



# Developing Science Gateways with Clowder for the Permafrost Discover Gateway

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**National Center for  
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# Science Gateway

- **Goal:** Make models and tools developed by the PDG accessible to a wider set of user, beyond the original creators
- A **science gateway** is a resource to simplify access to community specific tools, applications and data collections for researchers, educators and students using user friendly, online interfaces



# Data Pipelines

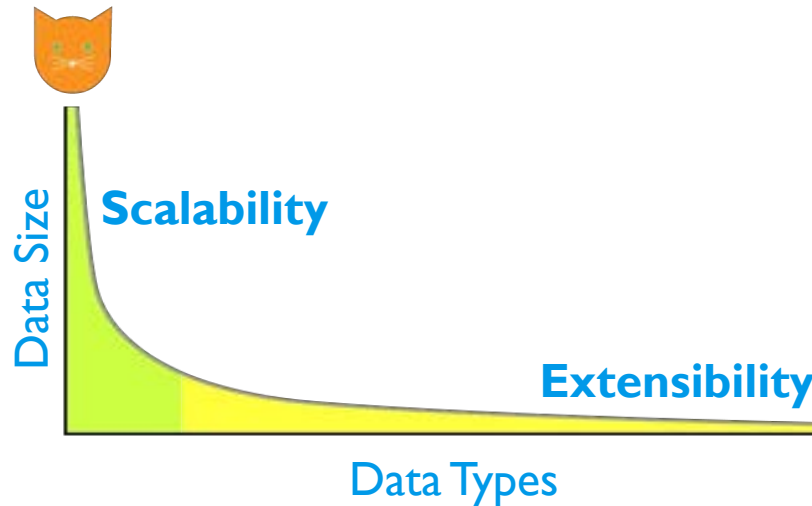
1. **Mapping Application for Arctic Permafrost Land Environment (MAPLE):** detects ice wedge polygons from high resolution optical imagery archived at Polar Geospatial Center. Uses Deep Learning Convolutional Neural Networks and requires GPU (Python, Tensorflow) (*Chandi Witharana and Team*)
2. **Permafrost Region Disturbances (PRDs):** Lake area change, fire scars and retrogressive thaw slumps from Landsat images pre-processed using Google Earth Engine, plus additional machine learning and geospatial analysis (Python, Javascript, scikit-learn) (*Ingmar Nitze*)
3. **Arctic Satellite Joint Product (ASJP):** distills the high volume of archived NASA public satellite data to the essential community variables of interest that can be made available quickly and efficiently (Fortran) (*Gala Wind*)



# Clowder and Long Tail Data

Clowder is a customizable and scalable data management framework to support any data format and multiple research domains

<https://clowderframework.org/>

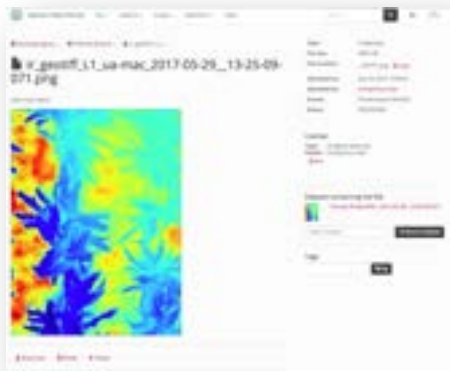


Heidorn, P. Bryan. "Shedding Light on the Dark Data in the Long Tail of Science." *Library Trends*, vol. 57 no. 2, 2008, p. 280-299. *Project MUSE*, [doi:10.1353/lib.0.0036](https://doi.org/10.1353/lib.0.0036).

