

Introduction

CASA0025:
Building Spatial Applications with Big Data



Ollie Ballinger

Outline

1. Spatial Applications
2. Handling big data by type
 1. Vector data
 2. Raster data
3. Module Overview



Remote Sensing of Environment

Volume 331, 15 December 2025, 115025



Open access battle damage detection via Pixel-Wise T-Test on Sentinel-1 imagery

Ollie Ballinger 

← Submissions with an Editorial Office Decision for Author

Page: 1 of 1 (1 total completed submissions)

Results per page 10

Action	Manuscript Number	Title	Initial Date Submitted	Current Status	Date Final Disposition Set	Final Disposition
Action Links	RSE-D-24-00882	Open Access Battle Damage Detection via Pixel-Wise T-Test on Sentinel-1 Imagery	03 Apr 2024	Completed - Reject	05 Apr 2024	Reject

Action	Manuscript Number	Title	Date Submission Began	Current Status
Action Links	RSE-D-24-02143R1	Open Access Battle Damage Detection via Pixel-Wise T-Test on Sentinel-1 Imagery	06 Jan 2025	Under Review

Abstract

In the context of recent, highly destructive conflicts in Gaza and Ukraine, reliable estimates of building damage are essential for an informed public discourse, human rights monitoring, and humanitarian aid provision. Given the contentious nature of conflict damage assessment, these estimates must be fully [reproducible ↗](#), explainable, and derived from open access data. This paper introduces a new method for building damage detection– the Pixel-Wise T-Test (PWTT)– that satisfies these conditions. Using a combination of freely-available synthetic aperture radar imagery and statistical change detection, the PWTT generates accurate conflict damage estimates across a wide area at regular time intervals. Accuracy is assessed using an original dataset of over 2 million labeled building footprints spanning 30 cities across Palestine, Ukraine, Sudan, Syria, and Iraq. Despite being simple and lightweight, the algorithm achieves building-level accuracy statistics ($AUC=0.87$ in the full sample) rivaling state of the art methods that use deep learning and high resolution imagery. The workflow is [open source ↗](#) and deployed entirely within the Google Earth Engine environment, allowing for the generation of interactive Battle Damage Dashboards for [Ukraine ↗](#) and [Gaza ↗](#) that update in near-real time, enabling the public and humanitarian practitioners to immediately get estimates of damaged buildings in a given area.

Gaza Damage Proxy Map

The colored overlay is a cumulative damage proxy map of the Gaza Strip. Click the button below and draw a box on the map to get estimates of the number of damaged buildings and the affected population in a given area.

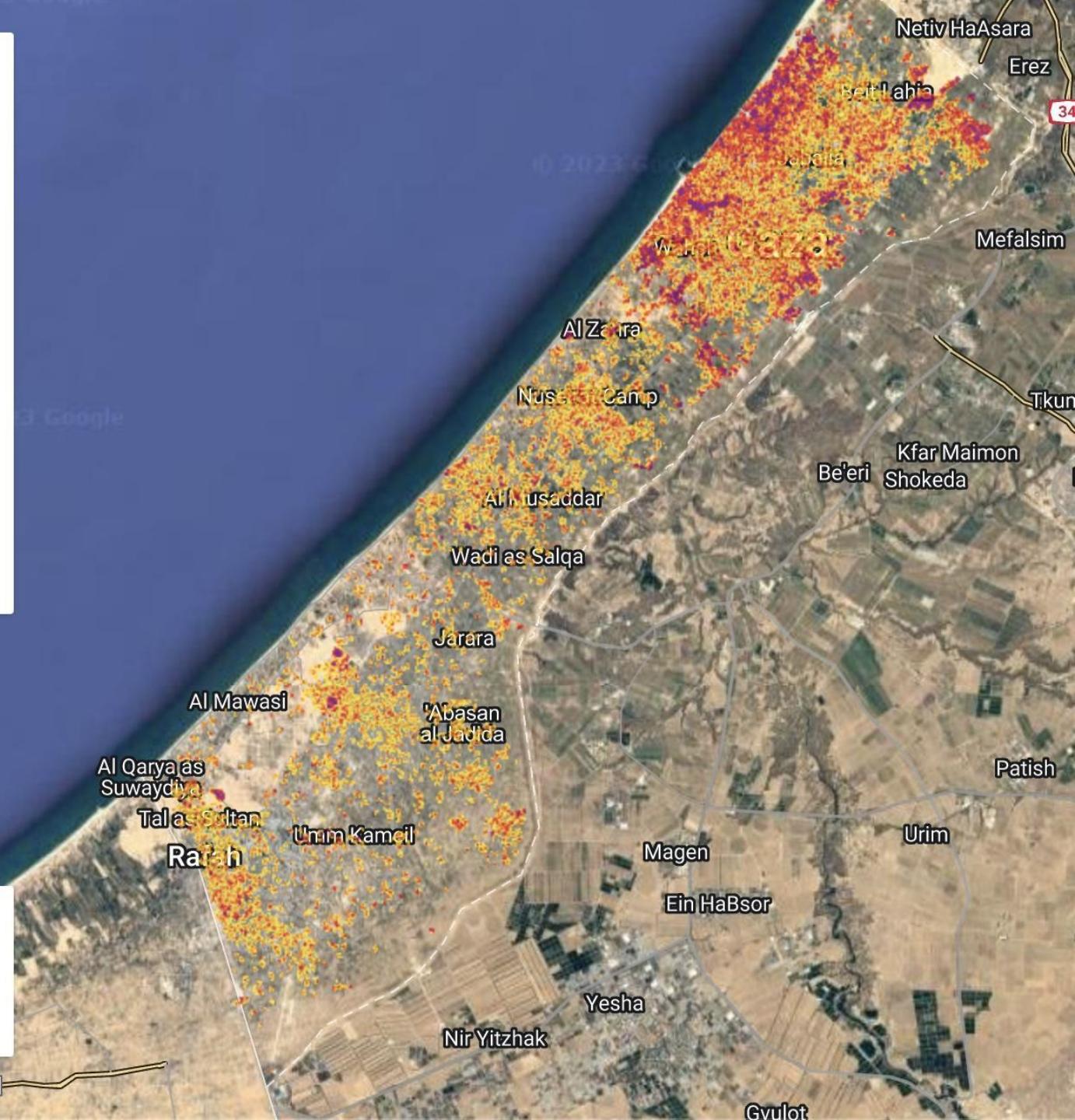
▲ Draw an Area of Interest

Geolocated footage can be added to the map as blue triangles, available under the "Layers" tab in the top right. Clicking on an event will show a brief description, the date, a link to the source media, and a link to the geolocation of the event. Verified damage points from UNOSAT can be added to the map as well.



Low (70%)

High (>98%)



Israel has destroyed more than 1,500 buildings in Gaza since ceasefire

BBC Verify's visual analysis of satellite imagery has found that the destruction of buildings in Gaza by the Israeli military has been continuing on a huge scale.

We used a change detection algorithm to analyse radar images taken before and after the ceasefire to highlight changes, which might indicate destruction, then manually counted visibly destroyed buildings.

"This is definitely a violation of the ceasefire," said Dr H A Hellyer, a RUSI Senior Associate Fellow. "But [Washington] DC is unwilling to recognise it as such, insisting that the ceasefire has to hold, even when it isn't actually holding."

Some analysts - such as Adil Haque, professor of law at Rutgers University - said Israel could be violating the laws of war, which prohibit the destruction of civilian property by an occupying power.

He noted that exceptions to the rule can only arise "from military operations, that is, from combat or direct preparations for combat," arguing that during a ceasefire "it is not plausible that such significant destruction of civilian property has been rendered absolutely necessary by military operations".



How BBC Verify tracked
Israel destroying
buildings after ceasefire

Israel and Hamas agreed to a
ceasefire on 10 October.

Article | [Open access](#) | Published: 03 January 2024

Satellite mapping reveals extensive industrial activity at sea

Abstract

The world's population increasingly relies on the ocean for food, energy production and global trade^{1,2,3}, yet human activities at sea are not well quantified^{4,5}. We combine satellite imagery, vessel GPS data and deep-learning models to map industrial vessel activities and offshore energy infrastructure across the world's coastal waters from 2017 to 2021. We find that 72–76% of the world's industrial fishing vessels are not publicly tracked, with much of that fishing taking place around South Asia, Southeast Asia and Africa. We also find that 21–30% of transport and energy vessel activity is missing from public tracking systems. Globally, fishing decreased by $12 \pm 1\%$ at the onset of the COVID-19 pandemic in 2020 and had not recovered to pre-pandemic levels by 2021. By contrast, transport and energy vessel activities were relatively unaffected during the same period. Offshore wind is growing rapidly, with most wind turbines confined to small areas of the ocean but surpassing the number of oil structures in 2021. Our map of ocean industrialization reveals changes in some of the most extensive and economically important human activities at sea.

FINANCIAL TIMES

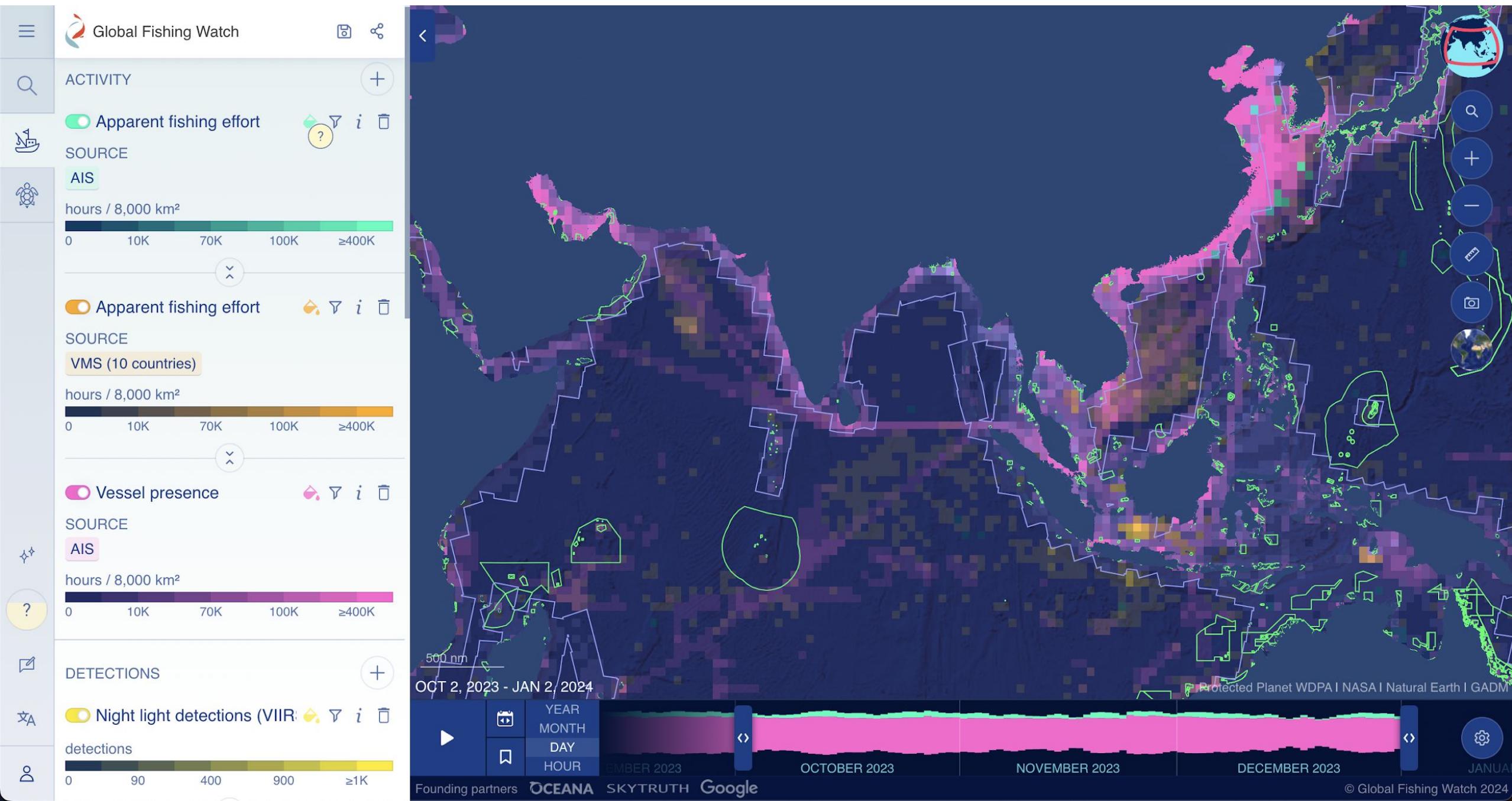
Off-radar fishing threatens efforts to preserve stocks, study warns

Majority of industrial vessels in world's oceans are not publicly tracked, says Global Fishing Watch



Human activity is powering 'a new industrial revolution' at sea, say experts

Researchers using AI and satellite imagery find 75% of industrial fishing is not being publicly tracked, while wind turbines now outnumber oil platforms



Satellite Data Reveals How the U.S. Navy Is Deployed Near Venezuela

The Times identified nearly 100 locations traversed by naval vessels across a two-and-a-half-month period to determine what the military pressure campaign against Venezuela looks like at sea.

