

Setting up an R development environment on github

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Table of contents

0.1	Introduction	2
0.2	The Problem/Data	2
0.2.1	Step 1: Initial Repo Setup	2
1	Debugging workflow	5
2	Debugging Workflow from chatGPT	6
2.1	Reproducibility	8
2.2	Next Steps	8
2.3	References	9
3	Debugging Errors with devtools::test()	9
3.1	Step 1: Understand the devtools::test() Workflow	9
3.2	Step 2: Interpret the Error Message	10
3.3	Step 3: Isolate the Problem	10
3.4	Step 4: Debugging Common Errors	11
3.4.1	Error 1: Assertion Failures	11
3.4.2	Error 2: Unexpected Errors or Warnings	12
3.4.3	Error 3: Missing Dependencies	13
3.5	Step 5: Add Debugging Output	13
3.6	Step 6: Rerun Tests	14
3.7	Step 7: Finalize Fixes	14
3.8	Example Workflow for Debugging	14

0.1 Introduction

Its often the case that a data scientist needs to share an R function with a co-worker or a student. This post describes a step by step methodology for wrapping the function in a package, that includes a number of support files, and sharing it either via github or CRAN. This may seem like overkill but over time a not having to address the many technical issues that can arise when sharing a function in a more ad hoc manner may be appreciated.

0.2 The Problem/Data

The end goal of this terminal is to create a directory (or repository) that contains the package contents. The top level elements of the package are the DESCRIPTION file, the NAMESPACE file, the R directory, the tests directory, and the man directory. Other files such as a README.md, a LICENSE file are optional but recommended. The DESCRIPTION file contains metadata about the package such as the package name, the version number, the author, and the license. The NAMESPACE file contains the export and import declarations. The R directory contains the R functions. The tests directory contains the unit tests.

0.2.1 Step 1: Initial Repo Setup

Start by using the various helpful tools in the devtools and usethis packages to facilitate the repository building process.

Open R from the shell prompt in your development directory. and run the command `usethis::create_package("my_package")` to create the package directory and the DESCRIPTION and NAMESPACE files. Assuming the package will be named `my_package`.



Figure 1: purrr

```
install.packages("devtools")
library(devtools)
usethis::create_package("my_package")
```

This creates the following directory structure.

```
my_package  tree --charset=ascii
.
|-- DESCRIPTION
|-- NAMESPACE
`-- R

my_package  more DESCRIPTION
Package: my_package
Title: What the Package Does (One Line, Title Case)
Version: 0.0.0.9000
Authors@R:
  person("First", "Last", , "first.last@example.com", role = c("aut", "cre"),
         comment = c(ORCID = "YOUR-ORCID-ID"))
Description: What the package does (one paragraph).
License: `use_mit_license()`, `use_gpl3_license()` or friends to pick a
        license
Encoding: UTF-8
Roxygen: list(markdown = TRUE)
RoxygenNote: 7.3.2
```

Next use the `usethis` package tools to generate repository support files.

```
usethis::use_git()
usethis::use_github()
use_gpl_license(version = 3, include_future = TRUE)
usethis::use_readme_md()
usethis::use_code_of_conduct("rgthomas@ucsd.edu")
usethis::use_tidy_contributing()
```

Next copy the R file containing the function to the R directory and add a `#'` roxygen comment block to the top of the file. Then call `devtools::document()` to generate the `man` directory containing the help page.

```
devtools::document()
```

At this point the directory structure looks like this.

```
julia dev/my_package tree --charset=ascii
.
|-- DESCRIPTION # package Metadata
|-- NAMESPACE  # Exports and imports declarations
|-- R          # R functions
|  `-- my_package.R
|-- man        # Documentation for the functions
|  `-- my_package.Rd
```

The next step is to set up testing.

```
usethis::use_testthat()
```

```
call inside R
usethis::use_test("my_package")
```

This open an editor. Enter the unit tests using the `test_that` function.

```
# Test: Empty dataframe error
test_that("t2f throws an error for empty dataframe", {
  empty_df <- data.frame()
  expect_error(my_package(empty_df, filename = "empty_table"), "`df` must not be empty")
})
```

Set up a new repository on github.

```
git init
git add .
git commit -m "Initial commit"
```

Add each dependency (e.g. `kableExtra`) (e.g. `kableExtra`)
(e.g. `kableExtra`)

```
usethis::use_package("kableExtra", type = "Imports")
```

Finally do a full check using `devtools::check()`. This reflects the checks that CRAN will perform when you submit the package.

```
devtools::build()
devtools::install()
devtools::test()
devtools::check()
```

1 Debugging workflow

```
git checkout -b fix-bug
```

Debug locally and isolate the issue.