# UPLOAD, SHARE, INDEX HETEROGENOUS DATA AND METADATA

- **Upload** files, tag, license and organize them in datasets and collections
- Selectively **share** datasets with collaborators and publish them to the Internet
- Add well structured **metadata** to files and datasets using JSON Linked Data (**JSON-LD**)
- Automatically trigger **execution** of custom code on files and datasets using a Cloud based extraction system
- **Visualize** data and metadata in the browser using custom previewers

File visualizations, License, tags



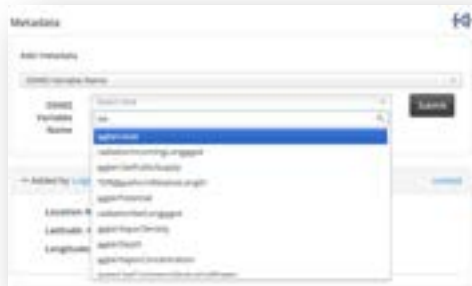
Structured metadata



User management



Controlled vocabularies



Advanced search





## Model Training - Frontera at TACC

Polar Geospatial  
Center

Satellite Imagery  
(Geotiff)

Labels: Manually  
drawn polygons  
and label each  
polygon  
(high center, low center,  
background)

Train MAPLE Model  
(Current: Trained Locally GPU  
10-12 GB)  
(Future? XSEDE? Radiant?  
HAL?)

Tensorflow Weights

## Model Execution - XSEDE Bridge2 PSC

MAPLE

## Model Execution - NCSA Radiant at NCSA

Polar Geospatial  
Center

(Geotiff)  
(20km x 75 km)

MAPLE Extractor

Ice-wedge polygons  
Shapefiles  
(class 1 high center, class 2  
low center)  
(one polygon 30m wide)

# MAPLE on XSEDE Bridges2

- Requires GPU resources not currently available on Radiant.
- Solution - use ssh to submit job to XSEDE Bridges2
- **File Extractor:** individual files are transferred to XSEDE. A slurm job is created, run when resources are available, and then uploaded.
- **Dataset Extractor:** Start with empty dataset and path to input files on bridges. Determine which files have not been run, create slurm jobs, then upload. Checks to see which files do not need to be uploaded or run.