Sightings of U.S. military ships in the Caribbean

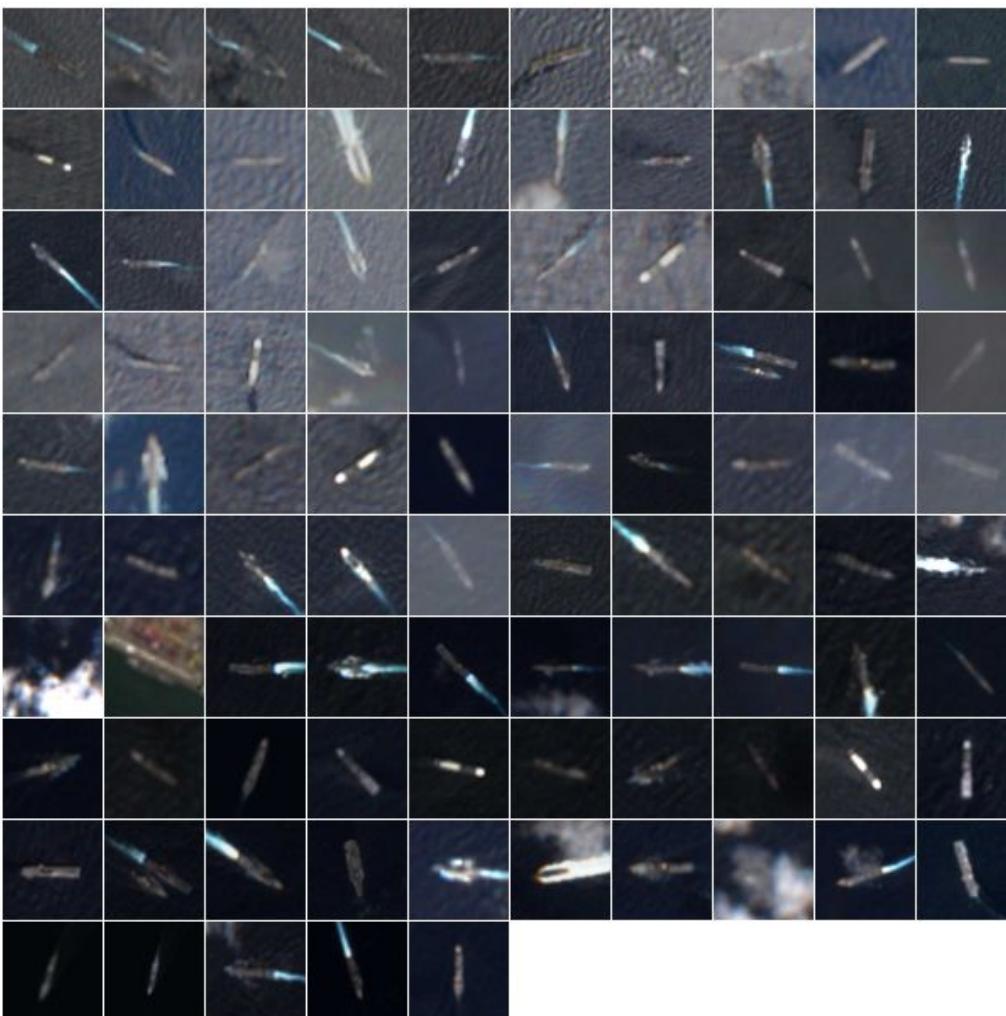


Ollie B.

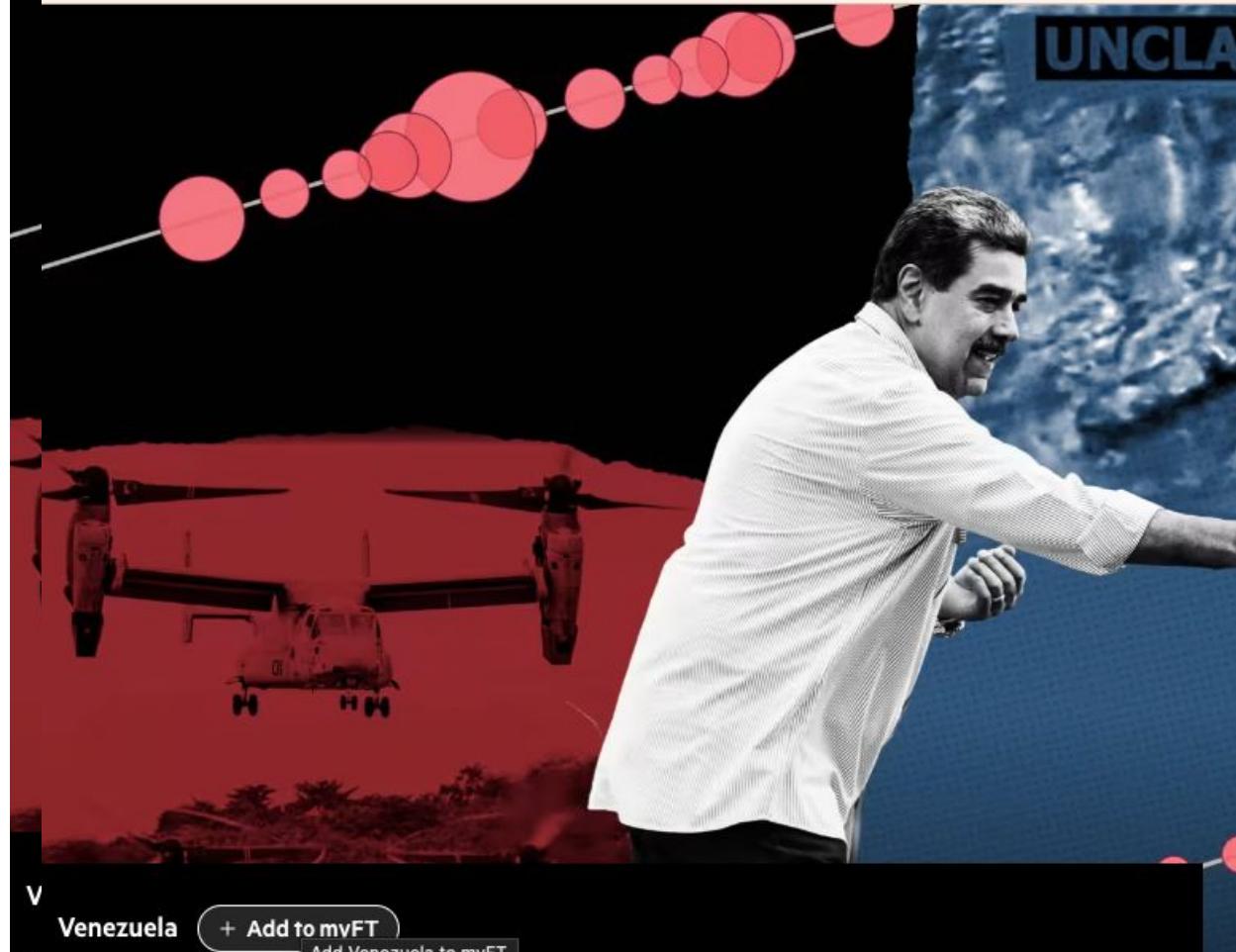
Sources: Vessel detections by Ollie Ballinger using satellite images by Copernicus via Sentinel Hub. Lazaro Gamio/The New York Times

To track the U.S. Navy ships, Ollie Ballinger, a lecturer in geocomputation at University College London's Center for Advanced Spatial Analysis, wrote a computer program that can search for specific vessels in free, publicly available satellite images.

U.S. military ship sightings in the Caribbean



Note: Some images contain more than one ship. Source: Satellite images by Copernicus via Sentinel Hub. Lazaro Gamio/The New York Times



Venezuela + Add to myFT
Add Venezuela to myFT

Satellite images track Donald Trump's squeeze on Venezuela

Photos, data and videos show increasing military pressure on Nicolás Maduro after US president refuses to rule out war

Dr Ollie Ballinger, a lecturer in geocomputation at University College London, identified the warships using image-matching technology, with each potential match then checked manually by the Financial Times.

US naval vessels in the Caribbean



Source: Vessel detections by Dr Ollie Ballinger using satellite images by Copernicus via Sentinel Hub, FT research • Last known ship locations in December
© FT

Visual Forensics

11 tankers under U.S. sanctions defy blockade in Venezuela, satellite imagery indicates

One ship was seized Friday in the Caribbean, and others were spotted steaming hundreds of miles into the Atlantic.

January 10, 2026

5 min Summary ↗ 248

Make us preferred on Google



Nine tankers sail east from Venezuela, captured by satellite imagery in early January. (European Union, contains modified Copernicus Sentinel data 2026, processed with Copernicus Browser)

Ollie E

The firm provided some of the satellite analysis for this article. Other analysis was provided by Ollie Ballinger, a lecturer in geocomputation at University College London's Center for Advanced Spatial Analysis. The Washington Post examined reference images of the ships to confirm the analyses.

■ Location of sanctioned ships as of January 5 and 6



Source: TankerTrackers.com and Ollie Ballinger

12 /44

U.S. Navy Is Chasing Several Oil Tankers in the Atlantic

A mass departure by sanctioned tankers, some switching to Russian flags, has triggered a fresh effort by the United States to pursue Venezuelan oil shipments at sea.

▶ Listen to this article · 3:08 min [Learn more](#)

Share full article



An image from video released by the U.S. military purporting to show the seizure of the oil tanker M Sophia in the Caribbean on Wednesday. U.S. Southern Command



Sources: Vessel detections by Ollie Ballinger using satellite images by Copernicus via Sentinel Hub. By Leanne Abraham

The ships were detected by a computer program written by Ollie Ballinger, a lecturer in geocomputation at University College London's Center for Advanced Spatial Analysis. The Times also analyzed the imagery, and verified Mr. Ballinger's findings.





COVID-19 Dashboard

by the Center for Systems Science and Engineering (CSSE) at Johns Hopkins University (JHU)



JHU Ceased Updates at:
3/10/2023, 8:21 AM
See Terms of Use for more info

Cases | Deaths by Country/Region/Sovereignty

US

28-Day: 959,794 | 9,451
Totals: 103,804,263 | 1,123,836

Japan

28-Day: 418,671 | 2,804
Totals: 33,329,551 | 73,046

Germany

28-Day: 355,168 | 2,275
Totals: 38,249,060 | 168,935

Russia

28-Day: 350,549 | 989
Totals: 22,086,064 | 388,521

Korea, South

28-Day: 290,039 | 396
Totals: 30,615,522 | 34,093

Taiwan*

28-Day: 214,024 | 778

Admin0

Total Cases
676,609,955

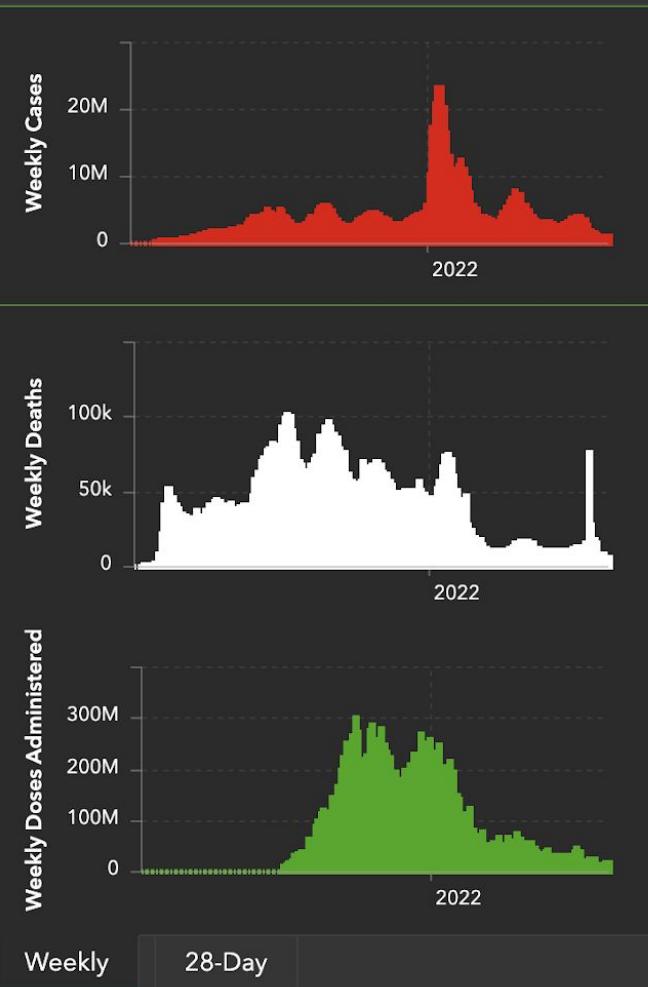
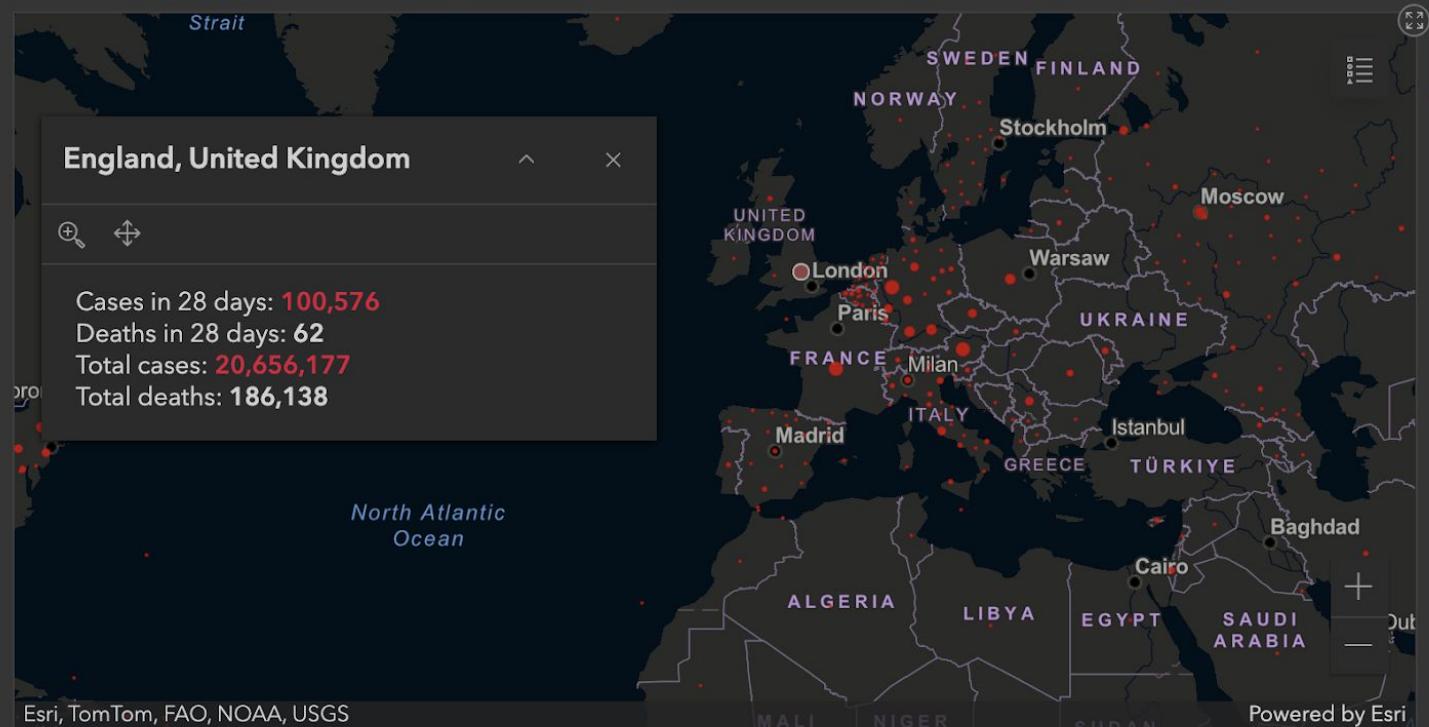
28-Day Cases
4,035,254

Total Deaths
6,881,955

28-Day Deaths
28,018

Total Vaccine Doses Administered
13,338,833,198

28-Day Vaccine Doses Administered
28,156,730



Best 2 min 6 min 2 min X

Saved

Recents

UCL Main Campus, Gower St, London WC

Work (90 Tottenham Ct Rd)

Add destination

Options

Send directions to your phone

via University St and Tottenham Ct Rd 6 min 0.2 mile

via Grafton Way and Tottenham Ct Rd 7 min 0.3 mile

All routes are mostly flat

Layers

Hospital Elizabeth...

