# rgee: An R package for interacting with Google Earth Engine

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## Summary

Google Earth Engine (Gorelick et al., 2017) is a cloud computing platform designed for planetary-scale environmental data analysis. Its multi-petabyte data catalog and computation services are accessed via an Internet-accessible API. The API is exposed through JavaScript and Python client libraries. Google provides a browser-based IDE for the JavaScript API, and while convenient and useful for rapid data exploration and script development, it does not allow third-party package integration, relying solely on Google Maps and Google Charts for data visualization, and proprietary systems for metadata viewing and asset management. In contrast, the Python and Node.js distributions offer much flexibility for developers to integrate with third-party libraries. However, without the structure of a dedicated IDE, casual users can be left directionless and daunted. A significant gap exists between these two offerings (Google-supported JavaScript IDE and base client libraries) where convenience and flexibility meet. We propose to fill this gap with an R package that wraps the Earth Engine Python API to provide R users with a familiar interface, rapid development features, and flexibility to analyze data using open-source, third-party packages.

rgee is an Earth Engine (EE) client library for R that allows users to leverage the strengths of the R spatial ecosystem and Google Earth Engine in the same workflow. All of the Earth Engine Python API classes, modules, and functions are made available through the reticulate package (JJ Allaire & Eddelbuettel, 2017), which embeds a Python session within an R session, enabling seamless interoperability. Additionally, rgee adds several new features such as (i) new I/O design, (ii) multiple user support, (iii) interactive map display, (iv) easy extraction of time series, (iv) asset manage interface, and (v) metadata display. In addition, rgee also makes it possible to execute Earth Engine Python code from within R, making the translation of large Python projects unnecessary.

## Features

### Enhanced I/O

rgee implements several functions to support download/upload of spatial objects (Table 1 and Table 2). For instance, to download vector (image) files one can use ee_as_sf (ee_as_raster or ee_as_stars). In rgee, all the functions from server to local side have the option to fetch data using an intermediate container (Google Drive or Google Cloud Storage) or through a REST call ("$getInfo"). Although the latter option performs a quick download, there is a request limit of 262144 pixels for ee.Image and 5000 elements for ee.FeatureCollection which

makes it unsuitable for large objects. Other download functions, from server-side to others (see Table 1), are implemented to enable more customized download workflows. For example, using ee_image_to_drive and ee_drive_to_local users could create scripts which save results in the .TFRecord rather than the .GeoTIFF format. The upload process follows the same logic (Table 2). rgee includes raster_as_ee and stars_as_ee for uploading images and sf_as_ee for uploading vector data. Large uploads are only possible with an active Google Cloud Storage account.

**Table 1:** Download functions provided by the rgee package.

|         |                   | FROM          | TO            | RETURN            |
|---------|-------------------|---------------|---------------|-------------------|
| Image   | ee_image_to_drive | EE server     | Drive         | Unstarted task    |
|         | ee_image_to_gcs   | EE server     | Cloud Storage | Unstarted task    |
|         | ee_image_to_asset | EE server     | EE asset      | Unstarted task    |
|         | ee_as_raster      | EE server     | Local         | RasterStack object |
|         | ee_as_stars       | EE server     | Local         | Proxy-stars object |
| Table   | ee_table_to_drive | EE server     | Drive         | Unstarted task    |
|         | ee_table_to_gcs   | EE server     | Cloud Storage | Unstarted task    |
|         | ee_table_to_asset | EE server     | EE asset      | Unstarted task    |
|         | ee_as_sf          | EE server     | Local         | sf object         |
| Generic | ee_drive_to_local | Drive         | Local         | object filename   |
|         | ee_gcs_to_local   | Cloud Storage | Local         | GCS filename      |

**Table 2:** Upload functions provided by the rgee package.

|         |                 | FROM          | TO            | RETURN        |
|---------|-----------------|---------------|---------------|---------------|
| Image   | gcs_to_ee_image | Cloud Storage | EE asset      | EE Asset ID   |
|         | raster_as_ee    | Local         | EE asset      | EE Asset ID   |
|         | stars_as_ee     | Local         | EE asset      | EE Asset ID   |
| Table   | gcs_to_ee_table | Cloud Storage | EE asset      | EE Asset ID   |
|         | sf_as_ee        | Local         | EE asset      | EE Asset ID   |
| Generic | local_to_gcs    | Local         | Cloud Storage | GCS filename  |

