

1 **rmacrostrat: An R package for accessing and retrieving** 2 **data from the Macrostrat geological database**

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18 **ABSTRACT**

19 The geological record is a vast archive of information that provides the only empirical data
20 about the evolution of the Earth. In recent years, concentrated efforts have been made to
21 compile macrostratigraphic data into the online centralized database Macrostrat
22 (<https://macrostrat.org>). Macrostrat is a global stratigraphic database containing information
23 regarding surface and subsurface rock units and their respective ages, lithologies, geographic
24 extents, and various other associated metadata. However, these data are currently only
25 accessible through the Macrostrat application programming interface (API), which is a barrier

26 to potential users that are less familiar with such services. This data accessibility hurdle
27 currently prevents full capitalization of the value offered by Macrostrat, particularly its
28 potential to improve understanding of the geological and biological evolution of the Earth.
29 Here, we introduce *rmacrostrat*, an R package which interfaces with the Macrostrat database,
30 to access and retrieve a variety of geological, paleontological, and economic data directly into
31 the R programming environment. In this article, we provide details about how the package can
32 be installed, its implementation, and potential use cases. For the latter, we showcase how
33 *rmacrostrat* can be used to visualize regional stratigraphic columns, produce regional geologic
34 outcrop maps, and estimate the proportion of North American carbonate and siliciclastic
35 sediments through time. We hope that this package will make geological data more readily
36 accessible, and in turn will facilitate new research utilizing Earth System data.

37 **INTRODUCTION**

38 Earth's geologic record provides a unique spatiotemporal archive of the evolutionary history
39 of the planet (Ernst and Youbi, 2017; Tetley et al., 2019; Cao et al., 2019; Scotese et al., 2021).
40 Historically, to understand macro-scale Earth System trends through geological time,
41 researchers were required to synthesize local or regional quantitative studies, predominantly
42 from data gathered in the form of regional geological maps, sections, and individual
43 sedimentary logs or boreholes (e.g., Ronov et al., 1980; Sestlavinskiy, 1991; Bosscher and
44 Schlager, 1993; Miall, 2022). However, the introduction of large online open-access databases,
45 in which a variety of complementary datasets are already digitized and synthesized, has
46 facilitated the development of macroscale analyses through both time and space. One such
47 database is Macrostrat (<http://macrostrat.org/>), a relational geospatial database that aims to
48 aggregate and synthesize field-derived geological data from geological maps and regional
49 geological columns into a dataset that describes the spatial distribution of geological units

50 within the Earth’s upper crust (Peters et al., 2018). Macrostrat contains information regarding
51 individual rock “units”, linked by unique identification numbers to associated lithological,
52 environmental, paleontological, and economic attributes, alongside information regarding their
53 respective chronostratigraphic context. These units are organized spatially into “columns”,
54 representing a cross-section of the upper crust within particular geological basins, and
55 temporally by Macrostrat’s internal chronostratigraphic age model (Figure 1). Sequentially
56 deposited units bounded by unconformities form geological “sections”, which also have their
57 own unique identification numbers. Additionally, Macrostrat units are linked by unique
58 identification numbers to geological mapping data amalgamated from a variety of sources, as
59 well as data from other large geoscience databases such as the Paleobiology Database (PBDB;
60 <http://paleobiodb.org/>) (Peters and McClennen, 2016; Uhen et al., 2023) and Mindat
61 (<http://mindat.org/>).

62
63 Since its initial compilation in 2005 from the American Association of Petroleum Geologists
64 Correlation of Stratigraphic Units of North America (COSUNA) charts (Peters, 2006),
65 Macrostrat has grown into a comprehensive and well-established database containing over
66 35,000 units and 1,500 geologic columns, all of which are publicly accessible. Macrostrat aims
67 to provide such data on a global scale, and while the abundance and resolution of available data
68 is currently geographically variable, improving spatial coverage is one of the major aims of the
69 project moving forwards (Quinn et al., 2024). Data hosted by Macrostrat have been used for a
70 wide variety of applications in scientific research, as well as science communication and
71 education. The broad temporal and spatial scale of data hosted by Macrostrat has facilitated a
72 diverse array of research related to Earth Systems through time, including in the fields of
73 sedimentology (Peters and Husson, 2017), stratigraphy (Tasistro-Hart and Macdonald, 2023),
74 igneous petrology (Peters et al., 2021), geochemistry (Husson and Coogan, 2023), and

75 paleobiology (Peters and Heim, 2010, 2011a, 2011b; Heim and Peters, 2011; Rook et al., 2013;
76 Nelsen et al., 2016; Peters et al., 2017; Balseiro and Powell, 2019, 2023; Ye and Peters, 2023).
77 Macrostrat has also collaborated with the extending Ocean Drilling Pursuits project (eODP;
78 <https://eodp.github.io/>) to integrate existing drill core data from sources such as the
79 International Ocean Drilling Program (IODP) into the database (Sessa et al., 2023). Geologic
80 map data held within Macrostrat is also displayed by a variety of software and mobile
81 applications that aim to enable usage of geologic materials by the wider scientific community,
82 the general public, and university education and teaching platforms (Cohen et al., 2018).
83 Macrostrat is also currently developing plans to expand and integrate community-led validation
84 of sections, ingestion of stratigraphic column data, and development of new software to
85 facilitate data collaboration (Quinn et al., 2024). As such, Macrostrat is a vital resource for
86 Earth scientists investigating a variety of issues related to both the geological history of our
87 planet and the impacts of geological processes today.