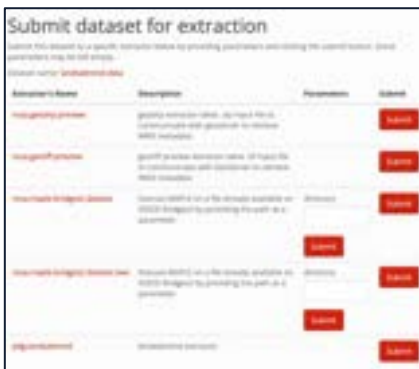


# MAPLE Clowder Extractor

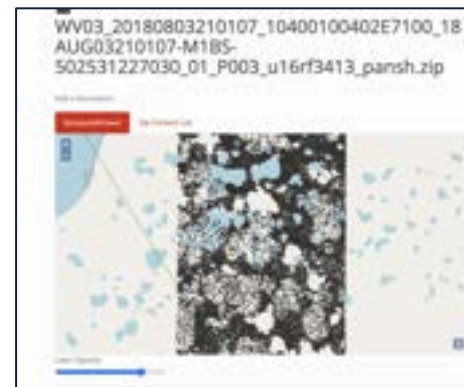
Input - GeoTiff



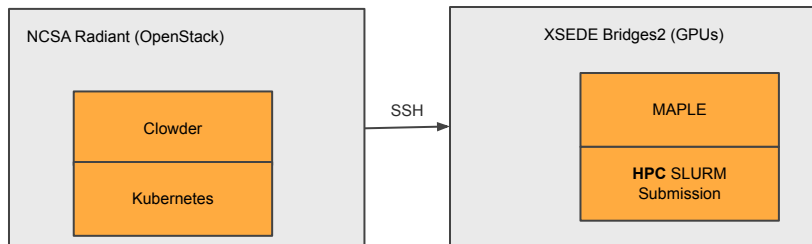
Submit



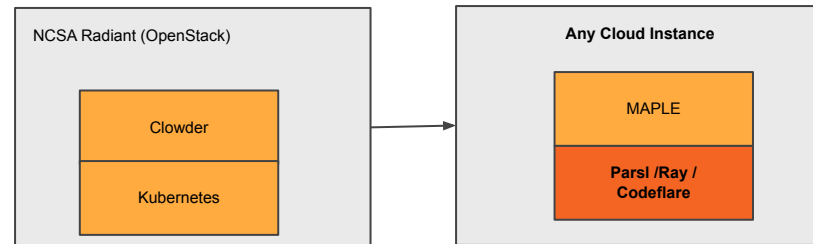
Output - Shapefile



Current



Future



# Portable Parallel Pipelines

- Prototyped a Clowder extractor to Launch Parsl jobs on local Kubernetes cluster
- Prototyped Ray extractor to launch Ray jobs on CodeFlare provisioned clusters



<https://parsl-project.org/>



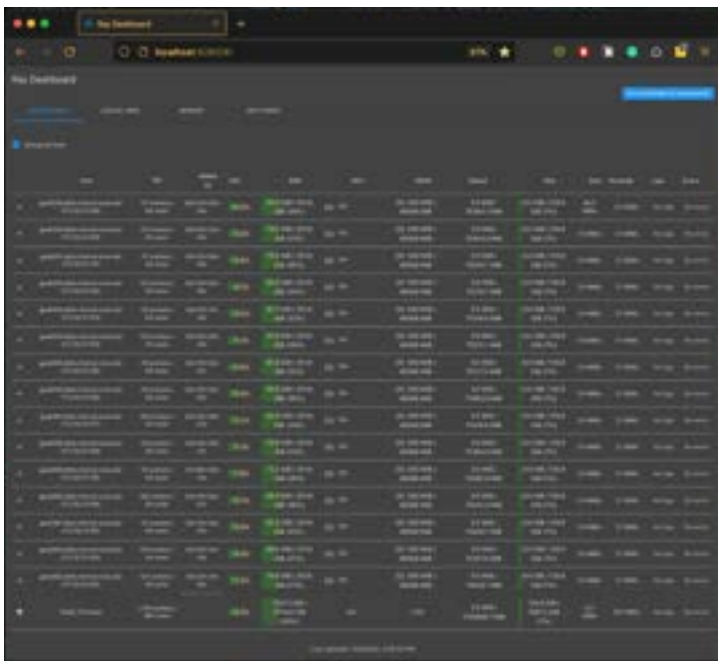
<https://www.ray.io/>



<https://codeflare.dev/>



# MAPLE Visualization Pipeline in Ray On Delta



<https://www.ncsa.illinois.edu/research/project-highlights/delta/>



## MAPLE file count

- 10s of thousands.
- Each image requires a dedicated GPU (or 8GB vram) for ML inference

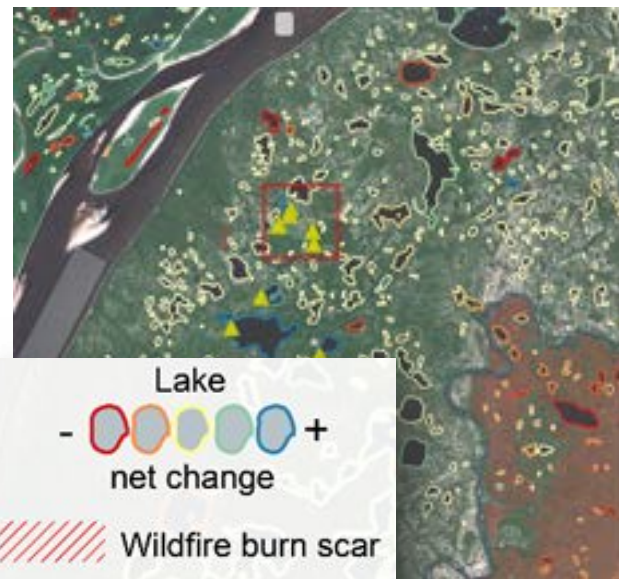
type	min_file_size	max_file_size	total_file_count
high_img	8.536743e-05	1643.908	22370
low_img	8.536743e-05	1253.986	1416
medium_img	8.536743e-05	1254.634	11209
water_clipped	8.536743e-05	1073.423	8807

