



Hobby Drones, Urban Forests

A Geospatial Journey To Greener Cities

Kevin Lacaille | July 8, 2024

A little bit about me



Astrophysics
Software engineering
Mountain biking, skiing, cooking



Today we will



Learn geospatial concepts
Python for image analysis
Apply to real-world example



An aerial photograph of a steep, rocky hillside covered in dense green coniferous trees. The hillside slopes down towards a dark, calm body of water. The sky above is filled with heavy, dark clouds.

The Why

Drone Imagery for Climate

How do we use it effectively?

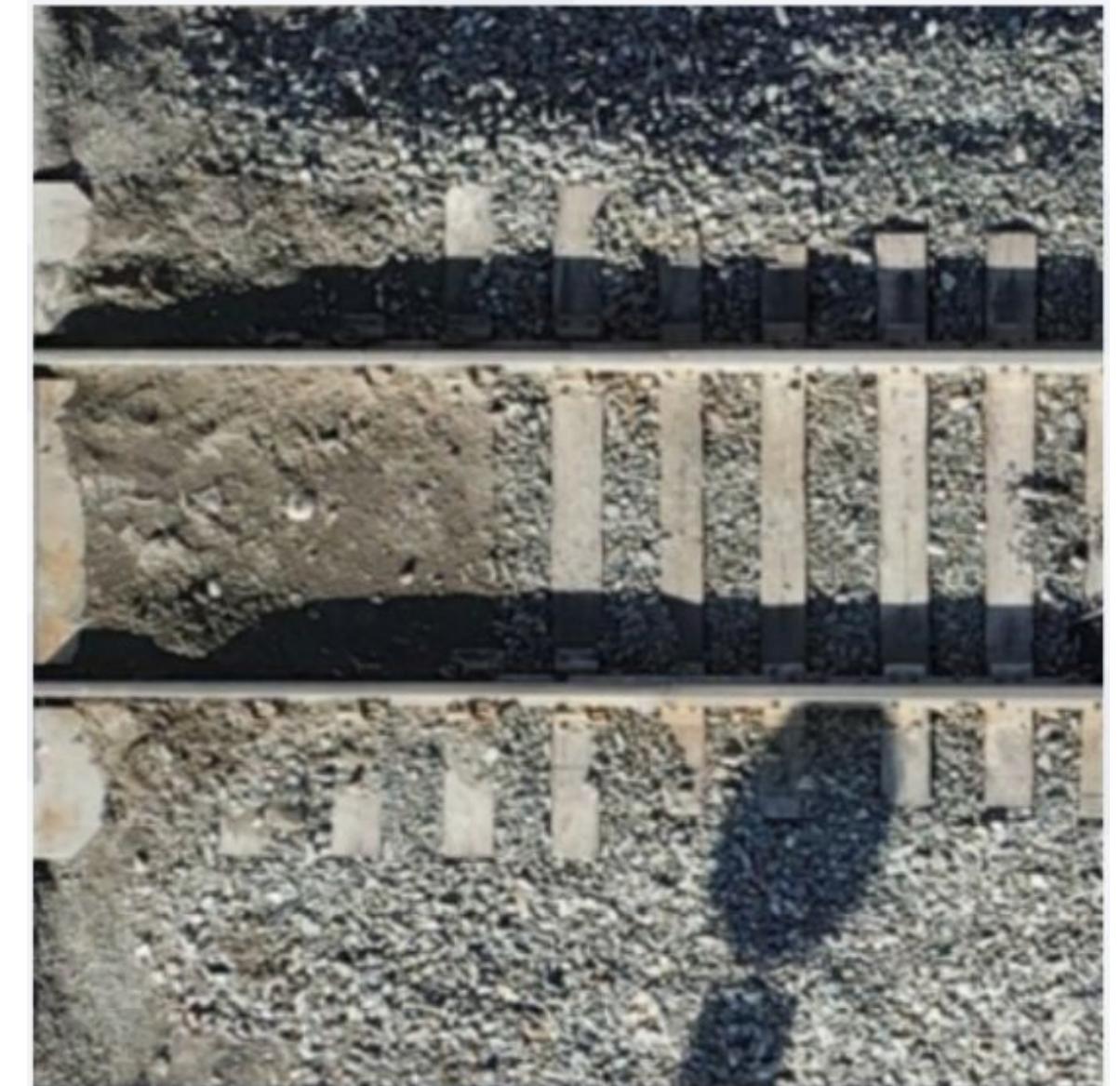
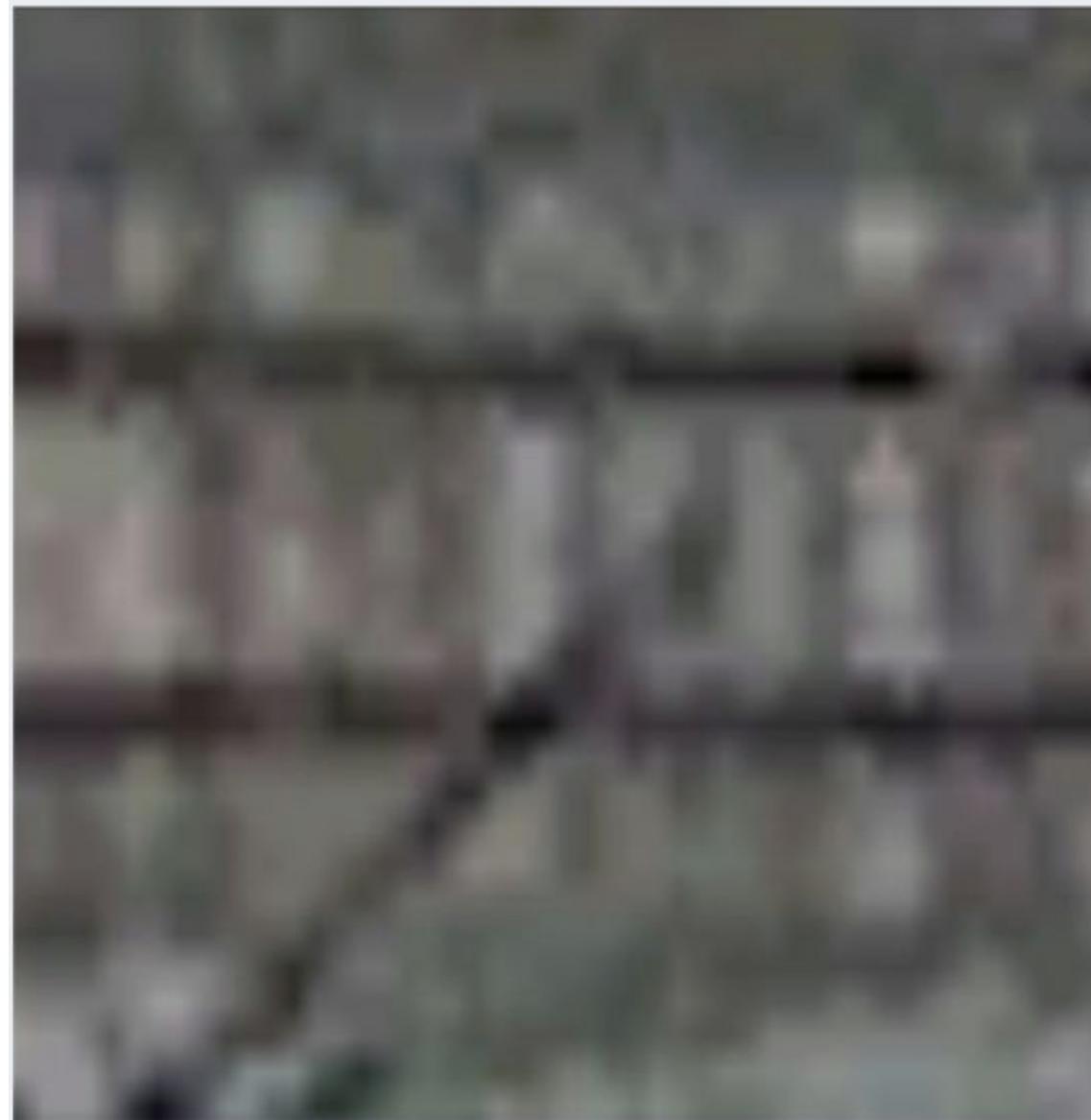
- Climate change urgency
- High-res imagery
- Accessible and cost effective
- Real time monitoring
- Urban planning and sustainability