88

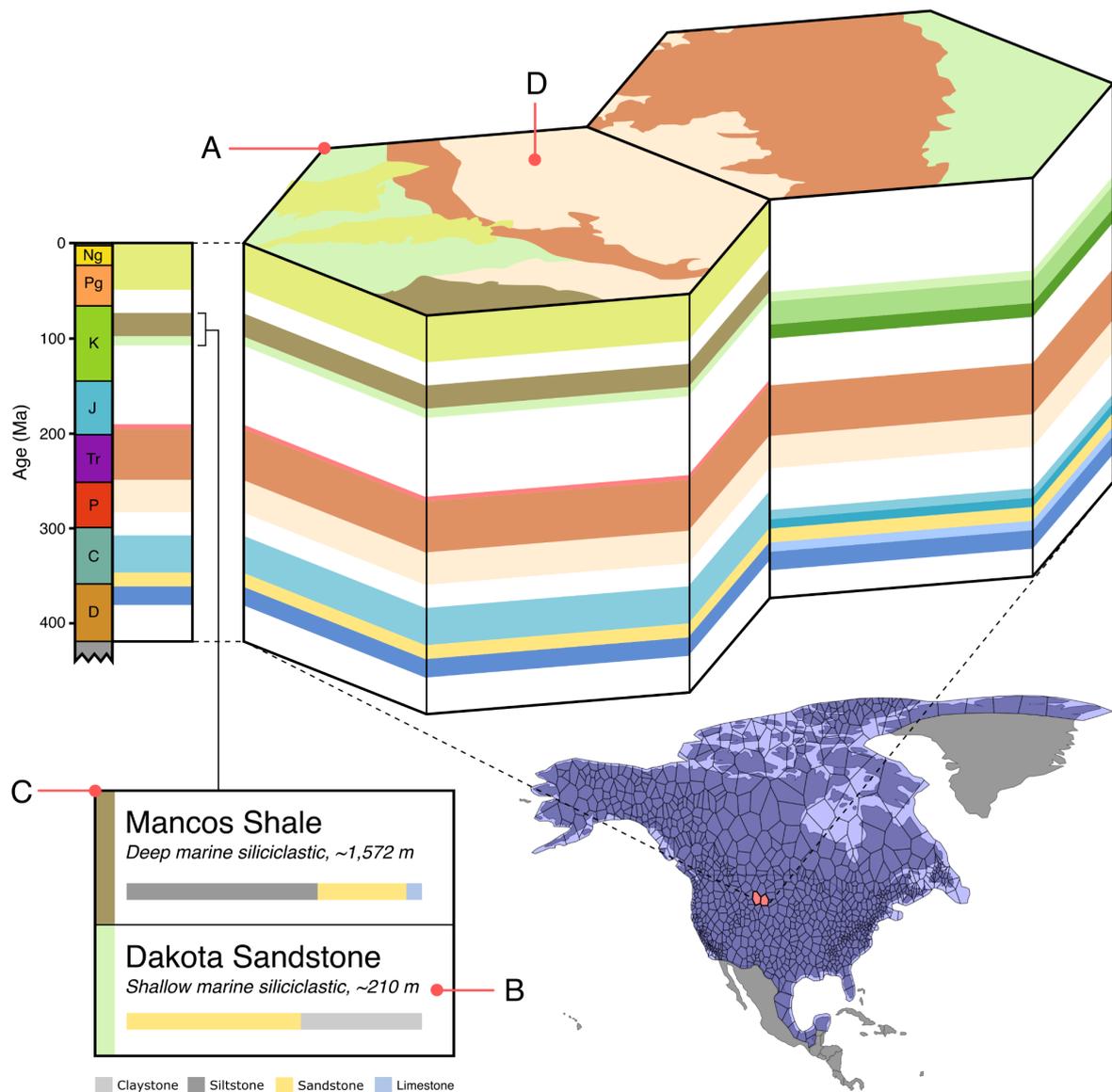
89 Despite the apparent opportunities offered by Macrostrat, its hosted data can currently only be
90 accessed via the database's Application Programming Interface (API). Although a powerful
91 resource, this single data access avenue means that familiarity with both the structure of the
92 database and how to interact with APIs is necessary in order to use the database. Those able to
93 overcome this data accessibility hurdle are still required to develop their own custom protocols
94 to integrate Macrostrat data into coding-based scientific workflows; this can inherently lead to
95 researchers 'reinventing the wheel', and producing code that is case-specific and difficult to
96 repurpose, inhibiting the reproducibility of research conducted using Macrostrat data. Such
97 processes are often carried out in the programming language R, which the Earth Science
98 community has broadly adopted to access, prepare, analyze, and plot data (e.g. Bell and Lloyd,
99 2015; Varela et al., 2015; Ortiz and Jaramillo, 2018; Barido-Sottani et al., 2019; Kocsis et al.,

100 2019; Jones et al., 2023; Gearty, 2024). In particular, several R packages have been developed
101 to interface with databases relevant to the geosciences through API services, supporting the
102 generation of readable, reusable, and reproducible workflows (Varela et al., 2015; Gearty and
103 Jones, 2023; Vidaña and Goring, 2023). However, until now, no such package has been
104 available for interacting with the Macrostrat database.

105

106 Here, we present *rmacrostrat*, a dedicated R package for interfacing with the geological
107 database Macrostrat. The package provides streamlined functionality for querying the database
108 via its API service and retrieving various geological data (e.g., lithostratigraphic units) and
109 definitions/metadata associated with the hosted data (e.g., lithological terms). First, we provide
110 instructions for installing the package and details on its implementation. We then demonstrate
111 the functionality available in *rmacrostrat* and provide typical usage examples. Finally, we
112 provide details about the resources we have made available to support *rmacrostrat* users. By
113 providing a programmatic solution to accessing the data hosted by Macrostrat, we endeavor to
114 facilitate new research across the Earth Sciences that is conducted in a streamlined, readable,
115 reusable, and reproducible manner.