Restaurants Coffee Groceries Things to do More

Main Campus

Cruciform Building

Gower St

Grafton Way

Tottenham Ct Rd

Maple St

Paramount Court

Rayne Institute

Rayne Building

Mortimer Market Centre

Bartlett School of Architecture

Fitness First

Shropshire House

Queens House

Gordon Mansions

Janet Poole House

Chenies Mews

Torrington Pl

Ridgmount

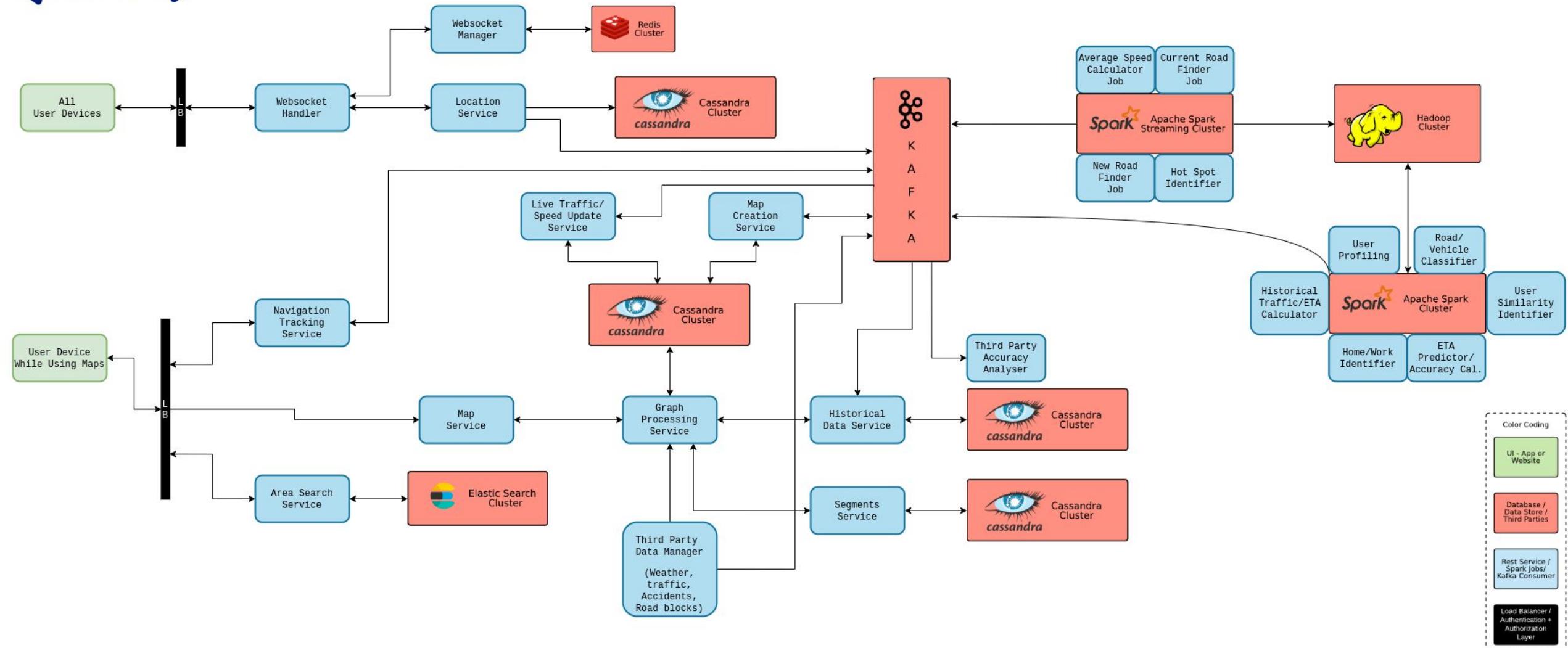
UCL

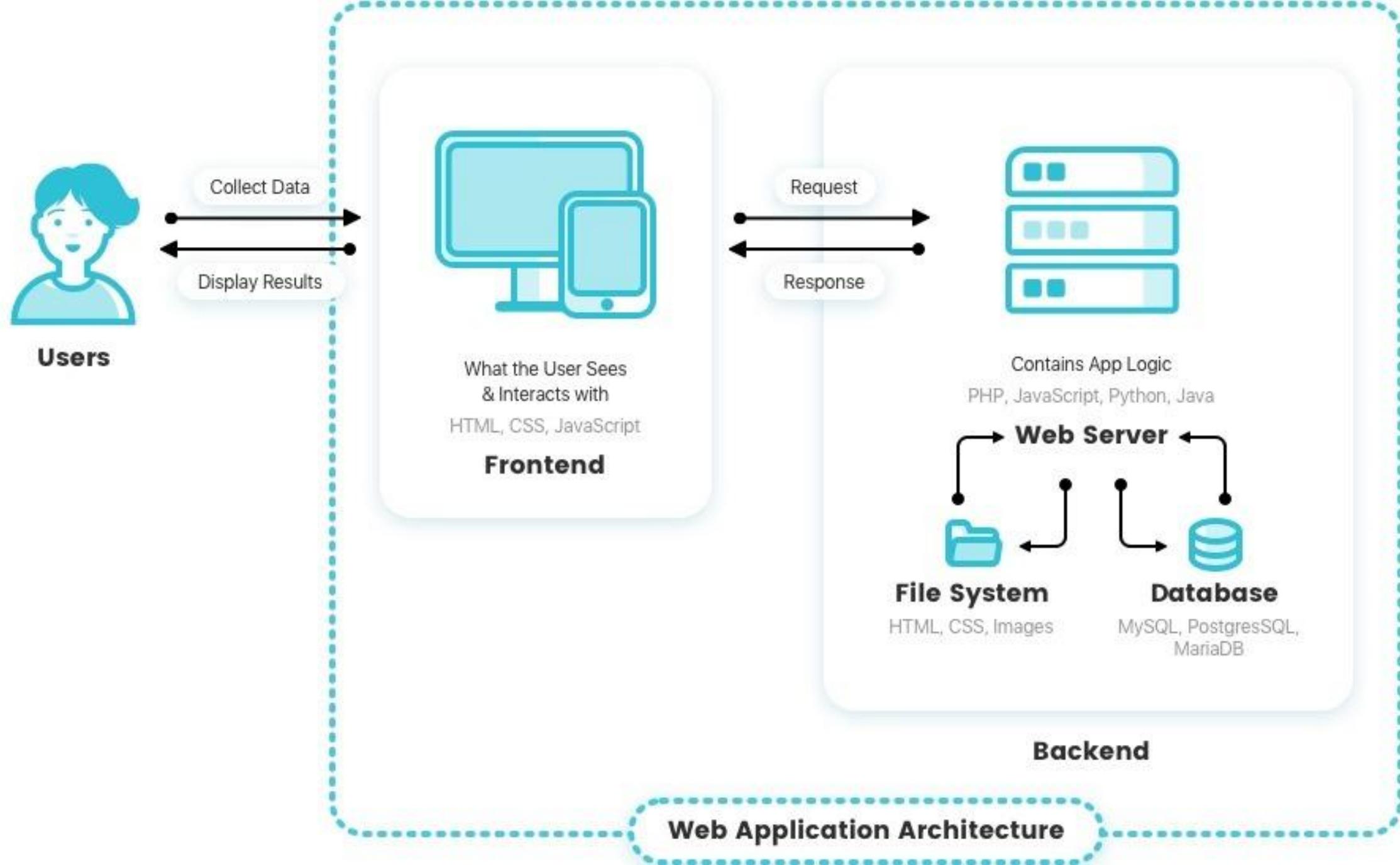
Google

Map data ©2024 Google United States Terms Privacy Send Product Feedback 100 ft

Google Maps System Design

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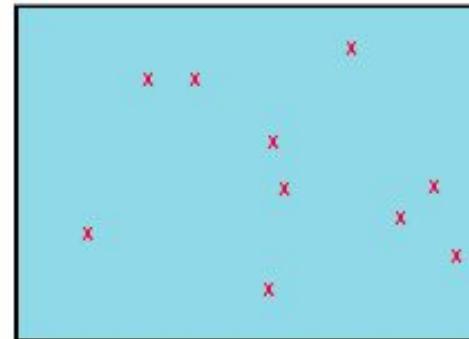




Spatial Data

- There are two main types of spatial data:
 - **Vector**
 - Tabular data with a geometry column
 - **Raster**
 - A digital image, i.e. a matrix of values
- **Handling “big” spatial data depends heavily on its type.**

Vectors



Vector Point Features

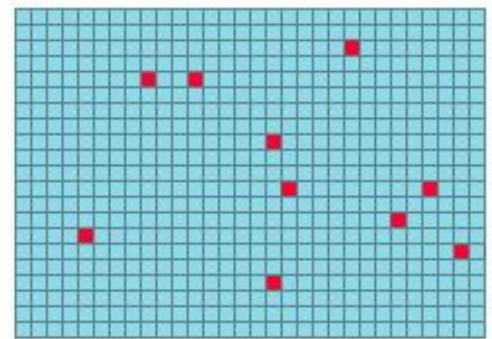


Vector Line Features

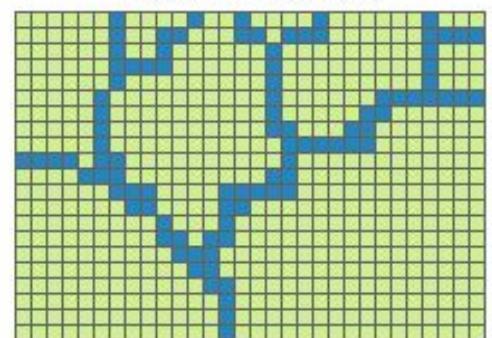


Vector Polygon Features

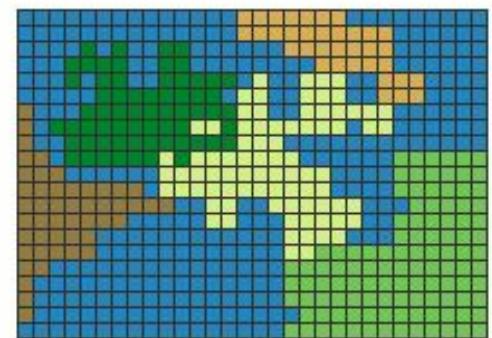
Rasters



Raster Point Features



Raster Line Features



Raster Polygon Features

Vector Data

Vector Data: Definition

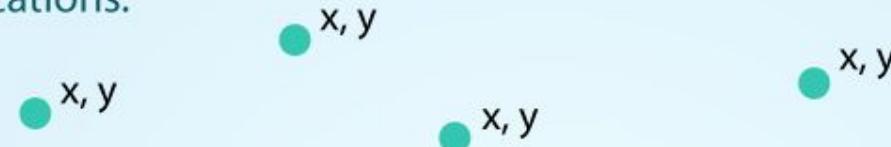
- Vector data can be thought of as a list of values. The features are recorded one by one, with shape being defined by the numerical values of the pairs of xy coordinates, so that:
 - A point is defined by a single pair of coordinate values.
 - A line is defined by a sequence of coordinate pairs defining the points through which the line is drawn.
 - An area is defined in a similar way, only with the first and last points joined to make a complete enclosure.
- The position and shape of a building is captured as a series of four pairs of numerical coordinates. To reproduce the building in a GIS the computer reads these values and draws a line linking the coordinate positions.
- The vector version can also store additional context information about these features – the **attributes** – a very important aspect.

Vector Data Types

- CSV
 - points
- Shapefile
 - Lines/polygons
- GeoJSON
 - Points/lines/polygons
- KML/GPX
 - Lines/points

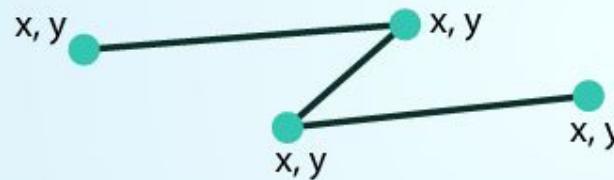
POINTS: Individual **x, y** locations.

ex: Center point of plot locations, tower locations, sampling locations.



