

Addressing Global Sustainability Challenges with Jupyter Technologies and Cloud-based Geospatial Data Platforms

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github.com/tylere/JupyterCon-2023

Background

We have a global grand challenge of making living on Earth sustainable, as it undergoes significant change due to human activity. The world's inhabitants are already experiencing many negative impacts, such as displacement from sea-level rise, ecosystem degradation, and increases in extreme weather events. People need access to better information so they can understand the impacts that are already occurring, mitigate changes we can prevent from occurring, predict what changes will occur in the future, and prepare to adapt to those changes that we cannot mitigate.

Fortunately, in recent decades there has been a rapid increase in data about the state of the Earth. Satellites have been continuously observing the globe for over 50 years, and Earth System Models (ESMs) generate weather predictions of the coming weeks and climate projections of what may happen in future decades. However, several barriers to progress remain:

- Data volume & velocity is a major challenge for most potential users. Satellites and ESMs generate terabytes/day and archives are petabytes in size.
- Raw data needs to be processed into actionable information that is appropriate for decision makers.

Geospatial Data Analysis Systems

Barriers are being addressed through the use of **data-proximate computing** systems that are optimized for geospatial data analysis. These systems move compute close to data storage, and are an improvement over the historical practice (and bottleneck) of downloading large datasets for local analysis.

Some examples of these systems include

- Pangeo (pangeo.io/) - an open source software stack consisting of Jupyter, Dask, and Xarray.
- Google Earth Engine (developers.google.com/earth-engine) - a cloud-based geospatial analysis system, including a >50 PB data catalog.

Jupyter Widgets

Widgets are “*eventful python objects that have a representation in the browser*” ([Jupyter Widget Docs](#)). Widgets enable bidirectional communication between the client side UI (JavaScript) and server side (ipython kernel) components.

Widgets can be used to enable rapid interaction with large (terabyte to petabyte) geospatial datasets via the following process:

1. Notebook users interact with widget representations in the browser, triggering one or more *widget events*.
2. Widget events trigger Python code that requests representations of geospatial data from a remote server. The remote server processes data to a small size (~100KB) before returning it to the browser.

Ideas for Future Improvements

1. Make it easier to interact with common spatio-temporal datasets by creating additional compound widgets that exploit the structure of the datasets. For example, compound widgets could be created for polar orbiting satellite imagery, stationary sampling stations, nested watersheds, etc.
2. Deploy examples/tutorials using JupyterLite, preloaded with widgets and geospatial libraries, to remove the need for a hosted geospatial server (for some use cases).

Algorithm
Development
Environment

JupyterLab
Jupyter Notebook
Google Colab
Microsoft Azure Notebooks
Amazon Sagemaker
etc.

Jupyter Widgets
bidirectional communication

Geospatial
Analysis

filter
aggregate
classify
join
etc.

Geospatial
Data

satellite observations
weather/climate data
physical & administrative
boundaries
digital elevation models
etc.