116



117

118 **Figure 1:** Schematic showing the relationships between geological and stratigraphic data
 119 stored in Macrostrat. Data are arranged spatially as ‘columns’ (A), which contain
 120 chronostratigraphic columns of distinct stratigraphic ‘units’ (B), distinguished by a variety of
 121 attributes (e.g. lithology, environment etc.) and organized temporally by Macrostrat’s time–
 122 age model. Packages of continuously deposited units bounded by unconformities can be subset
 123 into ‘sections’ (C). Macrostrat units can also be linked to Macrostrat’s geological map outcrop
 124 (D), amalgamated from a variety of sources. Note that columns are idealized over the region,
 125 and do not intersect accurately with surficial geological maps.

126 **INSTALLATION**

127 The *rmacrostrat* package can be installed from CRAN (**PENDING ACCEPTANCE**) using
128 the `install.packages()` function in R (R Core Team, 2023):

```
install.packages("rmacrostrat")
```

129 If preferred, the development version of *rmacrostrat* can be installed from GitHub via the
130 `remotes` R package (Csárdi et al., 2023):

```
remotes::install_github("palaeoverse/rmacrostrat")
```

131 Following installation, *rmacrostrat* can be loaded via the `library()` function in R:

```
library("rmacrostrat")
```

132 **Dependencies**

133 The current version of *rmacrostrat* (ver. 0.0.1) depends on R (≥ 4.0) (R Core Team, 2023),
134 and imports functions from `curl` (Ooms, 2024), `geojsonsf` (Cooley, 2022), `httr` (Wickham,
135 2023), `jsonlite` (Ooms, 2014), and `sf` (Pebesma and Bivand, 2023). The package was developed
136 with the support of the R packages `devtools` (Wickham et al., 2024b), `testthat` (Wickham, 2011)
137 and `roxygen2` (Wickham et al., 2024a).

138 **IMPLEMENTATION**

139 Functions are broadly grouped into two categories in *rmacrostrat*: (1) `def_*` and (2) `get_*`. The
140 `def_*` suite of functions provides access to the definitions (or metadata) associated with data
141 stored in Macrostrat, such as lithologies (`def_lithologies()`), measurements
142 (`def_measurements()`), or Macrostrat columns (`def_columns()`). A summary of this suite of
143 functions is provided in Table 1. The `get_*` suite of functions are for retrieving data from
144 Macrostrat, such as Macrostrat columns (`get_columns()`), Macrostrat units (`get_units()`), and

145 geological map outcrop objects (`get_map_outcrop()`). Detailed descriptions of these functions
 146 are provided in Table 2.

147

148 **Table 1:** Summary table of the definition suite of functions (`def_*`) currently available in the
 149 *rmacrostrat* R package.

Function	Description
<code>catalog()</code>	Wrapper function to retrieve all definitions within a given definition set (e.g., lithologies)
<code>def_columns()</code>	Retrieve definitions for Macrostrat columns
<code>def_drilling_sites()</code>	Retrieve metadata for variables associated with extending Ocean Drilling Pursuits (eODP)
<code>def_econs()</code>	Retrieve definitions for economic resources (e.g., coal)
<code>def_environments()</code>	Retrieve definitions for environments (e.g., dune)
<code>def_grain_sizes()</code>	Retrieve definitions for grain sizes (e.g., cobble)
<code>def_intervals()</code>	Retrieve definitions for time intervals (e.g., Cenozoic)
<code>def_lithologies()</code>	Retrieve definitions for lithologies (e.g., sandstone)
<code>def_lithology_att()</code>	Retrieve definitions for lithology attributes (e.g., tabular)
<code>def_measurements()</code>	Retrieve definitions of different measurements (e.g., porosity)
<code>def_minerals()</code>	Retrieve definitions of different minerals (e.g., Agate)
<code>def_plates()</code>	Retrieve definitions of tectonic plates (e.g., Eurasia)
<code>def_projects()</code>	Retrieve definitions of Macrostrat projects (e.g., eODP)
<code>def_references()</code>	Retrieve definitions for published references
<code>def_sources()</code>	Retrieve definitions for geological maps (e.g., USGS)
<code>def_strat_names()</code>	Retrieve definitions for stratigraphic names (e.g., Hell Creek)
<code>def_strat_name_concepts()</code>	Retrieve definitions for stratigraphic name concepts (e.g., Dakota)
<code>def_structures()</code>	Retrieve definitions for geological structures (e.g., antiform)
<code>def_timescales()</code>	Retrieve definitions for timescales (e.g., international periods)

150

151 **Table 2:** Summary table of the data retrieval suite of functions (get_*) currently available in
 152 the *rmacrostrat* R package.