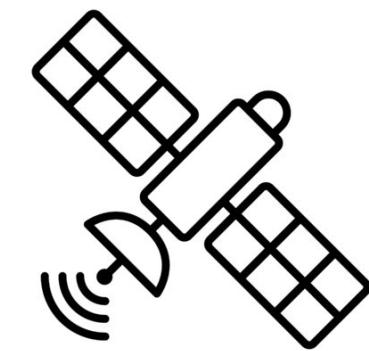
An aerial photograph of a steep, densely forested hillside. The trees are primarily evergreen, with varying shades of green and some yellow autumn foliage. The hillside slopes down towards a dark blue body of water. The sky above is filled with heavy, grey clouds.

Drone Imagery

Geospatial Data Source



Satellite



- Low res (0.3-10m)
- Long wait (~1 month)
- Limited by weather & atmosphere
- Large coverage
- Limited by satellite's orbit
- Requires professionals
- Expensive

Aerial



- Medium res (0.1-1m)
- Long wait (~1 week)
- Limited by weather
- Large coverage
- Can fly most places
- Requires professionals
- Expensive

Drone



- High res (0.03m)
- Almost no wait (~4hr)
- Limited by weather & battery life
- Small coverage
- Can fly most places
- Easy to operate
- Inexpensive

Micro Drones

A bit about hobby drones

DJI micro drones

Portable, cost effective, high-quality imaging

- Weight <250g
- 3cm (1 inch) resolution
- Flight time >30 min
- No FAA registration required
- <\$500



