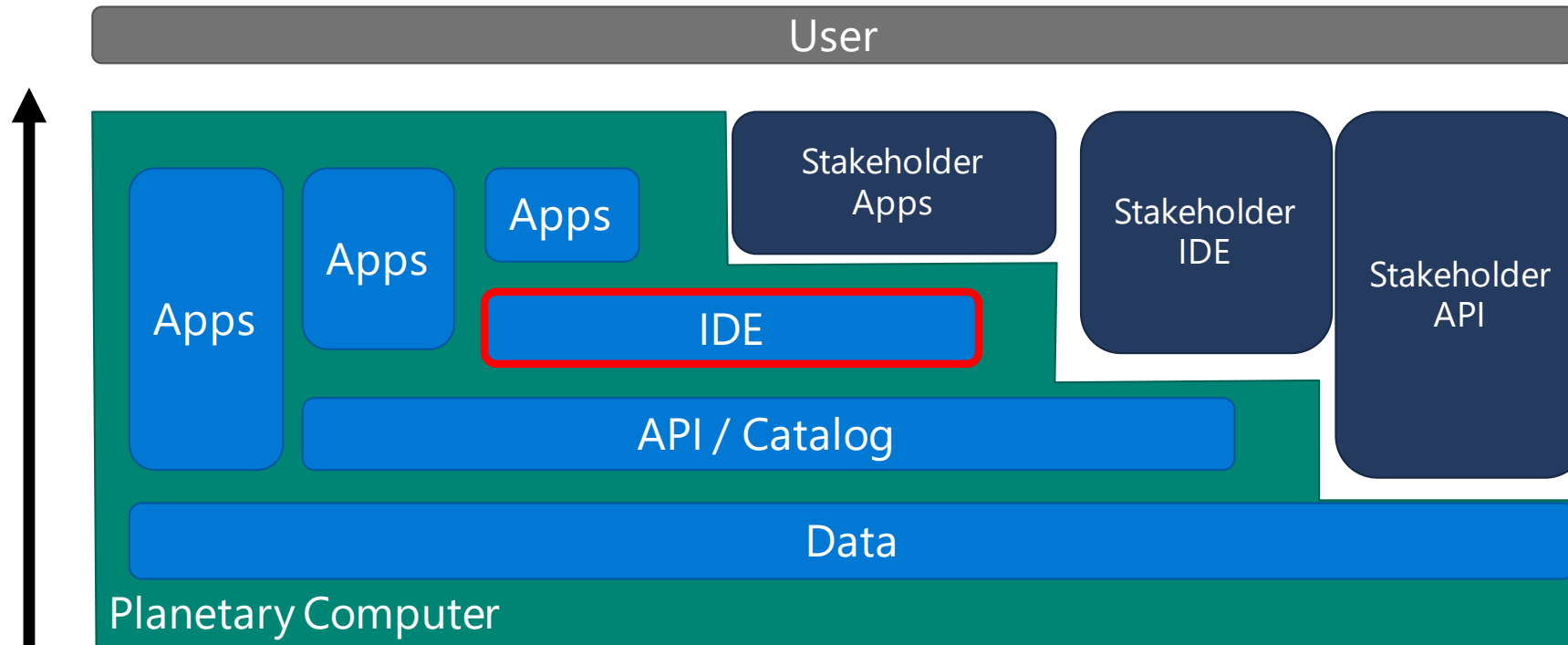
Planetary Computer

From research to
operations...

... to protect and restore
ecosystems through
cloud-based geospatial
analytics and AI.

Speed | Scale | Efficiency | Effectiveness



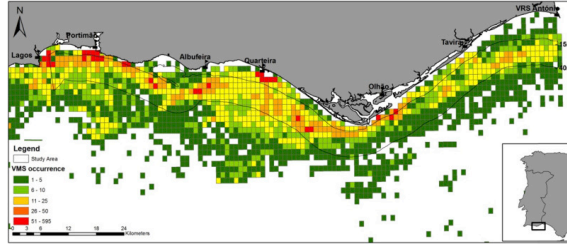


Microsoft's Planetary Computer

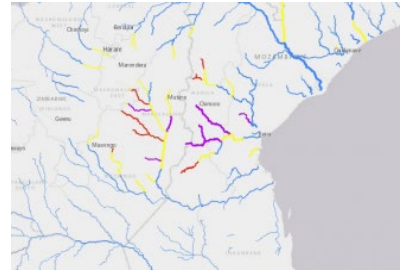
...protecting and restoring ecosystems through geospatial analysis and AI