Function	Description
get_units()	Get data for Macrostrat units
get_sections()	Get data for Macrostrat sections
get_columns()	Get data for Macrostrat columns
get_age_model()	Get information about the age models for Macrostrat columns
get_map_outcrop()	Get spatial polygon data for geologic map outcrop
get_map_points()	Get spatial point data for geologic map measurements (e.g., strike, dip)
get_map_legends()	Get information from geologic map legends, associated with outcrop and points
get_fossils()	Get Paleobiology Database collections data associated with Macrostrat entities
get_eodp()	Get extending Ocean Drilling Pursuits data associated with various drilling programs
get_measurements()	Get measurements relevant to making geological inferences
get_paleogeography()	Get paleogeographic geometries based on the Wright et al. (Wright et al., 2013) Global Plate Model
get_stats()	Get statistics about Macrostrat projects

153 **Definition Functions**

154 Definitions (or metadata) of the various data stored in Macrostrat are retrieved from the
 155 Macrostrat API service via the def_* suite of functions (Table 1). The coverage of each of these
 156 functions should hopefully be immediately recognizable via their naming convention (e.g.,
 157 def_lithologies() returns definitions of the lithologies used in Macrostrat). Data returned using
 158 the def_* suite of functions contains both categorical (and often hierarchical) information about
 159 data attributes of interest (e.g., def_lithologies() will return individual lithologies
 160 ["sandstone"], as well as the type ["siliciclastic"] and class ["sedimentary"] of the lithology)
 161 as well as unique identification numbers for individual attributes that can be used to query

162 Macrostrat. Without user-specified arguments, all `def_*` functions will return a `data.frame`
163 object containing the entire dataset of definitions associated with that function:

```
# Get all lithologies
def_lithologies()
# Get all minerals
def_minerals()
# Get all time intervals
def_intervals()
```

164 Alternatively, users can search for definitions of specific entities or groups of entities using the
165 specific arguments for each `def_*` function. This can generally be achieved using specific
166 unique identification numbers (integers) for those definitions, or via a name (character strings):

```
# Get all marine environments by name
def_environments(environ_class = "marine")
# Get specific environment by ID
def_environments(environ_id = 2)
```

167 For convenience, we have also provided a wrapper around all `def_*` functions via the `catalog()`
168 function. This function returns complete sets of definitions for each `def_*` function, which takes
169 the suffix of an individual `def_*` function for its argument:

```
# Get all geological timescales
catalog(type = "timescales")
# Get all geological structures
catalog(type = "structures")
```

170 We strongly recommend using the `def_*` suite of functions prior to retrieving data from
171 Macrostrat to better understand both the structure of the database and the utility offered by the
172 functions available in *rmacrostrat*. Due to the wide variety of data available in Macrostrat,
173 individual `get_*` functions include a large array of potential arguments which can differ
174 substantially between functions (see below). By using the specific `def_*` functions related to
175 potentially useful search criteria, users can efficiently identify arguments and parameters with
176 which to query the database via the `get_*` suite of functions. Examples of the utility of the `def_*`

177 functions are provided in the application section below, as well as in the available vignettes for
178 the package.

179 **Data Retrieval Functions**

180 Data can be retrieved from the Macrostrat database API directly into the R environment using
181 the `get_*` suite of functions (Table 2). These functions either return data related to specified
182 Macrostrat entities (e.g., Macrostrat columns, units, sections, and age definitions), geologic
183 map elements, or external data related to Macrostrat entities (e.g., PBDB collections, eODP
184 data, paleogeographies), and can be returned either as a standard `data.frame` or as a spatial
185 simple features (i.e., `sf`) object. In accordance with the `def_*` suite of functions, the purpose of
186 individual `get_*` functions are intended to be easily identifiable from their named suffix (e.g.,
187 `get_columns()` retrieves data for Macrostrat columns).

188

189 As opposed to the `def_*` suite of functions, the `get_*` suite of functions require at least one
190 supplied argument for a valid database query. Although the array of possible arguments differs
191 substantially between `get_*` functions, users can generally retrieve data based on several
192 categories. Firstly, users can search by unique identification number, for either the chosen data
193 type to retrieve, or based on another Macrostrat entity.

```
# Get specific column according to an ID  
get_columns(column_id = 45)  
# Get units and sections associated with a specific column ID  
get_units(column_id = 45)  
get_sections(column_id = 45)  
# Get map outcrop related to specific unit ID  
get_map_outcrop(unit_id = 1610)
```

194 Attribute information—such as lithostratigraphic name, lithology, environment or economic
195 source—can also be used independently, or in combination in some instances, to retrieve
196 subsets of Macrostrat data. These attributes can be specified either using their unique

197 identification number or by character string. Further information about each attribute to search
198 by can be found in the respective def_* functions (e.g., lithology attribute information can be
199 found in the def_lithologies() function).

```
# Get units inferred to be marine
get_units(environ_class = "marine")
# Get all sandstone units by name or ID
get_units(lithology = "sandstone")
get_units(lithology_id = 10)
```

200 Data can also be retrieved using temporal limits, either by specifying a specific interval name
201 as a character string (e.g., "Permian"), a unique identification number, a numeric value (e.g.,
202 275 Ma), or from providing constraints based on numerical limits (e.g., 251.9–298.9 Ma). All
203 Macrostrat entities which overlap with the specified parameter(s) in terms of their
204 chronostratigraphic range defined in the Macrostrat age model are returned.

```
# Get units by interval name
get_units(interval_name = "Aptian")
# Get units by interval ID
get_units(interval_id = 43)
# Get units by age
get_units(age = 200)
# Get units by age range
get_units(age_bottom = 250, age_top = 200)
```

205 Finally, some get_* functions allow the user to query the database using geographic or spatial
206 information. This can either be achieved by specifying coordinates in decimal
207 latitude/longitude degrees, or if continental scale resolution is desired, through the use of
208 Macrostrat projects. Macrostrat data is split into regional projects, such as North America
209 (project_id = 1) and Australia (project_id = 6); setting this argument will return all Macrostrat
210 entities associated with that regional project.

```
# Get sections which appear at a specific longitude & latitude
get_sections(lng = -105.15, lat = 37.89)
# Get map outcrop which appears at a specific longitude & latitude
```

```
get_map_outcrop(lng = -105.15, lat = 37.89)
# Get all Macrostrat unit data for the North American continent
get_units(project_id = 1)
```

211 As aforementioned, it is recommended that these arguments are used in tandem with the `def_*`
212 suite of functions to maximize search potential and data retrieval. For instance, a user interested
213 in retrieving units deposited in a specific paleoenvironment may want to use the
214 `def_environments()` function prior to their search to see the full variety of parameters with
215 which to search by.

