

# Supplement to Shiny in Production

*2019-01-13*



# Contents

<b>1 Shiny in Production Workshop @ RStudio Conf 2019</b>	<b>5</b>
<b>2 Introduction to Shiny in Production</b>	<b>7</b>
2.1 Workshop Objectives . . . . .	7
2.2 Workshop Infrastructure . . . . .	8
<b>3 Introduction to the Application</b>	<b>9</b>
3.1 Activity: Explore the Application . . . . .	9
<b>4 Application Testing: shinytest</b>	<b>11</b>
<b>5 Profiling: “The most important thing”</b>	<b>17</b>
<b>6 Deployment</b>	<b>21</b>
6.1 RStudio Connect . . . . .	21
6.2 Packrat . . . . .	21
6.3 Extended Topics and Resources . . . . .	21
<b>7 Connecting to Data in Production</b>	<b>23</b>
<b>8 Load Testing</b>	<b>25</b>
<b>9 Plot Caching</b>	<b>29</b>
9.1 Shiny Plot Caching Overview . . . . .	29
<b>10 Scaling</b>	<b>33</b>
10.1 Application Scaling 101 . . . . .	33
10.2 RStudio Connect Performance Settings . . . . .	33
<b>11 DevOps Philosophy &amp; Tooling</b>	<b>39</b>
<b>12 Production Case Studies</b>	<b>43</b>

<b>13 Alternatives to Shiny</b>	<b>45</b>
13.1 Plumber . . . . .	45
13.2 R Markdown . . . . .	47
<b>14 Shiny Async</b>	<b>51</b>

# Chapter 1

## Shiny in Production Workshop @ RStudio Conf 2019

This document is full of supplemental resources and content from the Shiny in Production Workshop delivered at rstudio::conf 2019.

The **bookdown** package can be installed from CRAN or Github:

```
install.packages("bookdown")
# or the development version
# devtools::install_github("rstudio/bookdown")
```



# Chapter 2

## Introduction to Shiny in Production

### Why are we here?

Shiny applications are being deployed in high-value, customer-facing, and/or enterprise-wide scenarios. Unfortunately, they are often being done without the benefit of best practices. This workshop will help you and/or your IT colleagues who support your data scientists learn how to accelerate a successful Shiny application deployment in production scenarios.

Over the past year, software developers at RStudio have been working hard to dispel rumors that Shiny “isn’t ready and can’t run in production”. They’ve built a bunch of cool new tools that are useful in preparing applications for production and understanding how to configure and scale them for optimized performance and user experience.

This workshop will cover all these new tools for shiny development as well as the equally important logistical pieces of a production story:

- What does production infrastructure and tooling look like for Shiny apps?
- How do we get Shiny apps from development into production?
- How are Shiny apps maintained in production?

When developers begin to think of infrastructure as part of their application, stability and performance become normative. - Jeff Geerling “Ansible for DevOps”

### 2.1 Workshop Objectives

#### Running Shiny in Production

- Is Shiny ready for production?
- What does it take to get there?
- What do you need to know?
- What tools are available?
- How do you develop a workflow?

#### Understand the importance of incremental changes and testing

- Version control

- Tests for package upgrades
- Use of separate environments for staging and production
- Incorporating automated testing into a development workflow: `shinytest`

### Data product tradeoffs

- What are the advantages to using Shiny vs Plumber vs R Markdown
- What is the difference between a stateless Plumber API and a Shiny Session?

### Development vs. Production environment considerations

- Defining a data model
  - Working with databases
- Tools for understanding application performance
  - `shinyloadtest`
  - `profvis`
- Tools for improving application performance
  - Plot caching
  - Synchronous vs asynchronous paradigms: `async`

### Deployment architecture and tools

- Introduction to analytic infrastructure
- Configuration management
- Resources for scaling horizontally

## 2.2 Workshop Infrastructure

- RStudio Connect
- PostgreSQL
- RStudio Server Pro

Instructions for accessing the classroom environment are available in the workshop slide deck.

# Chapter 3

## Introduction to the Application

### Every Application has an Origin Story

Data Scientists at RStudio University have discovered that there are trackable traits and behaviors students engage in that have been predictive of the desired 4-year graduation track.

They have built a shiny application that can be used by the very data-savvy advisors at this illustrious institution to identify students in need of guidance and show them the top behavioral factors driving individual predictions coming out of the model.

The POC was a smashing success - but now *our advisors actually want to use this thing for real.*

- We've developed a nice app
- We want to put it into production
- We want confidence that it will perform well in production, both now and in the future

### 3.1 Activity: Explore the Application

*First: Open app.R and Run the Application*

**Discussion:** Explore the Application

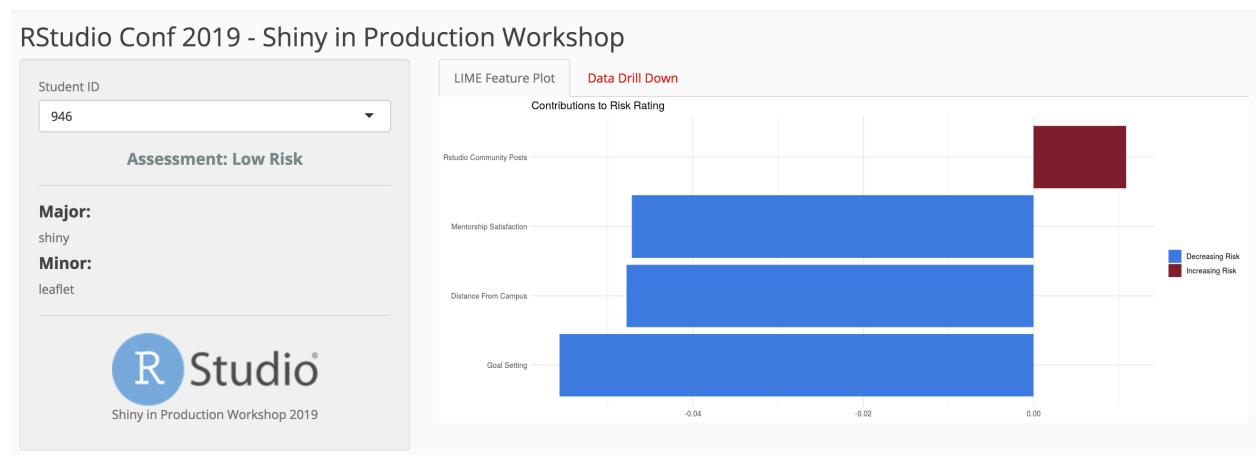


Figure 3.1: App Preview

## Checklist

- Tests
- Performance Optimization
- Environment (Packages) Management
- Data Access
- Deployment Hand Off
- Scaling
- Monitoring

Figure 3.2: Ideas for Getting Started: Checklist

- Are there any parts of the app code that don't make sense?
- Is this application ready for production?
- How would you define "production"?
- What insights would be useful to have about the application before we try to deploy it?

*Additional Discussion:*

- What is your current process for taking applications into production?

### Deliverable: Start a Plan

- Create a checklist
- Outline the steps you might take to put this application into production

### Checklist for Taking Applications into Production

A high-level Checklist to build off of:

# Chapter 4

## Application Testing: `shinytest`

- You've developed a nice app
- You've put it in production
- You want to be confident that it will keep running in the future

### Things that can change/break a Shiny application

- Modifying code
- Upgrading the `shiny` package
- Upgrading other packages
- Upgrading R
- External data source changes or fails

#### 4.0.0.1 Testing Options

- Manual testing
  - time intensive
  - inconsistent
- Automated testing (hard)
  - web browser
  - simulated user interactions
  - tests for graphical elements

Enter: Snapshot-based testing Webinar Slide Deck

- Can be easier to create tests
- Can test an entire application
- More sensitive to spurious changes

#### 4.0.0.2 Automated Testing for Shiny Apps

- Blog

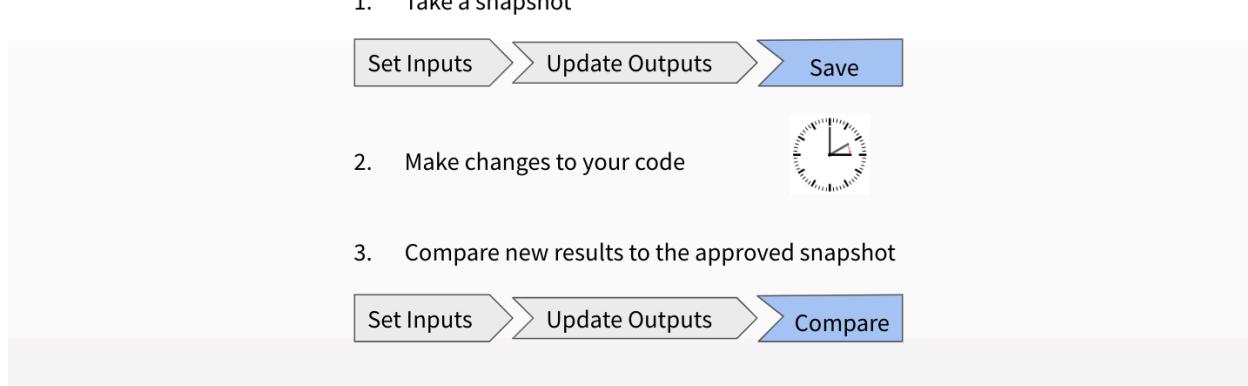


Figure 4.1: Snapshot Testing

- **Create a test:** Record user interactions with the app, saves them in a *test script*.
- **Make baseline (expected) snapshots:** Run the test script, which replays the interactions on a headless browser. This takes snapshots of application state along the way and saves them for later comparison.
- **Do your work:** Modify your app, modify your data, upgrade packages, upgrade R.
- **Re-run the test script and compare:** Run the test script again and take snapshots, then compare the new snapshots to the expected snapshots.