The following example illustrates how to integrate the rgee I/O module and ggplot2 (Wickham, 2011) to download and visualize metadata for the BLM AIM TerrestrialAIM dataset.

```r
library(tidyverse)
library(rgee)
library(sf)

ee_Initialize()

# Define a Region of interest
roi <- ee$Geometry$Point(-120.06227, 40.64189)$buffer(25000)

# Load TerrADat TerrestrialAIM Dataset
blocks <- ee$FeatureCollection("BLM/AIM/v1/TerrADat/TerrestrialAIM")
subset <- blocks$filterBounds(roi)

# Move an Earth Engine FeatureCollection to their local env
sf_subset <- ee_as_sf(x = subset)
```

```r
# Create a boxplot with ggplot2
gapPct <- c("_25_50" = "GapPct_25_50","_51_100"="GapPct_51_100",
            "101_200" = "GapPct_101_200","200_>" = "GapPct_200_plus")

sf_subset[gapPct] %>%
  st_set_geometry(NULL) %>%
  as_tibble() %>%
  rename(!!gapPct) %>%
  pivot_longer(seq_along(gapPct), names_to = "Range") %>%
  ggplot(aes(x = Range, y = value, fill = Range)) +
  geom_boxplot() +
  xlab("") + ylab("% of the plot's soil surface") +
  theme_minimal()
```
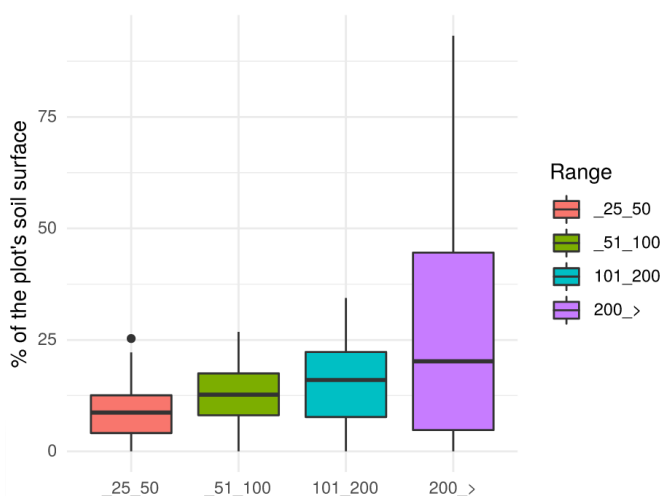


**Figure 1:** Gaps percentage between plant canopies of different sizes in a place near to Carson City, Nevada, USA.

## Authentication of multiple users

rgee makes it easy to authenticate and initialize the EE API through `ee_Initialize` a function which uses a file folder system to arrange multiple credentials (Google Earth Engine, Google Drive, and Google Cloud Storage) for multiple users. One of the main advantages of this is the possibility to distribute requests by accounts (i.e., parallelize on the client-side). For instance, if a research group wants to analyze deforestation, the script shown below will allow them to obtain results three times faster.

```r
library(foreach)
library(rgee)

ee_users()
ee_user_info()
google_account <- c("csaybar", "ryali93", "lbautista")

foreach(account = google_account, .combine = "c") %dopar% {
  ee_Initialize(gmail)
  ic_results <- deforestation_analysis(split_by = google_account)
  ee_imagecollection_to_local(ic_results)
} -> results
```

## Interactive Map Display

rgee offers interactive map display through Map\$addLayer, an R function mimicking the mapping module of the Earth Engine JavaScript Code Editor. Map\$addLayer takes advantage of the getMapId EE method to fetch and return an ID dictionary being used to create layers in a mapview (Appelhans, Detsch, Reudenbach, & Woellauer, 2016) object. Users can specify visualization parameters to Map\$addLayer by using the visParams argument, as demostrated below:

```r
library(rgee)
ee_Initialize()

# Load an ee.Image
image <- ee$Image("LANDSAT/LC08/C01/T1/LC08_044034_20140318")

# Centers the map view
Map$centerObject(image)

# Display the ee.Image
Map$addLayer(
  eeObject = image,
  visParams = list(bands = c("B4", "B3", "B2"), max = 10000),
  name = "SF"
)
```
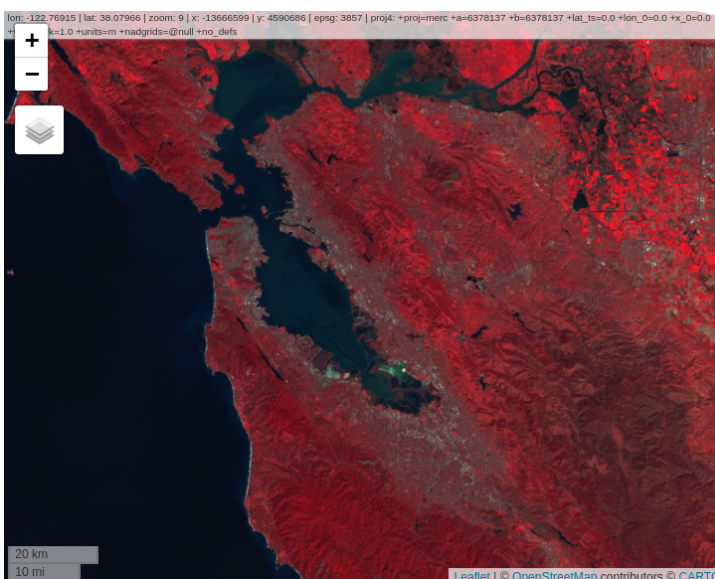


