

Connecting Google Colab and Earth Engine for Satellite Imagery Analysis and Vegetation Monitoring

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Natural

Environment



Towards the UK fire danger rating system (UKFDRS)

The project aims to establish and validate the scientific basis and essential components necessary for the development of a bespoke UK Fire Danger Rating System (FDRS).

This system will facilitate the evaluation of current and future fire regimes, enabling the assessment, management, and mitigation of wildfire impacts and the associated risk of uncontrollable wildfires throughout the UK.





Agenda

- Introduction
- Google Earth Engine (GEE) What it is? Benefits?
- Google Colab What it is? Benefits?
- Connecting GEE API in Google Colab
- Importing and visualising datasets
- Data analysis applications
- Practical overview of the Colab notebook
- Limitations
- Questions



Introduction

- The Earth is undergoing rapid changes, and environmental degradation is intensifying, significantly impacting vegetation patterns.
- Environmental monitoring, forest monitoring, land cover assessment, climate studies, fires and disaster management rely heavily on access to up-to-date satellite data.
- The ability to extract meaningful insights from these vast datasets is crucial for informed decision-making that help addressing environmental challenges.
- There is an urgent need for timely and efficient analysis of satellite imagery to monitor and understand these transformations.





Google Earth Engine

- A cloud-based geospatial platform developed by Google
- Allows their users to analyse and process large-scale Earth observation and geospatial data



Represents a valuable tool for researchers, scientists, environmentalists, and developers interested in connecting the power of geospatial data and analysis to address various Earth-related challenges and questions.

Source image: Google Earth Engine



Archive

Code editor

Geospatia

Analysis

Visualizatio n

Collaboratior

GEE Key Features and benefits

- Provides access to a massive archive of Earth observation and derived products .
 This data spans multiple decades and covers the entire globe in most of the cases
 - This data spans multiple decades and covers the entire globe in most of the cases.
 - You can write and run JavaScript or Python code within the Earth Engine Code Editor.
 - Offers a wide range of geospatial analysis tools and functions on large datasets
 - Examples: image processing, spatial analysis, time-series analysis, and machine learning.
 - You can visualise geospatial data and analyse results interactively on the map within the Code Editor, making it easier to explore and understand complex spatial patterns.
 - Supports collaboration by enabling users to share their scripts, visualisations, and analysis results
 - Collaborators can work on projects simultaneously and access shared data and code.





- It is a free cloud-based platform provided by Google
- It allows you to write and execute Python code in a web-based interactive environment
- It is presented in a Jupyter notebook format



Google Colab Key Features

	• Provides access to access to Graphics Processing Units (GPUs) and Tensor Processing Units (TPUs) at no
Free	cost.
access	
Google drive	 It is integrated with Google Drive, making easy to store and share notebooks
	 It comes with a wide range of pre-installed Python libraries
Libraries	 Example: pandas, scikit-learn, TensorFlow, PyTorch etc.
	• Notebooks are interactive documents that combine code execution cells with formatted text and visualisations.
Notebooks	• You can write and run Python code, add explanations, charts, and images, creating a narrative around their data analysis or machine learning projects.
	 Supports collaboration by enabling users to share their scripts, visualisations, and analysis results with others.
Collaboratio	• Collaborators can work on projects simultaneously and access shared data and code.
	• You can integrate Colab with version control (e.g. GitHub) to track changes, and edit.



Let's explore a Colab Jupyter notebook and how you can connect it with Google Earth Engine





STEP 1: Install your libraries

STEP 2: Import your libraries and link to your google drive

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Setting up 🛆 FOSS4G_UK.ipynb 🛛 🏠 File Edit View Insert Runtime Tools Help + Code + Text ≣ [2] # library to link Google Earth Engine !pip install earthengine-api $\{x\}$ # Other libraries that you might need for further analysis !pip install geemap !pip install matplotlib-venn import ee import geemap from google.colab import drive print("Setup completed") drive.mount("/content/drive")

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COLink with Google drive







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	GENERATE TOKEN	↑		Cancel	Continue	If you don't want to continue, close this window.