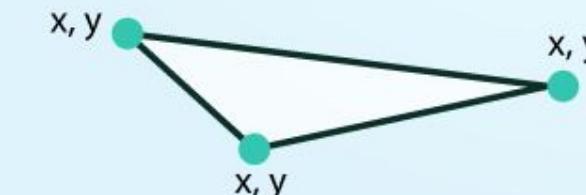
LINES: Composed of many (at least 2) vertices, or points, that are connected.

ex: Roads and streams.



POLYGONS: 3 or more vertices that are connected and **closed**.

ex: Building boundaries and lakes.

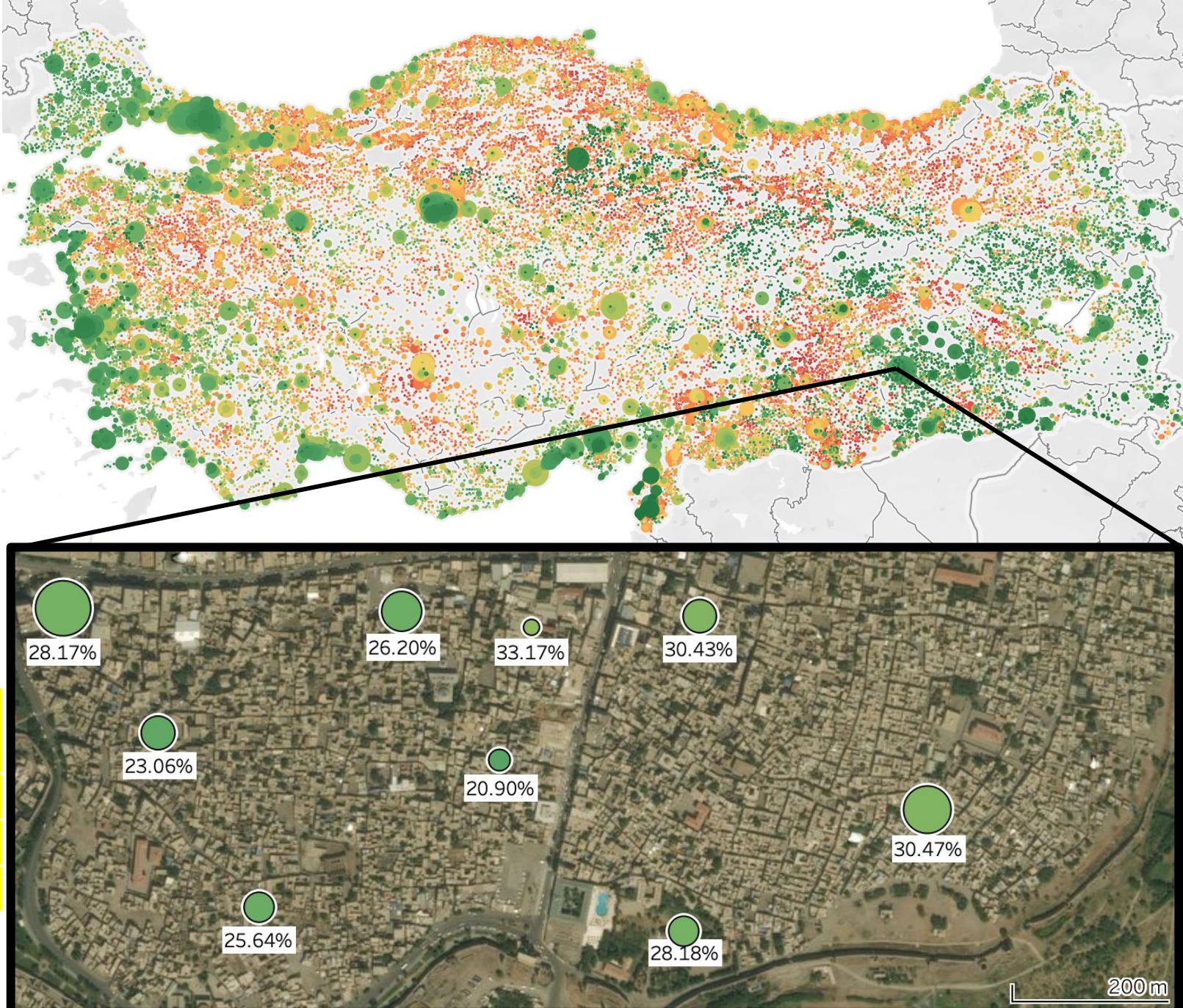


neon®

Point Data

- e.g. ballot box level election results
- Each point can have many attributes including vote share for different parties, etc.

Ballot Box	Party A Vote Share	latitude	longitude
Istanbul 1	0.39	34.349	-84.134
Istanbul 2	0.99	34.079	-84.189
Ankara 1	0.50	34.456	-84.521

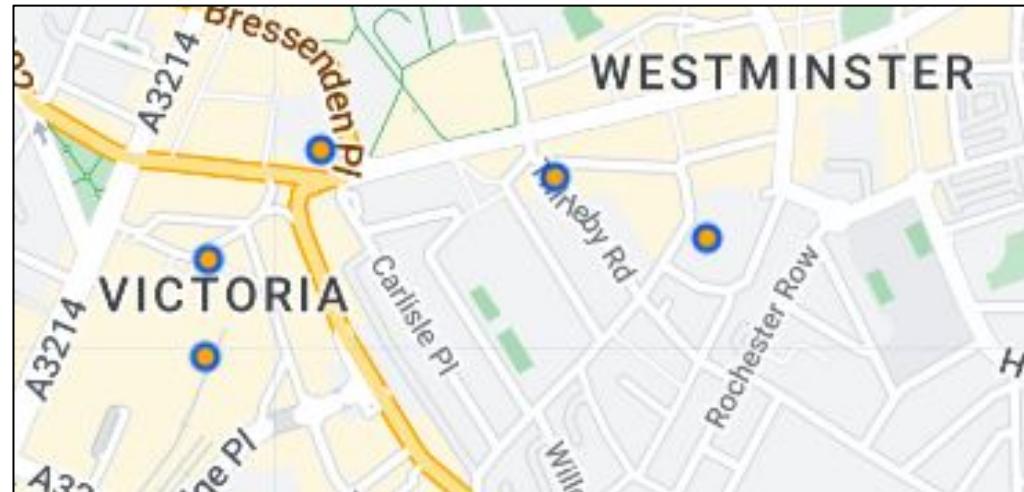


Point Data

device_id	timestamp	ip	user_agent	OS	OS_version	manufacturer	model	carrier	latitude	longitude
7b7ad340-630e	1/1/22 9:31	123.456.789	13.3.1	iOS		13 Apple	iPhone	giffgaff	34.349	-84.134
7b7ad340-630e	1/1/22 9:52	123.456.789	13.3.1	iOS		13 Apple	iPhone	giffgaff	34.079	-84.189
7b7ad340-630e	1/1/22 17:13	123.456.789	13.3.1	iOS		13 Apple	iPhone	giffgaff	34.456	-84.521



→



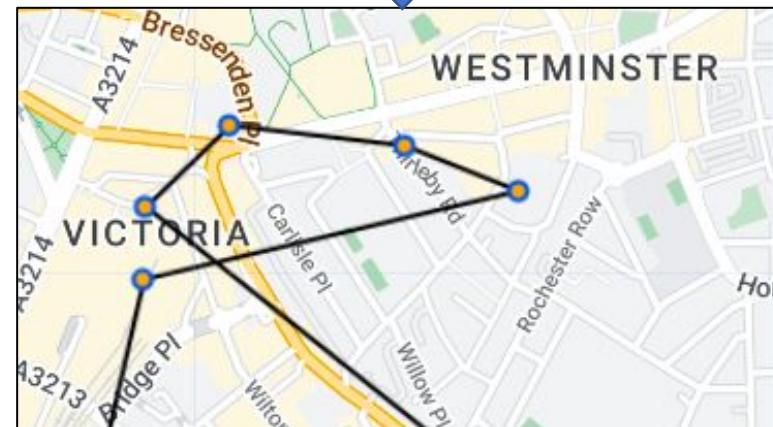
Producer
(app developer)

Line Data



device_id	timestamp	ip	user_agent	OS	OS_version	manufacturer	model	carrier	latitude	longitude
7b7ad340-630e	1/1/22 9:31	123.456.789	13.3.1	iOS		13 Apple	iPhone	giffgaff	34.349	-84.134
7b7ad340-630e	1/1/22 9:52	123.456.789	13.3.1	iOS		13 Apple	iPhone	giffgaff	34.079	-84.189
7b7ad340-630e	1/1/22 17:13	123.456.789	13.3.1	iOS		13 Apple	iPhone	giffgaff	34.456	-84.521

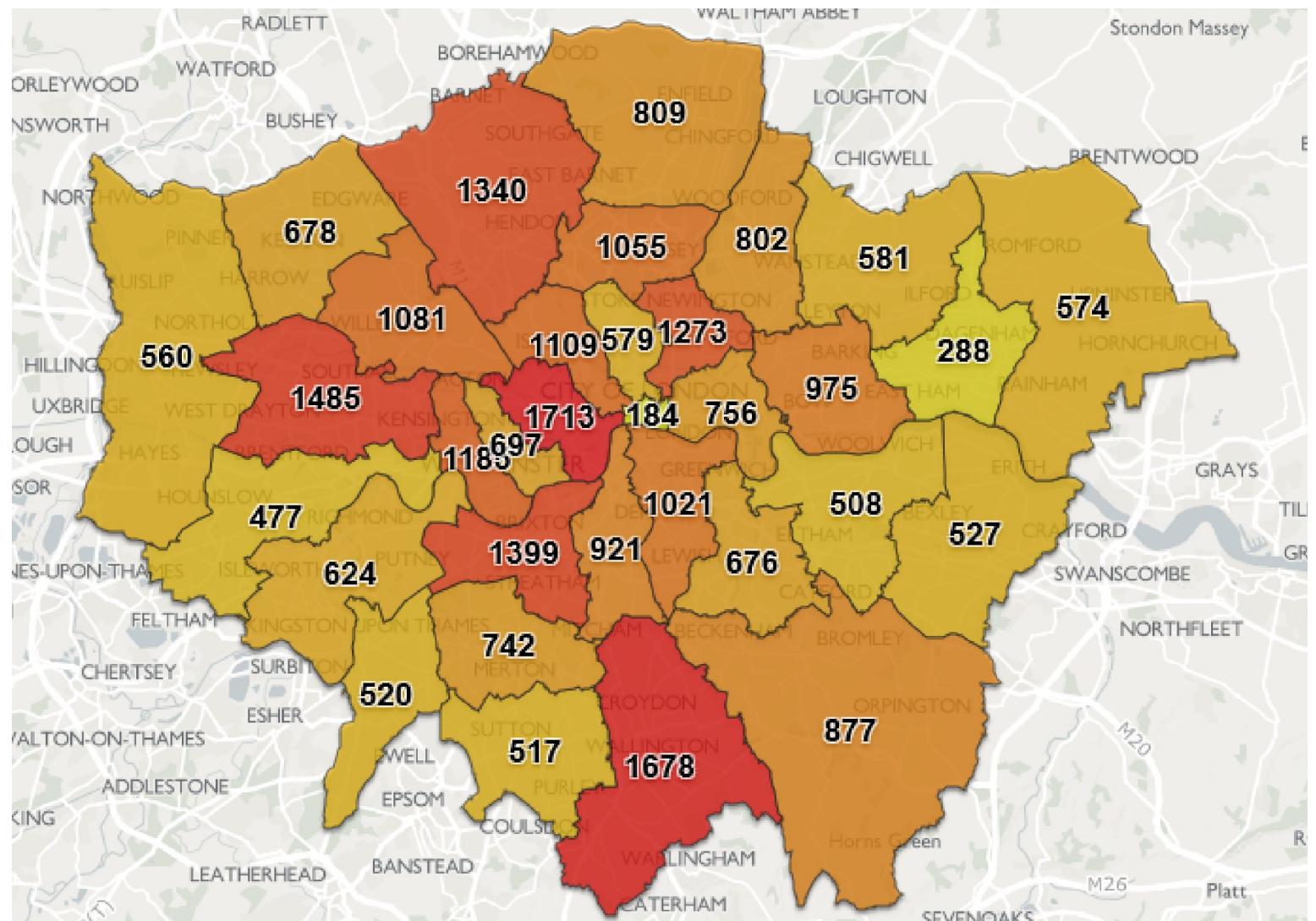
device_id	timestamp	ip	user_agent	OS	OS_version	manufacturer	model	carrier	Geometry
7b7ad340-630e	1/1/22 9:31	123.456.789	13.3.1	iOS		13 Apple	iPhone	giffgaff	LINESTRING(34.349 -84.134, 34.079 -84.189, 34.456 -84.521)



Producer
(app developer)

Polygon Data

Borough	Population	Number of Pubs	geometry
Hackney	280,000	2300	POLYGON(34.079 -0.381 ...)
Camden	279,000	1943	POLYGON(35.419 -0.352 ...)
Westminster	261,000	532	POLYGON(33.044 0.431 ...)



Building Footprints (Polygons)

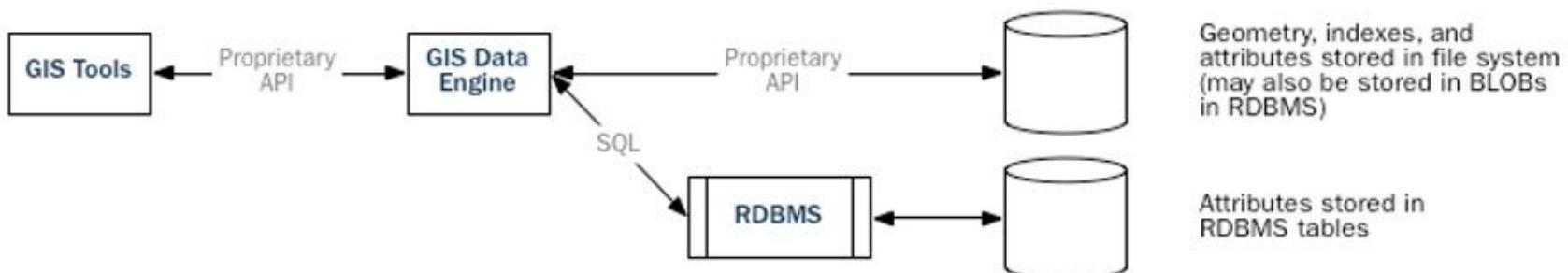


Evolution of GIS Architectures

First-Generation GIS:



Second-Generation GIS:



Third-Generation GIS:



What is a database?

