

# shinyr: A Framework for Building, Managing, and Stitching Shiny Modules into Reproducible Workflows

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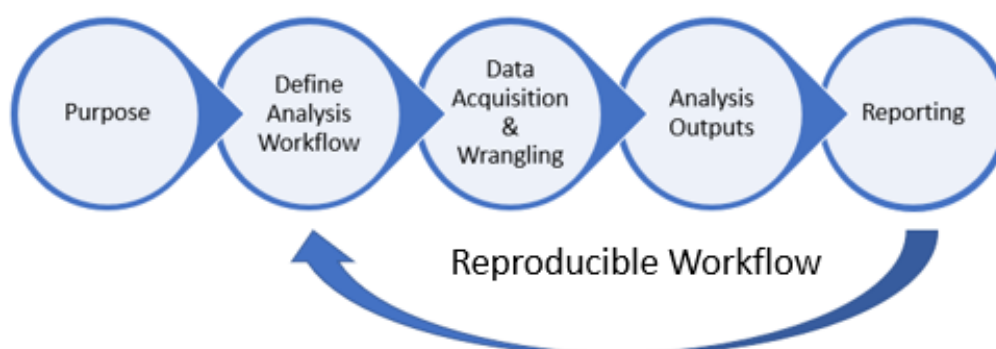
**Abstract** The R package shinyr provides a unifying framework that allows Shiny developers to create, manage, and deploy a master Shiny application comprised of one or more “apps”, where an “app” is a tab-based workflow that guides end-users through a step-by-step analysis. Each tab in a given “app” consists of one or more Shiny modules. The shinyr app builder allows developers to “stitch” Shiny modules together so that outputs from one module serve as inputs to the next, creating an analysis pipeline that is easy to implement and maintain. Apps developed using shinyr can be incorporated into R packages or deployed on a server, where they are accessible to end-users. Users of shinyr apps can save analyses as an RDS file that fully reproduces the analytic steps and can be ingested into an RMarkdown or Quarto report for rapid reporting. In short, developers use the shinyr framework to write Shiny modules and seamlessly combine them into Shiny apps, and end-users of these apps can execute reproducible analyses that can be incorporated into reports for rapid dissemination. A comprehensive overview of the package is provided by 12 learnr tutorials.

## 1 Introduction

The [shiny](#) R package allows users to build interactive web apps straight from R, without advanced knowledge of HTML or JavaScript ([Chang et al., 2022](#)). A *shiny* web app can permit an expedient analysis pipeline or workflow. Ideally, the pipeline can produce outputs that are fully reproducible ([Peng, 2011](#); [Gentleman and Lang, 2007](#); [Alston and Rick, 2021](#)). Moreover, the pipeline can permit rapid reporting to convey the results of an analysis workflow to a target audience ([Stoudt et al., 2021](#)) (Figure 1).

*shiny* applications range from simple to complex, each with an intended purpose developed for an intended user audience. Several R packages provide a development framework for building multi-faceted master applications, including [shinysum](#) for prototyping ([Fay and Rochette, 2020](#)), [golem](#) ([Fay et al., 2021](#)), and [rhino](#) ([Żyła et al., 2023](#)).

From the developer’s perspective, complex *shiny* applications can result in many lines of code, creating challenges for collaborating, debugging, streamlining, and maintaining the overall product. *shiny* modules are a solution to this problem. As stated by Winston Chang ([shi, 2020](#)), “A *shiny* module is a piece of a *shiny* app. It can’t be directly run, as a *shiny* app can. Instead, it is included as part of a larger app . . . Once created, a *shiny* module can be easily reused – whether across different apps, or multiple times in a single app.” *shiny* modules, and modularization in general, are a core element of agile software development practices ([Larman, 2004](#)). Several authors have contributed R packages for distributing pre-written *shiny* modules for general use, including the [datamods](#) ([Perrier et al., 2022](#)), [shinyreglog](#) ([Kosinski, 2022](#)), [periscope](#) ([Brett and Neuhaus, 2022](#)), [shinyauthr](#) ([Campbell, 2021](#)), and



**Figure 1:** Stages of a reproducible workflow, a process that moves an inquiry from raw data to insightful contribution.

`jsmodule` (Kim and Lee, 2022) packages.

However, as the number of available modules increases, there is a pressing need for documenting available *shiny* modules and easily incorporating them into new workflows. For example, consider a toy modular-based app that guides a user through an analysis of the famous “Iris Dataset,” which contains 150 records of 3 species of iris, including measurements of the length and width of the flowers’ sepals and petals (Fisher, 1936). The app, called “Iris Explorer,” consists of 5 tabs to be worked through in sequence (Figure 2, top).

Tab 1 displays instructions for use, while tab 2 performs a *k*-means clustering of the data, where *k* is specified by the user. The resulting clusters are displayed with two variables of the user’s choosing as depicted in Figure 2. In tab 3, the user will choose a value *n*, indicating the number of rows by which to randomly subset the data, and in tab 4 the user selects a single variable to be plotted as a bar chart. Finally, in tab 5 the user can save their outputs as an RDS file. This contrived example includes some key elements of a typical workflow in that the five tabs introduce a dataset, guide the user through light data wrangling, produce analysis outputs, and offer the ability to save the results.

The app’s blueprint (Figure 2, bottom) identifies the *shiny* modules in each tab, showing how outputs from one module can serve as inputs to the next. Note that while this example shows a single module in each tab with differing inputs/outputs, in the general case tabs can contain an arbitrary number of *shiny* modules (including multiple instances of the same module) and each module can have multiple inputs/outputs.

While two of the *shiny* modules within the “iris\_explorer” app pertain to the iris dataset specifically (“iris\_intro” and “iris\_cluster”), the remaining *shiny* modules (“subset\_rows”, “single\_column\_plot”, and “save”) may be incorporated into other apps.



**Figure 2:** Top: The ‘iris\_explorer’ app guides a user through an analysis of the iris dataset in a tab-based sequence. Bottom: A blueprint of the ‘iris\_explorer’ app shows the 5 tabs, each containing a single module identified by name within blue ovals. Some of the shiny modules require inputs and generate outputs as identified in gray polygons.