Core Geospatial Concepts

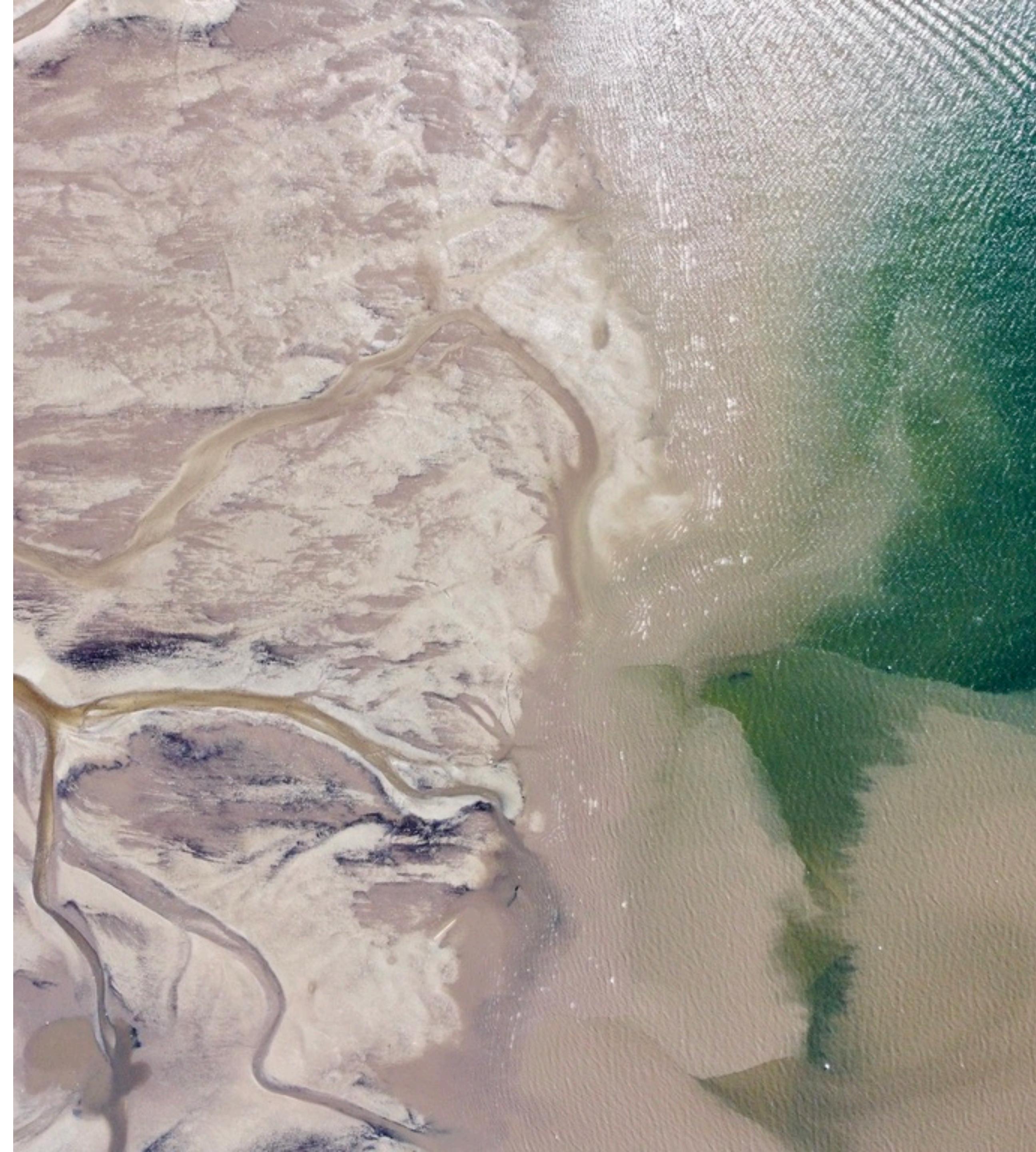


Spatial Data is Special

Unique qualities of drone imagery require special consideration

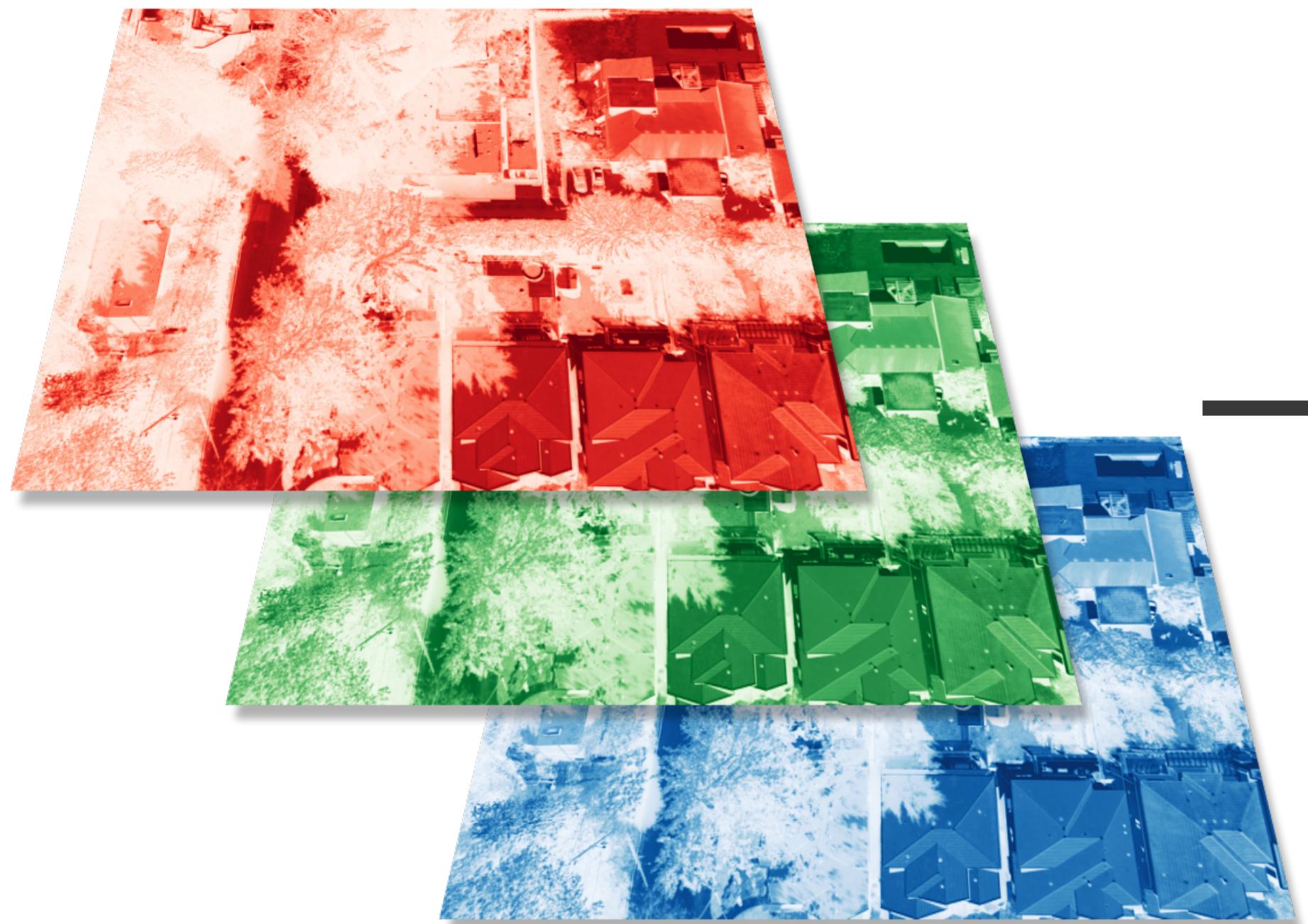
Geospatial Raster Imagery

- Is spatially referenced
- Has spatially-relevant metadata
- Is multispectral



Multispectral Imagery

An image contains 3 spectral bands



Drone imagery requires special consideration

Using non-geospatial Python imagery libraries can have unexpected results

Rasters, Pixels, and 2D Arrays

Representing Remote-Sensing Data

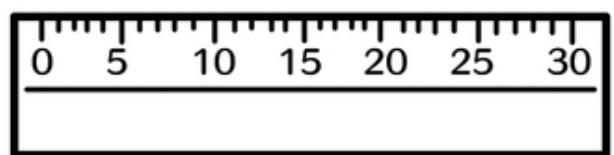


Pixels in a raster dataset are arranged in a grid

Pixels representing equivalent data have the same value.