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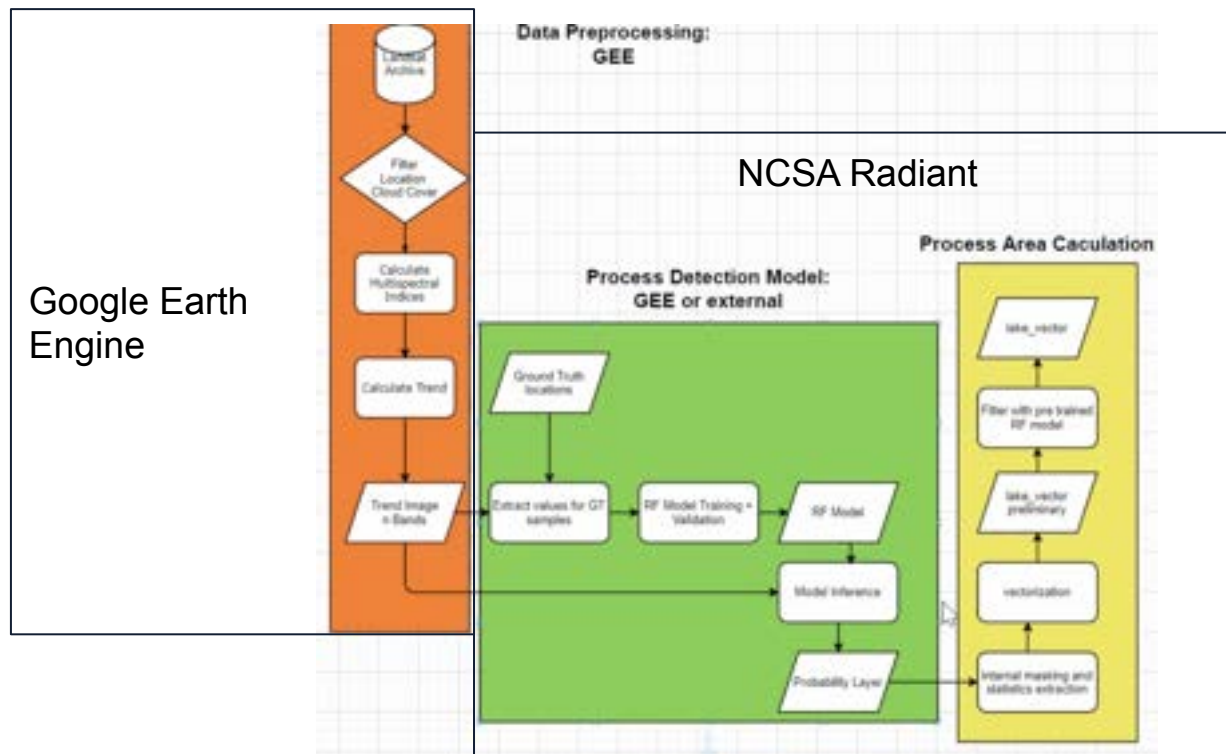


# LandsatTrends - Permafrost Region Disturbances (PRD)

- Input Landsat & Sentinel
- Identify
  - Thaw slumps
  - Fire scars
  - Lake changes



# LandsatTrends Pipeline



# LandsatTrends Extractors

## 1. Preprocessing Extractor

- a. submits preprocessing steps to Google Earth Engine using Python API
- b. Data from Google Earth Engine is transferred to Google Cloud using GC API
- c. Data in Google Cloud is uploaded to Clowder