Figure 4.2: shinytest procedure

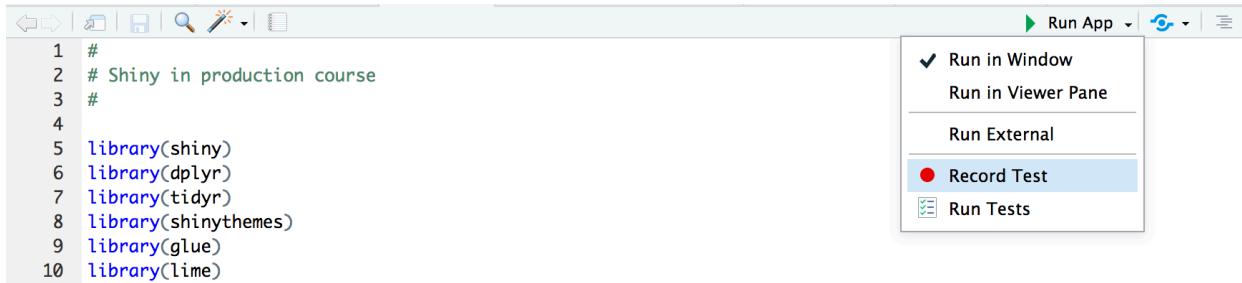


Figure 4.3: Record Test Button

**shinytest** is a package (available on CRAN) to perform automated testing for Shiny apps.

Basic **shinytest** procedure:

Support for **shinytest** is available in RStudio v1.2 preview.

### Installation

```
'install.packages("shinytest")'
```

**Note:** When running **shinytest** for the first time, you may be prompted by the RStudio IDE or package warning messages to install some dependencies. **shinytest** requires a headless web browser (PhantomJS) to record and run tests.

- To install it, run `shinytest::installDependencies()`
- If it is installed, make sure the `phantomjs` executable can be found via the PATH variable.

### Record Tests

- Run `recordTest()` to launch the app in a test recorder.
- Create the tests by interacting with the application - this will allow the recorder to snapshot the application state at various points.
- Quit the test recorder. This action will trigger the following events:
  - The test script will be saved as a .R file in a subdirectory of the application named `tests/`.
  - If you are running in the RStudio IDE, it will automatically open this file in the editor.
  - The test script will be run, and the snapshots will be saved in a subdirectory of the `tests/` directory.

To record tests from R, run the following:

```
library(shinytest)

recordTest("path/to/the/app")  #Replace with the correct path
```

To record tests from RStudio v1.2, when an application file (app.R, server.R, ui.R or global.R) is open in the editor, a button labeled *Run App* will appear at the top of the editor pane. Click on the small black triangle next to this button to reveal the menu of extended options.

This launches the Shiny application to be tested in a separate R process. We'll refer to this as the **target app**. At the same time, the current R process launches a special Shiny application which displays the target app in an iframe along with some controls. We'll refer to this as the **recorder app**. You should see something like this:

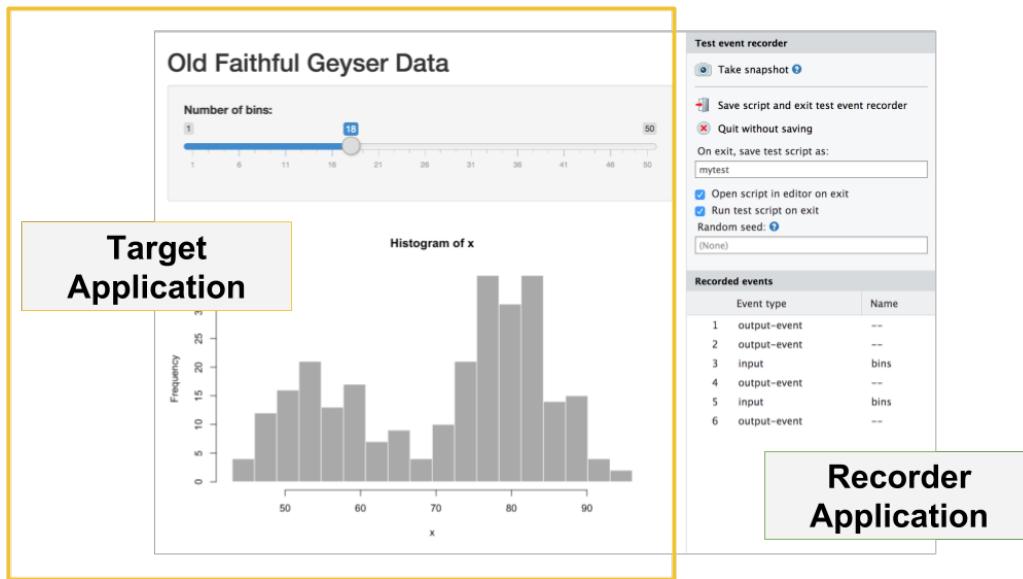


Figure 4.4: Target and Recorder App iframe

The panel on the right displays some controls for the test recorder, as well as a list of **recorded events**. As you interact with the target app, you will see those interactions appear in the recorded events list.

For testing a Shiny application, interacting with the inputs is only one part of the equation. It's also necessary to check that the application produces the correct outputs. This is accomplished by taking **snapshots** of the application's state.

To take a snapshot of the application's state, click the *Take snapshot* button on the recorder app. This will record all input values, output values, and exported values.

### Running Tests

When you quit the test recorder, it will automatically run the test script. There are three separate components involved in running tests:

1. First is the **test driver**. This is the R process that coordinates the testing and controls the web browser. When working on creating tests interactively, this is the R process that you use.
2. Next is the **Shiny process**, also known as the **server**. This is the R process that runs the target Shiny application.
3. Finally, there is the **web browser**, also known as the **client**, which connects to the server. This is a headless web browser - one which renders the web page internally, but doesn't display the content to the screen (PhantomJS).

So, when you exit the test recorder, it will by default automatically run the test script and print something like this:

```
Saved test code to /path/to/app/tests/mytest.R
Running mytest.R
===== Comparing mytest ...
No existing snapshots at mytest-expected/. This is a first run of tests.
```

```
Updating baseline snapshot at tests/mytest-expected
```



Figure 4.5: View Failed Tests

```
Renaming tests/mytest-current
=> tests/mytest-expected.
```

This is the result of running `testApp()`, which can also be manually run by providing the desired application and test like this:

```
testApp("exampleApp", "mytest")
```

The built-in integration with RStudio v1.2 provides `Run Tests` as a drop down menu option in your Shiny app source file (see it located under the *Record Test* option in the screenshot above).

### Subsequent Test Runs

After the initial test run, you can continue to run the tests to check for changes in application behavior.

If there are any differences between current and expected results, the test output will look something like this:

```
Running mytest.R
=====
Comparing mytest ...
Differences detected between mytest-current/ and mytest-expected/:

  Name      Status
  001.json  != Files differ
  001.png   != Files differ
Would you like to view the differences between expected and current results [y/n]?
```

To view failed tests in the RStudio IDE, go to the *Build tab* and make sure the *issues toggle* is selected:

For each test with different results, you can see the differences between the expected and current results.

### Testing Code

The `shinytest` package was created for testing Shiny applications on the interaction-level. To test Shiny code and functions, we suggest using the `testthat` package.

[Info GitHub](#)

Learn about testing and how to setup test workflow and structure: R packages by Hadley Wickham

#### 4.0.0.3 Workshop Exercise: `shinytest`

**First: Open `app.R` and use ‘Record Test’**

**Discussion:**

*Understanding shinytest*

- What does the recording file create?
- What challenges might our app pose to testing?
- Can you run the tests and get a success?
- How might you automate tests?

**Deliverable:** Try `shinytest`

- A set of tests that can catch unintentional changes to our app.

# Chapter 5

## Profiling: “The most important thing”

Profvis is a tool for helping you to understand how R spends its time. It provides a interactive graphical interface for visualizing data from `Rprof`, R’s built-in tool for collecting profiling data.

- Profvis Home

To install profvis from CRAN: `install.packages("profvis")`

The RStudio IDE includes integrated support for profiling with profvis. These features are available in the current Preview Release of RStudio.