216 **APPLICATION**

217 Herein, we provide three example applications of the *rmacrostrat* package. These examples
218 are greatly expanded in step-by-step vignettes provided alongside the package, available online
219 via the associated package website (<https://rmacrostrat.palaeoverse.org/articles/>) and are also
220 bundled with the package, accessible via:

```
browseVignettes(package = "rmacrostrat")
```

221 **Constructing Stratigraphic Columns**

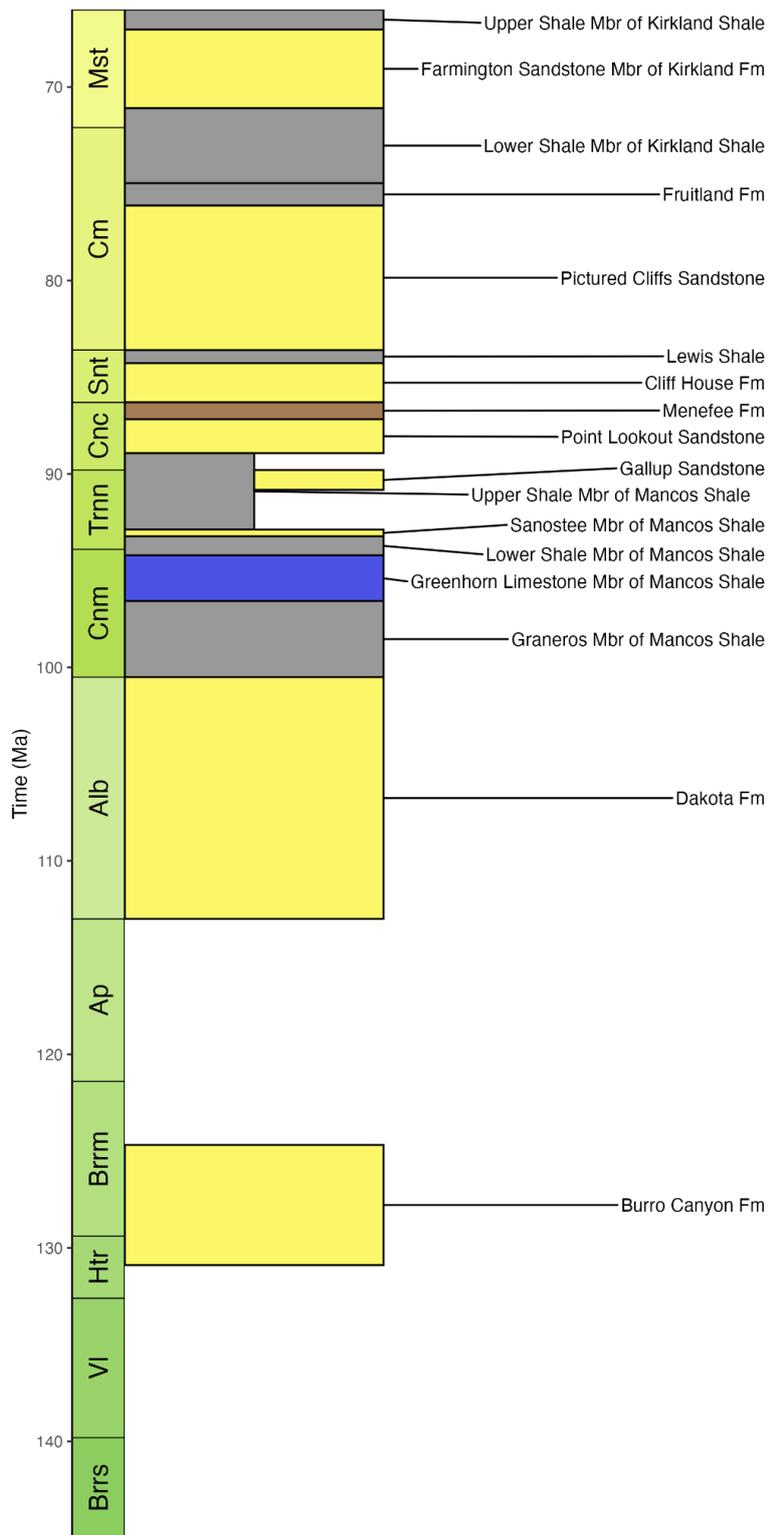
222 An understanding of stratigraphy—that is, the relationships between adjacent geological
223 units—is fundamental to accurately reading the geological record, enabling researchers to put
224 relative ages to lithological units and making temporal and spatial correlations. Using
225 *rmacrostrat*, the geological data within the Macrostrat database can be easily retrieved and
226 used to generate a stratigraphic column for a specific location and/or time interval. Below we
227 provide an example showing how to retrieve and plot a stratigraphic column for the San Juan
228 Basin, an asymmetric structural basin in northwestern New Mexico and southwestern Colorado
229 (Four Corners region of Southwestern United States), containing sedimentary rocks ranging
230 from Cambrian to Holocene in age (Fassett and Hinds, 1971). For this example, we restrict our