System for storage and random access of relationally (tables of rows and columns) structured data, providing the following capabilities for that data.

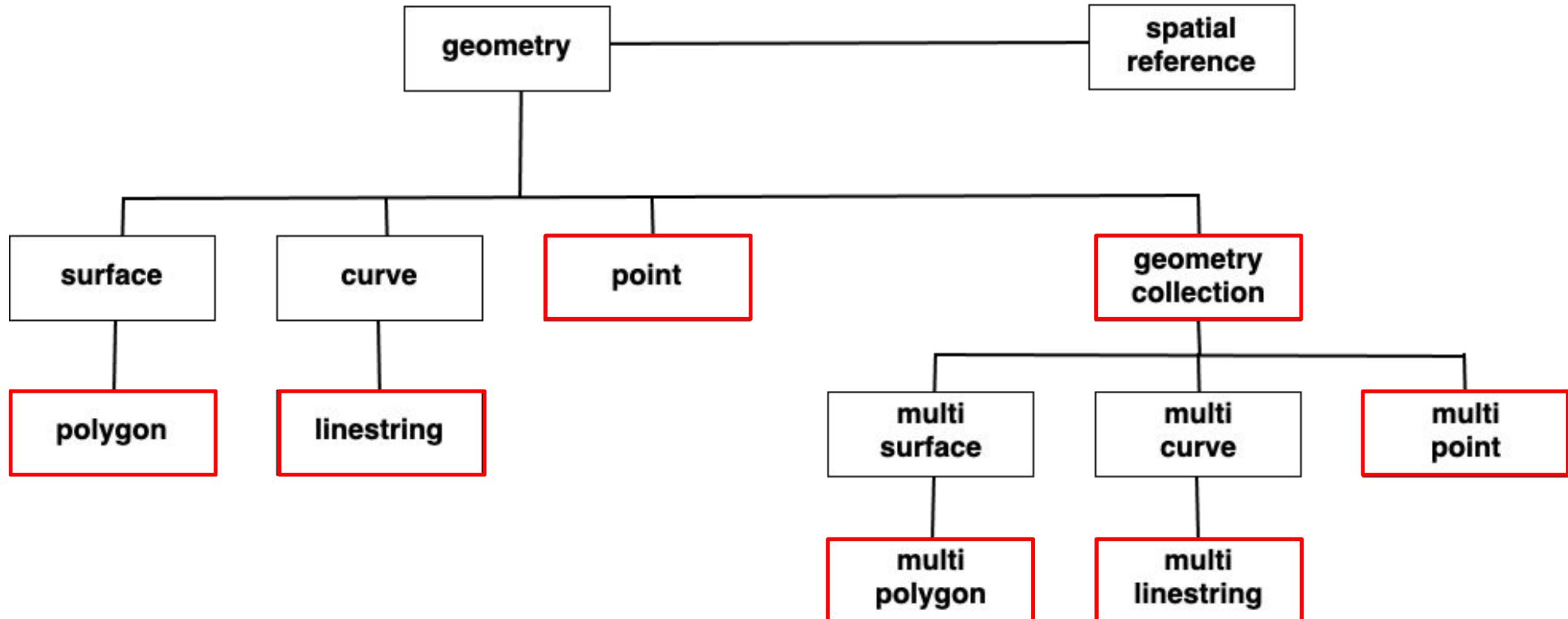
- **Data Types**
 - number, date, and string
- **Indexes**
 - b-tree, hash
- **Functions**
 - `strlen(string)`, `pow(float, float)`, `now()`

What is a **spatial** database?

System for storage and random access of relationally (tables of rows and columns) structured data, providing the following capabilities for that data.

- **Data Types** including **Spatial Types**
 - number, date, string, **geometry**, **geography** and **raster**
- **Indexes** including **Spatial Indexes**
 - b-tree, hash, **rtree**, quadtree
- **Functions** including **Spatial Functions**
 - **strlen(string)**, **pow(float, float)**, **now()**, **ST_Area()**, **ST_Distance()**

Spatial Types



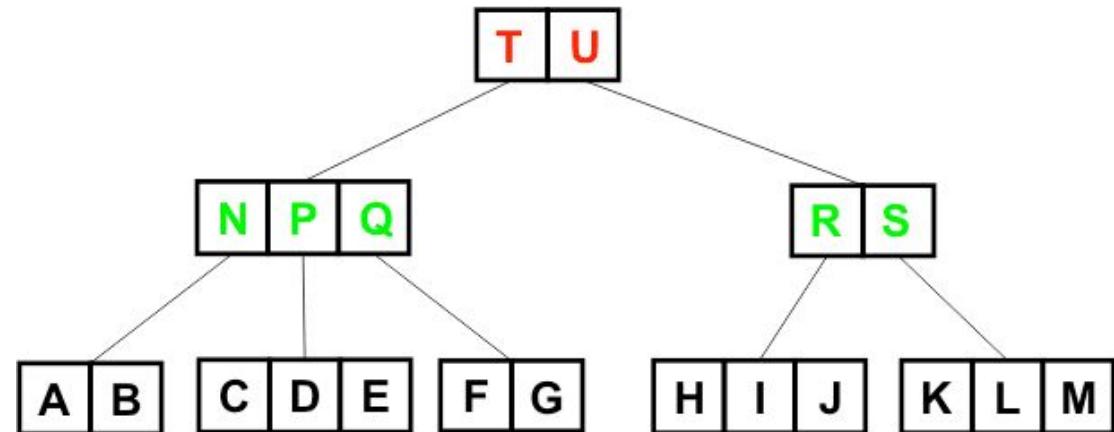
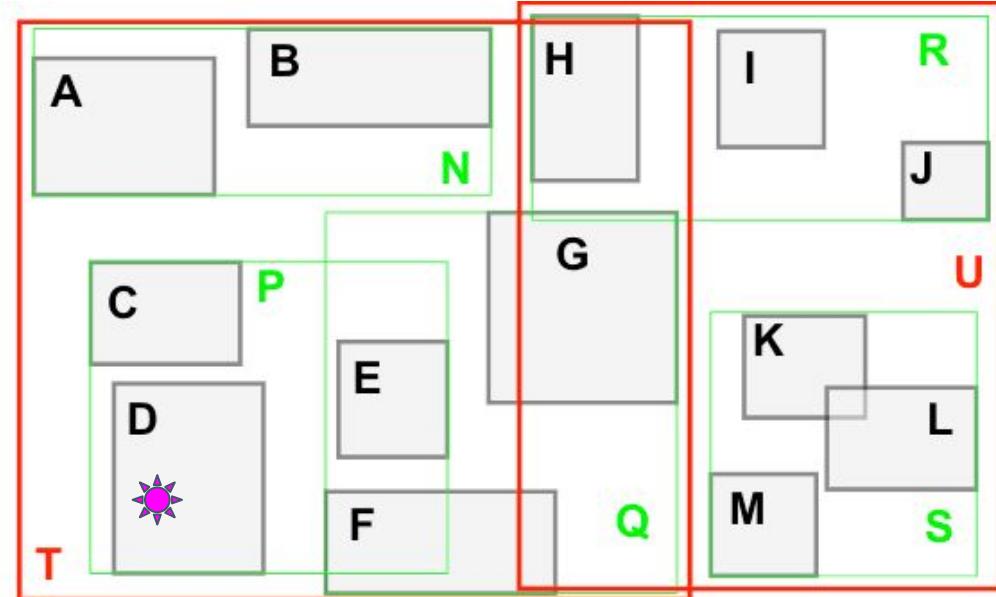
Spatial Indexes

This R-Tree organizes the spatial objects so that a spatial search is a quick walk through the tree.

To find what object contains  ?

- The system first checks if it is in **T** or **U** (**T**)
- Then it checks if it is in **N**, **P** or **Q** (**P**)
- Then it checks if it is in **C**, **D** or **E** (**D**)

Only 8 boxes have to be tested. A full table scan would require *all 13 boxes* to be tested.
The larger the table, the *more powerful* the index is.

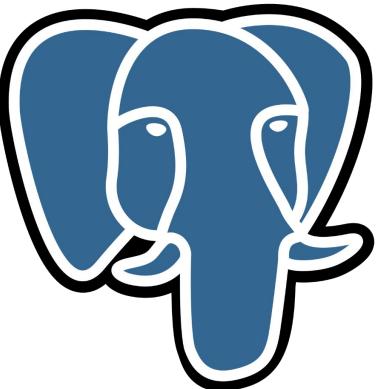


Spatial Functions

For example:

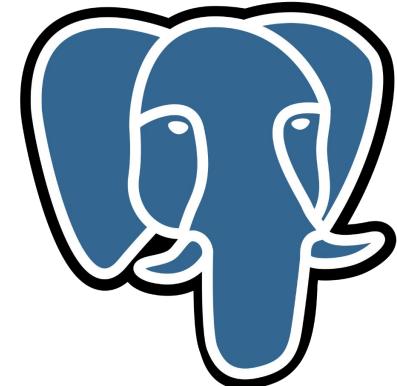
- `ST_GeometryType(geometry) → text`
- `ST_Area(geometry) → float`
- `ST_Distance(geometry, geometry) → float`
- `ST_Buffer(geometry, radius) → geometry`
- `ST_Intersection(geometry, geometry) → geometry`
- `ST_Union([geometry]) → geometry`

What is PostGIS?

 =  + **CREATE
EXTENSION
postgis;**

What is PostgreSQL?

- Enterprise RDBMS
- Functionally equivalent to Oracle / MSSQL
- Multi-vendor open source community
- Multi-platform support, and available on all clouds
- Highly extensible by design
 - Types, functions, indexes, replication slots, foreign data, core hooks
 - What makes PostGIS possible



Why not files?

- Shape, FGDB, GeoPackage?
- No lingua franca for file access, every format has its own library
 - Database has SQL
 - Databases have JDBC, ODBC, others
- Multi-user access to files results in either
 - Global file locking and performance issues (best case)
 - File corruption and data loss (worst case)
- All database-style queries have to be implemented on client
 - Joins, aggregations, set-based logic



PostGIS reference users - private sector



Ball Aerospace
& Technologies Corp.



PostGIS reference users - government



Natural Resources
Canada

Canada

Ressources naturelles
Canada



Ordnance Survey

FC

PostGIS 3rd party integration

Desktop



and more...

Middle



GeoServer

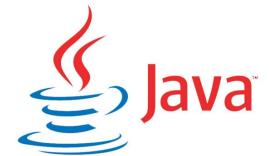


mapnik



and more...

Language

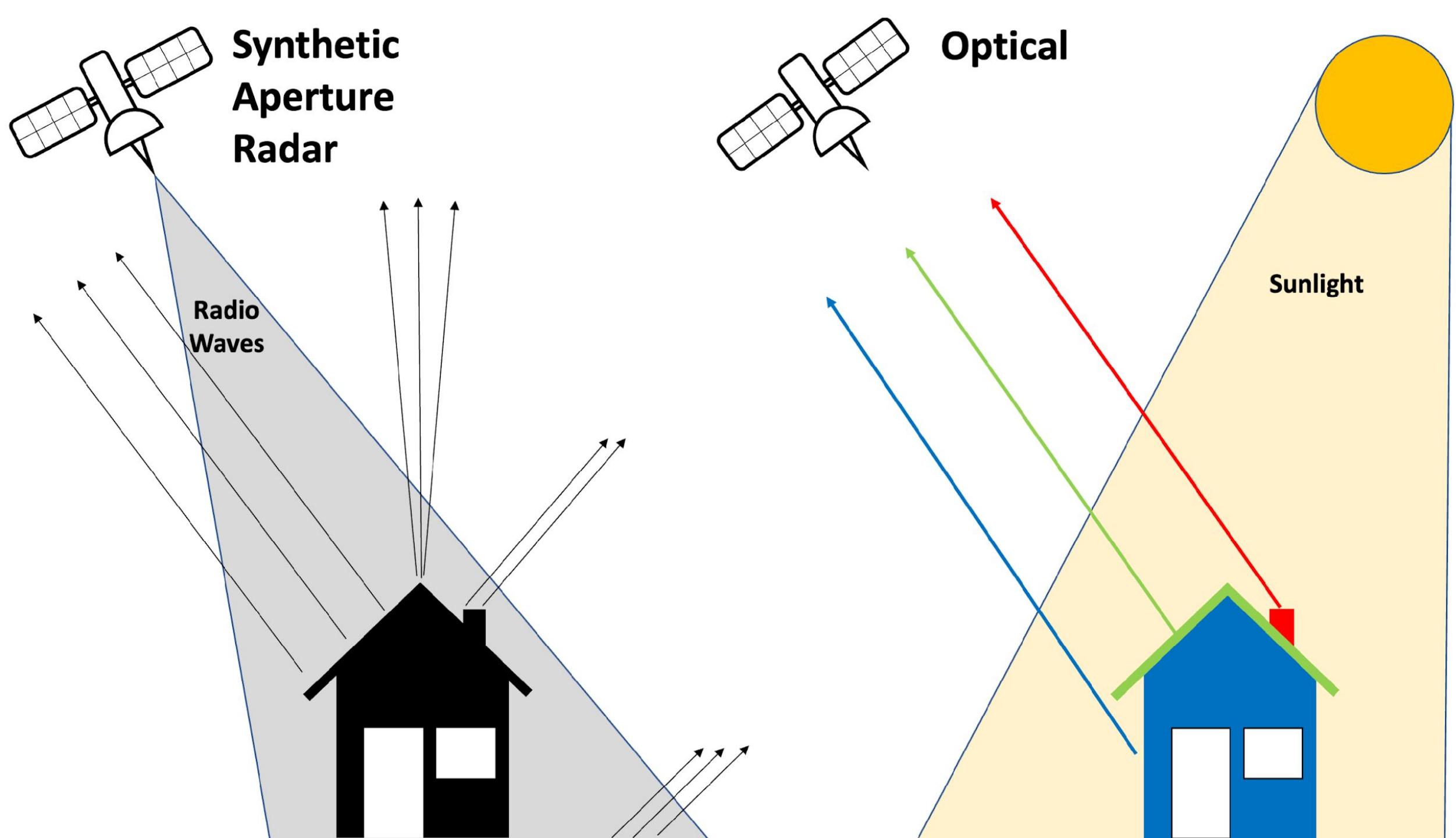


and more...