Data	Data access APIs	Computing environment	Applications and services	
Selected data sets Remote sensing data Landsat 4, 5, 7, 8 Sentinel-1, -2, -3, -5P MODIS GOES-16, -17 NAIP ASTER Weather/climate data CMIP6, ERA5 GFS, ISD, NEXRAD, GHE Land cover data CCI, Corine, CCAP, NLCD, CDL, USGS GAP	Key features Spatiotemporal queries ...for tiles ...for stitched pixels STAC catalog Tiling/visualization Change notifications Access to Esri Living Atlas and user-imported data	Key features Jupyter front-end Synchronous batch processing Asynchronous batch processing Single login/auth for user-provided data and access to Living Atlas/AGOL data Data export to blob storage	Classification What exists at (x/y/t), for $t \leq 0$? Initial applications... Land use assessment Biodiversity surveys Human modification evaluation	Forecasting What might exist at (x/y/t), for $t > 0$? Initial applications... Flood risk mapping Deforestation prediction Water availability prediction
			Planning What do we <i>do</i> about a change in $f(x,y)$? Initial applications... Conservation spatial planning Prioritizing carbon offset projects	Diagnostics What is causing $f(x,y)$ to change or not change? Initial applications... GHG emissions monitoring Carbon offset program evaluation

Some of our launch applications



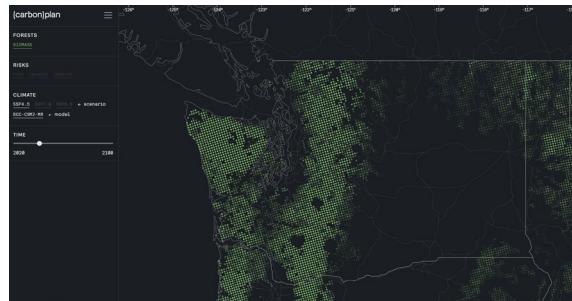
Conservation planning



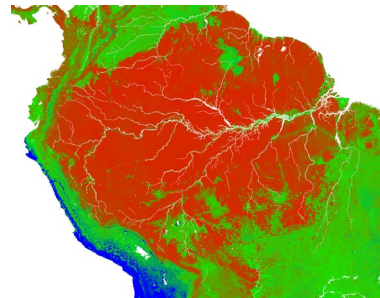
Water availability forecasting



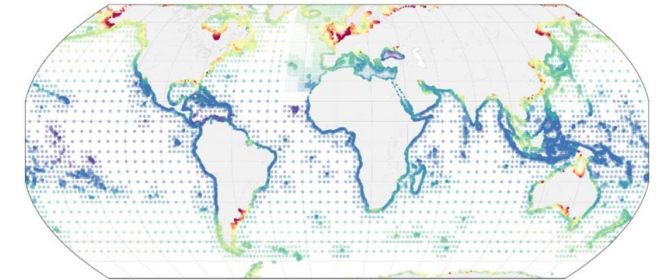
Land cover mapping



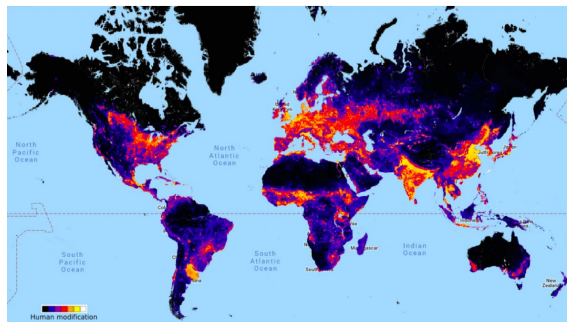
Forest carbon offset risk



Deforestation prediction



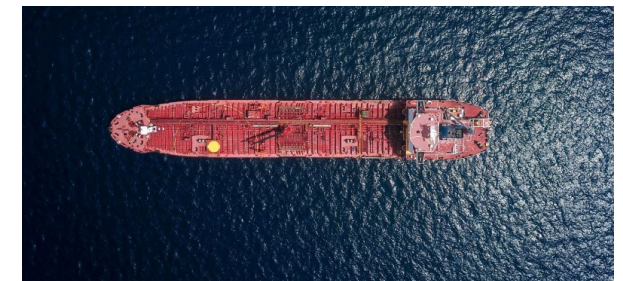
Flood risk assessment



Human footprint mapping



Accelerating biodiversity surveys



Shipping emissions estimation

Planetary Computer Data

...hosting geospatial data sets on Azure that are critical to sustainability and environmental science