231 column data to the Cretaceous, but this approach could equally be applied to any other basin
232 or temporal interval. As the most broad-scale geological entity available within Macrostrat is
233 a column, `def_columns()` is first used to identify the column associated with the San Juan Basin.
234 The unique column identification number can then be used to get data for all appropriate units
235 via `get_units()`. As the example focuses only on the Cretaceous, additional arguments available
236 in `get_units()` can be used to further filter the queried data. With the returned data—Cretaceous
237 lithostratigraphic units within the San Juan Basin—a stratigraphic column can be generated.

```

# Load packages
library(rmacrostrat)
library(ggplot2)
library(ggrepel)
library(deeptime)
# Get the column definition of the San Juan Basin
column_def <- def_columns(column_name = "San Juan Basin")
# Using the column ID, retrieve all units of Cretaceous age
san_juan_units <- get_units(column_id = column_def$col_id,
                           interval_name = "Cretaceous")
# Specify x_min and x_max in dataframe
san_juan_units$x_min <- 0
san_juan_units$x_max <- 1
# Tweak values for overlapping units
san_juan_units$x_max[10] <- 0.5
san_juan_units$x_min[11] <- 0.5
# Add midpoint age for plotting
san_juan_units$m_age <- (san_juan_units$b_age +
                        san_juan_units$t_age) / 2
# Plot stratigraphic column
ggplot(san_juan_units, aes(ymin = b_age, ymax = t_age,
                          xmin = x_min, xmax = x_max)) +
  # Plot units, colored by rock type
  geom_rect(fill = san_juan_units$color, color = "black") +
  # Add text labels
  geom_text_repel(aes(x = x_max, y = m_age, label = unit_name),
                 size = 3.5, nudge_x = 1.5) +
  # Reverse direction of y-axis
  scale_y_reverse(limits = c(145, 66), n.breaks = 10,
                 name = "Time (Ma)") +
  # Theming
  theme_classic() +
  theme(legend.position = "none",
        axis.line.x = element_blank(),
        axis.title.x = element_blank(),
        axis.text.x = element_blank(),
        axis.ticks.x = element_blank()) +
  # Add geological time scale
  coord_geo(pos = "left", dat = list("stages"), rot = 90)

```



239

240 **Figure 2:** A stratigraphic column of Cretaceous lithostratigraphic units from the San Juan

241 Basin. Lithostratigraphic units were fetched from the Macrostrat database

242 (<https://macrostrat.org/>) using the *rmacrostrat* package ver. 0.0.1.

243 **Plotting Geologic Outcrop Maps**

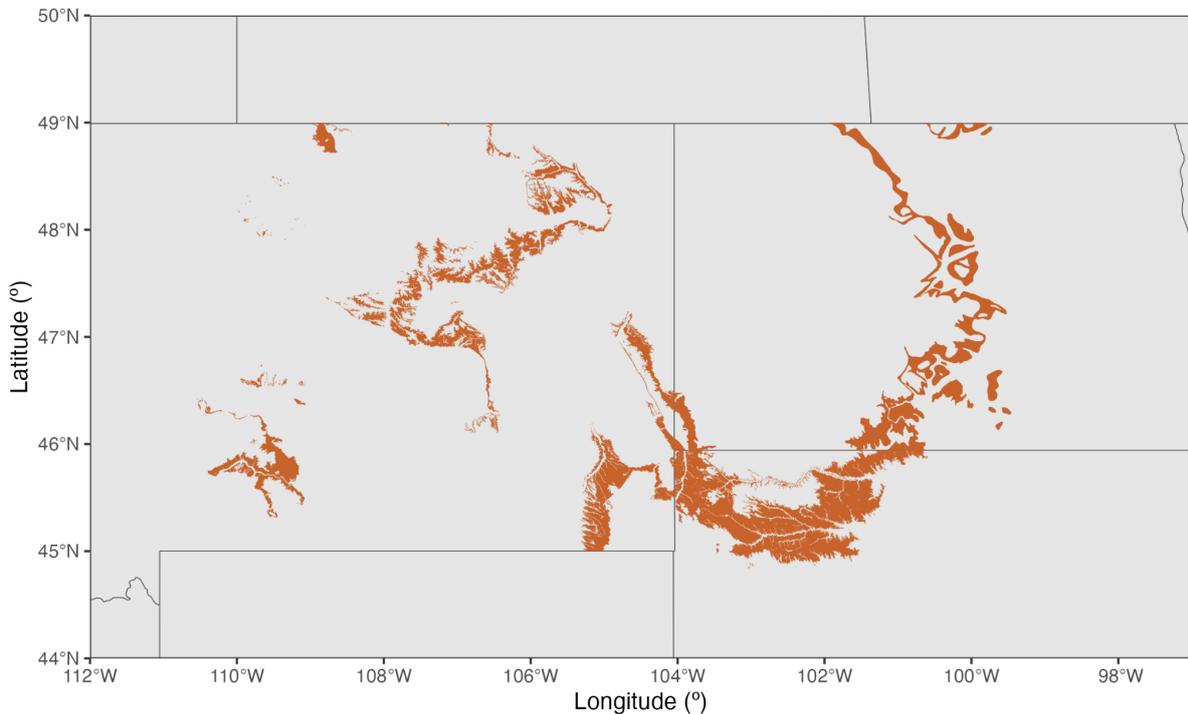
244 A commonly required figure across a range of disciplines within the geosciences is a
245 geographic map of the outcrop for a specific geologic formation. These figures can be easily
246 generated using the `get_map_outcrop()` function of *rmacrostrat*, which retrieves geospatial data
247 associated with lithostratigraphic units. Below we provide an example for constructing a map
248 of outcrop for the Hell Creek Formation, a geologic formation from the latest Cretaceous and
249 early Paleogene of North America, which is found cropping out across Montana, and North
250 and South Dakota, in the United States (Johnson et al., 2002; Fastovsky and Bercovici, 2016).
251 As outcrop spatial data is compiled from various map sources, the definition function
252 `def_strat_names()` is first used to find the appropriate identification numbers for any
253 stratigraphic names of formations that include the Hell Creek. This information can then be
254 used with the `get_map_outcrop()` function to retrieve geospatial data for the formation as a
255 simple features (sf) object. These data can be plotted to produce a geological map (Figure 3).

```
# Load libraries
library(rmacrostrat)
library(rnaturalearth)
library(ggplot2)
library(sf)
# Get data for chosen formation, specifying by stratigraphic rank
hc_def <- def_strat_names(strat_name = "hell creek",
                        rank = "Fm")
# Get spatial outcrop data for the chosen formation based on ID
hc <- get_map_outcrop(strat_name_id = hc_def$strat_name_id,
                    sf = TRUE)
# Load background maps
n_a <- ne_states(country = "united states of america",
                returnclass = "sf")
ca <- ne_states(country = "canada",
                returnclass = "sf")
# Plot the map
ggplot() +
  geom_sf(data = n_a) +
  geom_sf(data = ca) +
```

```

geom_sf(data = hc,
        fill = '#C7622B',
        lwd = 0) +
coord_sf(xlim = c(-112, -97), ylim = c(44, 50),
        expand = FALSE) +
labs(x = "Longitude (°)", y = "Latitude (°)") +
theme_bw()