Datasets available in GEE achive

Category	Datasets	Description
Optical Satellite Data	Landsat 8 Surface Reflectance, Sentinel-2 Surface Reflectance, MODIS Land Surface Temperature	Satellite imagery datasets providing reflectance, temperature, and other Earth observation data.
Radar Satellite Data	ALOS PALSAR, Sentinel-1	Radar satellite datasets offering various applications including terrain mapping and monitoring.
Topography	SRTM (Shuttle Radar Topography Mission)	Elevation and topography data derived from radar measurements.
Land Cover Data	Copernicus Global Land Cover Layers, MODIS Land Cover Type	Global land cover classification datasets at various resolutions and time periods.
Climate Datasets	ERA5 Climate Reanalysis, CHIRPS Precipitation Data	Climate datasets offering temperature, precipitation, and other meteorological variables.
Derived Maps	Global Forest Change, NASA Global Fire Maps, GEDI Biomass	Maps derived from satellite data, including forest change, fire occurrence, and biomass.

See more at: <u>Earth Engine Data Catalog</u> <u>Google for Developers</u>



Define parameters you want to consider

- Scale
- Area
- Time period

```
# Study area
polygon = ee.Geometry.Polygon(coordinates)
```

```
# Time period
start_date = "2017-01-01"
end_date = "2022-12-31"
```



Define functions that you want to apply to your imagery

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- Cloud masking
- Vegetation indices

🝐 FOSS4G_UK.ipynb 🛛 🕁 File Edit View Insert Runtime Tools Help All changes saved + Code + Text C def maskL8sr(image): """ # define cloud mask function""" # Bits 3 and 5 are cloud shadow and cloud, respectively. cloudShadowBitMask = (1 << 3)cloudsBitMask = (1 << 5)# Get the pixel QA band. qa = image.select('pixel qa') # Both flags should be set to zero, indicating clear conditions. mask = qa.bitwiseAnd(cloudShadowBitMask).eq(0) \ .And(qa.bitwiseAnd(cloudsBitMask).eq(0)) return image.updateMask(mask) def L8 indices(image): # Rename bands image renamed = image.select('B1').rename('L8 SR') \ .addBands(image.select('B2').rename('L8 Blue')) \ .addBands(image.select('B3').rename('L8 Green')) \ .addBands(image.select('B4').rename('L8 Red')) \ .addBands(image.select('B5').rename('L8 NIR')) \ .addBands(image.select('B6').rename('L8_Swir_1')) \ .addBands(image.select('B7').rename('L8_Swir_2')) # Calculate vegetation indices for Landsat 8 18 ndvi = image renamed.normalizedDifference(['L8 NIR', 'L8 Red']).rename('L8 ndvi')

Add indices to the original image
l8_with_indices = image_renamed.addBands([l8_ndvi])
return l8_with_indices



Loading Satellite data

Define functions that you want to apply to your imagery

		Edit View Insert Runtime Tools Help <u>All changes saved</u>
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5		
s	0	<pre>l8 = ee.ImageCollection('LANDSAT/LC08/C01/T1_SR') \ .filterDate(start_date, end_date) \ rest(rest(201))</pre>
		<pre>.map(maskL8sr) \ .map(L8 indices) \</pre>
		.filterBounds(polygon)
		<pre># Compute the median composite of the Landsat 8 imagery stack l8stack = l8.median().float().clip(polygon)</pre>
		l8stack
		▼ Image (12 bands)
		type: Image
		▼ bands: List (12 elements)
		▶ 0: "L8 SR", float, EPSG:4326, 13x13 px
		▶ 1: "L8 Blue", float, EPSG:4326, 13x13 px
		▶ 2: "L8 Green", float, EPSG:4326, 13x13 px
		▶ 3: "L8_Red", float, EPSG:4326, 13x13 px
		▶ 4: "L8_NIR", float, EPSG:4326, 13x13 px
		▶ 5: "L8_Swir_1", float, EPSG:4326, 13x13 px
		▶ 6: "L8_Swir_2", float, EPSG:4326, 13x13 px
		▶ 7: "L8_ndvi", float, EPSG:4326, 13x13 px
		▶ 8: "L8_nbr", float, EPSG:4326, 13x13 px
		▶ 9: "L8_nbr2", float, EPSG:4326, 13x13 px
		▶ 10: "L8_ndmi", float, EPSG:4326, 13x13 px
		▶ 11: "L8_savi", float, EPSG:4326, 13x13 px
		properties: Object (1 property)



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Visualisation of other available datasets and plotting histograms

display(m)



SRTM

✓ [22] # Get the elevation band from the SRTM image elevation_band = srtm.select('elevation')