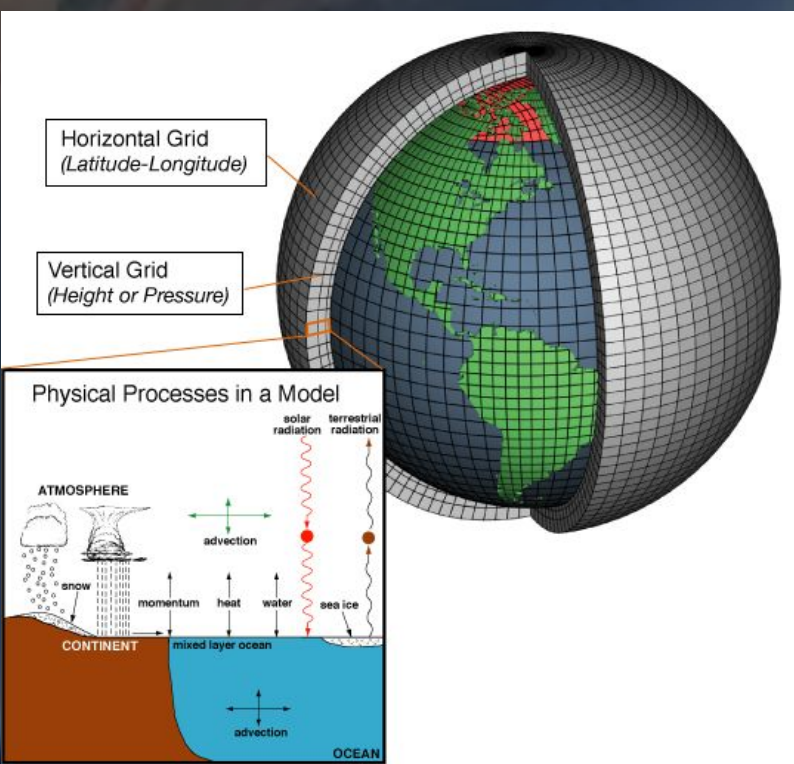
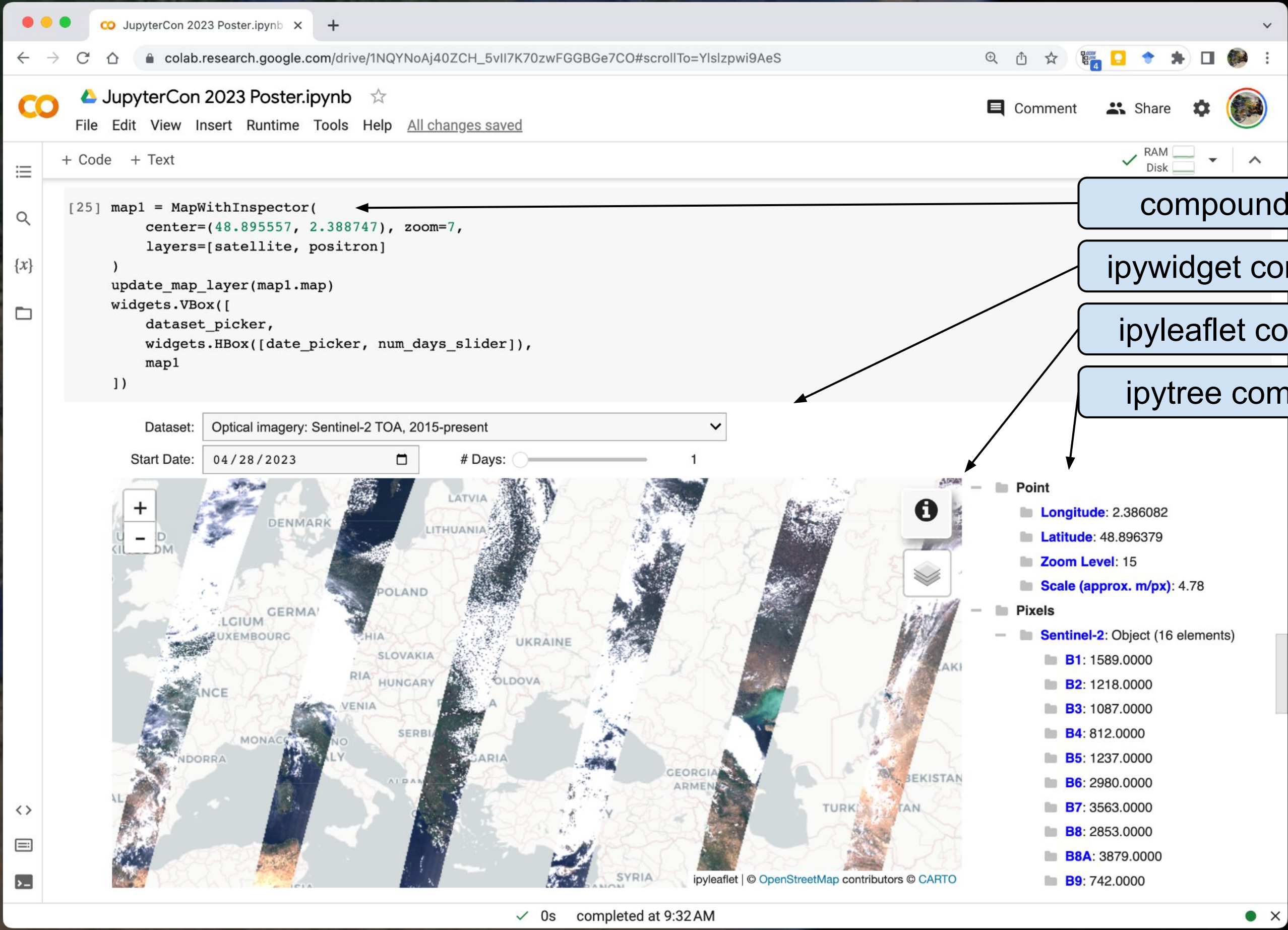