## 2. Inference Extractor

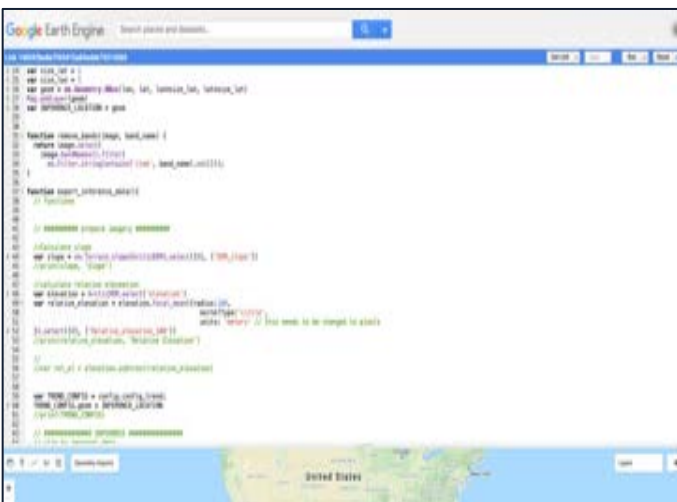
- a. Run model on local files



# Google Earth Engine API

- Instead of using the google earth console or Google drive, images can be downloaded directly to a location on disk.

GEE Javascript Console



```
Google Earth Engine - Search places and datasets.
// Import the Landsat 7 satellite imagery
var Landsat7 = ee.ImageCollection('LANDSAT/LT05/C01/T11');

// Filter for the year 2015
var Landsat7_2015 = Landsat7.filterDate('2015-01-01', '2015-12-31');

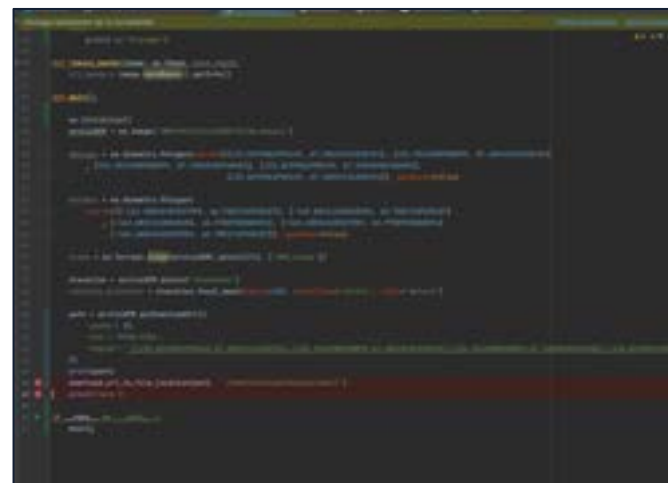
// Select the best image for each location
function selectBestImage(location) {
  return Landsat7_2015
    .select(location)
    .filterDate('2015-01-01', '2015-12-31')
    .max();
}

// Create a feature collection of the best images
var bestImages = ee.FeatureCollection(
  // Loop over the locations
  ee.List([
    // Location 1
    ee.Feature(ee.Geometry.Point(-122.5, 45.5), {
      'location': 'San Francisco'
    })
  ])
);

// Download the best image for each location
var download = ee.Function({
  // Input: location
  args: ['location'],
  // Function body
  script: '
    var location = ee.Feature(arguments[0]);
    var bestImage = selectBestImage(location);
    var filename = ee.String(location.geometry().coordinates().get(0, 1))
      .concat('_')
      .concat(ee.String(bestImage.get('system:time_start').slice(0, 4))
        .replace(/-/g, ''))
      .concat('.tif');
    ee.Image(bestImage)
      .export({
        fileName: ee.String(filename)
      });
  ',
  // Output: list of download URLs
  return: '
    ee.List([
      ee.String(filename)
    ])
  '
});

// Run the function
var downloadUrls = download.execute(bestImages);
```