Developers who utilize the same *shiny* modules within different apps will naturally be faced with several questions:

1. Which *shiny* modules have been written? Are they well documented with unit testing?
2. What are the module's inputs (arguments) and outputs (returns)?
3. Where are the *shiny* modules stored?
4. How can *shiny* modules be combined into a cohesive, well-documented app?
5. How can production-ready apps be deployed for end-users?

Users of an app created with the *shiny* framework may wish to know:

6. Can analysis outputs be saved as a fully reproducible workflow?
7. Can outputs be ingested into a *Rmarkdown* or *Quarto* template for rapid reporting?

## 1.1 Introducing *shiny*mgr

The R package, *shiny*mgr, was developed to meet these challenges (Clarfeld et al., 2024). The *shiny*mgr package includes a general framework that allows developers to create *shiny* modules, stitch them together as individual “apps” that are embedded within the master *shiny* application, and then deploy them on a *shiny* server or incorporate them into R packages. *shiny*mgr was motivated from our first-hand experience in our work building tools that assist scientists in remote wildlife monitoring with the R package *AMMonitor* (Balantic and Donovan, 2020). Dependencies of *shiny*mgr include the packages *DBI* (R Special Interest Group on Databases (R-SIG-DB) et al., 2022), *reactable* (Lin, 2022), *RSQLite* (Müller et al., 2022), *renv* (Ushey, 2023), *shiny* (Chang et al., 2022), *shinyjs* (Attali, 2021), and *shinydashboard* (Chang and Borges Ribeiro, 2021).

From the developer's perspective, an “app” consists of an ordered set of tabs, each of which contain specified *shiny* modules. *shiny* modules are the basic element in the *shiny*mgr framework; they can be used and re-used across different tabs and different apps. Information about each module and app is stored in a SQLite database (Hipp, 2020). The *shiny*mgr app builder “stitches” *shiny* modules together so that outputs from one module serve as inputs to the next, creating an analysis pipeline that is easy to implement and maintain. When apps are production-ready, developers can deploy a stand-alone *shiny* application independent of *shiny*mgr on a server or within an R package. From the end-user's perspective, an “app” created with the *shiny*mgr framework consists of an ordered series of *shiny* tabs, establishing an analysis. Users can save their inputs and outputs as an RDS file to ensure full reproducibility. Furthermore, the RDS file may be loaded into an R Markdown (Rmd) or Quarto (qmd) template for rapid reporting. We are unaware of existing packages that unify the elements of modularization, documentation, reproducibility, and reporting in a single framework.

We introduce *shiny*mgr in sections 2-4 below. In section 2 we describe how developers can create apps using the *shiny*mgr framework. In section 3 we describe how developers can deploy a *shiny*mgr project on a local machine, server, or within an R package. In section 4 describes the end-user experience, where end-users execute an “app” and store results for reproducibility and reporting. The package tutorials and cheat sheet are described in section 5. The *shiny*mgr package comes with a series of *learnr* (Schloerke et al., 2020) tutorials described at the end of the paper.

## 2 Developing *shiny*mgr apps

### 2.1 Setting up *shiny*mgr

The canonical home of *shiny*mgr is <https://code.usgs.gov/vtcfwru/shinymgr/> where *shiny*mgr users may post merge requests and bug fix requests. *shiny*mgr may also be downloaded from CRAN.

```
install.packages("shinymgr")
```

The development version can be downloaded with:

```
remotes::install_gitlab(
  repo = "vtcfwru/shinymgr",
  auth_token = Sys.getenv("GITLAB_PAT"),
  host = "code.usgs.gov",
  build_vignettes = FALSE)
```

Once installed, a new *shiny*mgr project can be created within a parent directory:

```
# set the directory path that will house the shinymgr project
parentPath <- getwd()
```

```
# set up raw directories and fresh database
shinymgr_setup(
  parentPath = parentPath,
  demo = TRUE)
```

The `shinymgr_setup()` function produces the following directory structure within the primary “shinymgr” directory. This structure consists of 3 files that make up the “master” app (`global.R`, `server.R`, and `ui.R`), and 9 directories. If the argument `demo` is set to `FALSE`, these directories will be largely empty, except for the “modules\_mgr” and “database” directories, which will contain *shiny* modules for rendering *shinymgr*’s UI and an empty SQLite database, respectively. If the argument `demo` is set to `TRUE`, each directory will include several demo files as shown, including a pre-populated database. Here, we highlight a subset of the demo files related to the “iris\_explorer” app to guide developers through the key elements of *shinymgr* (additional demo files come with package but are omitted here for clarity).

```
shinymgr
+-- analyses
|   \-- iris_explorer_Gandalf_2023_06_05_16_30.RDS
+-- data
|   \-- iris.RData
+-- database
|   \-- shinymgr.sqlite
+-- global.R
+-- modules
|   +-- iris_cluster.R
|   +-- iris_intro.R
|   +-- single_column_plot.R
|   \-- subset_rows.R
+-- modules_app
|   \-- iris_explorer.R
+-- modules_mgr
|   +-- add_app.R
|   +-- add_mod.R
|   +-- add_report.R
|   +-- add_tab.R
|   +-- app_builder.R
|   +-- my_db.R
|   +-- new_analysis.R
|   +-- new_report.R
|   +-- queries.R
|   +-- save_analysis.R
|   +-- stitch_script.R
|   \-- table.R
+-- reports
|   \-- iris_explorer
|       \-- iris_explorer_report.Rmd
+-- server.R
+-- tests
|   +-- shinytest
|   |   +-- test-iris_explorer-expected
|   |   |   +-- 001.json
|   |   |   +-- 001.png
|   |   |   +-- 002.json
|   |   |   \-- 002.png
|   |   \-- test-iris_explorer.R
|   +-- shinytest.R
|   +-- testthat
|   |   +-- test-iris_cluster.R
|   |   \-- test-subset_rows.R
|   \-- testthat.R
+-- ui.R
```

```
\-- www
  +-- dark_mode.css
  \-- shinymgr-hexsticker.png
```

The directory structure produced by `shinymgr_setup()` includes the following:

- The **analyses** directory provides the developer an example of a previously run analysis that was created using the *shinymgr* framework (an RDS file). An analysis file name includes the app name (e.g. “iris\_explorer”), the name of the person who ran the analysis (e.g. “Gandalf”), and the date and time of the analysis (e.g., “iris\_explorer\_Gandalf\_2023\_06\_05\_16\_30.RDS”).
- The **data** directory stores RData files that can be used by various *shinymgr* apps (e.g., “iris.RData”).
- The **database** directory stores the *shinymgr* SQLite database, named “shinymgr.sqlite.” The database is used by the developer to track all *shiny* modules, their arguments (inputs), returns (outputs), and how they are combined into *shinymgr* apps.
- The **modules** directory stores stand-alone *shiny* modules. These files are largely written by the developer with the help of the `mod_init()` function, and are registered in the database with the `mod_register()` function. Four of the example *shiny* modules listed are used in the “iris\_explorer” app.
- The **modules\_app** directory stores *shiny* modules that are *shinymgr* “apps” – the stitching together of *shiny* modules into a tab-based layout that provides an analysis workflow (Figure 2 shows the “iris\_explorer” app layout). Files within the “modules\_app” directory are not written by hand - instead, they are created with the *shinymgr* “app builder.”
- The **modules\_mgr** directory stores *shiny* modules that build the overall *shinymgr* framework.
- The **reports** directory provides an example of an *RMarkdown* (Rmd) template (e.g., “iris\_explorer\_report.Rmd”), allowing for rapid reporting by an end-user.
- The **tests** directory stores both *testthat* (Wickham, 2011) and *shinytest* (Chang et al., 2021) code testing scripts.
- The **www** directory stores images that may be used by a *shiny* app.
- In addition to these directories, three files are created for launching the master *shinymgr shiny* application:
  1. **ui.R** - This file contains code to set the user interface for the master *shinymgr* app.
  2. **server.R** - The master server file.
  3. **global.R** - The global.R file is sourced into the server.R file at start-up. It sources all of the *shiny* modules within the *shinymgr* framework so they are available when *shinymgr* is launched.

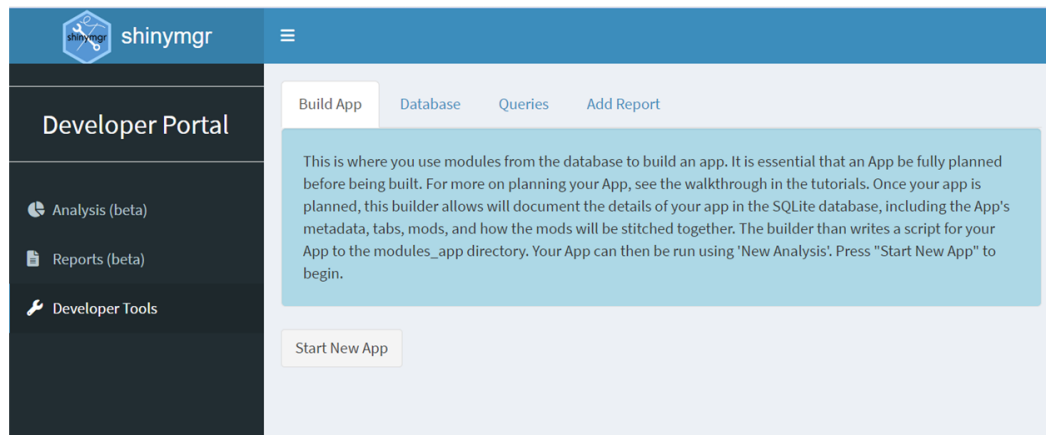
## 2.2 The *shinymgr* developer’s portal

Once set-up is complete, the `launch_shinymgr()` function will launch the *shinymgr* “Developer’s Portal” UI, allowing developers to create and test new *shinymgr* apps.

```
# launch shinymgr
launch_shinymgr(shinyMgrPath = paste0(parentPath, "/shinymgr"))
```

The portal is recognizable by the *shinymgr* logo in the upper left corner (Figure 3). The portal consists of three main tabs in the left menu. The “Developer Tools” tab is used to create apps, view the *shinymgr* database, and register reports, while the “Analysis (beta)” and “Reports (beta)” tabs allow developers to evaluate apps from the user’s perspective.

The “Developer Tools” section includes 4 tabs for app development: The “Build App” tab allows the developer to create new *shinymgr* apps from existing modules using the *shinymgr* app builder; the “Database” tab displays the *shinymgr* database tables, the “Queries” tab contains a set of standard database queries, and the “Add Reports” tab allows the developer to link a report (Rmd or qmd) to a given *shinymgr* app (Figure 3), as described below.



**Figure 3:** The shinymgr Developer Portal consists of a sidebar panel where developers can create new shiny modules and new apps, and test-drive analyses and reports from the user’s perspective. The main panel shows the ‘Build App’ tab within the ‘Developer Tools’ section.

### 2.3 The *shinymgr* database

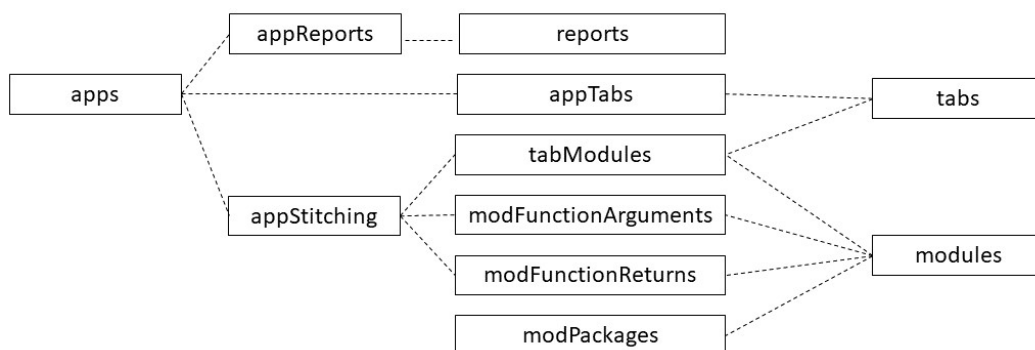
The *shinymgr* SQLite database (“shinymgr.sqlite”) is a single file created by the `shinymgr_setup()` function. The database tracks all *shiny* modules, their arguments (inputs), returns (outputs), their package dependencies and version numbers, how they are combined into an “app,” and any reports that are associated with apps. The database tables are populated via dedicated *shinymgr* functions.