Raster Data

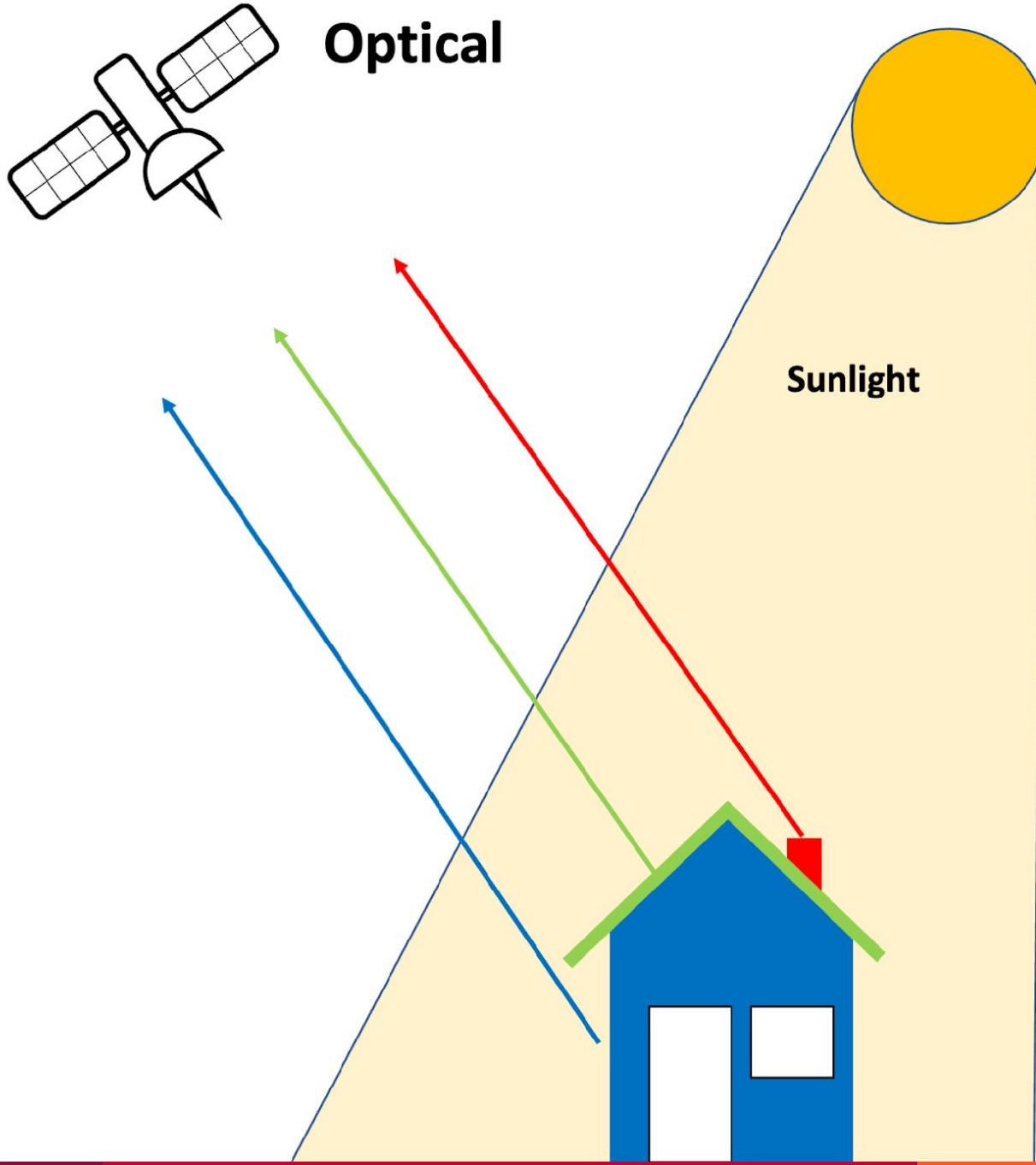
Raster Data: Definition

- Similar to a digital photograph. The entire area of the map is subdivided into a grid of tiny cells, or pixels. A value is stored in each of these cells to represent the nature of whatever is present at the corresponding location on the ground.
- The major use of raster data involves storing map information as digital images, in which the cell values relate to the pixel colors of the image. To reproduce the image the computer reads each of these cell values one by one and applies them to the pixels on the screen.



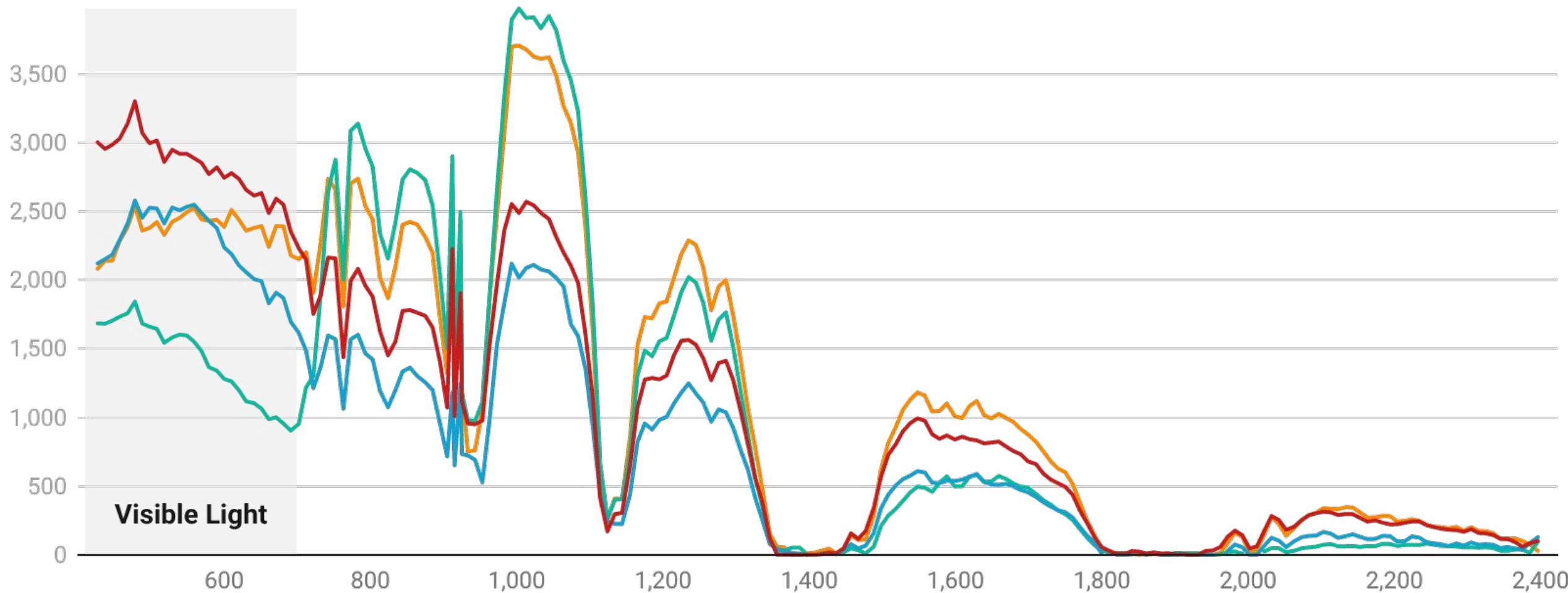
Raster Data: Examples

- Satellite imagery
 - Optical
 - High resolution (<1m/pixel)
 - MAXAR, PLEIADES
 - Medium resolution (10-30m per pixel)
 - [ESA Sentinel-2](#), [NASA Landsat](#)
 - Low resolution (>30m/pixel)
 - MODIS, VIIRS, GOES
 - Radar
 - High resolution (<1m/pixel)
 - Capella Space, ICEYE
 - Medium resolution (10-30m per pixel)
 - ESA Sentinel-1
 - Low resolution (>30m/pixel)



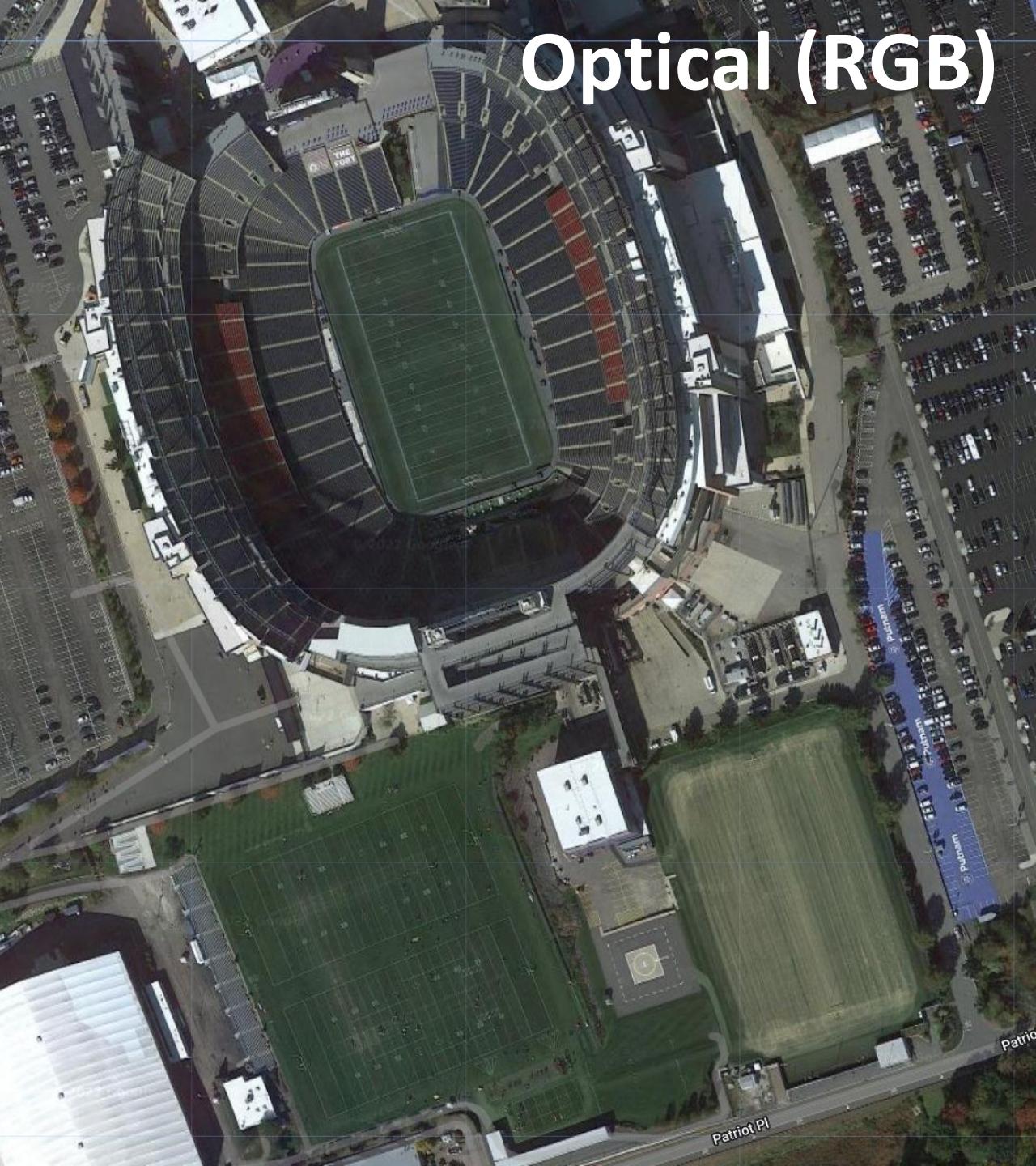
Spectral Profiles of Different Materials