*Profvis: Profiling tools for faster R code*

- How do I make my R code faster?
- Why is my R code slow?

Each block in the flame graph represents a call to a function, or possibly multiple calls to the same function. The width of the block is proportional to the amount of time spent in that function. When a function calls another function, another block is added on top of it in the flame graph.

The profiling data has some limitations: some internal R functions don’t show up in the flame graph, and it offers no insight into code that’s implemented in languages other than R (e.g. C, C++, or Fortran).

Profvis is interactive. You can try the following:

- As you mouse over the flame graph, information about each block will show in the info box.
- Yellow flame graph blocks have corresponding lines of code on the left. (White blocks represent code where profvis doesn’t have the source code – for example, in base R and in R packages. But see this FAQ if you want package code to show up in the code panel.) If you mouse over a yellow block, the corresponding line of code will be highlighted. Note that the highlighted line of code is where the yellow function is called from, not the content of that function.
- If you mouse over a line of code, all flame graph blocks that were called from that line will be highlighted.
- Click on a block or line of code to lock the current highlighting. Click on the background, or again on that same item to unlock the highlighting. Click on another item to lock on that item.
- Use the mouse scroll wheel or trackpad’s scroll gesture to zoom in or out in the x direction.
- Click and drag on the flame graph to pan up, down, left, right.
- Double-click on the background to zoom the x axis to its original extent.

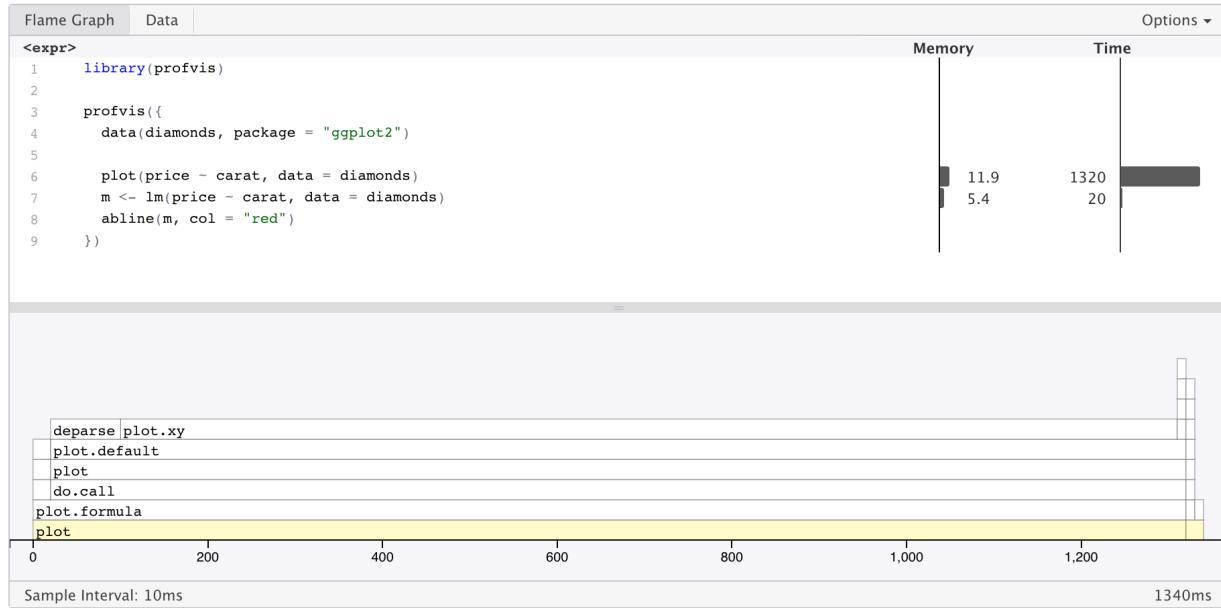


Figure 5.1: Profvis in Use

- Double-click on a flamegraph block to zoom the x axis the width of that block.

#### Things to remember:

- Understand how the sampling profiler works
- Understand the `profvis` interface
- Sometimes performance bottlenecks are counterintuitive!

#### 5.0.0.1 Workshop Exercise: Profiling

##### First: Create a profile of our app

##### Discussion:

*Understanding app performance*

- Where does our app spend most of its time?
- Do any parts of the profile surprise you?
- If you test the app again, do you get the same results?

##### *Additional Group Discussion*

- Is the call to `gt` slow?
- Why does `getStudentNum` take so much longer than `getStudentBin`? How could we speed this up?

#### Deliverable: Optimization Recommendations

- One recommendation for how we could speed up our app.

### 5.0.0.2 References and Resources

- Make Shiny Fast by Alan Dipert
- Profvis Webinar Slides
- Shiny Dev Center Article: Profiling your Shiny app



# Chapter 6

## Deployment

### 6.1 RStudio Connect

It doesn't matter how great your analysis is unless you can explain it to others: you need to communicate your results. - Grolemund & Wickham in *R For Data Science*

#### 6.1.0.1 Workshop Exercise: Initial Deployment

**First:** Login to RStudio Connect

**Discussion:**

*Pre-deployment Brainstorm*

- What is our goal in deploying this code?
- What does our code depend on locally?
- What needs to deploy with our code?

**Deliverable:** Deployed App

Press the publish button in RStudio:

- Link to your Connect account
- Select the files to publish (do NOT use the default to publish all the files)
- Poke around at the results

*Note: We are expecting an error!*

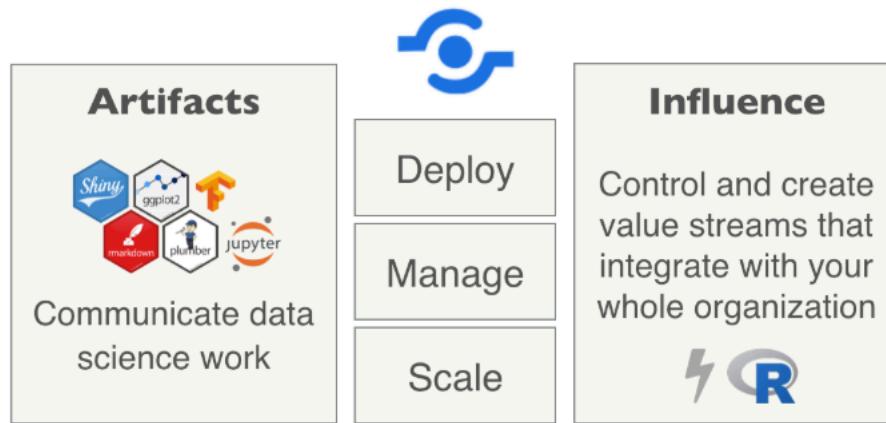
---

### 6.2 Packrat

### 6.3 Extended Topics and Resources

#### 6.3.1 Programmatic Deployment

- RStudio Connect



**RStudio Connect is a publishing platform that makes sharing data science simple**

Figure 6.1: RStudio Connect

### 6.3.2 Collaborative Publishing

- RStudio Connect

User Guide Resource: Collaboration

Some data products will have multiple authors and collaborators who are responsible for managing the content deployed to RStudio Connect. The first step to collaboration is sharing and working together on code. We recommend using a version control tool like Git to coordinate collaboration across many users.

# Chapter 7

## Connecting to Data in Production

### 7.0.0.1 Databases

### 7.0.0.2 The `config` package

### 7.0.0.3 Workshop Exercise: Databases

**First: Read This!**

**Discussion:**

*Data Management*

- How does your organization connect to data?
- What data are we exposing to viewers? Is it appropriate for all viewers?
- What else could we manage with the `config` package?

**Deliverable: Running, Deployed App**

- Update config.yml and Redeploy
- Edit the Environment Variables in RStudio Connect

### 7.0.0.4 Revisit the Deployment Checklist

## Deploy Checklist

- ☒ Supporting Files (`use_models.R`, `*.RDS`)
- ☒ Database Connection
  - ☒ Database Driver
  - ☒ Configuration (Credentials, DSN, `config.yml`, etc)
  - ?
  - Think about data sensitivity & viewership

- R Version
- R Packages

Figure 7.1: Database Checklist

# Chapter 8

## Load Testing

Load testing helps developers and administrators estimate how many users their application can support. If an application requires tuning, load testing and load test result analysis can be used to identify performance bottlenecks and to guide changes to infrastructure, configuration, or code.

It's a common misconception that "*Shiny doesn't scale*". In actuality, properly-architected Shiny applications can be scaled horizontally, a fact which Sean Lopp was recently able to demonstrate at rstudio::conf 2018. We used `shinycannon` to simulate 10,000 concurrent users interacting with an application deployed to AWS. You can see a recording of Sean's talk and the load test demonstration here: [Scaling Shiny](#)

To perform a load test you'll need two pieces of software: `shinyloadtest` and `shinycannon`.

- `shinyloadtest` is an R package used to generate recordings and analyze results. You should install it on your development machine. [GitHub](#)
- `shinycannon` is the command-line replay tool. You can install it on your development machine for testing, but for best results we recommend installing it on a server, and preferably not the one the application under test is also on. [GitHub](#)

See the Load Testing Quickstart [Here](#).