The *shinymgr* database consists of 11 tables in total (Figure 4). These tables are connected to each other as a typical relational database, with primary keys establishing unique records in each table, and foreign keys that reference primary keys in other tables (see Appendix A for a full database schema and the “database” *learnr* tutorial for additional information).

The “apps,” “appReports,” “reports,” “appTabs,” and “tabs” tables largely store information on what a user would see when they run an analysis. The table “apps” stores information about apps such as “iris\_explorer.” Apps consist of tabs, which are listed in the “tabs” table. Tabs are linked to apps via the “appTabs” table. The table “reports” lists any Rmd or qmd files that serve as a report template, and the table “appReports” links a specific report with a specific app.

The remaining 6 tables in Figure 4 are “modules,” “modFunctionArguments,” “modFunctionReturns,” “modPackages,” “tabModules,” and “appStitching.” These tables largely store information about *shiny* modules that a developer creates, i.e., what *shiny* modules have been written, what are their arguments and returns, and what packages they use. The “tabModules” table identifies which tabs call which *shiny* modules (with a single tab capable of calling multiple *shiny* modules), and the “appStitching” table specifies how *shiny* modules are “stitched” together, i.e., which module returns are passed in as arguments to downstream *shiny* modules.

Four of the 11 database tables focus on modules, highlighting that *shiny* modules are basic building blocks of any *shinymgr* app. Developers create new *shiny* modules with the `mod_init()` function, which copies a *shinymgr* module template (an R file template) that includes a header with key-value



**Figure 4:** The 11 tables of the shinymgr SQLite database. Lines indicate how the tables are related to each other.

that describe the module, including the module name, display name, description, citation, notes, and module arguments and returns (if any). For example, the header of the `iris_cluster` module is:

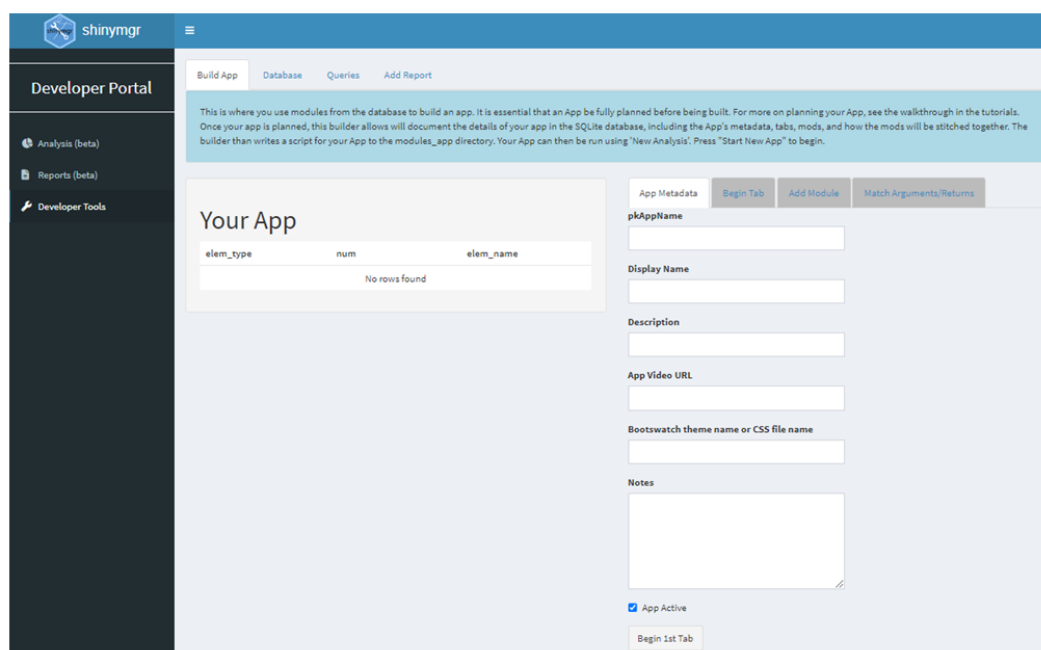
```
#!/ ModName = iris_cluster
#!/ ModDisplayName = Iris K-Means Clustering
#!/ ModDescription = Clusters iris data based on 2 attributes
#!/ ModCitation = Baggins, Bilbo. (2023). iris_cluster. [Source code].
#!/ ModNotes = Demo module for the shinymgr package.
#!/ ModActive = 1
#!/ FunctionReturn = returndf !! selected attributes and their assigned clusters !! data.frame
```

The module code is written beneath the header (see Appendix B for an example). Function calls within the module code should be written with `package::function()` notation, making explicit any R package dependencies. Once the module is completed, unit tests can be written and stored in the *shinymgr* project's "tests" directory. The final module file is saved to the "modules" directory and registered into the database with the `mod_register()` function. The `mod_register()` function populates the modules, "modFunctionArguments", and "modFunctionReturns" SQLite database tables. Further, it uses the *renv* package to identify any package dependencies and inserts them into the `modPackages` table. Readers are referred to the "modules" "tests", and "shinymgr\_modules" *learnr* tutorials that come with the *shinymgr* package for more details.

Once modules are registered in the database, the developer can incorporate them into new apps. As *shiny* modules and apps in the database represent files that contain their scripts, deleting a module or an app from the database will delete all downstream database entries as well as (optionally) the actual files themselves. Deletion of a module will fail if it is being used in other apps. Module updates can be versioned by creating a new module and then referencing its precursor in the "modules" database table.

## 2.4 The *shinymgr* app builder

Once developers create and register their own stand-alone *shiny* modules, apps are generated with *shinymgr*'s app builder (Figure 5).



**Figure 5:** The *shinymgr* Developer Portal layout, showing the app builder in the Developer Tools.

Developers are guided through a process where they design their app from *shiny* modules they have registered. The builder then populates the *shinymgr* database with instructions on how to construct the app and writes the app's script based on those instructions. The newly created script is saved to the "modules\_app" directory. Through this structured process, apps produced by the builder are well-documented and generate highly reproducible analyses. Readers are encouraged to peruse the tutorial, "apps", for more information.