> # Create a histogram of elevation values histogram = elevation_band.reduceRegion(reducer=ee.Reducer.histogram(100), # Number of bins for the histogram geometry=polygon, scale=30, # Resolution in meters maxPixels=1e9

Extract the histogram data as a list hist_data = histogram.get('elevation').getInfo()

Get the elevation values (X-axis)
elevation_values = [float(x) for x in hist_data['bucketMeans']]

Get the frequency/count of elevation values (Y-axis)
elevation_counts = [int(x) for x in hist_data['histogram']]

Plot the histogram

plt.figure(figsize=(12, 6))
plt.bar(elevation_values, elevation_counts, width=10)
plt.tile('SRTM Elevation Histogram')
plt.xlabel('Elevation (meters)')
plt.ylabel('Frequency')
plt.grid(True)
plt.tight_layout()
plt.show()





Applications

Application	Description	Google Earth Engine Functions
Temporal Analysis	Analysing changes over time in satellite data to detect trends and patterns.	ee.Reducer.mean(), ee.Reducer.median(), ee.Reducer.stdDev(), ee.ImageCollection.filterDate()
Land Cover Mapping	Classifying land use, tracking deforestation, monitoring urbanisation, and assessing vegetation changes.	ee.Image.clip(), ee.Image.normalizedDifference(), ee.Image.select(), ee.Image.expression()
Climate Studies	Studying climate change, weather patterns, and climate variables like temperature and precipitation.	ee.ImageCollection.filter() ee.ImageCollection.mean(), ee.ImageCollection.reduce()
Disaster Management	Responding to natural disasters by analysing pre- and post-disaster imagery to identify affected areas.	ee.ImageCollection.filterBounds(), ee.ImageCollection.mosaic(), ee.Image.reduceRegion()
Machine Learning	Leveraging machine learning algorithms like Random Forest, SVM, and others for classification and analysis.	ee.Classifier.smileRandomForest(), ee.Classifier.train(), ee.Classifier.predict(), ee.FeatureCollection()
Topography Analysis	Deriving elevation, slope, aspect, and other topographic information from elevation datasets.	ee.Image.gradient(), ee.Image.hillshade(), ee.Terrain.slope(), ee.Terrain.aspect()
Climate and Weather Data	Accessing climate data, weather data, and climate indices for research and analysis.	ee.ImageCollection("NASA/GLDAS/V021/NOAH/ G025/T3H"), ee.ImageCollection("UCSB-CHG/CHIRPS/DAILY"), ee.ImageCollection("NASA/ORNL/DAYMET_V4")
Derived Maps	Generating thematic maps such as biomass, vegetation height, forest change, and fire risk.	Custom algorithms using ee.Image.expression(), ee.Image.updateMask(), and ee.Reducer functions.



Google Earth Engine

Google Colab

Ease of Use	Requires proficiency in Earth Engine's JavaScript- based code.	Familiar Python interface for users.
Learning Curve	Steeper learning curve for newcomers.	More accessible for Python developers.
Access to Data	Extensive satellite and geospatial datasets available.	Limited to publicly available datasets.
Customisation	Highly customizable, but requires coding skills.	Customization through Python libraries.
Scalability	Excellent for handling large geospatial datasets.	Limited by Colab's CPU/GPU resources.
Interactive Analysis	Offers an interactive map-based interface.	Limited interactivity, primarily code.
Storage	Data is hosted on Google's servers, storage is no ⁻ an issue.	t Limited storage, temporary session data.
Cost	Free tier available, but costs can increase with extensive use.	Limited free GPU hours, additional costs for extended use.
Offline Use	Internet connection required to access datasets.	Internet connection required for usage.



Let's have a practical showcase of a Colab notebook to explore optical satellite data and derive a time series of the normalised vegetation index NDVI

FOSS4G_UK.ipynb - Colaboratory (google.com)



Do you want to learn more?

- Explore the Geemap Github where you can find, videos, and tutorials that will helpful <u>Tutorials geemap</u>
- Explore more about Google Colab at <u>Welcome to</u> <u>Colaboratory - Colaboratory (google.com</u>)





Natural Environment Research Council



Any questions?

Thank you

Contact us:

Email: <u>ana.pachecopascagaza@manchester.ac.uk</u> Website: <u>https://ukfdrs.com/</u> Twitter: @ukfdrs GitHub: <u>https://github.com/UKFDRS</u>