Remote sensing data

- Landsat 4, 5, 7, 8
- Sentinel-1, -2, -3, -5P
- GOES-16, -17
- MODIS, NAIP, ASTER



Weather/climate data

- CMIP6, ERA5, GFS, ISD, NEXRAD, GHE



Land cover data

- CCI, Corine, CCAP, NLCD, CDL, USGS GAP

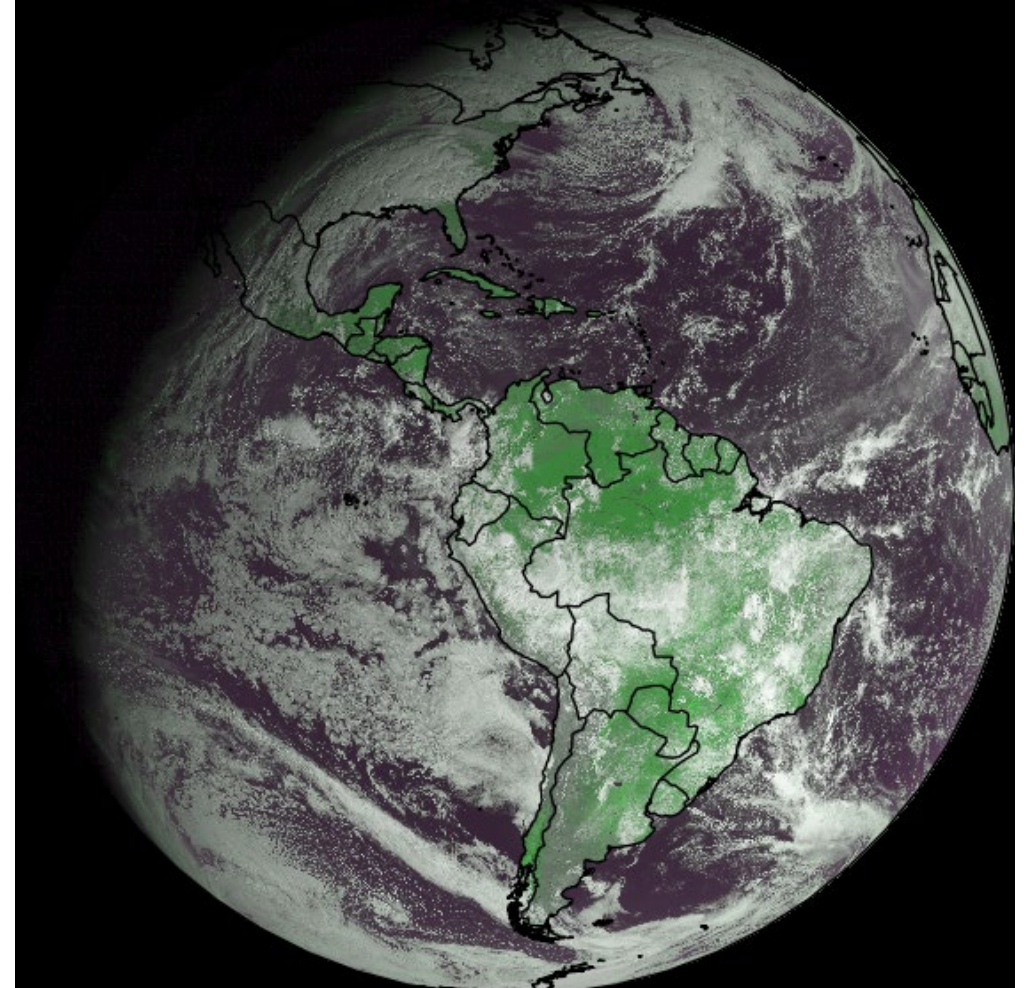


External data sources

- Esri Living Atlas



User-contributed data



Planetary Computer APIs

...facilitating discovery and access for large geospatial data catalogs

Spatiotemporal queries for images

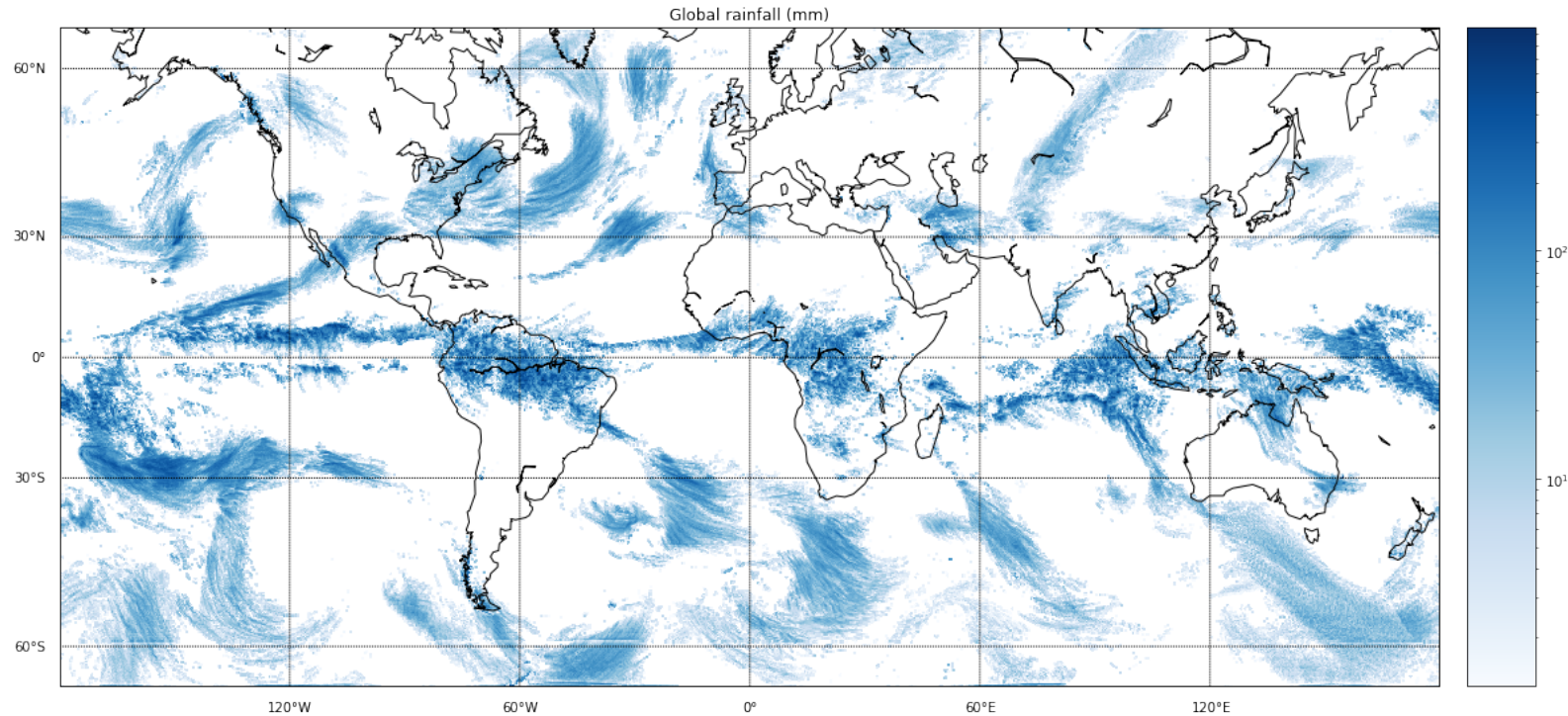
Spatiotemporal queries for stitched pixels

Data discovery via a public STAC catalog

Tiling/visualization services

Change notifications

Resampling to new grids



Planetary Computer Computing Environment

...abstracting away the complexity of distributed processing for global-scale analysis



Jupyter front-end



Compute provisioning



Synchronous processing



Asynchronous batch processing and job management



Single login/auth for Microsoft-hosted data, user-provided data, and Esri Living Atlas data