```



256

257 **Figure 3:** Outcrop of the Hell Creek Formation across Montana, and North and South Dakota.

258 Outcrop data were fetched from the Macrostrat database (<https://macrostrat.org/>) using the

259 *rmacrostrat* R Package ver. 0.0.1.

260 Macrostratigraphic Time Series Analyses

261 Initial publications using data from the Macrostrat database quantified how the number and

262 proportion of Macrostrat entities, as well as different lithostratigraphic unit types associated

263 with different paleoenvironments (e.g., marine, marginal, mixed, terrestrial), varied throughout

264 the Phanerozoic (Peters and Heim, 2010). *rmacrostrat* facilitates access to these types of data,

265 and allows for similar analyses to be conducted. Below we provide an example of such an

266 analysis, in this case estimating the number of siliciclastic versus carbonate lithostratigraphic
267 units in North America throughout the Phanerozoic.

268

269 For this example, the relevant lithostratigraphic unit data from Macrostrat is first fetched using
270 the `get_units()` function from *rmacrostrat*. For this query, several filters are applied to retrieve
271 the appropriate data. First, the `environ_type` argument is used to filter for carbonate and
272 siliciclastic sediments. Second, the `interval_name` argument is used to filter to units only from
273 the Phanerozoic. Finally, the `project_id` argument is used to filter results to units from the North
274 American geological record:

```
# Load library
library(rmacrostrat)
# Get units by environment type, interval, and project ID
carbonate <- get_units(environ_type = "carbonate",
                      interval_name = "Phanerozoic",
                      project_id = 1)
siliciclastic <- get_units(environ_type = "siliciclastic",
                          interval_name = "Phanerozoic",
                          project_id = 1)
# Add column of sediment type
carbonate$sediment_type <- "Carbonate"
siliciclastic$sediment_type <- "Siliciclastic"
# Bind data
units <- rbind.data.frame(carbonate, siliciclastic)
```

275 With this data, the number of siliciclastic and carbonate lithostratigraphic units for each
276 international geological stage (i.e., time bin) through time can be calculated. Functionality
277 available in the *palaeoverse* R package can be used to retrieve relevant information about
278 geological stages (Jones et al., 2023):

```
# Load libraries
library(palaeoverse)
# Generate stage-level time bins
bins <- time_bins(scale = "international ages")
# Rename age columns in units to be consistent with our bins
```

```

colnames(units)[which(colnames(units) == "b_age")] <- "max_ma"
colnames(units)[which(colnames(units) == "t_age")] <- "min_ma"
# Bin data
units <- bin_time(occdf = units, bins = bins,
                 min_ma = "min_ma", max_ma = "max_ma",
                 method = "all")
# Calculate the number of environment classes per bin assignment
counts <- group_apply(occdf = units,
                     group = c("sediment_type",
                               "bin_assignment"),
                     fun = nrow)
# Rename columns to ease reading and merging
colnames(counts) <- c("count", "sediment_type", "bin")
# Merge datasets by bin number
counts <- merge(x = bins, y = counts, by = "bin")

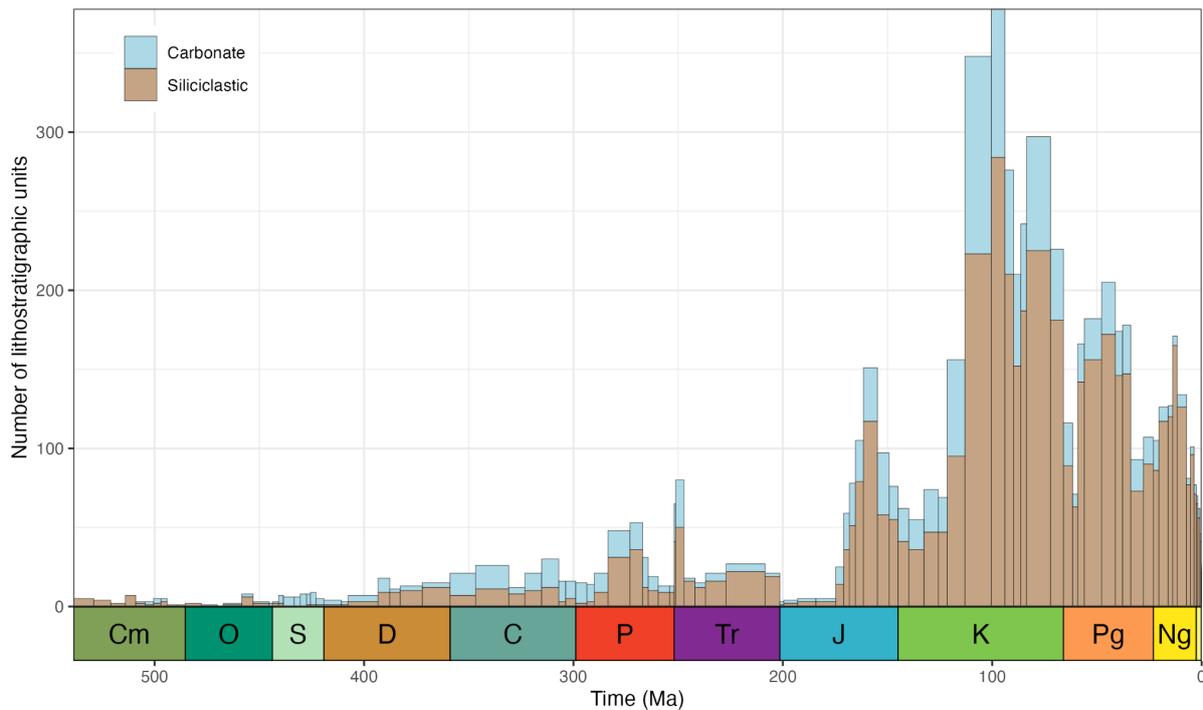
```