Local Python Code



```
import ee
import urllib2
import os

# Import the Landsat 7 satellite imagery
Landsat7 = ee.ImageCollection('LANDSAT/LT05/C01/T11')

# Filter for the year 2015
Landsat7_2015 = Landsat7.filterDate('2015-01-01', '2015-12-31')

# Select the best image for each location
def selectBestImage(location):
    return Landsat7_2015.select(location).filterDate('2015-01-01', '2015-12-31').max()

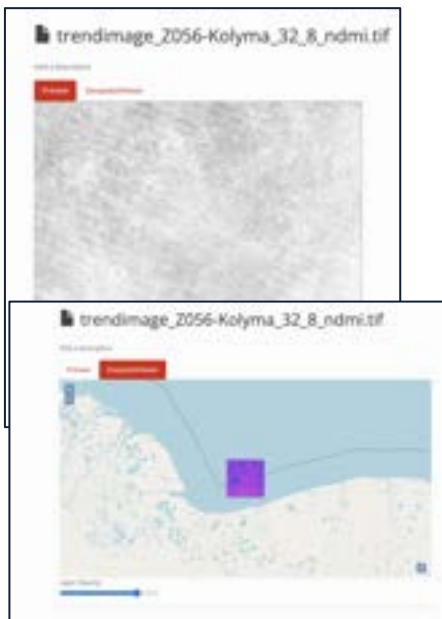
# Create a feature collection of the best images
bestImages = ee.FeatureCollection([
    ee.Feature(ee.Geometry.Point(-122.5, 45.5), {'location': 'San Francisco'})
])

# Download the best image for each location
def download(location):
    bestImage = selectBestImage(location)
    filename = '%s_%s.tif' % (location.geometry().coordinates().get(0, 1),
                              bestImage.get('system:time_start').slice(0, 4).replace('-', ''))
    bestImage.export({'fileName': filename})

# Run the function
download(bestImages)
```

# LandsatTrends Detection Model and Area Calculation

Inputs - GeoTiffs



Submit



Outputs - GeoTiffs and Shapefiles





# Arctic Satellite Joint Product (ASJP)

Some Arctic satellite data for these interesting variables would be nice



NASA Public Data Archives

No problem. Here is your order.



Archived data is very available, but difficult to navigate. Many variables stored together in large files. Entire file or entire sets of files must be downloaded for study. Access assumes very high speed internet connection.



Clowder

Docker

Arctic Satellite Joint Product (ASJP)



Data volume  
1.6Mb/day



Arctic Satellite Joint Product with help of Docker and Clowder distills the high volume of archived satellite data to the essential community variables of interest that can be made available quickly and efficiently.