### 8.0.0.1 Optimization Loop Methodology

- **Benchmark:** Use `shinyloadtest::record_session()` to record interaction, `shinycannon` to simulate users
- **Analyze:** Visualize and interpret the metrics
- **Recommend:** Propose ways for the capacity of the application to be increased
- **Optimize:** Implement recommendations and benchmark again. Repeat until satisfied

*Workflow:*

- Use 'shinyloadtest' to record a session with a Shiny app
- Generate load with `shinycannon`
- Analyze metrics with `shinyloadtest`
- Make changes to the Shiny application
- Generate load and analyze again

Load Testing Shiny Applications

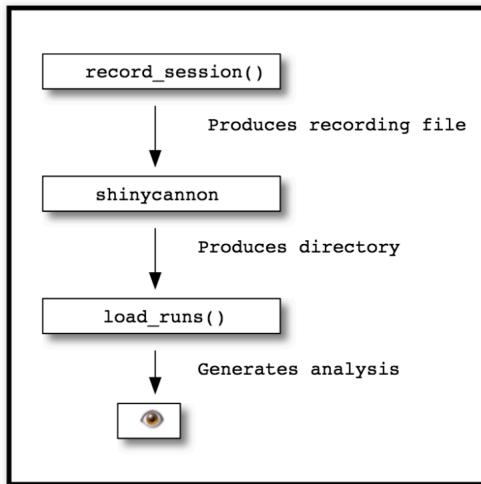


Figure 8.1: Load Test Workflow



Figure 8.2: Load Test Steps

### 8.0.0.2 Workshop Exercise: Load Testing

**First:** Open `runloadtest.R`, do the pre-run checklist

**Discussion:**

*Preparing for the load test*

- First, what is up with the pre-run checklist? Any idea why these steps are (currently) necessary?
- The `runloadtest.R` file is going to start with a baseline test of 1 and then a test of 25. Why is the baseline important?

**Deliverable:** Run the load test

- Follow the first set of commands in `runloadtest.R`
- How did the experience for 1 user compare to the experience for 25 users?

### 8.0.0.3 References and Resources

**Webinar:**

- Load testing Shiny by Alan Dipert
- Webinar Slides

**Vignettes:**

- Analyzing Load Test Logs
- Case Study: Scaling an Application
- Limitations of `shinyloadtest`
- Load Testing Authenticated Apps



# Chapter 9

## Plot Caching

### 9.1 Shiny Plot Caching Overview

Blog: Shiny 1.2.0: Plot caching - November 18, 2018

The Shiny 1.2.0 package release introduced *Plot Caching*, an important new tool for improving performance and scalability in Shiny applications.

**In a nutshell:** The term “caching” means that when time-consuming operations are performed, we can save (cache) the results so that the next time that operation is requested, instead of re-running that calculation, we instead go fetch the previously cached result. When applied appropriately, this “fetch” operation should take less time than the original calculation and improve application performance (and user experience) overall.

Shiny’s reactive expressions do some amount of caching by default, and you can use more explicit techniques to cache various data operations. Examples include use of the `memoise` package, or manually saving intermediate data frames to disk as CSV or RDS.

Plots are a very common and (potentially) expensive to compute type of output object in Shiny applications, which makes them a great candidate for caching. In theory, you could use `renderImage` to accomplish this, but because Shiny’s `renderPlot` function contains a lot of complex infrastructure code, it’s actually quite a difficult task.

Shiny v1.2.0 introduces a new function, `renderCachedPlot`, that makes it much easier to add plot caching to your application.

#### 9.1.1 When to use plot caching

A shiny app is a good candidate for plot caching if:

1. The app has plot outputs that are time-consuming to generate
2. These plots are a significant fraction of the total amount of time the app spends thinking
3. Most users are likely to request the same few plots

#### 9.1.2 Using `renderCachedPlot`

The following example is a simple, but computationally expensive, plot output:

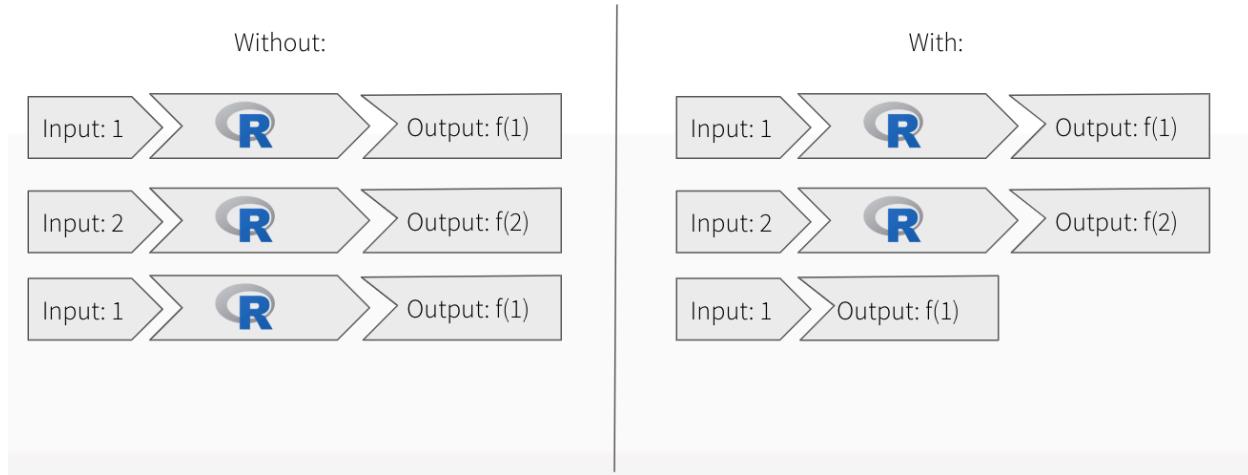


Figure 9.1: renderChachedPlot()

```
output$plot <- renderPlot({
  ggplot(diamonds, aes(carat, price, color = !!input$error_by)) +
    geom_point()
})
```

The `diamonds` dataset has 53,940 rows. This plot might take roughly 1580 milliseconds (1.58 seconds) to generate depending on the resources available. For a high traffic Shiny application in production, 1.58 seconds is likely slower than ideal.

Plot caching can be implemented in two steps:

1. Change `renderPlot` to `renderCachedPlot`
2. Provide a suitable `cacheKeyExpr`. This is an expression that Shiny will use to determine which invocations of `renderCachedPlot` should be considered equivalent to each other. In the example case, two plots with different `input$error_by` values can't be considered the "same" plot, so the `cacheKeyExpr` needs to be `input$error_by`

The example plot code would be updated like this:

```
output$plot <- renderCachedPlot({
  ggplot(diamonds, aes(carat, price, color = !!input$error_by)) +
    geom_point()
}, cacheKeyExpr = { input$error_by })
```

With these code changes, the first time a plot with a particular `input$error_by` is requested, it will take the normal amount of time. But the next time it is requested, it will be almost instant, as the previously rendered plot will be reused.

### 9.1.3 Workshop Exercise: Plot cache benchmarking

**First:** Update your app code to use `renderCachedPlot`

**Discussion:** *caching + shinyloadtest*

- How will introducing `renderCachedPlot` affect our load test experience?

*What is the performance comparison between tests?*

**Deliverable:** Re-run Load Test

- Redeploy the version of the app with `renderCachedPlot`
- Re-run the load test and compare the outputs (continue to follow along with `runloadtest.R`)

#### 9.1.4 Extended Topics and Resources

Shiny Article: Plot Caching by Winston Chang

##### 9.1.4.1 Plot Caching on RStudio Connect

Shiny can store cached plots in memory, on disk, or with another backend like Redis. There are also a number of options for limiting the size of the cache. Applications deployed on RStudio Connect should use a disk cache and specify a subdirectory of the application directory as the location for the cache. To do so, add this code to the top of your application:

```
library(shiny)
shinyOptions(cache = diskCache("./cache"))
```

This option ensures that cached plots will be saved and used across the multiple R processes that RStudio Connect runs in support of an application. In addition, this configuration results in the cache being deleted and reset when new versions of the application are deployed.



# Chapter 10

## Scaling

### 10.1 Application Scaling 101

### 10.2 RStudio Connect Performance Settings

#### Scaling and Performance Tuning in RStudio Connect

- Article Reference

RStudio Connect is built to scale content. Publishers and administrators have access to runtime settings to help tune and scale their applications and APIs. The primary concern for scaling content in RStudio Connect is understanding when and how R code will be executed on the server.

Content published on RStudio Connect can broadly fall into two categories:

1. Static / Batch Content

Static content is content a user can visit without requiring a running process. Examples include PDF documents, plots, and HTML files. HTML files can include interactive elements where the interactivity occurs on the client. RStudio Connect is able to update static content on a schedule.