- Create a local branch (`fix-bug`) for the fix.
- `git checkout -b fix-bug`
- Make and test the changes.
- Run `devtools::test()` to confirm all tests pass.
- Use `devtools::check()` to validate the package.
- `git add .`
- `git commit -m "Fix issue with my_package function"`
- `git push`
- Merge the branch into the main branch and clean up.
- `git checkout main`
- `git merge fix-bug`
- `git branch -d fix-bug`
- `git push`

- Open a Pull Request.
- Update the version number
- `usethis::use_version("patch")`
- and push the final changes.
- `git push`

2 Debugging Workflow from chatGPT

Follow these steps to debug and fix issues in your R package:

1. Debug Locally

- Isolate the issue using R debugging tools like `browser()`, `traceback()`, or `debug()`.

2. Create a Local Git Branch

- Create a branch for the fix to isolate your changes:

```
git checkout -b fix-bug
```

3. Make and Test Changes

- Modify your code to fix the issue and add or update unit tests as needed.
- Run tests to confirm functionality:

```
devtools::test() # Confirm all tests pass  
devtools::check() # Validate the package complies with CRAN standards
```

4. Commit Your Changes

- Stage and commit your changes:

```
git add .  
git commit -m "Fix issue with my_package function"
```

5. Push the Branch

- Push the branch to GitHub for collaboration or to prepare for merging:

```
git push origin fix-bug
```

6. Open a Pull Request

- Open a Pull Request (PR) on GitHub to merge the fix into the main branch. Include a clear description of the changes.

7. Merge and Clean Up

- After review and approval, merge the branch into the main branch:

```
git checkout main  
git merge fix-bug
```

- Delete the branch locally and remotely:

```
git branch -d fix-bug  
git push origin --delete fix-bug
```

8. Test the Main Branch

- Ensure the main branch passes all tests:

```
devtools::test() # Confirm functionality  
devtools::check() # Validate compliance
```

9. Update the Version Number

- Increment the package version using `usethis::use_version()`:

```
usethis::use_version("patch") # Use "patch", "minor", or "major"
```

- Commit and push the version update:

```
git add DESCRIPTION  
git commit -m "Bump version to 1.0.1"  
git push
```

2.1 Reproducibility

```
# Print session info for reproducibility
sessionInfo()
```

```
R version 4.4.2 (2024-10-31)
Platform: aarch64-apple-darwin20
Running under: macOS Sequoia 15.2
```

```
Matrix products: default
```

```
BLAS: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRblas.0.dylib
LAPACK: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRlapack.dylib; LA
```

```
locale:
```

```
[1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
```

```
time zone: America/Los_Angeles
```

```
tzcode source: internal
```

```
attached base packages:
```

```
[1] stats      graphics  grDevices datasets  utils      methods    base
```

```
other attached packages:
```

```
[1] here_1.0.1      shiny_1.9.1      styler_1.10.3    quarto_1.4.4     pacman_0.5.1
[11] dplyr_1.1.4     purrr_1.0.2      readr_2.1.5      tidyr_1.3.1      tibble_3.2.1
[21] janitor_2.2.0   datapasta_3.1.0  ggthemes_5.1.0   conflicted_1.2.0 DT_0.33
```

```
loaded via a namespace (and not attached):
```

```
[1] tidymodels_1.2.1 viridisLite_0.4.2 R.utils_2.12.3    fastmap_1.2.0    promises_1.3.2
[11] processx_3.8.4    magrittr_2.0.3    compiler_4.4.2    rlang_1.1.4      tools_4.4.2
[21] miniUI_0.1.1.1    R.cache_0.16.0    withr_3.0.2      R.oo_1.27.0      grid_4.4.2
[31] cli_3.6.3         generics_0.1.3    remotes_2.5.0    rstudioapi_0.17.1 tzdb_0.4.0
[41] hms_1.1.3         visdat_0.6.0      systemfonts_1.1.0 glue_1.8.0        ps_1.8.1
[51] htmltools_0.5.8.1 R6_2.5.1          rprojroot_2.0.4  evaluate_1.0.1   R.methodsS3_1.8.2
[61] xfun_0.49         fs_1.6.5          pkgconfig_2.0.3
```

2.2 Next Steps

- Suggest areas for further exploration

- Mention potential improvements
- Invite reader engagement

2.3 References

- Cite your sources
 - Link to relevant documentation
 - Credit other contributors
-

3 Debugging Errors with `devtools::test()`

This guide provides a step-by-step process to debug errors occurring during `devtools::test()` for R packages.

3.1 Step 1: Understand the `devtools::test()` Workflow

1. What It Does:

- Runs all test scripts in the `tests/testthat/` directory using the `testthat` package.
- Reports errors, warnings, and failed assertions.