279 Using additional R packages for visualization, such as ggplot2 (Villanueva and Chen, 2019)
 280 and deeptime (Gearty, 2024), Phanerozoic stage-level counts of North American
 281 lithostratigraphic units can be plotted by sediment type (Figure 4):

```

# Load libraries
library(ggplot2)
library(deeptime)
# Generate a plot of lithostratigraphic units through time
ggplot(counts, aes(fill = sediment_type, y = count, x = mid_ma)) +
  # Stacked bar chart with width specified by interval duration
  geom_bar(position = "stack", stat = "identity",
           width = counts$duration_myr,
           color = "black", linewidth = 0.1) +
  # Label y-axis
  scale_y_continuous("Number of lithostratigraphic units") +
  # Label x-axis and reverse direction
  scale_x_reverse("Time (Ma)") +
  # Data plotting colors
  scale_fill_manual(values = c("#add8e6", "#C4A484")) +
  # Theming
  theme_bw() +
  theme(legend.title = element_blank(),
        legend.position = c(0.1, 0.9)) +
  # Add geological time scale
  coord_geo()

```



283

284 **Figure 4:** Phanerozoic lithostratigraphic units. The number of Macrostrat lithostratigraphic
 285 units throughout the Phanerozoic (541–0 Ma) binned by stratigraphic stage-level bins and
 286 grouped by sediment type (carbonate and siliciclastic). Units are counted for all time bins they
 287 overlap with. Lithostratigraphic units were fetched from the Macrostrat database
 288 (<https://macrostrat.org/>) using the *rmacrostrat* package ver. 0.0.1.

289 RESOURCES

290 We have made several resources available for our users. First, we have built a package website
 291 (<http://rmacrostrat.palaeoverse.org>) that provides information on how to use and contribute to
 292 *rmacrostrat*, how to report issues and bugs, and a contributor code of conduct. We have also
 293 made available three vignettes/tutorials for the package, which provide user-friendly usage
 294 guides (<https://rmacrostrat.palaeoverse.org/articles>). Through *rmacrostrat*, we hope to further
 295 foster collaboration and the sharing of resources within the Earth Science community. With
 296 this goal in mind, we warmly welcome the community to join and follow our community

297 spaces, such as our GitHub organization page (<https://github.com/palaeoverse>) and Google
298 Group (<https://groups.google.com/g/palaeoverse>), where users can share ideas and resources,
299 advertise opportunities, and network with colleagues.

300 **FUTURE PERSPECTIVES**

301 The development of *rmacrostrat* expands upon the current suite of software toolkits available
302 within the Palaeoverse (<https://palaeoverse.org/>) ‘universe’ (Jones, 2022; Gearty and Jones,
303 2023; Jones et al., 2023). Through *rmacrostrat*, we hope to improve accessibility to the vast
304 geological data available within the Macrostrat database (<https://macrostrat.org/>) and facilitate
305 new research across the Earth Sciences. The *rmacrostrat* R package offers researchers the
306 opportunity to streamline their research by providing a bridge between Macrostrat and the R
307 environment, as well as supporting the capacity to generate fully-reproducible pipelines. We
308 hope that these benefits will encourage the community to further capitalize on the value offered
309 by Macrostrat, and may ultimately lead to higher data quality through peer review. As we have
310 demonstrated with our example applications, *rmacrostrat* can be used to support the efficient
311 plotting of stratigraphic columns, mapping of geological outcrop, and quantification of
312 temporal dynamics in available lithostratigraphic units. However, we envision that *rmacrostrat*
313 can also be used to support a wide range of additional analyses across the Earth Sciences, such
314 as economic resource exploration, comparisons between deep-time diversity dynamics and
315 environmental change, and hazard mapping.

316 **AUTHOR CONTRIBUTIONS**

317 Lewis A. Jones conceived the project. All authors contributed to developing the project. Lewis
318 A. Jones and William Gearty developed the core functionality of the *rmacrostrat* package. All
319 authors contributed to developing the full functionality and documentation available, as well

320 as testing and reviewing the code. Lewis A. Jones and Christopher D. Dean drafted the
321 manuscript. Lewis A. Jones, Christopher D. Dean, and Bethany J. Allen produced the figures.
322 All authors contributed to the final editing and checking of the manuscript.

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