Defining Resolution in Drone Imagery

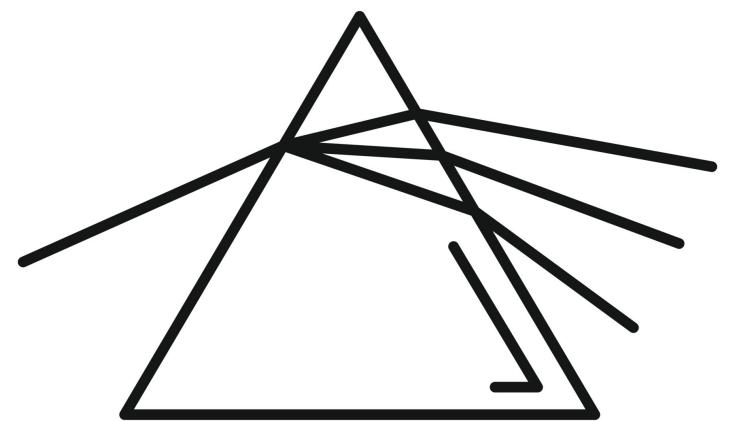
Spatial resolution: Pixel size on the ground



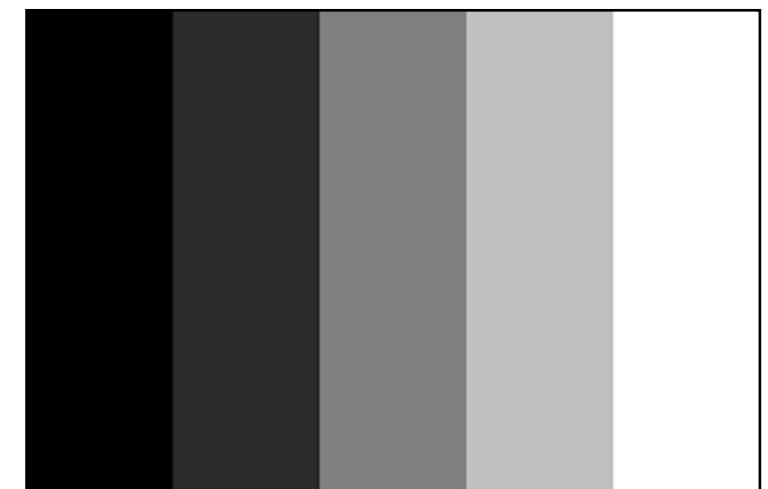
Temporal resolution: Frequency between image captures



Spectral resolution: Number of spectral bands measured



Radiometric resolution: Number of shades of a colour



Geospatial Tools & Libraries



Python Libraries

Geospatial, math, data viz, & computer vision

Rasterio

- Read and write geospatial raster data represented as 3D arrays

PyExifTool

- Extract meta-information from images

Numpy

- Read, write, and manipulate multi-dimensional arrays and matrices

Matplotlib

- Plotting library

OpenCV

- Computer vision library used for image processing, object detection, segmentation, etc.

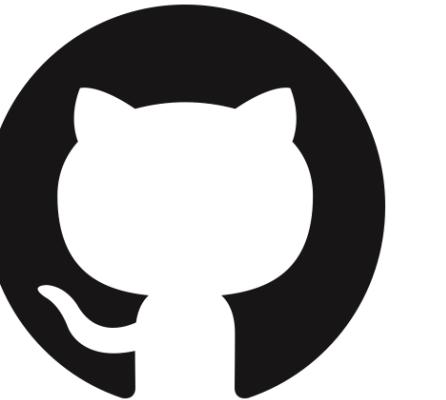


Workshop Setup

Good news! Open source tools exist to make this setup easy.

You can access an instance of this workshop's Jupyter Notebook collection

[github.com/kevinlacaille/presentations/
tree/main/scipy2024](https://github.com/kevinlacaille/presentations/tree/main/scipy2024)



The background image shows a steep, densely forested hillside on the right, with dark green coniferous trees. In the center-right, a dark blue body of water is visible. The sky is filled with heavy, greyish-blue clouds.