NASA  
TERRA  
MODIS

pdg-upload

flat  
list

by day

Upload  
Dataset

Run  
ASJP

ASJP  
Daily  
Product

17GB /  
day

1.7 MB /  
day



# Flat list of files

MOD06_L2.A2022156.1145.061.2022158133116.hdf	MOD14.A2022150.0600.061.2022151092320.hdf	MOD29E1D.A2022148.061.2022151212855.hdf
MOD06_L2.A2022156.1150.061.2022158132342.hdf	MOD14.A2022150.0605.061.2022151090022.hdf	MOD29E1D.A2022149.061.2022151221800.hdf
MOD06_L2.A2022156.1155.061.2022158132334.hdf	MOD14.A2022150.0610.061.2022151090442.hdf	MOD29E1D.A2022150.061.2022151223616.hdf
MOD06_L2.A2022156.1200.061.2022158132329.hdf	MOD14.A2022150.0615.061.2022151090650.hdf	MOD29E1D.A2022151.061.2022152234746.hdf
MOD06_L2.A2022156.1205.061.2022158132440.hdf	MOD14.A2022150.0620.061.2022151090110.hdf	MOD29E1D.A2022152.061.2022153140412.hdf
MOD06_L2.A2022156.1210.061.2022158132337.hdf	MOD14.A2022150.0625.061.2022151090543.hdf	MOD29E1D.A2022153.061.2022154094511.hdf
MOD06_L2.A2022156.1215.061.2022158132505.hdf	MOD14.A2022150.0630.061.2022151085940.hdf	MOD29E1D.A2022153.061.2022158013908.hdf
MOD06_L2.A2022156.1220.061.2022158132446.hdf	MOD14.A2022150.0635.061.2022151090739.hdf	MOD29E1D.A2022153.061.2022159224825.hdf
MOD06_L2.A2022156.1225.061.2022158132455.hdf	MOD14.A2022150.0640.061.2022151090812.hdf	MOD29E1D.A2022154.061.2022158014936.hdf
MOD06_L2.A2022156.1230.061.2022158132504.hdf	MOD14.A2022150.0645.061.2022151090248.hdf	MOD29E1D.A2022155.061.2022156085054.hdf
MOD06_L2.A2022156.1235.061.2022158132440.hdf	MOD14.A2022150.0650.061.2022151090121.hdf	MOD29E1D.A2022156.061.2022158014700.hdf
MOD06_L2.A2022156.1240.061.2022158133421.hdf	MOD14.A2022150.0655.061.2022151091117.hdf	MOD29E1D.A2022157.061.2022158090314.hdf
MOD06_L2.A2022156.1245.061.2022158133822.hdf	MOD14.A2022150.0700.061.2022151090007.hdf	MOD29E1D.A2022158.061.2022159080709.hdf
MOD06_L2.A2022156.1250.061.2022158133724.hdf	MOD14.A2022150.0705.061.2022151090633.hdf	MOD29E1D.A2022159.061.2022160075838.hdf
MOD06_L2.A2022156.1255.061.2022158134130.hdf	MOD14.A2022150.0710.061.2022151090757.hdf	MOD29E1D.A2022160.061.2022161080643.hdf
MOD06_L2.A2022156.1300.061.2022158133235.hdf	MOD14.A2022150.0715.061.2022151090405.hdf	MOD29E1D.A2022161.061.2022162092018.hdf
MOD06_L2.A2022156.1305.061.2022158134313.hdf	MOD14.A2022150.0720.061.2022151085812.hdf	MOD29E1D.A2022162.061.2022163084023.hdf
MOD06_L2.A2022156.1310.061.2022158133842.hdf	MOD14.A2022150.0725.061.2022151085744.hdf	MOD29E1D.A2022163.061.2022164080310.hdf
MOD06_L2.A2022156.1315.061.2022158133914.hdf	MOD14.A2022150.0730.061.2022151085752.hdf	MOD29E1D.A2022164.061.2022165101941.hdf
MOD06_L2.A2022156.1320.061.2022158134158.hdf	MOD14.A2022150.0735.061.2022151090432.hdf	MOD29E1D.A2022165.061.2022166075322.hdf



# Folders by day

```
lmarini@pdg-upload:/data/modaps_by_day$ ls
142 143 144 145 146 147 148 149 151 152 153 154 155 156 157 158 159 160 161
162 163 164
```

```
lmarini@pdg-upload:/data/modaps_by_day/163$ ls
MOD06                                MOD14
MOD09CMG.A2022161.061.2022163043428.hdf  MOD29E1D.A2022162.061.2022163084023.hdf
MOD10C1.A2022161.061.2022163054356.hdf
```

MOD09CMG.A2022**161**.061.2022163043428.hdf



# Raw Data per Day

Datasets in the Space

Viewing most recent datasets [View All Datasets](#)



<b>ASJP Day 153</b> + Raw data for day 153 1 0 0 0 0 1 0 0	<b>ASJP Day 161</b> + Raw data for day 161 2 5 0 0 0 0 0 0	<b>ASJP Day 152</b> + Raw data for day 152 2 8 0 0 0 0 0 0
<b>ASJP Day 160</b> + Raw data for day 160 2 7 0 0 0 0 0 0	<b>ASJP Day 164</b> + Raw data for day 164 2 3 0 0 0 1 0 0	<b>ASJP Day 146</b> + Raw data for day 146 2 6 0 0 0 1 0 1
<b>ASJP Day 151</b> + Raw data for day 151 2 5 0 0 0 1 0 0	<b>ASJP Day 163</b> + Raw data for day 163 2 3 0 0 0 2 0 1	<b>ASJP Day 145</b> + Raw data for day 145 2 5 0 0 0 0 0 0