The `qry_app_flow()` function will query the database to return a list of the *shiny* modules and tabs included in a specified app, such as “iris\_explorer”:

```
# look at the appTabs table in the database
qry_app_flow("iris_explorer", shinyMgrPath = paste0(getwd(), "/shinymgr"))
```

	fkAppName	fkTabName	tabOrder	fkModuleName	modOrder
1	iris_explorer	IE_intro	1	iris_intro	1
2	iris_explorer	IE_iris_data	2	iris_cluster	1
3	iris_explorer	IE_subset_rows	3	subset_rows	1
4	iris_explorer	IE_plot_data	4	single_column_plot	1

As shown in Figure 2, this app has 5 tabs, and each tab features a single module. The “Save” tab is the final tab in all *shinymgr* apps and is not listed in the query result.

Developers can “beta test” apps prior to deployment by selecting the Analysis (beta) tab in the Developer’s Portal (Figure 3). They can also create *RMarkdown* or *Quarto* report templates that accept the outputs from an analysis and incorporate them into a report. Report metadata are logged in the “reports” table of the database, and then linked with a specific app in the “appReports” table. An end-user will run an analysis and render a report, a process described more fully in the “Using *shinymgr* Apps” section below.

To summarize this section, developers use the `shinymgr_setup()` function to create the directory structure and underlying database needed to build and run *shiny* apps with *shinymgr*. Developers use the `mod_init()` and `mod_register()` functions to create modules and make them available for inclusion in new apps built with the *shinymgr* app builder. A developer can create as many *shinymgr* projects as needed. In each case, the *shinymgr* project is simply a fixed directory structure with three R files (`ui.R`, `server.R`, and `global.R`), and a series of subdirectories that contain the apps and *shiny* modules created by the developer, along with a database for tracking everything.

### 3 Deploying *shinymgr* projects

Once development is completed, developers can deploy their *shinymgr* project on a server or within an R package by copying portions of the *shinymgr* project to a new location while retaining the original project for future development. Once deployed, a *shinymgr* project no longer requires the *shinymgr* package or database to be run. Thus, the files and directories to be copied for deployment include only:

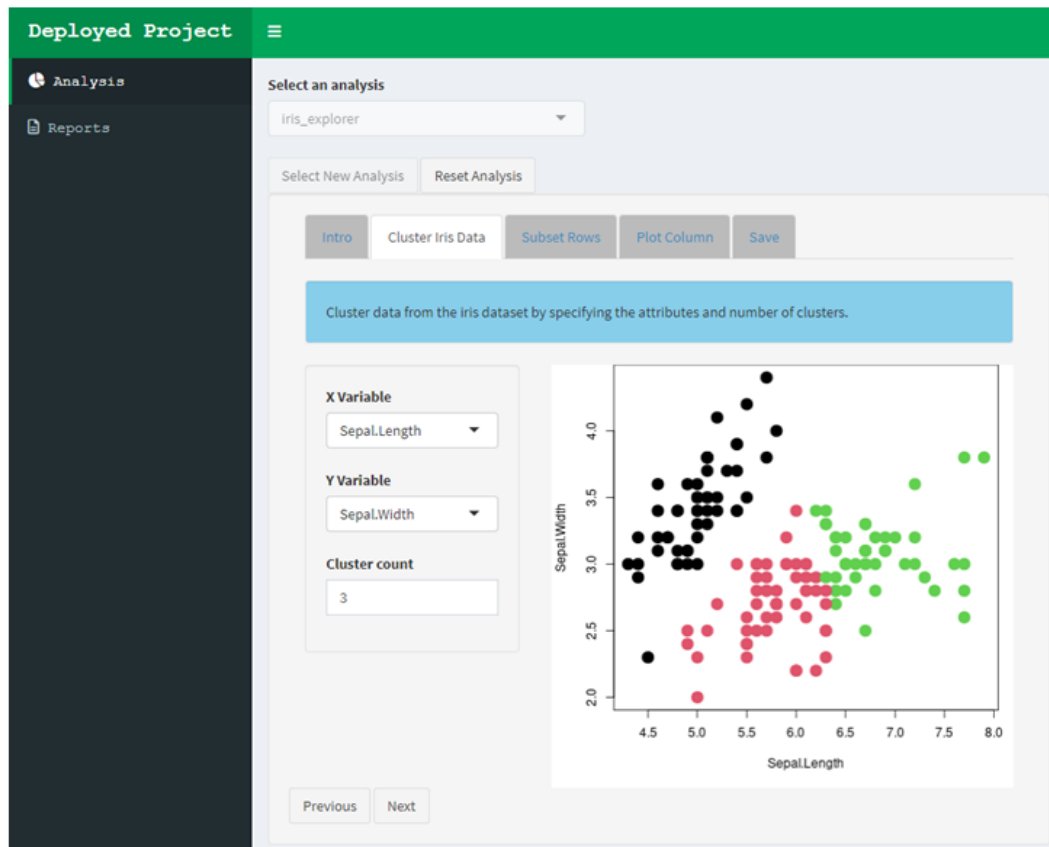
```
shinymgr
+-- data
+-- global.R
+-- modules
+-- modules_app
+-- modules_mgr
+-- reports
+-- server.R
+-- ui.R
\-- www
```

The master app files, `ui.R`, `global.R`, and `server.R`, are needed to run the *shinymgr* framework.

When deploying a *shinymgr* project within an R package, objects within the data folder should be copied into the package’s “data” folder. The remaining files should be copied into a directory within the package’s “inst” folder that will house the master *shiny* application. Deployment on a server such as shinyapps.io will require similar adjustments.

After files are copied to the correct location, a few key adjustments are needed. First, the “modules\_app” directory should contain only those apps (and dependent modules and reports) that can be used by end-users; unused apps, modules, and reports can be deleted. Second, the `new.analysis.R` script within the `modules_mgr` folder will require minor updates to remove dependencies on the *shinymgr* database. Third, the `ui.R` and `server.R` scripts should be updated to no longer showcase *shinymgr* and the Developer’s Portal; rather, it should be customized by the developer to create their own purpose-driven apps. For example, Figure 6 shows a hypothetical deployment of the master app titled “Deployed Project” that is based on the *shinymgr* framework. Notice the absence of the Developer Tools tab and the absence of references to *shinymgr*. The “deployment” *learnr* tutorial provides more in-depth discussion.





**Figure 6:** An example of a deployed shinyr app. The deployed version excludes the Developers Tools tab and is an example of what the end user sees when using a deployed app.