Data export to blob storage

```
return destination_filename
```

Access and plot a MODIS tile

```
In [3]: # Files are stored according to:
# http://modis.blob.core.windows.net/[product]/[htile]/[vtile]/[year][day]/filename

# Surface reflectance
product = 'MCD43A4'

# Let's look at the tile containing Chicago, IL, on May 15, 2019 (day of year 135)
h,v = lat_lon_to_modis_tile(41.881832,-87.623177)
daynum = '2019135'
folder = product + '/' + '{:0>2d}/{:0>2d}'.format(h,v) + '/' + daynum

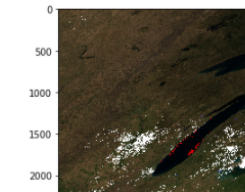
# Find all .tiff files from this tile on this day, one file per channel
files = list_tiff_blobs_in_folder(modis_container_name, folder)

norm_value = 4000

# Channel 7 in a MCD43A4 file corresponds to MODIS band 1.
#
# Let's map bands 1, 4, and 3 (channels 7,10,9) to RGB.
channels = [7,10,9]
image_data = []
for ifn in channels:
    remote_fn = files[ifn]
    url = modis_blob_root + '/' + remote_fn
    fn = download_url(url)
    raster = rasterio.open(fn, 'r')
    band_array = raster.read(1)
    raster.close()
    band_array = band_array / norm_value
    image_data.append(band_array)
rgb = np.dstack((image_data[0], image_data[1], image_data[2]))
np.clip(rgb, 0, 1, rgb)
plt.imshow(rgb)
```

```
Downloading file MCD43A4.A2019135.h11v04.006.2019149220457.hdf_08.tiff...done, 11546274 bytes.
Downloading file MCD43A4.A2019135.h11v04.006.2019149220457.hdf_11.tiff...done, 11546274 bytes.
Downloading file MCD43A4.A2019135.h11v04.006.2019149220457.hdf_10.tiff...done, 11546274 bytes.
```

```
Out[3]: <matplotlib.image.AxesImage at 0x223970bbc48>
```



Planetary Computer Application Area: Classification

What is where, and how much is there?

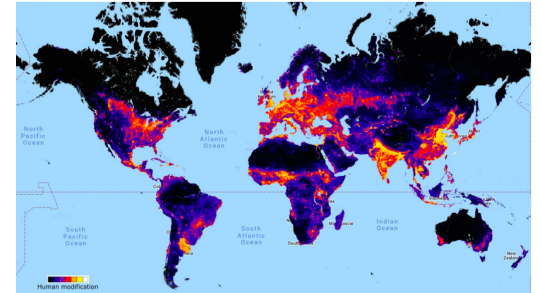
Example application: land use and land cover assessment

Accelerating geospatial image analysis with cloud-based AI



Example application: human modification evaluation

Mapping the footprint of human land use



Example application: accelerating biodiversity surveys

Leveraging AI to help conservation biologists spend more time planning conservation, and less time annotating data



Planetary Computer Application Area: Forecasting

What might be where in the future?



Example application: flood risk mapping

Leveraging historical data and modeled climate data to assess flood risk zones



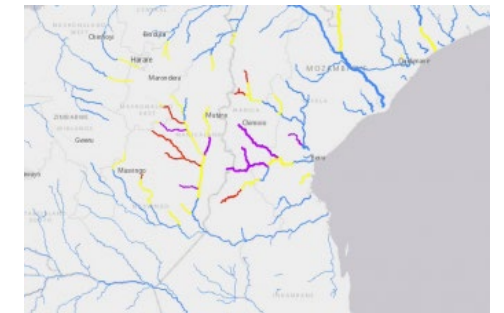
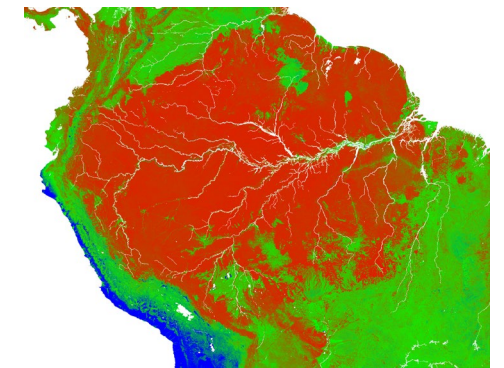
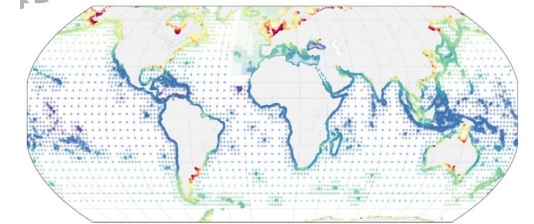
Example application: deforestation prediction