2. Interactive Content

A powerful component of RStudio Connect is the ability to host data products that require backend processes during a user's visit. Examples include shiny applications and R Markdown documents with `runtime::shiny`, `plumber` APIs, and `TensorFlow` APIs. In the case of Shiny, an end user's browser (the client) is connected with an R process running on the server. When a user changes an input, the client sends a message to the server and a portion of R code is re-run. The server sends back the result as output. In the case of APIs, a client makes a request which is sent to a running process, and the results are sent back to the client.

For Interactive Content, RStudio Connect enables scaling through parameters that determine the content's scheduler.

## Scaling 101



Figure 10.1: Kitchen Scaling

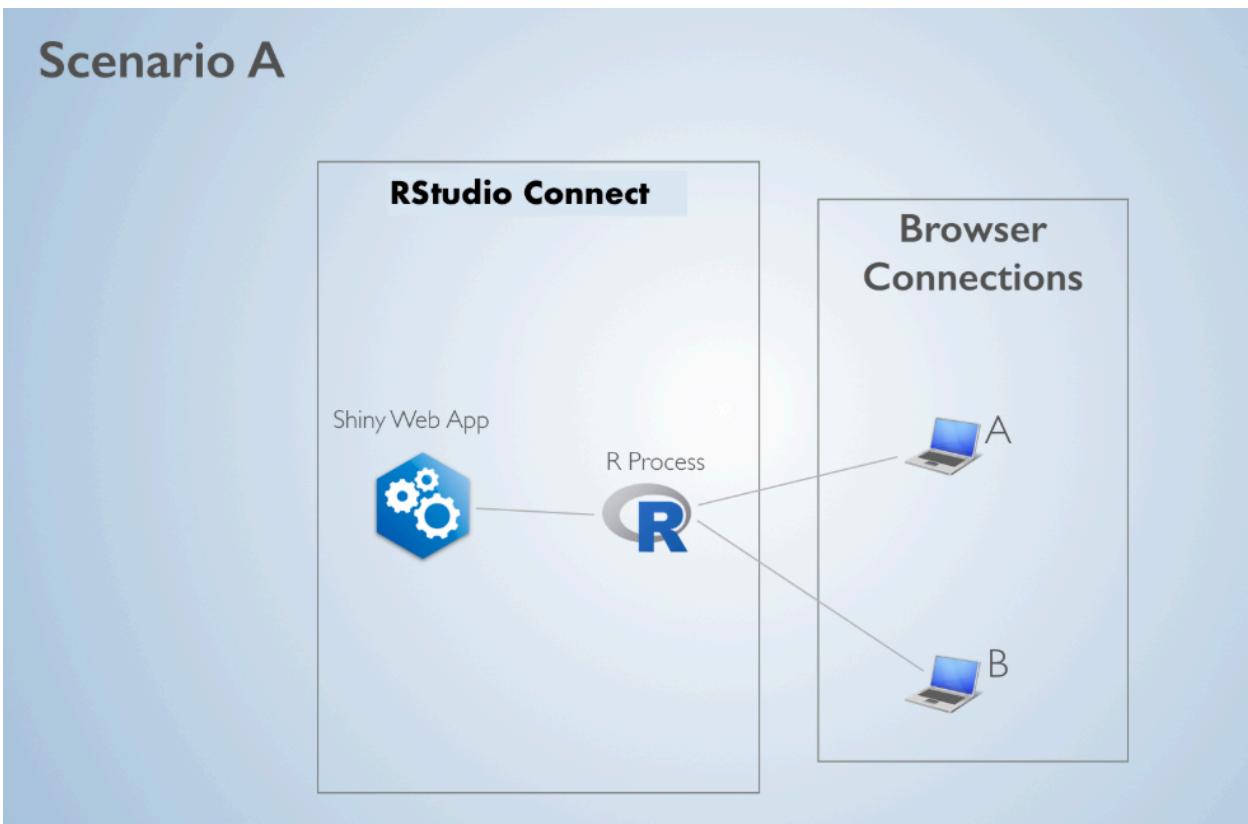


Figure 10.2: Scaling A

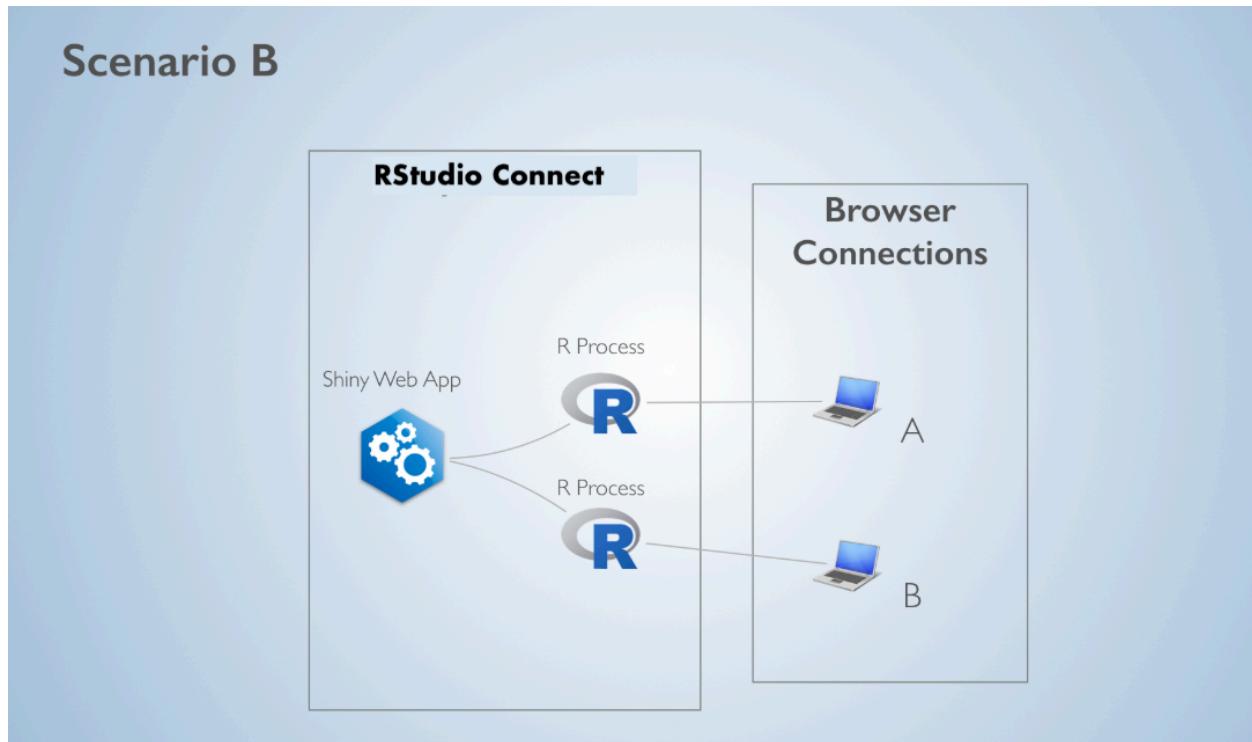


Figure 10.3: Scaling B

### 10.2.1 Content Scheduler

When a user requests content with Shiny components, RStudio Connect opens a channel between the user and an R process. Now, suppose a second user requests the same content. There are two potential scenarios:

In Scenario A, both users are linked to the same R process. Because R is single-threaded, if user A and user B both change an input and trigger R calculations, their requests will be handled sequentially. (User B will have to wait for user A's calculation to complete, and then for their own calculation to complete before they will see an updated output).

In Scenario B, Connect will link each user to their own R process. If user A and user B both change an input, their calculations will happen simultaneously.

Why wouldn't RStudio Connect always select Scenario B? The answer has to do with memory and initial load time. When 2 users are connected to the same R process, they get to share everything that is loaded outside of the server function.

To see this, consider when the different pieces of Shiny application code are executed:

This works because Shiny makes use of R's unique scoping rules (read more here). In Scenario B, all of the shiny code has to be re-run, including loading any globally available data. This means the memory usage is 2x what it would be in Scenario A. Additionally, spinning up an R process and executing all of the shiny code takes time. While the application is more responsive to both users after the web page is loaded, it will take longer for them to connect initially.

The scheduling parameters tell RStudio Connect to act somewhere in between Scenario A and Scenario B, to maximize the trade-off between app responsiveness and memory consumption/load time. The parameters also specify how long R processes and user connections should remain idle before timing out.