# Private Dataset per Day

Files Metadata Extractions Visualizations Comments (0)

Marked Files (0) +

 **MOD06**  
0 288  [Download](#)

 **MOD14**  
0 288  [Download](#)

 **MOD10C1\_A2022161.061.2022163054356.hdf** +  
application/x-hdf  
Jun 13, 2022  
5.0 MB  
2 0 0 2  [Download](#)  
[Follow](#)  
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<a href="#">iris-pdg-esp</a>	Arctic Specimens joined Product	<a href="#">Submit</a>
<a href="#">pdg-lambdatrend</a>	lambdatrend extractor	<a href="#">Submit</a>
<a href="#">pdg-lambdatrend-keanalyzer</a>	lake analysis lambdat extractor	<a href="#">Submit</a>
<a href="#">pdg-magik-2.0</a>	MAPL2.0 adapted as extractor	<a href="#">Submit</a>



# Public Data




**ASJP Outputs**  
Owner: [Luigi Marini](#) All Rights Reserved Luigi Marini  
Created on Mar 25, 2022

Add creator(s)  
Add a description

[+ Add Files](#) [Download All Files](#) [Archive All](#) [Unarchive All](#) [Delete](#) [★ Follow](#)  
[✔ Create Folder](#) [Submit for extraction](#)

Files Metadata Extractions Visualizations Comments (0)

Marked Files (0) +

 **ASJP Day 164.nc4**  
application/octet-stream  
Jun 15, 2022  
1.0 MB  
1 0 1 0

[Download](#)  
[★ Follow](#)

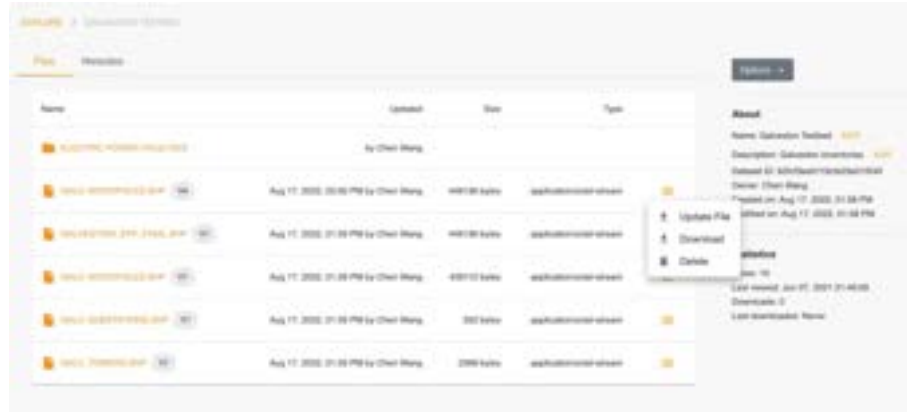
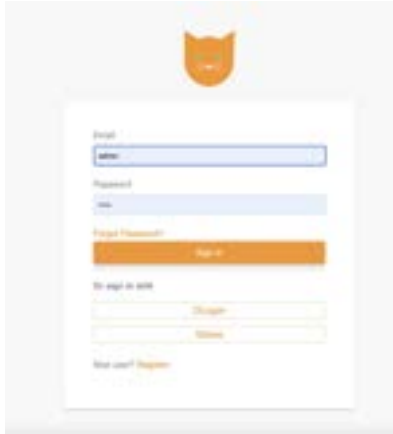






# Clowder v2 - Join the Development!

- <https://github.com/clowder-framework/clowder2>



# Thank you!

<https://permafrost.arcticdata.io>

<https://clowderframework.org/>

## Permafrost Discovery Gateway



Navigating  
the New Arctic

Awards #

1927872, 1927723,  
1927729, 1927720,  
1927920, & 2052107



NCSA | NATIONAL CENTER FOR SUPERCOMPUTING APPLICATIONS