Leveraging remote sensing data to facilitate rapid response to expected tree canopy loss



Example application: water availability prediction

Leveraging weather imagery and ground station data to predict surface water availability



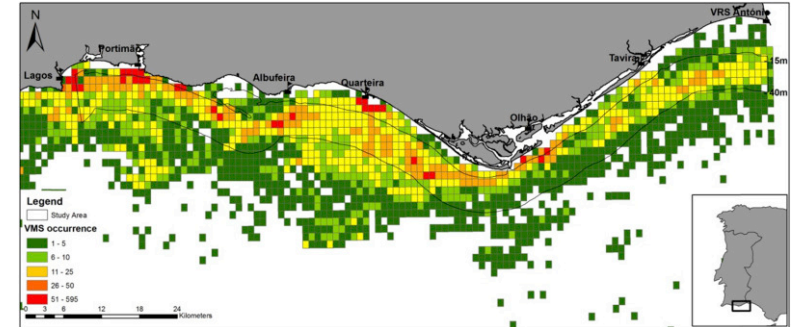
Planetary Computer Application Area: Planning

What land planning policies will optimally protect natural carbon sinks and global biodiversity?



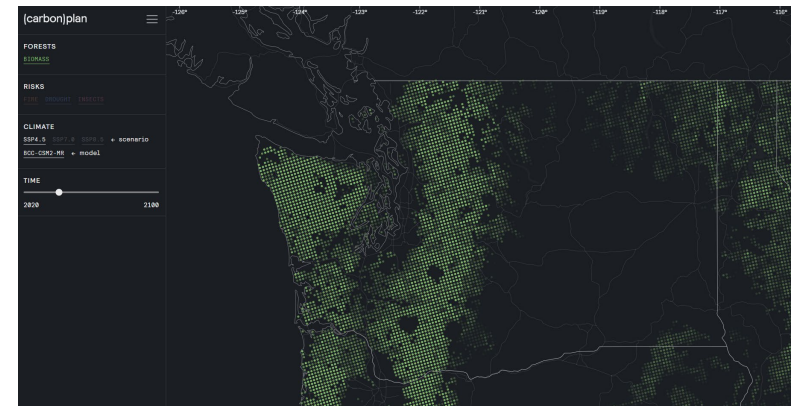
Example application: conservation spatial planning

Using global biodiversity data and remote sensing data to plan conservation policy in critical ecosystems



Example application: prioritizing carbon offset projects

Leveraging climate and fire risk models to inform decisions about forest protection projects



Planetary Computer Application Area: Diagnostics

What was the actual impact of intentional or unintentional human intervention on ecosystems and climate?

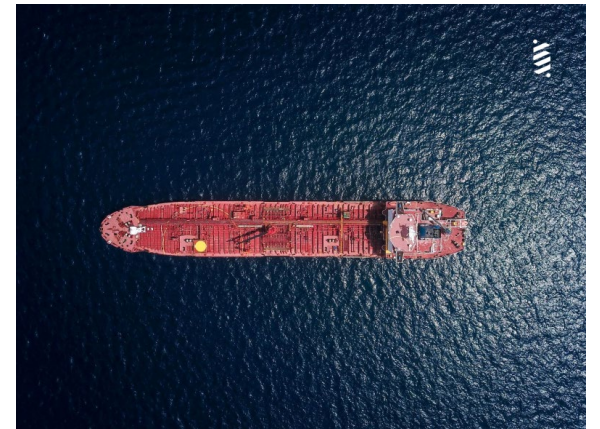
Example application: carbon offset credit retrospective analysis

Reviewing carbon offset programs in light of new data, to quantify the true utility of individual programs



Example application: GHG emissions monitoring

Attributing GHG emissions responsibility to operators in the global shipping industry



Planetary Computer

From research to
operations...

... to protect and restore
ecosystems through
cloud-based geospatial
analytics and AI.

Speed | Scale | Efficiency | Effectiveness