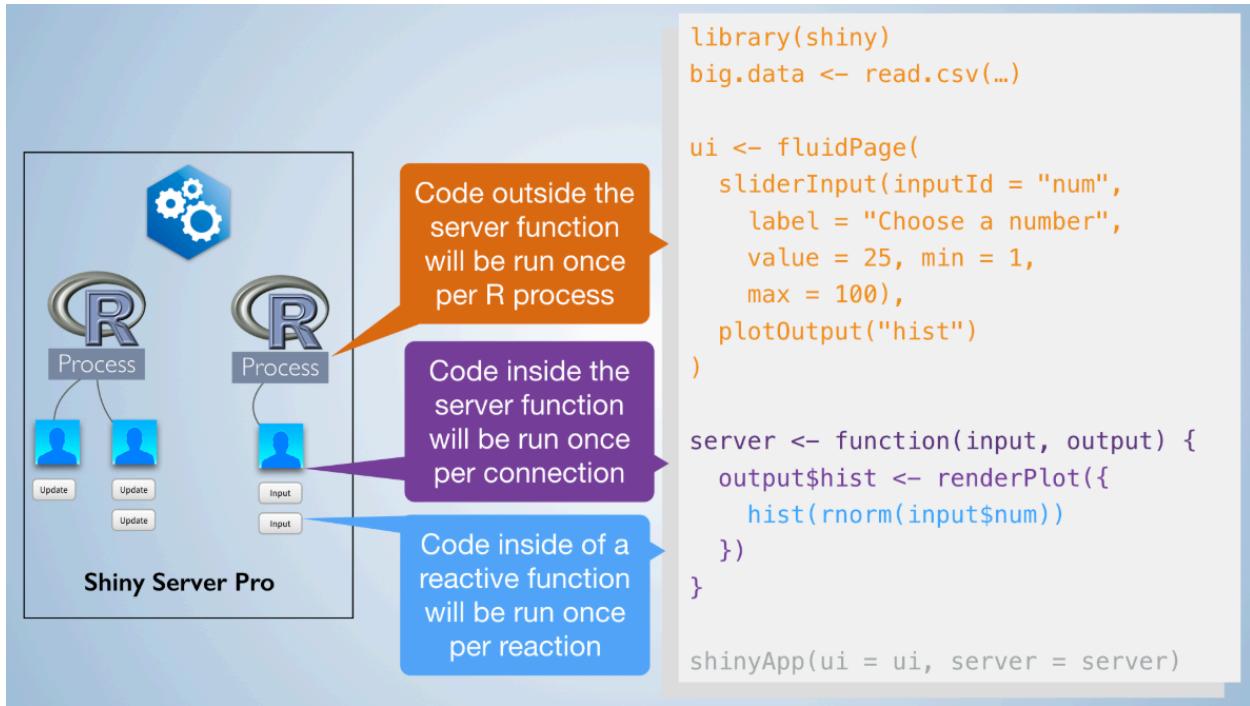


Figure 10.4: App Execution

### 10.2.2 Scheduling Parameters

**Max processes** - Determines the maximum number of processes that will be created. Max processes x Max connections per process = total number of concurrent connections. Default value is 3.

*Pick a value that will support the expected number of concurrent users or requests.*

**Min processes** - Determines the minimum number of processes that will always be running. For Shiny applications and plumber APIs, user requests only execute specific functions. R code outside of those functions can be run when the process starts before user requests are made, which can dramatically speed up response time.

*Pick a number close to Max processes if your application or API pre-loads a large amount of data to be shared by every user. Pick a small number to minimize the amount of memory consumed.*

**Max connections per process** - The maximum number of connections per process. Default value is 20.

*Pick a small number if your content involves heavy computation. Pick a larger number if your content shares data between users. (e.g. Pick a large number if your content takes a long time to load, but after loading is very fast.)*

**Load factor** - Determines how aggressively new processes will be created. A value close to 0 means new processes will be spun up aggressively to try and keep the number of connections per process small. A value close to 1 means the number of connections per process will be close to max connections. Default value is 0.5.

*Pick a small number if your content loads quickly but involves expensive computation. Pick a number closer to 1 if your content loads slowly, but after loading is fast OR if you want to minimize the amount of memory.*

#### 10.2.2.1 Workshop Exercise 1

First: Open the runtime settings for your app

**Discussion:**

- What do each of the settings mean?
- What changes to the settings might affect our app?

**Deliverable: Re-run Load Test**

Test your hypothesis by changing the runtime settings and then re-run the load test. Compare the results to the prior test (continue to follow along with `runloadtest.R`)

**10.2.2.2 Workshop Exercise 2****First: Open the admin dashboard in RStudio Connect****Discussion:***Multi-Tenant Environment*

- What are the tradeoffs for modifying the scaling settings?
- What else might you need to consider when sharing infrastructure?
- How does these concerns change over time?

**Deliverable: Update checklist**

Create one pre-deployment checklist step and one every-2-months checklist step based on your discussion.

**10.2.3 High Availability and Horizontal Scaling**

Other considerations for widely accessed content are high availability and horizontal scaling. Both require content to be hosted by more than one server. RStudio Connect supports a cluster setup.

**10.2.4 R Markdown Documents with `runtime::shiny`**

R Markdown documents are a bit different. In essence, an Rmd with `runtime::shiny` specified places everything (data loading, UI creation, etc) inside of the `server` function. The implication is an Rmd with `runtime::shiny` will always consume more memory as users connect.

Prerendered Shiny Documents alleviate this problem.



# Chapter 11

## DevOps Philosophy & Tooling

DevOps is a self-help philosophy for IT and software development teams that work together. It seeks to address a core problem: **The Fear of Taking Code into Production**

The whole point of this workshop is to get you some experience with what it takes to get Shiny Applications running in

- Deployments/releases are high-risk events
- It takes a long time to deliver new features to customers
- There is no visibility into how code is deployed into production, therefore it is impossible to know what to optimize or fix in the system and/or people who seek to improve portions of workflow are likely not addressing true bottlenecks.
- Dev and Ops have seemingly conflicting goals and objectives

The Fix (ref: The DevOps Handbook) “The principles underpinning DevOps”:

1. Accelerate the delivery of work from Development to Operations to customers
  2. Commit to evolving ever safer systems of work through feedback
  3. Promote a high-trust culture and embrace continual learning and experimentation (risk-taking) in daily work
- 

### Flow

- Make work visible
- Limit Work in Progress (WIP)
- Reduce Batch Sizes
- Reduce the number of handoffs
- Continually identify and elevate constraints
- Eliminate hardships and waste in the value stream

Typical DevOps Transformation (pg. 22) - Environment creation - Code Deployment - Test setup and run - Overly tight architecture

### Feedback

- Working safely within complex systems

- Complex work is managed so that problems in design and operations are revealed
- Problems are swarmed and solved
- New local knowledge is exploited globally throughout the organization
- Leaders create leaders
  
- See problems as they occur
- Swarm and solve problems to build new knowledge
- Keep pushing quality closer to the source
- Enable optimizing for downstream work centers

## Continual Learning & Experimentation

- Enable organizational learning and a safety culture
- Institutionalize the improvement of daily work
- Transform local discoveries into global improvements
- Inject resilience patterns into daily work

Generative organizations (pg. 39) // What type of organization do you have? Comparison Activity - Information is actively sought - Messengers are trained - Responsibilities are shared - Bridging between teams is rewarded - Failure causes inquiry - New ideas are welcomed

Even more important than daily work is the improvement of daily work –Mike Orzen, Lean IT

### 11.0.0.1 DevOps Vocab 101

Deployment Lead Time

Hardships and Waste: - Partially done work - Extra processes - Extra features - Task switching - Waiting - Motion - Defects - Nonstandard or manual work - Heroics

### 11.0.0.2 Integrating Data Science and DevOps

#### 11.0.0.2.1 How does data science fit in with the DevOps philosophy?

It's not *your job* to understand operations and systems administration. And it's not IT's job to understand R programming and shiny app development. But it takes shared goals and a little empathy from both sides to get less painful results.

We can leverage (steal) the DevOps philosophy and apply it to manage code deployment handoffs between data science and IT.

Foundations:

What do your dev, test and production environments look like? - Do developers have access to production-like environments on their own workstations? - Can these environments be assembled on-demand? - Version Control - Infrastructure as Code

#### 11.0.0.2.2 Design and Management of Analytic Infrastructure (alternative architectures)

#### 11.0.0.2.3 Programmatic Deployment into RStudio Connect

#### 11.0.0.2.4 Deployment Release Patterns

- Environment based release patterns
- Application based release patterns

#### 11.0.0.3 Feature Toggles (Dark Launch)

##### 11.0.0.3.1 User meta-data

Shiny applications can access the username and groups of the current user through the session parameter of the shinyServer function.

Groups meta-data is populated when using password, or OAuth authentication. It is not populated with LDAP, PAM, or proxy authentication.

```
shinyServer(function(input, output, session) {
  output$username <- reactive({
    session$user
  })

  output$groups <- reactive({
    session$groups
  })
})
```

Your application could use this information to display customized messages or to enable functionality for a specific subset of users.

#### 11.0.0.4 Extended Topics & Resources

##### Shiny Application Usage (Instrumentation)

RStudio Connect exposes usage information about shiny applications that will tell you which ones have been used, by whom and for how long. It is available via an API that may be used by administrators or publishers. Publishers may retrieve information about their own applications only.

More information about using the API is included in the historical information chapter of the RStudio Connect Admin Guide.



## Chapter 12

# Production Case Studies

*Workshop Exercise: Case Study Discussions*

**First:** Pick one case study

**Discuss:**

- What tools does your admin team use?
- How would you explain the goals of a shiny app to DevOps?

**Deliverable:**

- Whiteboard (in a medium of your choice) what an architecture might look like for your case study.
- Delineate who (and what tool) is responsible for code, R, R packages.