Notebook Time!

github.com/kevinlacaille/presentations/tree/main/scipy2024

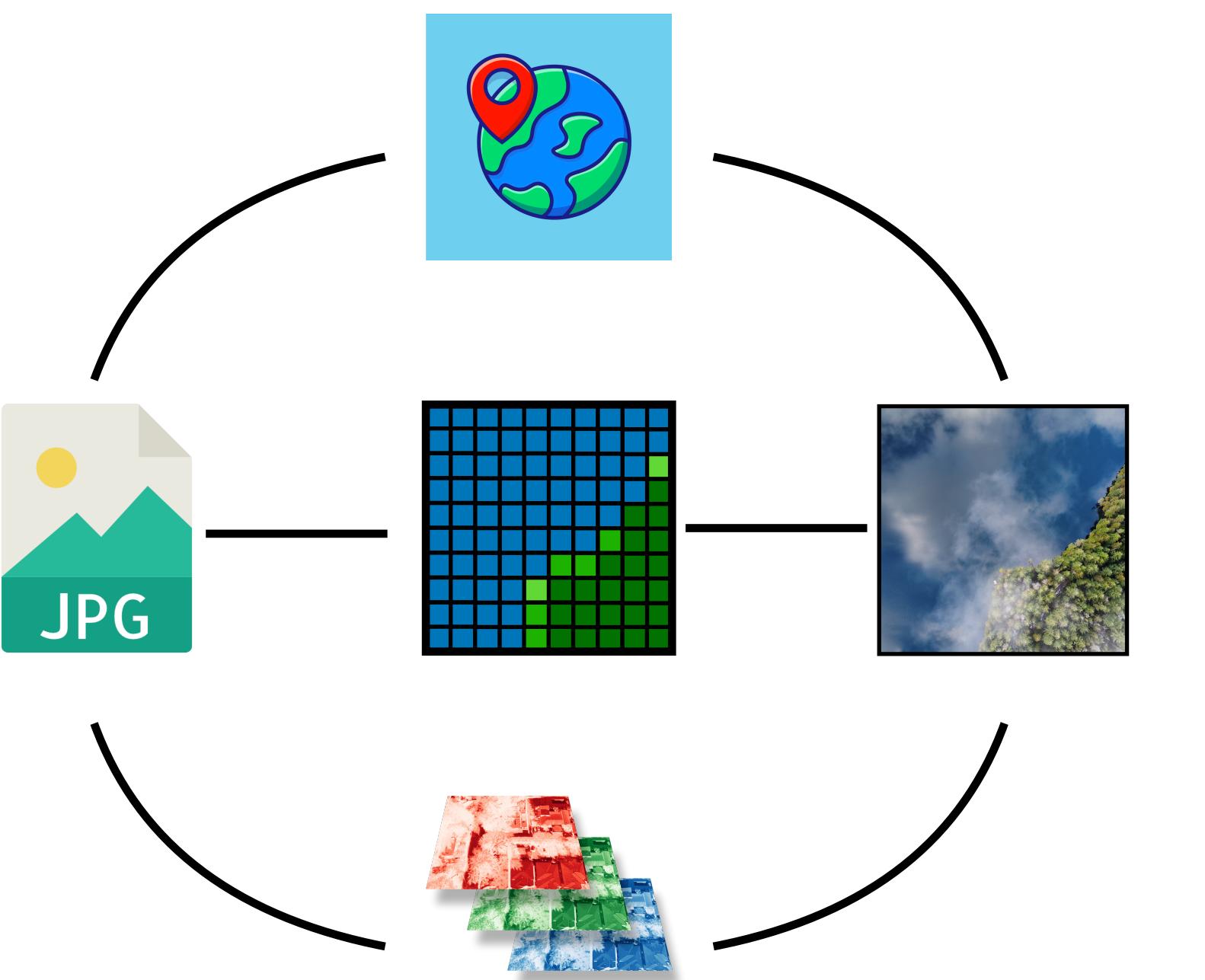
Extract Bands & Compose Scene

Use `rasterio` to extracting metadata from an image, including

- Coordinate reference system
- Resolution
- Bands

Decomposing a 3-band image

Use `numpy` to combine bands and create visual image



Reference: `1_rasters.ipynb`



Extract and Utilize EXIF Data

Extracting metadata from an image
