

- 1 rmap: An R package to plot and compare tabular data
- 2 on customizable maps across scenarios and time
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Software

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Summary

rmap is an R package that allows users to easily plot tabular data (CSV or R data frames) on maps without any Geographic Information Systems (GIS) knowledge. Maps produced by rmap are ggplot objects and thus capitalize on the flexibility and advancements of the ggplot2 package (Wickham, 2011) and all elements of each map are thus fully customizable. Additionally rmap automatically detects and produces comparison maps if the data has multiple scenarios or time periods as well as animations for time series data. Advanced users can load their own shapefiles if desired. rmap comes with a range of pre-built color palettes but users can also provide any R color palette or create their own as needed. Four different legend types are available to highlight different kinds of data distributions. The input spatial data can be both gridded or polygon data. rmap is designed in particular for comparing spatial data across scenarios and time periods and comes preloaded with standard country, state, and basin maps as well as custom maps compatible with the Global Change Analysis Model (GCAM) spatial boundaries (Calvin et al., 2019). rmap has a growing number of users and its products have been used in multiple multisector dynamics publications (Khan et al., 2021; Wild, Khan, Clarke, et al., 2021; Wild, Khan, Zhao, et al., 2021) as well as a required dependency in other R packages such as rfasst (Sampedro, 2021) and metis (Khan et al., 2020). rmap's automatic processing of tabular data using pre-built map selection, difference map calculations, faceting, and animations offers unique functionality which makes it a powerful and yet simple tool for users looking to explore multi-sector, multi-scenario data across space and time.

Statement of need

rmap is meant to help users having limited to no GIS knowledge use R for spatial visualization of tabular spatial data. rmap is not meant to be a replacement for spatial manipulation or cartographic software but focuses on the simple plotting of polygon and gridded data for spatio-temporal visualization of tabular data with a focus on comparing across scenarios and time periods. Several existing R packages such as tmap (Tennekes, 2018), cartography (Giraud & Lambert, 2016), rworldmap (South, 2011), GISTools (Brunsdon et al., 2015), choroplethr (Lamstein & Johnson, 2020), sp (E. Pebesma & Bivand, 2005) and sf (E. J. Pebesma, 2018), have been developed to conduct spatial visualization and analytics in R without depending on external software such as ArcGIS (ESRI, 2020), GRASS (GRASS Development Team, 2020) or QGIS (QGIS Development Team, 2021). rmap enhances the following key capabilities which are limited in these existing packages:



- 1. **Pre-built maps**: Existing packages come with only a few examples of built-in maps as package data. rmap comes with a growing collection of country, state, river basin, as well as other customized maps that are added into the package data based on user needs and requests. While built-in maps increase the size of the package, having direct access to these allows for automated searching and quick deployment of relevant shapefiles without the need to download any data. A list of pre-built maps in rmap can be found in the Built-in Maps section of the user guide.
- 2. Direct data table to map: Existing packages are not able to plot a map directly given only a simple data frame or a CSV file as an input. rmap has an automatic map_find function that searches for the appropriate built-in map based on the regions provided in a subRegion column and values in a value column. The sub-regions in the subRegion column must be one of the sub-regions in the existing set of rmap built-in maps. This truly frees users from the need for any other spatial data needs and they can simply map() their own data tables directly.
- 3. Difference maps: Existing packages do not produce difference maps to compare across scenarios or time periods. rmap provides this functionality by automatically recognizing multiple scenarios and time periods to produce difference maps across these dimensions. An important aspect of spatial data is exploring the difference between two scenarios or time periods and rmap makes this a seamless process.
- 4. **Post-process customization**: Existing packages do not produce output objects that can be saved and then customized. Customization of the maps in existing packages is limited to package specific functionality and arguments. rmap produces ggplot objects in which every element (axis, grids, titles, colors, line widths, facets) can all be customized after the map has been produced. This allows users to capitalize on existing knowledge of the widely used ggplot2 package and its arguments.

55 Functionality

- A detailed User Guide walks users step-by-step through all the available functionality of rmap.
- 67 A simpler Cheatsheet is also provided to help users remember some of the key functionality in
- a single sheet. The following few simple examples demonstrate the simplicity of using rmap.

69 Basic map



Figure 1: Basic map plotted on data provided. rmap automatically chooses a country map based on the names in the 'subRegion' column in the data table.

Map with labels and underlayer

```
library(rmap)

data = data.frame(subRegion = c("CA","FL","ID","MO","TX","WY"),

value = c(5,10,15,34,2,7))

map(data,

underLayer = mapUS52Compact,

crop_to_underLayer = T,

labels = T)
```

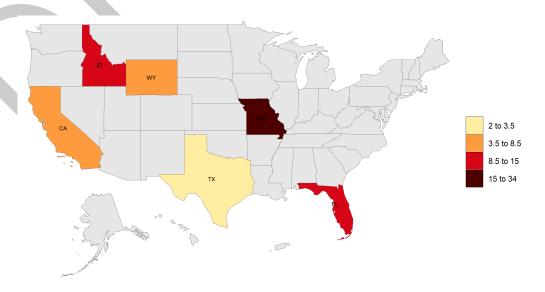


Figure 2: Map with chosen underLayer and labels for data provided. rmap automatically chooses the U.S. state map based on the names in the 'subRegion' column in the data table.



Compare Scenarios

```
library(rmap)
   data = data.frame(subRegion = c("Spain", "Germany", "Austria", "Greece", "Italy",
87
                                     "Spain", "Germany", "Austria", "Greece", "Italy",
                                     "Spain", "Germany", "Austria", "Greece", "Italy"),
                      value = c(5,10,15,34,2,
                               15,50,34,50,20,
91
                               1,2,7,13,5),
92
                      scenario = c("scen1","scen1","scen1","scen1","scen1",
                                    "scen2", "scen2", "scen2", "scen2", "scen2",
                                    "scen3", "scen3", "scen3", "scen3"))
   map(data,
       underLayer = "mapCountries",
       scenRef = "scen1",
       background = T)
```



Figure 3: Maps showing absolute values for all three scenarios in data table.

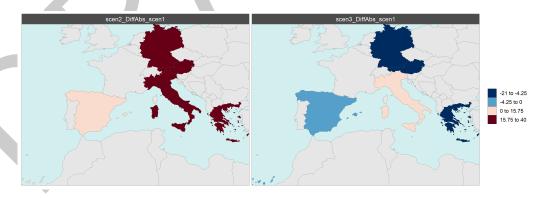


Figure 4: Maps showing absolute difference between chosen reference scenario ("scen1") and remaining scenarios ("scen2" and "scen3") in data table.

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