### 12.0.0.1 Case Study A: Dev/Test/Prod

Before publishing an update to our app, we want to run user acceptance tests.

### 12.0.0.2 Case Study B: CI, Git, Chef

All of our production code must be deployed from Git.

### 12.0.0.3 Case Study C: Docker

We scale applications through Docker - how does Shiny fit in?

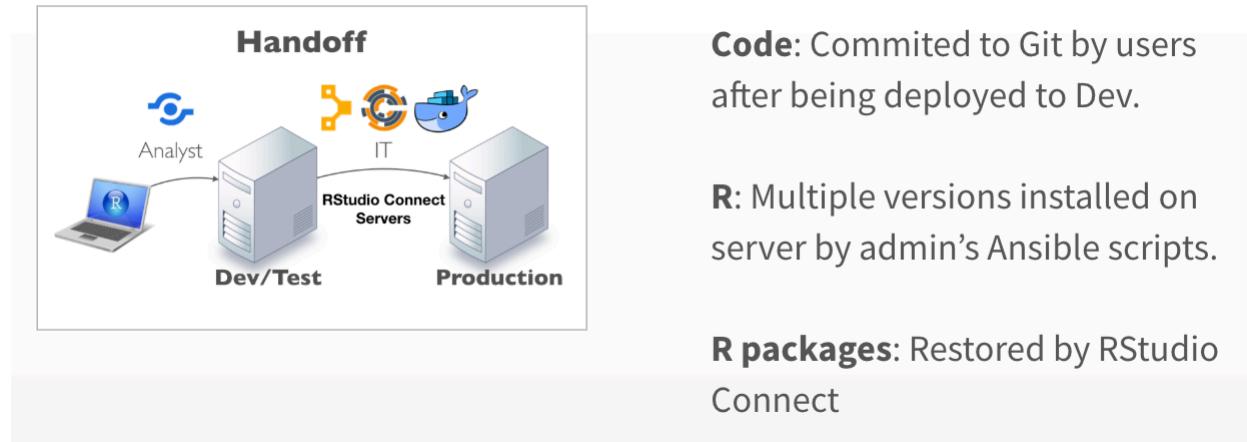


Figure 12.1: Case A Considerations

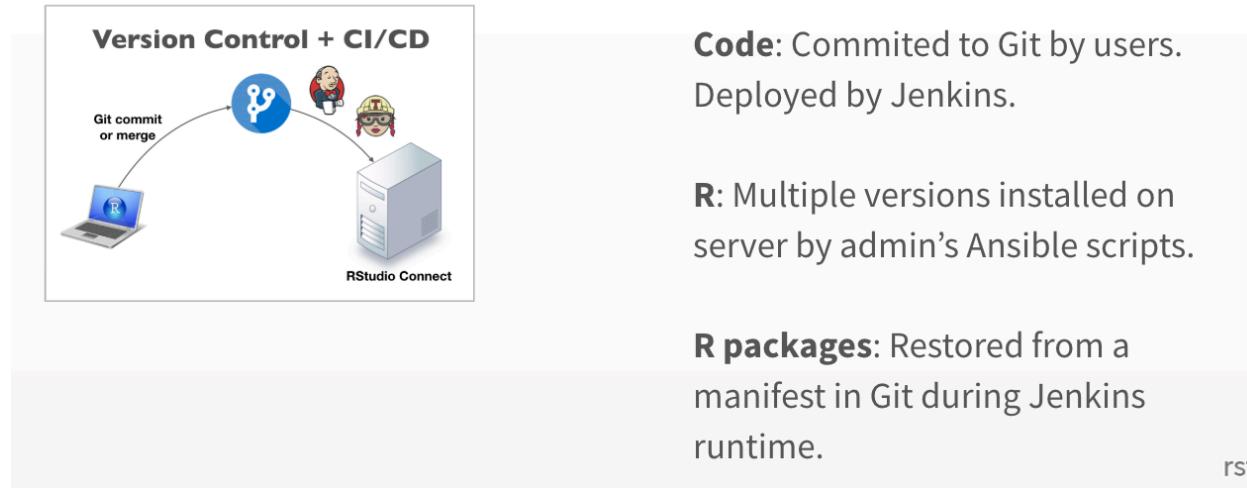


Figure 12.2: Case B Considerations

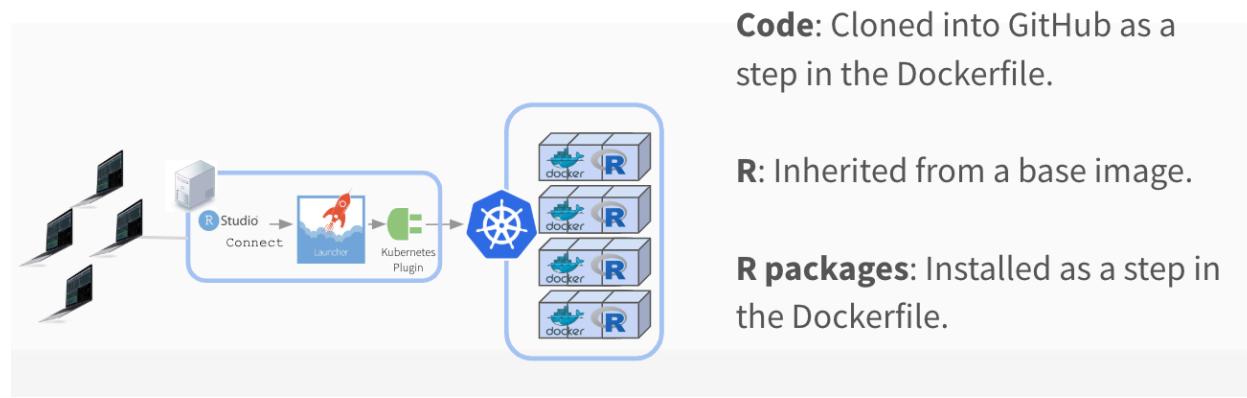


Figure 12.3: Case C Considerations

# Chapter 13

## Alternatives to Shiny

### 13.1 Plumber

*Could our student “scoring” be hosted outside of the Shiny app?*

#### 13.1.1 Intro to Plumber

##### Resources:

Plumber is an R package that converts your existing R code to a web API using a handful of special one-line comments.

What are Web APIs? For some, APIs (Application Programming Interface) are things heard of but seldom seen. However, whether seen or unseen, APIs are part of everyday digital life. In fact, you’ve likely used a web API from within R, even if you didn’t recognize it at the time! Several R packages are simply wrappers around popular web APIs, such as `tidycensus` and `gh`. Web APIs are a framework for sharing information across a network, most commonly through HTTP.

You can install the latest stable version from CRAN using the following command:

```
install.packages("plumber")
```

These comments allow plumber to make your R functions available as API endpoints. You can either prefix the comments with `#*` or `#'` but we recommend the former since `#'` will conflict with the Roxygen package.

Basic example:

1. Create a File: `plumber.R`

```
# plumber.R

#* Echo back the input
#* @param msg The message to echo
#* @get /echo
function(msg=""){
  list(msg = paste0("The message is: '", msg, "'"))}
```

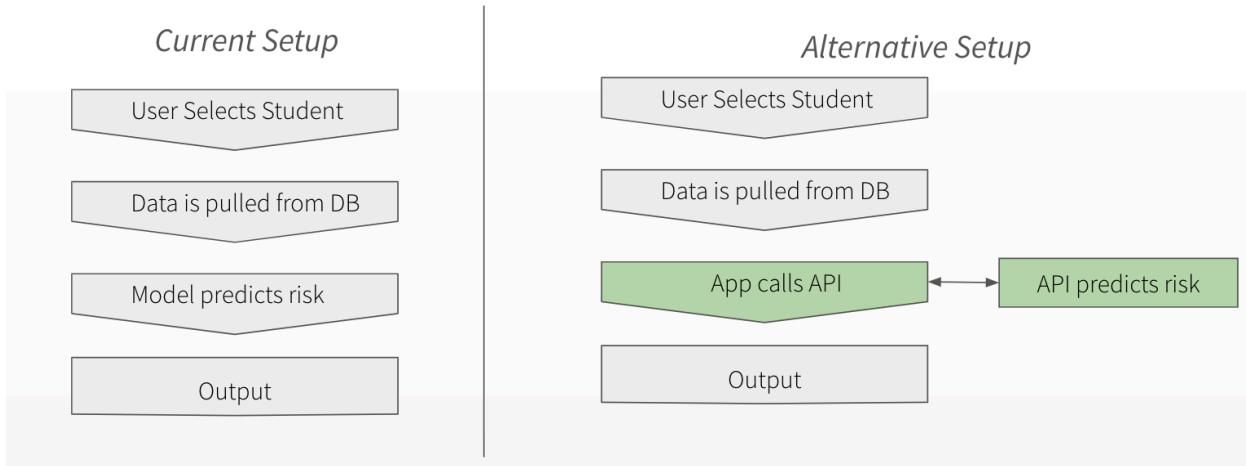


Figure 13.1: Plumber Scoring

```
}
##* Plot a histogram
##* @png
##* @get /plot
function(){
  rand <- rnorm(100)
  hist(rand)
}
```

```
##* Return the sum of two numbers
##* @param a The first number to add
##* @param b The second number to add
##* @post /sum
function(a, b){
  as.numeric(a) + as.numeric(b)
}
```

2. Serve the `plumber.R` file from the R console:

```
library(plumber)
r <- plumb("plumber.R") # Where 'plumber.R' is the location of the file shown above
r$run(port=8000)
```

You can visit this URL using a browser or a terminal to run your R function and get the results. For instance `http://localhost:8000/plot` will show you a histogram, and `http://localhost:8000/echo?msg=hello` will echo back the 'hello' message you provided.