To summarize this section, deploying the *shinyr* framework involves copying key elements of the *shinyr* developer project into package or server directories, updated as needed for use by end-users. Readers are referred to the “deployment” tutorial for further information.

## 4 Using *shinyr* apps

Apps built with *shinyr* can appeal to various types of end-users. When deployed as part of an R package, end-users would be anyone who uses that package. Apps may also be distributed as stand-alone scripts, or hosted on a server, as described above. Developers may also use *shinyr* to produce apps for their own use (i.e., the developer *is* the end-user). Regardless of who the intended end-user is, this section discusses that user’s experience after the master app is deployed.

Whoever the intended audience for the app, this section discusses how an app can be used *after* it has been deployed.

### 4.1 Reproducible analyses

The final tab in any *shinyr* app provides the opportunity to save the analysis itself. Reproducibility is a core tenet of *shinyr*. Therefore, a robust set of metadata are saved as an RDS file to allow a user to understand and replicate their results. An example of a completed analysis is the file, “iris\_explorer\_Gandalf\_2023\_06\_05\_16\_30.RDS,” which stores a user’s analytic steps for a run of the “iris explorer” app. The code below reads in this example file, and shows the structure (a list with 23 elements):

```
rds_filepath <- paste0(getwd(), "/shinyr/analyses/iris_explorer_Gandalf_2023_06_05_16_30.RDS")
old_analysis <- readRDS(rds_filepath)
str(old_analysis, max.level = 2, nchar.max = 20, vec.len = 15)
```

List of 23

```

$ analysisName      : chr "iri"| __truncated__
$ app               : chr "iris_explorer"
$ username          : chr "Gandalf"
$ mod2-clusters     : int 3
$ mod2-xcol         : chr "Sepal.Length"
$ mod2-ycol         : chr "Petal.Length"
$ mod3-full_table__reactable__pageSize : int 10
$ mod3-resample     : 'shinyActionButtonValue' int 1
$ mod3-full_table__reactable__pages    : int 15
$ mod3-subset_table__reactable__page   : int 1
$ mod3-full_table__reactable__page     : int 1
$ mod3-sample_num   : int 20
$ mod3-subset_table__reactable__pages   : int 2
$ mod3-subset_table__reactable__pageSize: int 10
$ returns           :List of 3
..$ data1:List of 1
..$ data2:List of 1
..$ data3:List of 2
$ notes             : chr "Thi"| __truncated__
$ timestamp         : POSIXct[1:1], format: "202"| __truncated__
$ metadata          :List of 6
..$ appDescription: chr "Clu"| __truncated__
..$ mod1            :List of 7
..$ mod2            :List of 7
..$ mod3            :List of 7
..$ mod4            :List of 7
..$ lockfile        :List of 2
$ app_code           : chr "# T"| __truncated__
$ iris_intro_code    : chr "#!!"| __truncated__
$ iris_cluster_code  : chr "#!!"| __truncated__
$ subset_rows_code   : chr "#!!"| __truncated__
$ single_column_plot_code : chr "#!!"| __truncated__

```

The list stores a great deal of information:

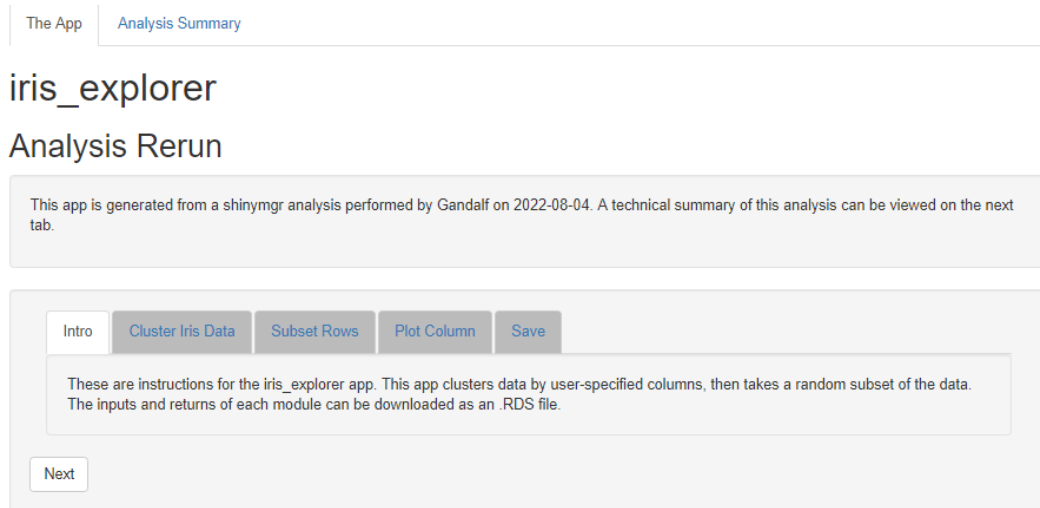
- **analysisName** is the name of the analysis and is equivalent to the filename of the RDS file (without the extension)
- **app** is the name of the app that produced the saved analysis results.
- **username** was entered in the “Save” tab when the analysis was performed.
- **mod#-value** indicate the values of each *shiny* module’s arguments (inputs), if any exist, at the time the analysis was saved.
- **returns** includes values of all outputs (returns) of each module.
- **notes** were entered in the “Save” tab when the analysis was performed.
- **timestamp** is the date/time when the analysis was saved.
- **metadata** includes robust information about each module, including the app description and the description of each module as it was originally stored in the *shinygr* database tables. The metadata list element also includes an *reno* “lockfile”: a list that describes the R version and R package dependencies (including *shinygr*) used by the app itself. The lockfile captures the state of the app’s package dependencies at the time of its creation; in the case of *shinygr*, it contains the dependencies used by the developer who created the app. Each lockfile record includes the name and version of the package and their installation source.
- **\*\_code** attributes with this format contain the source code for the app.

The code list element allows an end user to revisit the full analysis with *shinygr*’s `rerun_analysis()` function, supplying the file path to a saved *shinygr* analysis (RDS file).

```
rerun_analysis(analysis_path = rds_filepath)
```

The `rerun_analysis()` function will launch a *shiny* app with two tabs (Figure 7); it can only be run during an interactive R session, with no other *shiny* apps running.