2. Output Format:

- Lists the file and test name where the error occurred.
 - Provides the expected and actual outputs (if applicable).
 - Shows the error message and traceback for debugging.
-

3.2 Step 2: Interpret the Error Message

Run `devtools::test()` in your R console or IDE terminal and note the output. Focus on:

1. Test File and Test Name:

- Example:

```
Failure (test-zzlongplot.R:31:3): parse_formula handles grouping and faceting variables
result$y (`actual`) not equal to "y" (`expected`).
```

- This indicates:

- The error is in `test-zzlongplot.R`.
- It occurred in the test named `parse_formula handles grouping and faceting variables`.

2. Error Details:

- Example:

```
`actual`: "y ~ x | group"
`expected`: "y"
```

- The test expected `result$y` to be "y", but the actual output was `"y ~ x | group"`.

3. Traceback:

- The traceback provides a stack of function calls leading to the error. Use `traceback()` immediately after the test failure to get additional context.

3.3 Step 3: Isolate the Problem

1. Run the Test Manually:

- Extract the failing test from the test file and run it manually:

```
library(testthat)
source("path/to/zlongplot/R/parse_formula.R")
result <- parse_formula(y ~ x | group ~ facet_y + facet_x)
expect_equal(result$y, "y")
```

- This helps verify if the error occurs outside the testing framework.

2. Verify Input Data:

- Ensure that the inputs to the function are as expected. For example, check if the formula passed to `parse_formula` matches the expected format.

3. Use Debugging Tools:

- Insert `browser()` into the failing function to step through its execution:

```
parse_formula <- function(formula) {
  browser()
  # Function logic...
}
```

- When the code pauses at `browser()`, inspect the environment using:

```
ls()
print(formula)
```

3.4 Step 4: Debugging Common Errors

3.4.1 Error 1: Assertion Failures

Example:

```
result$y (`actual`) not equal to "y" (`expected`).
```

Steps: 1. Check the test code: - Verify if the `expect_equal()` or similar assertion accurately reflects the intended behavior.
- Example: `r expect_equal(result$y, "y")` - If the test expectation is wrong, update it to match the correct behavior.

2. Check the function output:

- Run the function manually with the same inputs and inspect its output:

```
parse_formula(y ~ x | group ~ facet_y + facet_x)
```

3. Fix the function logic:

- If the function output is incorrect, debug the function implementation.

3.4.2 Error 2: Unexpected Errors or Warnings

Example:

Error: Input must be a formula object

Steps: 1. Reproduce the Error: - Identify the exact input that triggers the error.

2. Add Input Validation:

- Validate inputs at the start of the function to catch issues early.

```
if (!inherits(formula, "formula")) {  
  stop("Input must be a formula object")  
}
```

3. Check for Edge Cases:

- Test the function with edge cases, such as missing or malformed inputs.

3.4.3 Error 3: Missing Dependencies

Example:

Error: could not find function "mutate"

Steps: 1. Check Imports: - Ensure the missing function's package is listed under Imports in DESCRIPTION.

2. Explicitly Load Dependencies:

- Use the `::` operator to call functions explicitly:

```
dplyr::mutate(...)
```

3. Add `requireNamespace()`:

- Dynamically load namespaces if not attached:

```
if (!requireNamespace("dplyr", quietly = TRUE)) {  
  stop("dplyr is required for this function.")  
}
```

3.5 Step 5: Add Debugging Output

1. Print Debugging Information:

- Add `print()` or `cat()` statements to inspect variables:

```
cat("Parsed y:", y_var, "\n")
```

2. Use Logging:

- Use a logging package like **logger** to add structured debug messages.
-

3.6 Step 6: Rerun Tests

- After making changes:
1. Save the modified code and test files.
 2. Re-run the tests: `r devtools::test()`
-

3.7 Step 7: Finalize Fixes

1. **Remove Debugging Code:**

- Remove `browser()`, `print()`, and other debug artifacts.

2. **Re-run Full Checks:**

- Run `devtools::check()` to ensure the package passes all CRAN checks.
-

3.8 Example Workflow for Debugging

1. Identify the failing test:

```
devtools::test()
```

2. Manually isolate and debug the test:

```
result <- parse_formula(y ~ x | group ~ facet_y + facet_x)
print(result)
```

3. Add debugging statements:

```
parse_formula <- function(formula) {
  print(formula)
  browser()
  # Logic...
}
```

4. Fix the issue and re-test:

```
devtools::test()
```

5. Remove debugging artifacts and run final checks:

```
devtools::check()
```

3.9 Conclusion

Debugging `devtools::test()` errors involves interpreting error messages, isolating failing tests, and systematically diagnosing issues in your code. With these steps, you can identify and fix problems effectively.