— Oil — Concrete — Vegetation — Water

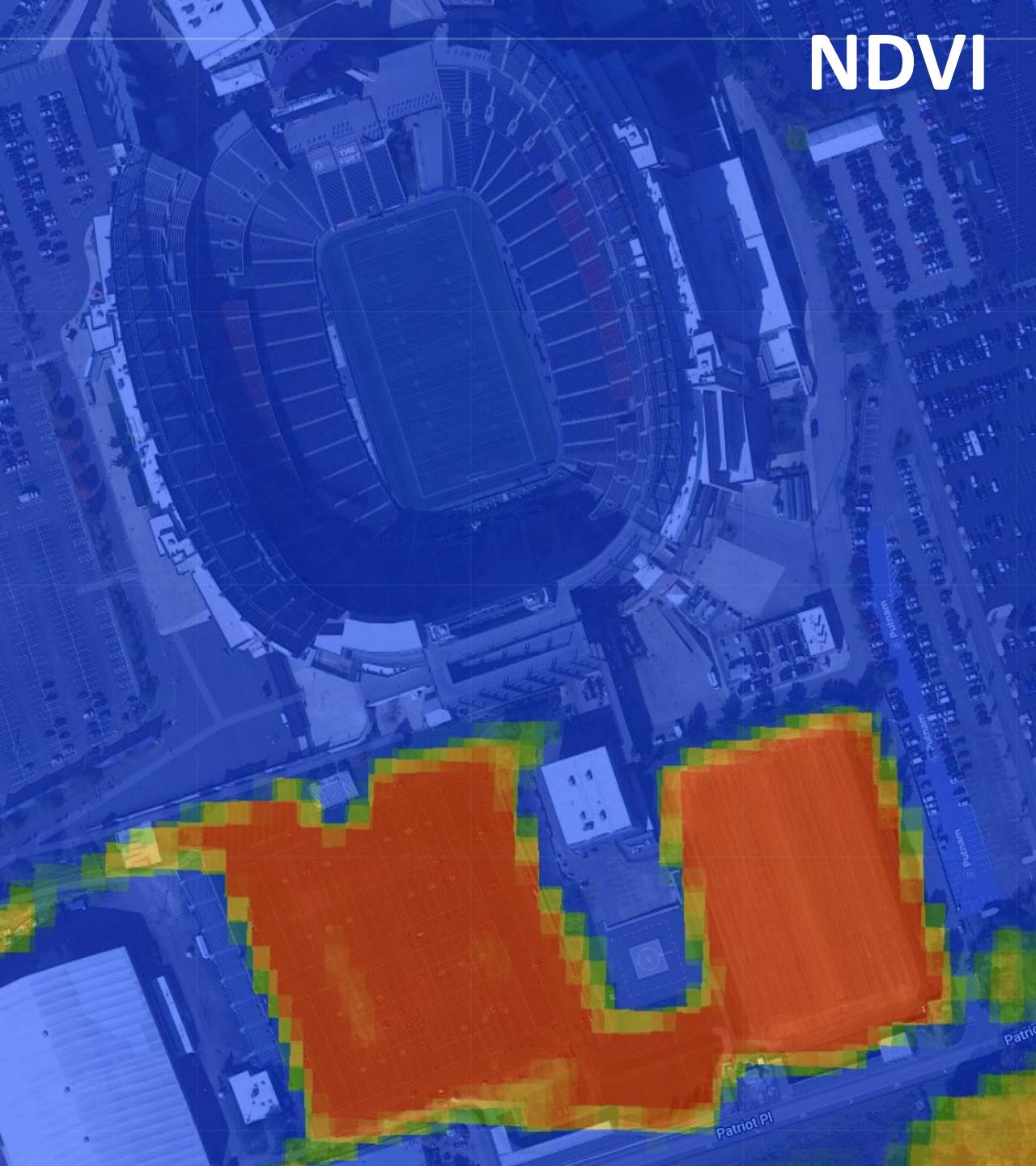


Created with Datawrapper

Optical (RGB)



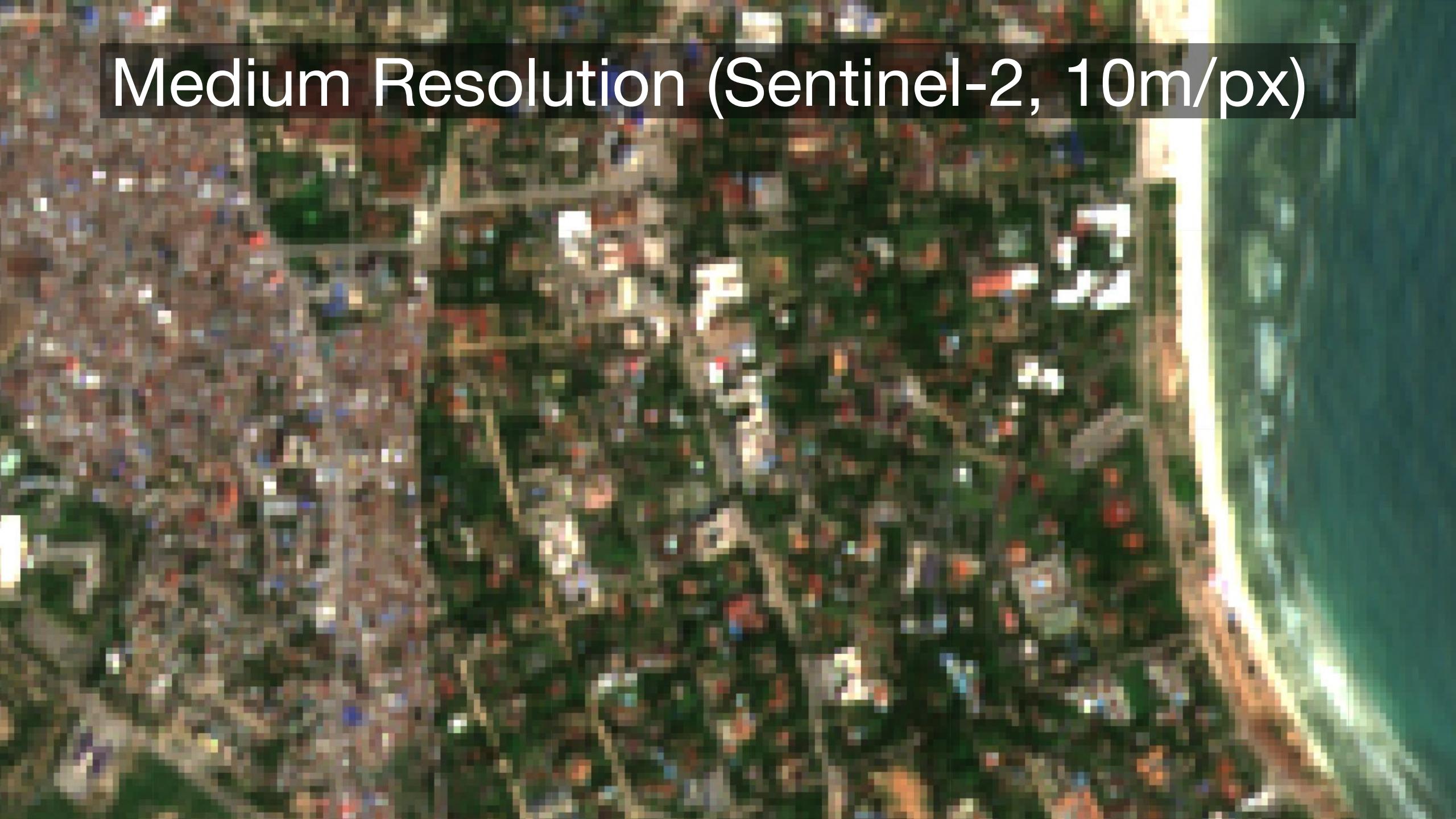
NDVI



Medium Resolution (Landsat 8, 30m/px)



Medium Resolution (Sentinel-2, 10m/px)



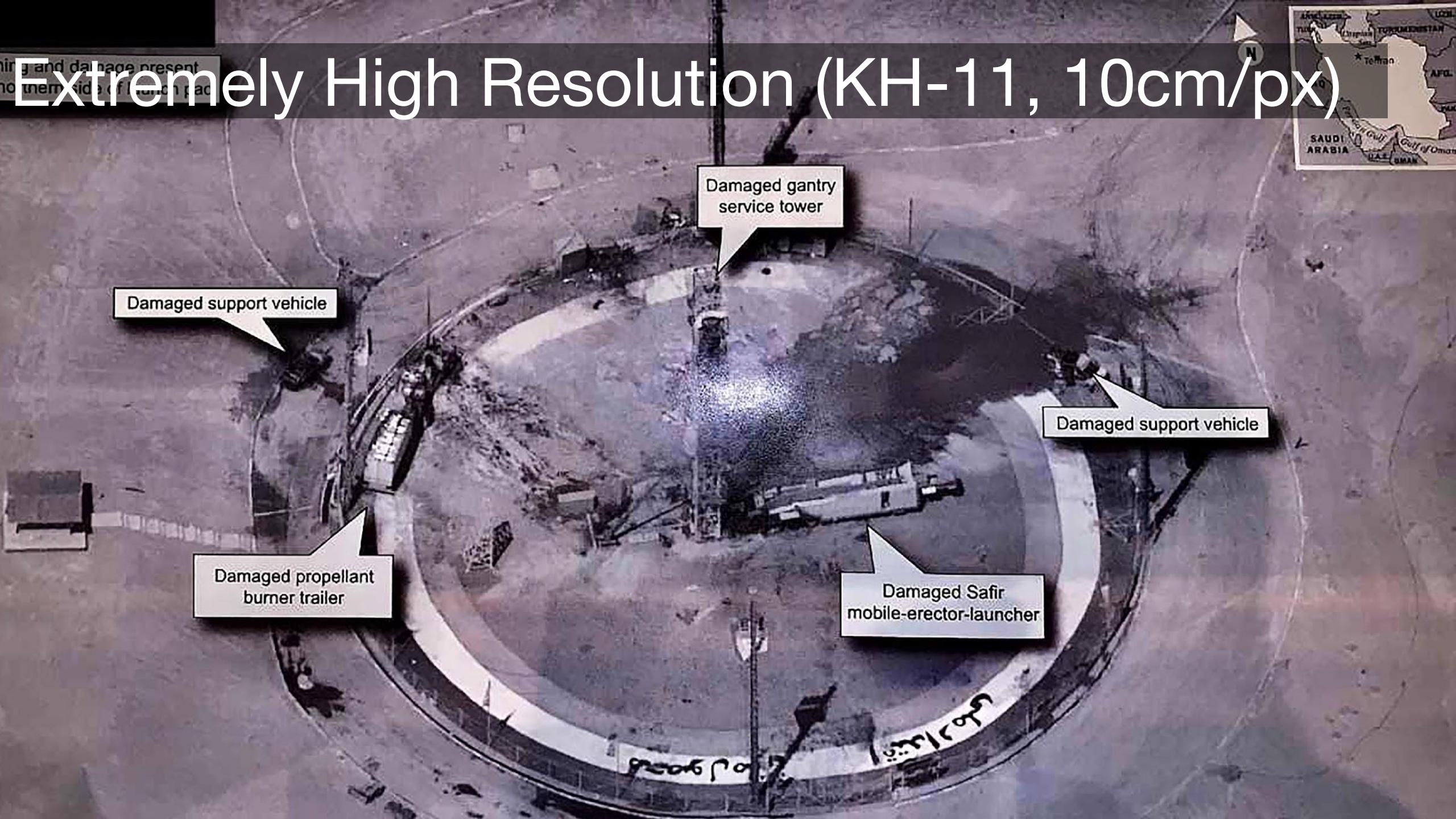
High Resolution (Maxar, 50cm/px)

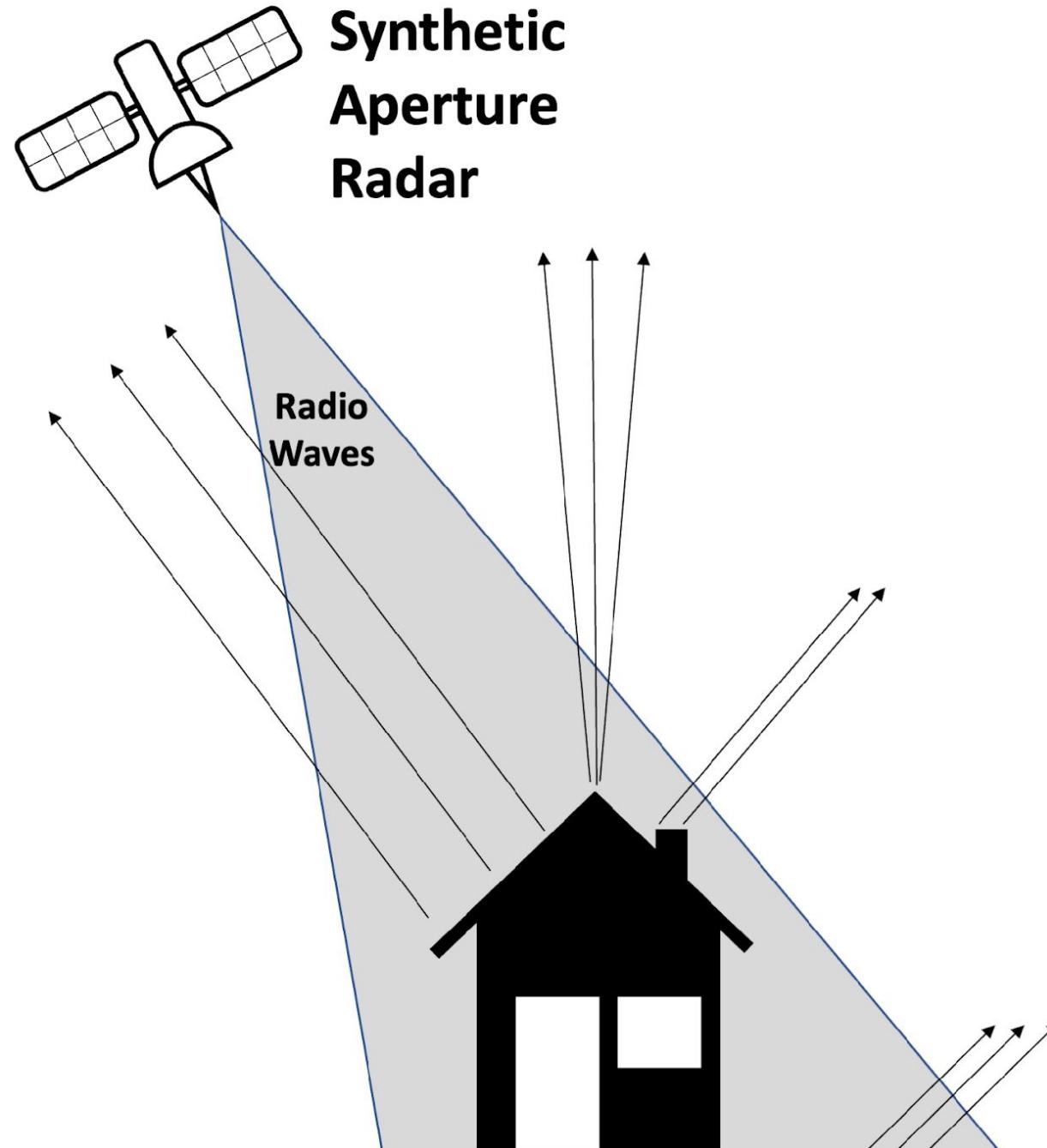


High Resolution (Maxar, 50cm/px)