The first tab is called “The App”, and will be visible when the `rerun_analysis()` function is called. It contains a header with the app’s name, a subheading of “Analysis Rerun,” and a fully functioning, identical copy of the *shiny* app used to generate the saved analysis. Below that, a disclaimer appears, indicating the app was produced from a saved analysis. A summary of the analysis is presented on the second tab that displays the values used to produce the given analysis output.



**Figure 7:** A screenshot of the `rerun_analysis()` function, as called on the saved analysis from the `iris_explorer` app (RDS file). The active tab, called 'The App', allows a user to rerun a previously executed analysis. The 'Analysis Summary' tab displays the values of all module arguments and returns, captured when the analysis was saved, along with a detailed description of the app, its modules, the App's source code, and all package dependencies.

If the `rerun_analysis()` function fails, it could be due to a change in R and package versions currently installed on the end-user's machine. To that end, the lockfile that is included in the metadata section of the RDS file can be used to restore the necessary R packages and R version with the `restore_analysis()` function. This function will attempt to create a self-contained *renv* R project that includes all of the packages and the R version used by the developer when the app was created. The analysis RDS is added to this new project, where the `rerun_analysis()` function can be attempted again. Readers are referred to the "analyses" tutorial for further information.

## 4.2 Rapid reporting

Another important feature of *shinymgr* is the ability to share results of an analysis with others in a friendly, readable format with *RMarkdown* or *Quarto*. Apps produce an RDS file, which may be passed into an Rmd or qmd file as a parameterized input. For example, the demo database includes a report template called "iris\_explorer\_report.Rmd." This file, with code shown below, allows users to navigate to the RDS file produced by the "iris explorer" app and render the rapid report.

```
---
title: 'Annual Report for Iris Explorer'
output: html_document
params:
  user:
    label: "User"
    value: "Bilbo"
    placeholder: "Enter user name"
  year:
    label: "Year"
    value: 2017
    input: slider
    min: 2010
    max: 2018
    step: 1
    sep: ""
  file:
    input: file
    label: "Choose RDS"
    value: ""
    multiple: FALSE
```

```
    buttonLabel: "Browse to analysis output..."
```

```
---
```

```
```${r setup, include=FALSE}
knitr::opts_chunk$set(echo = FALSE)
library(knitr)
ps <- readRDS(params$file)
```
```

This report summarizes an analysis of iris data by ``r params$user`` conducted in ``r params$year``. Iris data was clustered into ``r ps$'mod2-clusters'`` groups based on ``r ps$'mod2-xcol'`` and ``r ps$'mod2-ycol'``. A random sample of ``r ps$'mod3-sample_num'`` records were collected, with sample sizes shown in the pie chart below:

```
```${r}
pie_data <- table(ps$returns$data2$subset_data$cluster)
pie(
  x = pie_data,
  labels = as.character(pie_data),
  col = rainbow(length(pie_data)),
  main = "Number of random samples by cluster"
)
legend(
  x = "topright",
  legend = names(pie_data),
  fill = rainbow(length(pie_data))
)
```
```

Some things to note about this analysis are: ``r ps$notes``

Respectfully submitted,

Gandalf

Reports may be run within the deployed version of *shinygr* (e.g., left menu of Figure 6), or may be run directly in R by opening the Rmd file and navigating to the RDS as a file input. Users who run a report can download it to their local machine as a HTML, PDF, or Word file, where they can further customize the output.

To summarize this section, users of *shinygr* “apps” created with the *shinygr* framework are presented with a series of *shiny* tabs that establish an analysis workflow. Users can save their inputs and outputs as an RDS file to ensure full reproducibility. Further, the RDS file may be loaded into an R Markdown (Rmd) or Quarto (qmd) template for rapid reporting.

## 5 Tutorials and cheatsheet

with the package. Below is a list of current tutorials, intended to be worked through in order:

Available tutorials:

```
* shinygr
- intro           : "shinygr-01: Introduction"
- shiny           : "shinygr-02: Shiny"
- modules         : "shinygr-03: Modules"
- app_modules     : "shinygr-04: App modules"
- tests           : "shinygr-05: Tests"
- shinygr         : "shinygr-06: shinygr"
- database        : "shinygr-07: Database"
- shinygr_modules : "shinygr-08: shinygr_modules "
- apps            : "shinygr-09: Apps"
```

- analyses : "shinymgr-10: Analyses"
- reports : "shinymgr-11: Reports"
- deployment : "shinymgr-12: Deployment"

The “intro” tutorial gives a general overview. Tutorials 2-5 are aimed at developers who are new to *shiny*, while tutorials 6 – 12 focus on the *shinymgr* package.

Launch a tutorial with the `learnr::run_tutorial()` function, providing the name of the module to launch. The tutorial should launch in a browser, which has the benefit of being able to print the tutorial to PDF upon completion:

```
learnr::run_tutorial(  
  name = "modules",  
  package = "shinymgr")
```

Additionally, the package cheatsheet can be found with:

```
browseURL(paste0(find.package("shinymgr"), "/extdata/shinymgr_cheatsheet.pdf"))
```

Contributions are welcome from the community. Questions can be asked on the issues page at <https://code.usgs.gov/vtcfwru/shinymgr/issues>.