using exiftool

- Estimate size of objects using context of the scene



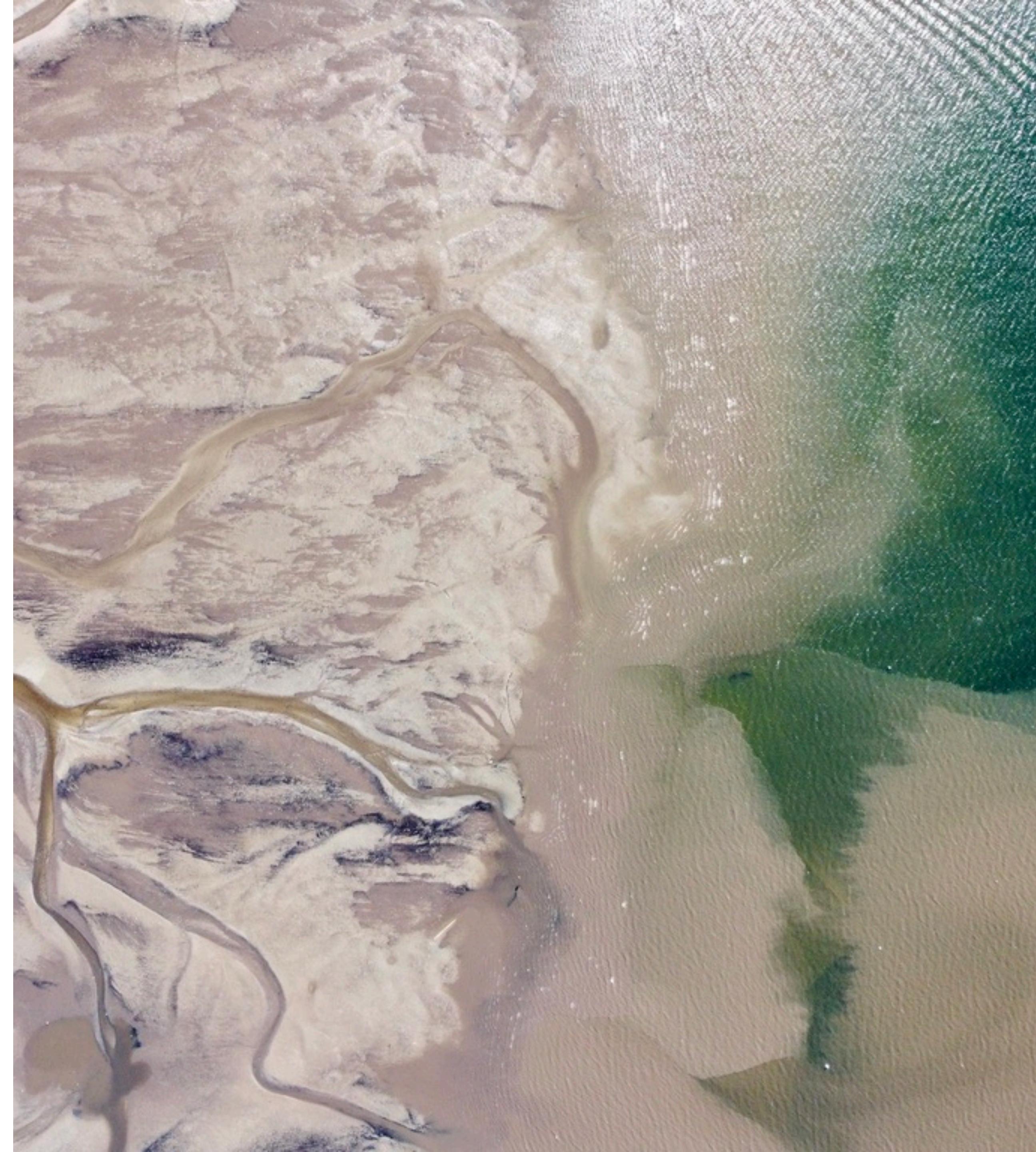
Reference: 2_exif.ipynb

Compute VARI

Visible Atmospherically Resistant Index (VARI)

- Emphasizes vegetation in the visible portion of the spectrum
- Mitigates illumination differences
- Hobby drone sensors are un calibrated - need to do it ourselves

$$\text{VARI} = \frac{\text{green} - \text{red}}{\text{green} + \text{red} - \text{blue}}$$