**Figure 2:** Landsat 8 false color composite of San Francisco bay area, California, USA.

## Extraction of time series

rgee can extract values from ee.Image and ee.ImageCollection objects at a certain location based on ee.Geometry, ee.Feature, ee.FeatureCollection and sf objects. If the geometry is a polygon, users can summarize the values using built-in Earth Engine reducer functions. The code below explains how to extract the average areal rainfall from North Carolina counties using the TerraClimate dataset.

```r
library(tidyverse)
library(rgee)
library(sf)

ee_Initialize()

# Image or ImageCollection (mean composite)
terraclimate <- ee$ImageCollection("IDAHO_EPSCOR/TERRACLIMATE")$
  filterDate("2001-01-01", "2002-01-01")$
  map(function(x) x$select("pr"))

# Define a geometry
nc <- st_read(system.file("shape/nc.shp", package = "sf"))

# Extract the average areal rainfall
ee_nc_rain <- ee_extract(terraclimate, nc, sf = FALSE)
colnames(ee_nc_rain) <- sprintf("%02d", 1:12)
ee_nc_rain$name <- nc$NAME

ee_nc_rain %>%
  pivot_longer(-name, names_to = "month", values_to = "pr") %>%
  ggplot(aes(x = month, y = pr, group = name, color = pr)) +
  geom_line(alpha = 0.4) +
  xlab("Month") +
  ylab("Precipitation (mm)") +
  theme_minimal()
```
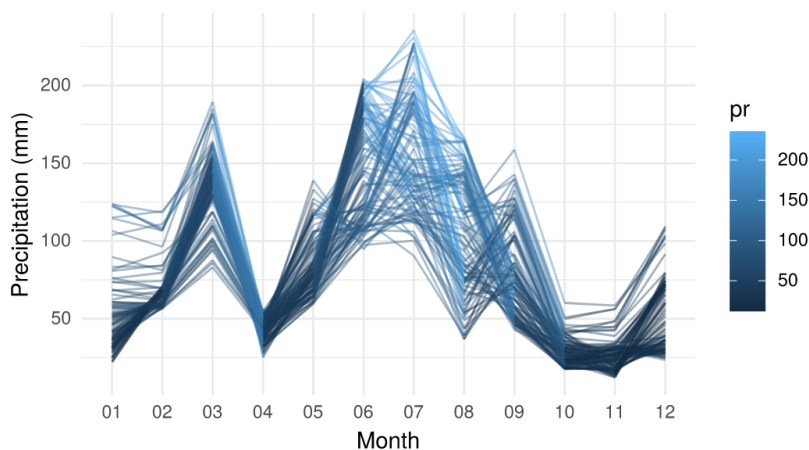


**Figure 3:** Average areal rainfall in counties of North Carolina for the year 2011 according to the TerraClimate dataset.

## Asset Management Interface

rgee implements an interface to batch actions on assets extending capabilities of the existing EE data module (ee.data.\*). The interface allows users to create and eliminate folders, move and copy assets, set and delete properties, handle access control lists, and manage or cancel tasks. For example, users can copy a Landsat 8 image to their personal EE assets as follows:

```
library(rgee)
ee_Initialize()

server_path <- "LANDSAT/LC08/C01/T1/"
user_asset_path <- ee_get_assethome()

ee_manage_copy(
  path_asset = paste0(server_path,"/LC08_044034_20140318"),
  final_path = paste0(user_asset_path,"/LC08_044034_20140318")
)
```

## Metadata display

The `ee_print` function can save and display all metadata related to EE spatial objects. With `ee_print`, users can retrieve information about the number of images or features, number of bands or geometries, number of pixels, geotransform, datatype, properties and approximate object size. `ee_print` can be used inside debugging pipelines (e.g. linking with `ee.Image.aside`).

```
library(rgee)

ee_Initialize()
l8 <- ee$Image("LANDSAT/LC08/C01/T1/LC08_044034_20140318")
ee_print(l8)
```

```
───────────────────────────── Earth Engine Image ──
Image Metadata:
  - Class                   : ee$Image
  - Number of Bands         : 12
  - Bands names             : B1 B2 B3 B4 B5 B6 B7 B8 B9 B10 B11 BQA
  - Number of Properties    : 117
  - Number of Pixels*       : 715030200
  - Approximate size*       : 1.07 GB
Band Metadata (img_band = B1):
  - EPSG (SRID)             : 32610
  - proj4string            : +proj=utm +zone=10 +datum=WGS84 +units=m +no_defs
  - Geotransform            : 30 0 460785 0 -30 4264215
  - Nominal scale (meters)  : 30
  - Dimensions              : 7650 7789
  - Number of Pixels        : 59585850
  - Data type               : INT
  - Approximate size        : 90.92 MB
```

**Figure 4:** Metadata for a Landsat 8 Image.

# Availability

rgee is an open-source software package made available under the Apache 2.0 license. It can be installed through CRAN using: install.packages("rgee"). rgee can also be installed from its GitHub repository using the remotes package: remotes::install_github("r-spatial/rgee"). A series of examples for using rgee are available at https://r-spatial.github.io/rgee.

# Acknowledgments

# References

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