Extremely High Resolution (KH-11, 10cm/px)

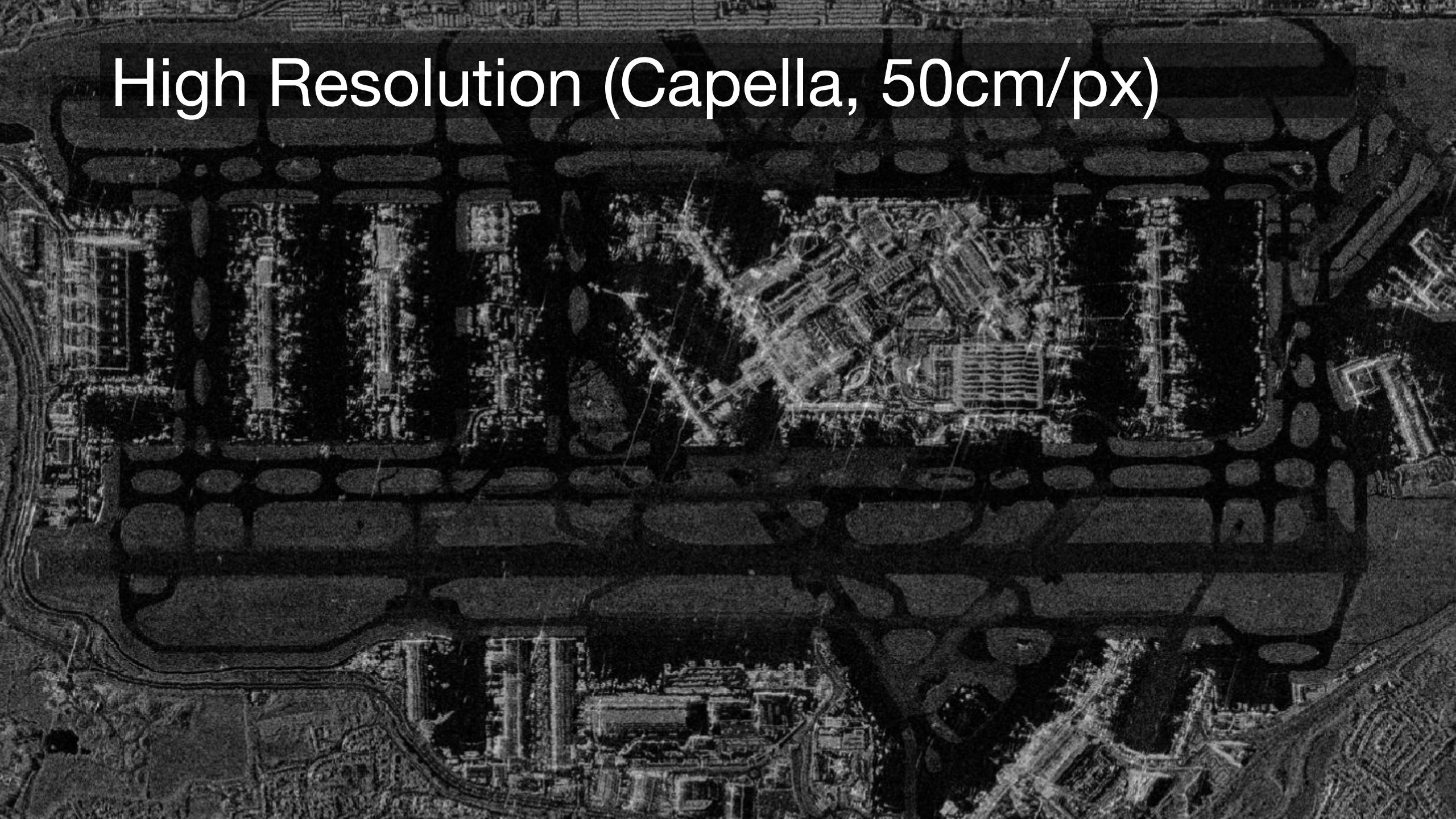




Medium Resolution (Sentinel-1, 10m/px)

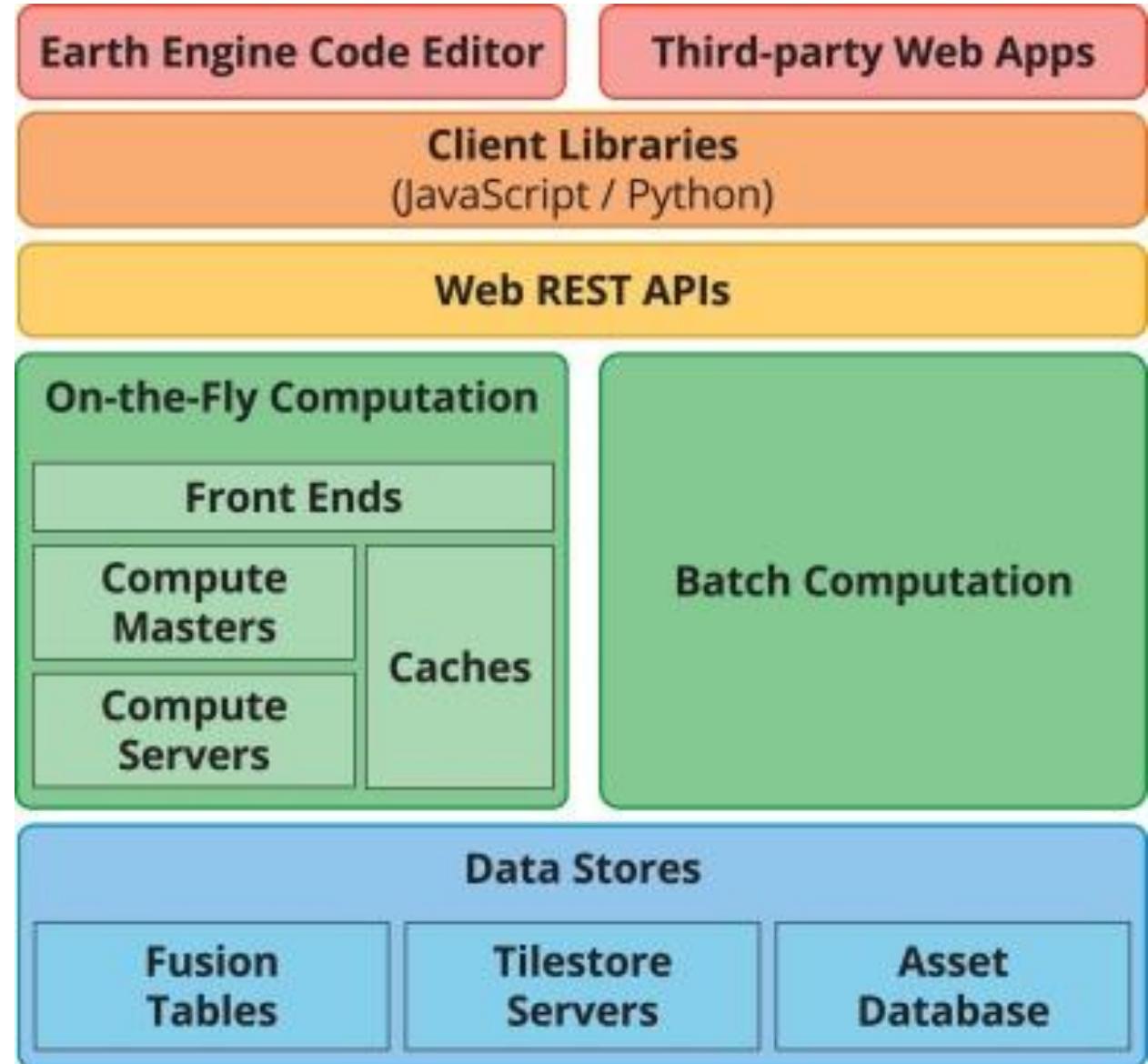


High Resolution (Capella, 50cm/px)

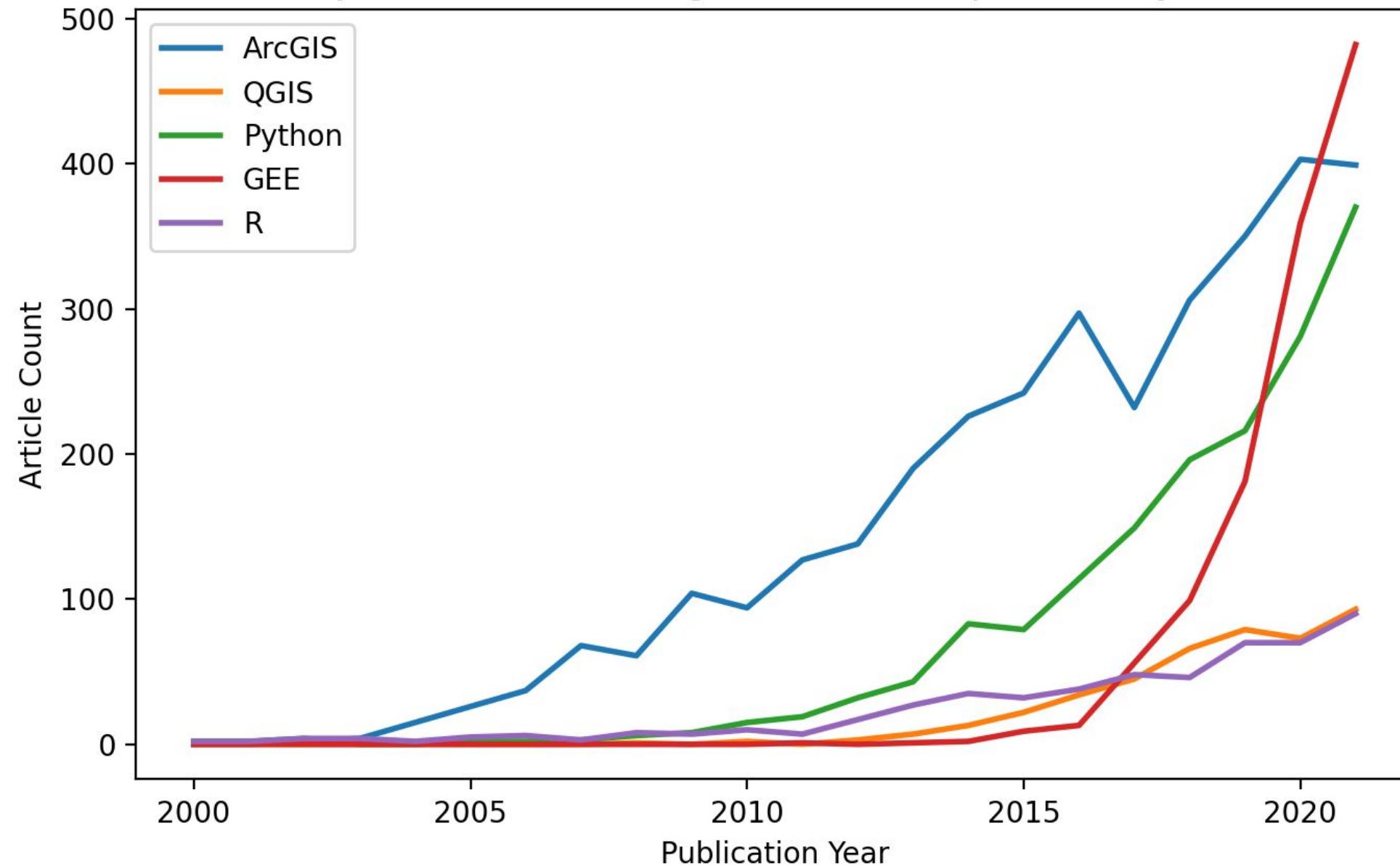


Google Earth Engine

- Asset catalog of over 30 petabytes of geophysical data, mostly sat imagery
 - If you stored this on macbook pros and laid them end-to-end in a straight line, they would stretch from here to Heathrow airport
- Runs large spatial computations on google servers
- Allows for the building and deployment of apps, all in one place



Number of Journal Articles Using Different Geospatial Analysis Softwares



Module Overview

Module Outline

Week	Lecture	Tools
1	Introduction	DuckDB
2	Databases and SQL	DuckDB
3	Spatial Databases I	DuckDB
4	Spatial Databases II	DuckDB
5	Guest Lecture: GFW	DuckDB
6	Intro to Earth Engine	Google Earth Engine
7	Classification I	Google Earth Engine
8	Classification II	Google Earth Engine
9	Synthetic Aperture Radar	Google Earth Engine
10	User Interface Design	Google Earth Engine

Lectures

- Lectures will provide the conceptual foundations for different aspects of application design.
- Weeks 1-5 will focus on handling large vector datasets using Databases and SQL.
 - Week 5 will feature a guest lecture from Global Fishing Watch
- Weeks 6-10 will introduce analytical concepts in Google Earth Engine
 - Lectures 6-9 will be co-taught with CASA0023 Remote Sensing

Workshops

- Practical sessions involving the application of concepts learned during the lecture
- Aim is to hone our coding skills
- Weeks 1-5
 - We'll be learning SQL using DuckDB in Jupyter Notebooks
- Weeks 6-10
 - We'll be working in Google Earth Engine's browser-based IDE
 - Weeks 6 and 7 will be joint with CASA0023 Remote Sensing, location TBD

Assessments

- **Database Quiz - 30%**

- **10th February 2024**

- Administered during the Workshop on Week 5. It will last one hour. Students will be provided with a dataset and will have to answer 10 questions.

- **Group Application - 50%**

- **27th April 2024**

- The application will import data and apply analysis to one of the datasets from the Earth Engine catalogue, with optional integration of third-party data. The application must allow users to interactively query results.

- **Presentations - 20%**

- **27th April 2024**

- Each group is expected to produce a presentation showing off the application, how the group carried out the analysis of the dataset, the limitations of the analysis and how the interactive tool works under the hood

Workshop