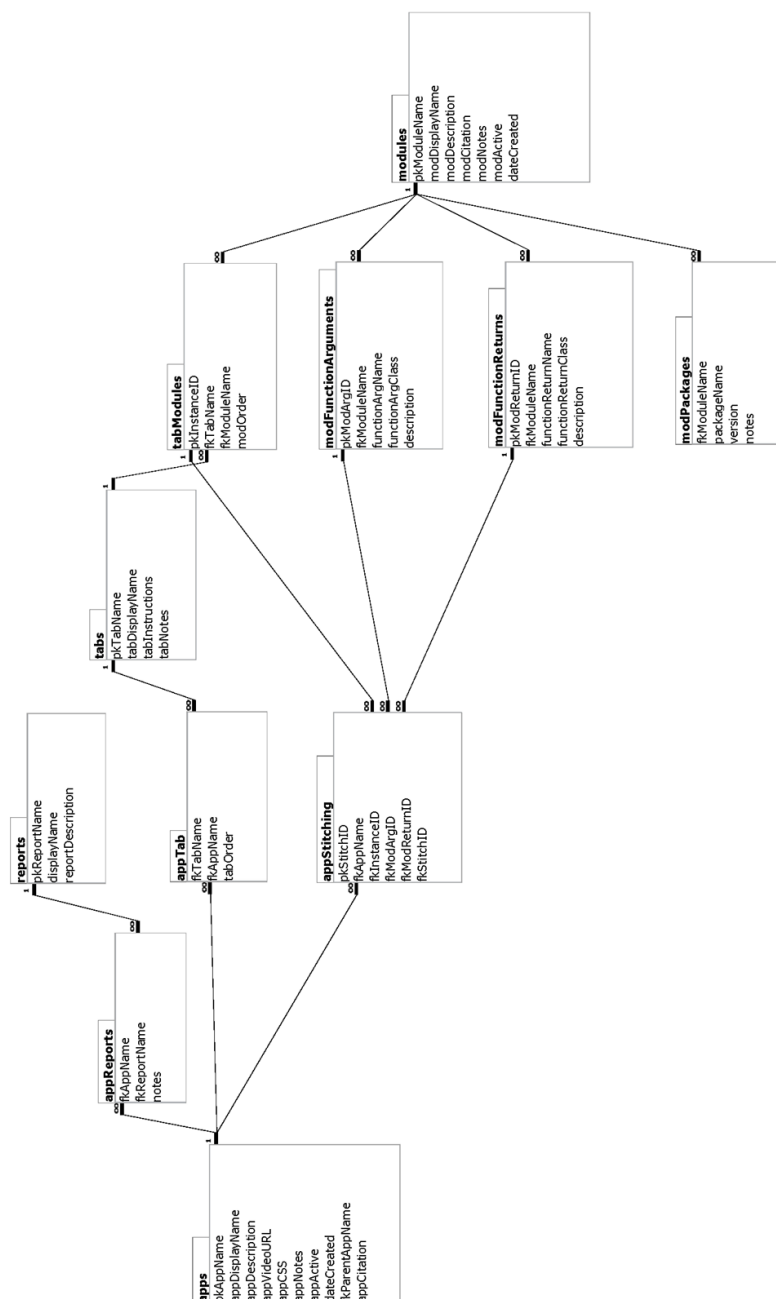
## 6 Acknowledgments

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## 7 Bibliography

## 8 Appendix A

Entity relationship diagram for the *shinymgr* database, which tracks all components of an apps and modules (Figure 8). The database consists of 11 tables. Primary keys are referenced with a “pk” prefix, while foreign keys are referenced with an “fk” prefix. A full description of the database is contained in the “database” *learnr* tutorial that comes with the *shinymgr* package



**Figure 8:** Entity relationship diagram for the *shinymgr* database, which tracks all components of an apps and modules. The database consists of 11 tables. Primary keys are referenced with a ‘pk’ prefix, while foreign keys are referenced with an ‘fk’ prefix. A full description of the database is contained in the ‘database’ *learnr* tutorial that comes with the *shinymgr* package.



## 9 Appendix B

Modules in *shiny* are written by developers for their own purposes. The `shiny::mod_init()` function creates a template for module development. The header is a series of key-value pairs that the developer fills out (typically after the module code is written and tested). The “iris\_cluster” module is presented below as an example. The module consists of two paired functions: here, `iris_cluster_ui(id)` and `iris_cluster_server()`. The UI is a function with an argument called `id`, which is turned into module’s “namespace” with the `NS()` function. A namespace is simply the module’s identifier and ensures that function and object names within a given module do not conflict with function and object names in other modules. The Id’s for each input and output in the UI must be wrapped in a `ns()` function call to make explicit that these inputs are assigned to the module’s namespace. All UI elements are wrapped in a `tagList()` function, where a `tagList` allows one to combine multiple UI elements into a single R object. Readers should consult the “modules,” “tests,” and “shinyr\_modules” tutorials for additional information.

```
##! ModName = iris_cluster
##! ModDisplayName = Iris K-Means Clustering
##! ModDescription = Clusters iris data based on 2 attributes
##! ModCitation = Bagins, Bilbo. (2022). iris_cluster. [Source code].
##! ModNotes =
##! ModActive = 1
##! FunctionReturn = returndf !! selected attributes and their assigned clusters !! data.frame

iris_cluster_ui <- function(id){
  # create the module's namespace
  ns <- NS(id)

  tagList(
    sidebarLayout(
      sidebarPanel(
        # add the dropdown for the X variable
        selectInput(
          ns("xcol"),
          label = "X Variable",
          choices = c(
            "Sepal.Length",
            "Sepal.Width",
            "Petal.Length",
            "Petal.Width"
          ),
          selected = "Sepal.Length"
        ),

        # add the dropdown for the Y variable
        selectInput(
          ns("ycol"),
          label = "Y Variable",
          choices = c(
            "Sepal.Length",
            "Sepal.Width",
            "Petal.Length",
            "Petal.Width"
          ),
          selected = "Sepal.Width"
        ),

        # add input box for the cluster number

        numericInput(
          ns("clusters"),
          label = "Cluster count",
          value = 3,
          min = 1,
          max = 9
        )
      )
    )
  )
}
```

```

    ), # end of sidebarPanel

    mainPanel(
      # create outputs
      plotOutput(
        ns("plot1")
      )
    ) # end of mainPanel
  ) # end of sidebarLayout
} # end of tagList
} # end of UI function

iris_cluster_server <- function(id) {

  moduleServer(id, function(input, output, session) {

    # combine variables into new data frame
    selectedData <- reactive({
      iris[, c(input$xcol, input$ycol)]
    })

    # run kmeans algorithm
    clusters <- reactive({
      kmeans(
        x = selectedData(),
        centers = input$clusters
      )
    })

    output$plot1 <- renderPlot({
      par(mar = c(5.1, 4.1, 0, 1))
      plot(
        selectedData(),
        col = clusters()$cluster,
        pch = 20,
        cex = 3
      )
    })

    return(
      reactiveValues(
        returndf = reactive({
          cbind(
            selectedData(),
            cluster = clusters()$cluster
          )
        })
      )
    )
  }) # end of moduleServer function
} # end of irisCluster function

```

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