Image credit: Climate.gov

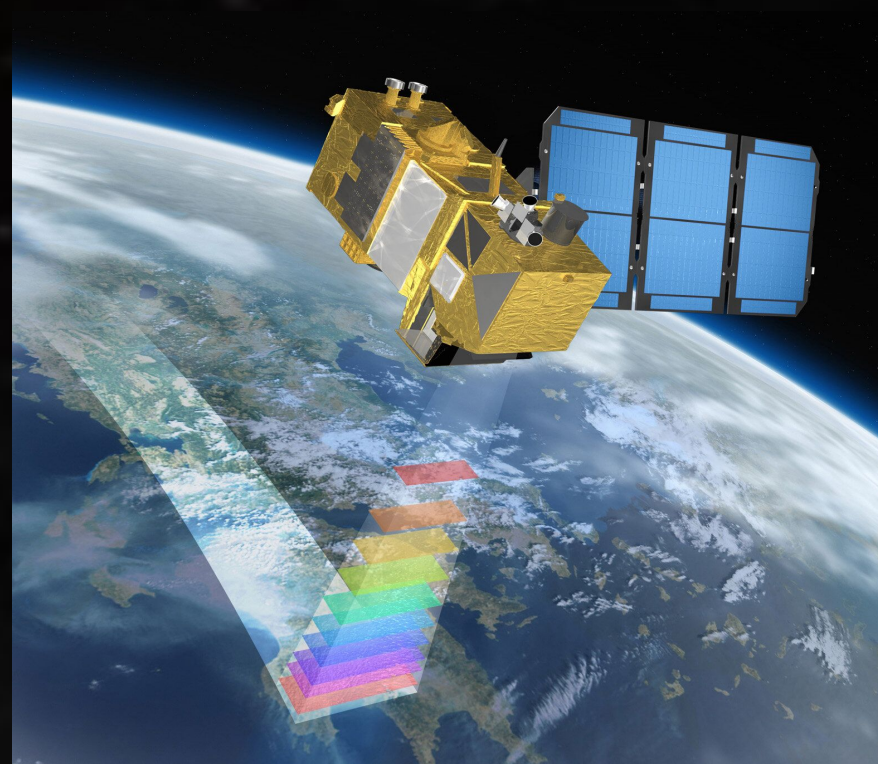
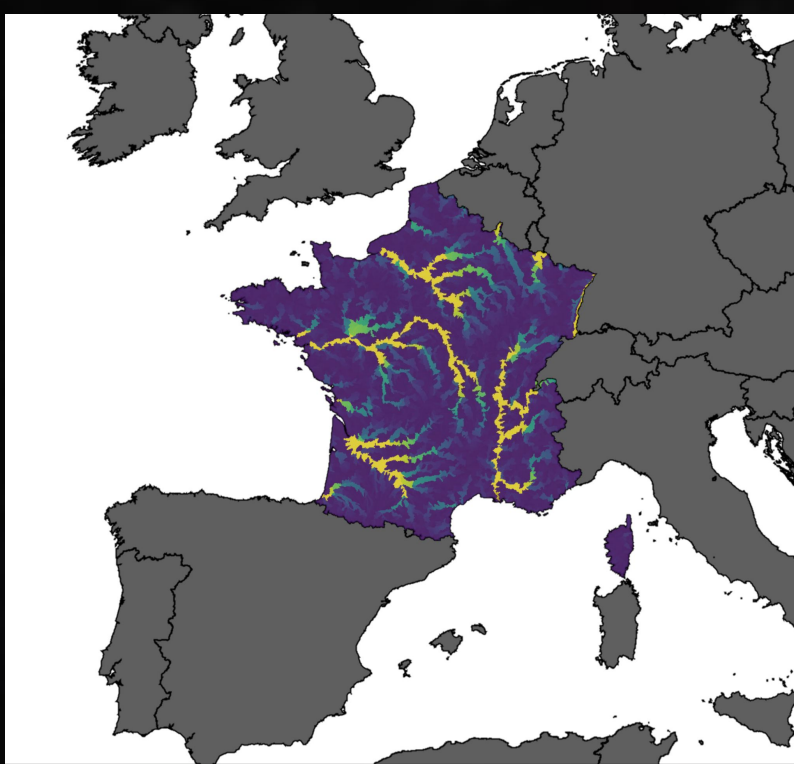


Image credit: ESA