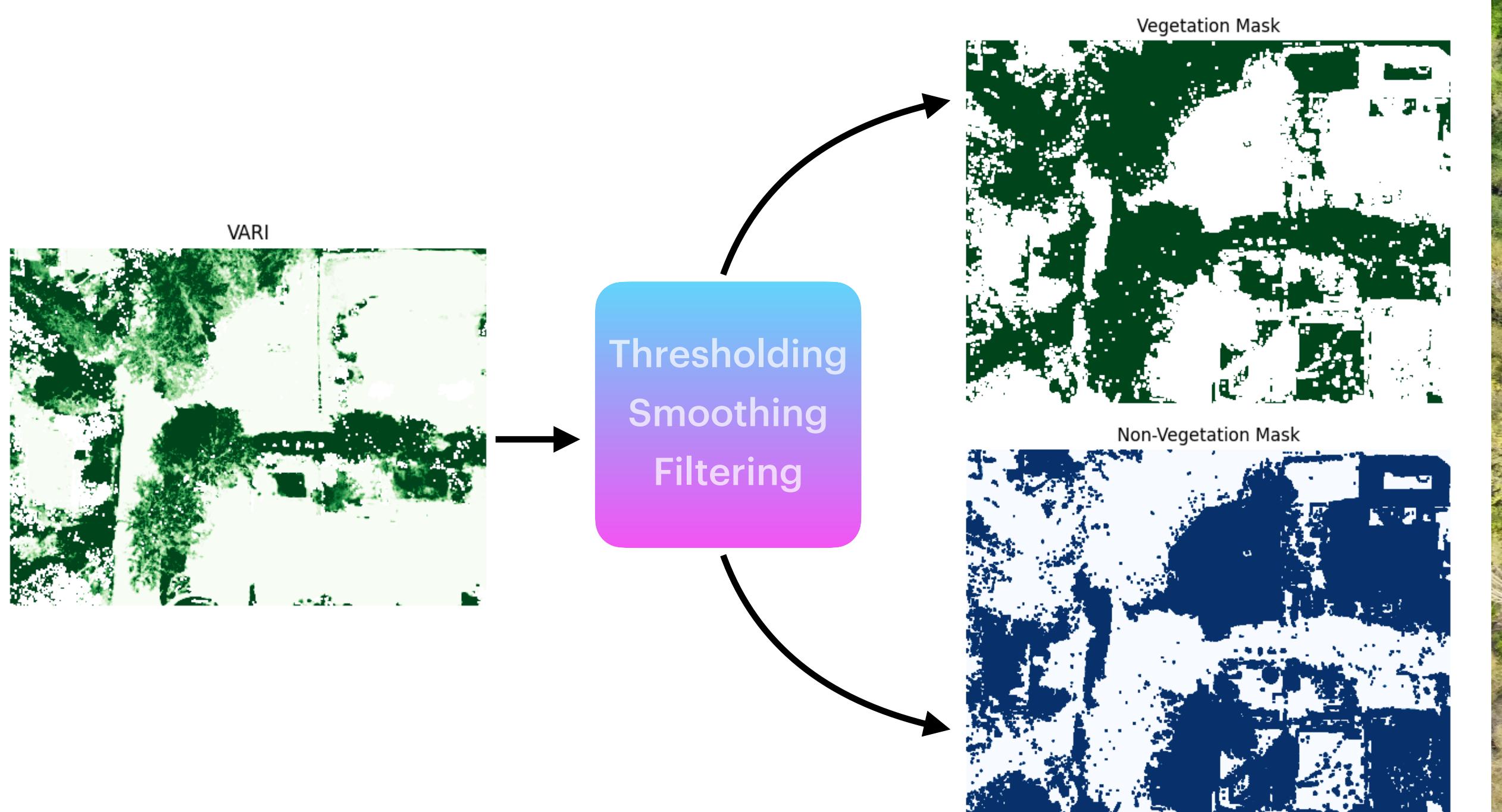


Reference: 3_vari.ipynb

Masks & filters

Apply a series of image processing techniques to VARI image

- Image thresholding
- Gaussian blurring
- Morphological filtering (opening and closing filters)



Reference: 4_masks_and_filters.ipynb



Remote sensing based segmentation

Don't have tons of annotated data?

Don't have time or resources to train an ML model?

Don't know how ML works?

Use science and context of a scene to our advantage!

- VARI, drone altitude, true size of objects, time of day, camera properties, etc.

vegetation segmentation



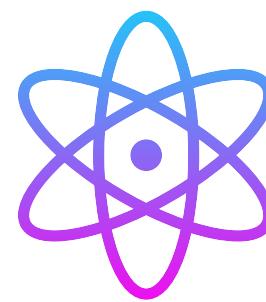
non-veg segmentation



Hands-on Exercise

A geospatial & computer vision analysis of urban trees?

1. Import and process drone imagery
2. Create an image processing pipeline
3. Develop a segmentation algorithm to find all trees in a scene
4. Estimate carbon capture for city
5. Drawbacks



What Can I do?

Become a pilot and contribute!



The background image is an aerial photograph of a steep, densely forested hillside. The trees are primarily evergreen, with varying shades of green and some yellowing, suggesting autumn. The hillside slopes down towards a dark blue body of water. The sky above is filled with heavy, grey clouds.

Get in touch
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