3. Hit the endpoints (example: use `curl` from mac/linux terminal)

```
$ curl "http://localhost:8000/echo"
>{"msg":["The message is: ''"]}

$ curl "http://localhost:8000/echo?msg=hello"
```

```
{"msg": ["The message is: 'hello'"]}

$ curl --data "a=4&b=3" "http://localhost:8000/sum"
[7]
```

### 13.1.1.1 References and Resources

- Reference: Plumber Docs
- Plumber Integration in RStudio 1.2
- RVViews Blog: REST APIs and Plumber by James Blair
- Video: Turning your R code into an API by Jeff Allen
- Webinar: Plumbing APIs with Plumber by Jeff Allen

### 13.1.2 Workshop Exercise: Plumber

**First:** Try creating your own plumber API in the IDE (New File)

**Discussion:**

*Consider our new architecture*

Does my R functionality need to be accessed by other systems?

- Why might a RESTful API be easier to scale than a Shiny app?
  - Requests are stateless, so it is easier for R processes to come and go.
- What benefits come from pulling the modeling code out of our app?
  - We can update the model independently of the app, e.g. if we wanted to retrain.

**Deliverable:**

- Add an item to our production checklist to consider plumber.
- Add 2 decision criteria to the checklist to decide when a plumber API would be useful.

## 13.2 R Markdown

R Markdown is an open-source R package that turns your analyses into high quality documents, reports, presentations and dashboards. An R Markdown document is written in markdown (an easy-to-write plain text format) and contains chunks of embedded R code. R Markdown documents are fully reproducible and support dozens of output formats including HTML, PDF, and Microsoft Word documents.

R Markdown files are designed to be used with the `rmarkdown` package. `rmarkdown` comes installed with the RStudio IDE, but you can acquire your own copy of `rmarkdown` from CRAN with the command:

```
install.packages("rmarkdown")
```

R Markdown reports rely on three frameworks:

1. `markdown` for formatted text
2. `knitr` for embedded R code
3. `YAML` for render parameters

*Scheduled data updates, email distribution, and client side interactivity.*

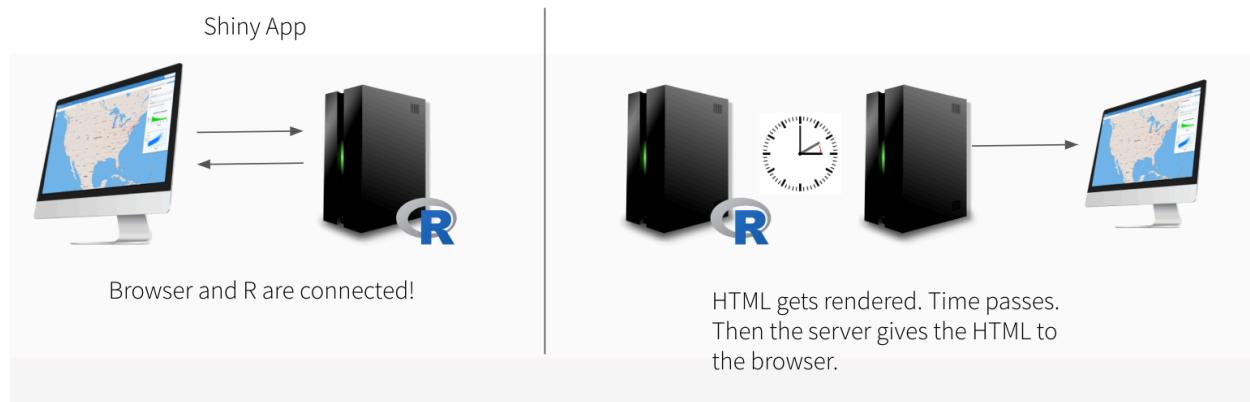


Figure 13.2: Publishing R Markdown

To create an R Markdown report, open a plain text file and save it with the extension .Rmd. You can open a plain text file in your scripts editor by clicking File > New File > Text File in the RStudio toolbar.

### Publishing R Markdown to RStudio Connect

- **Publishing Destination**

When publishing documents to RStudio Connect, you may encounter other deployment options depending on your content.

- RPubs documents are (1) always public, (2) always self-contained, and (3) cannot contain any Shiny content.
- Choose “RStudio Connect” to publish to your RStudio Connect server.

- **Publish Source Code**

Publishing the document with source code means that your R Markdown file (.Rmd) will be deployed to RStudio Connect. This file is rendered (usually to HTML) on the server.

Publishing only the finished document means that the HTML file you rendered locally is deployed to RStudio Connect.

We recommend publishing your documents with source code, as it allows you to re-render the document with RStudio Connect (on a weekly schedule, for example). If the document cannot be rendered by RStudio Connect because of files or data sources that are unavailable on the server, choose “Publish finished document only” so others can view your work.

- **Document Selection**

It is possible to link together multiple R Markdown documents to make a multi-page document, so this is your chance to indicate that you’ve done this, and to publish all the documents at once. In most cases however you’ll want to publish just the current document.

- RStudio Connect Publishing Guide
- R Markdown

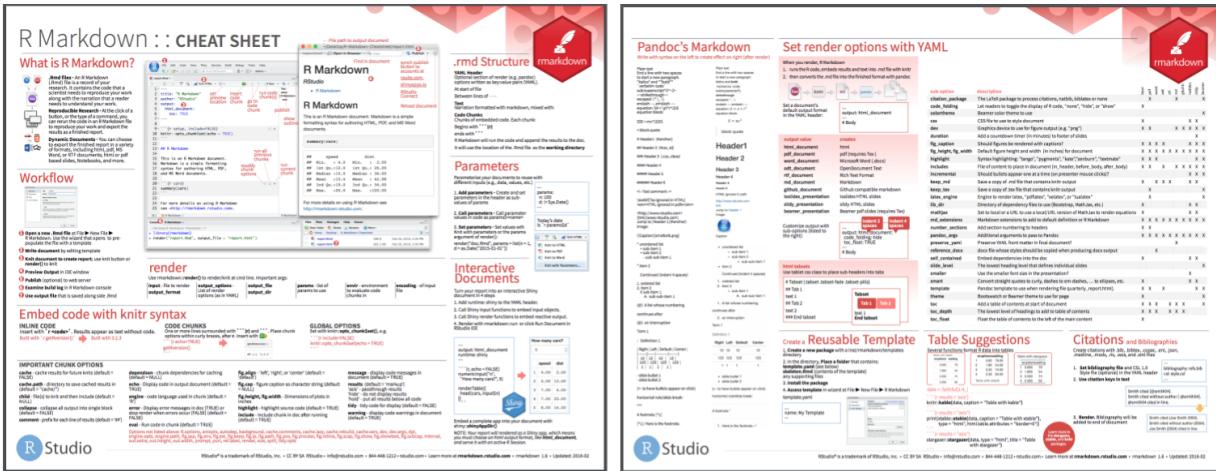


Figure 13.3: R Markdown Cheatsheet Preview

### 13.2.0.1 References and Resources

- R Markdown in the RStudio Connect User Guide
  - R Markdown Reference Articles
  - RVViews Blog: Communicating results with R Markdown by Nathan Stephens
  - RVViews Blog: Enterprise Dashboards with R Markdown by Nathan Stephens

### 13.2.1 Workshop Exercise: R Markdown

First: Create your own R markdown document (New File)

### **Discussion:**

## *Consider R Markdown*

- How does an R Markdown document scale?
  - What types of interactivity can be added to R Markdown?

## **Deliverable:**

- Create a decision matrix comparing Shiny to R Markdown

<i>Alternatives Criteria</i>	Shiny	R Markdown
# of Inputs	Users need multiple levels of dependent inputs to get an answer	All inputs can be specified up front
Update Frequency	Updates need to happen in real time	Updates occur on a regular basis
Output	Users come to the system to do exploratory Q&A, the outputs are the results	Users want a copy of the results or want results distributed to their email
Interactivity	Interactivity depends on lots of data	Interactivity can be based on data sent to the browser

Figure 13.4: Rmd Exercise

# Chapter 14

## Shiny Async

### 14.0.0.1 References and Resources

- Webinar - Scaling Shiny apps with asynchronous programming
- Webinar Slides
- Shiny Dev Center Article: Improving scalability with async programming