## Connectivity

Tableau comes with two options with regards to connecting to data. First, you can connect to a local file – remote server. The options there are far too many to list, so here's a screenshot of the current offerings (I



So yeah, there are a lot of options with Tableau. On the other side, R Shiny uses R as the programming Is competitor, the options here are endless. A simple Google search will yield either a premade library or an can sometimes have an edge. For instance, with R Shiny you can load and analyze gene expression data

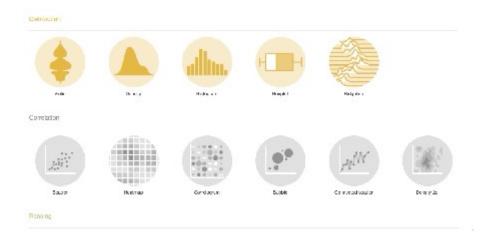
Winner (Connectivity): Tie, with an edge for R Shir

## **Chart Types**

Once again, Tableau doesn't come up short here, with a plethora of visualization options such as bar cha geographical plots. Refer to the image below for the full list:

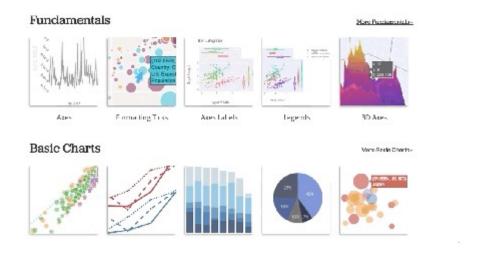


Still, R Shiny demolishes the competition in terms of optionality for chart types. The most widely used visu visualizations you can make with ggplot2:



Source: https://www.r-graph-gallery.com

And here are some cool things that you can do with Plotly:



Source: https://plotly.com/r/

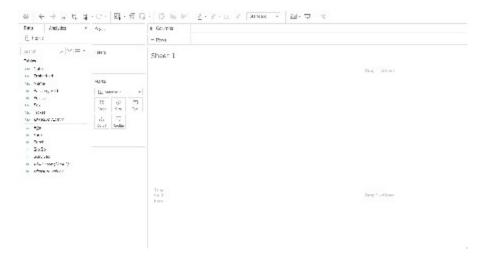
It's a no brainer that *R Shiny* wins this battle in terms of chart options overall.

Winner (Chart Types): R

# **Ease of Use: Simple Charts**

We'll now see the extent of the trouble you'll have to go through to create a simple bar chart, both in Table download the CSV file linked here if you're following along. To be more precise, we'll create a bar chart co

Let's start with Tableau. After the data imports, it's relatively easy to create a visualization. Just follow the



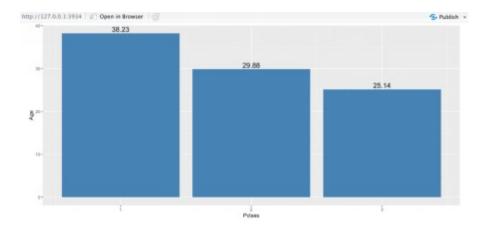
That was very easy, as Tableau was designed to be intuitive and simple to use for non-technical people.

Creating a simple chart is quite a different story with R Shiny. We need to write actual code to get a simple data in a concise, visualizable format. Next, we need to introduce some amount of boilerplate to manage projects, but a real obstacle for someone non-technical to create a simple chart.

#### Refer to the code below:

```
library(shiny)
library(ggplot2)
ui <- fluidPage(
    plotOutput('bar')
server <- function(input, output) {</pre>
    output$bar <- renderPlot({</pre>
        pclass age data <- titanic %>%
            select(Pclass, Age) %>%
            group by(Pclass) %>%
            summarise(Age = round(mean(Age, na.rm=TRUE), 2))
        ggplot(data=pclass_age_data, aes(x=Pclass, y=Age)) +
            geom bar(stat='identity', fill='steelblue') +
            geom_text(aes(label=Age), vjust=-0.3, size=5)
    })
}
shinyApp(ui=ui, server=server)
```

And here are the results:



As you can see, there's no point in using *R Shiny* if all you need are simple and straightforward data visual margin. If you still want to use R for charts and you're willing to put the effort in to have something fully cu simple charts with R can be an onerous task for non-technical people. We recommend Tableau for these

Winner (Simple Charts): 1

### **Ease of Use: Simple Dashboards**

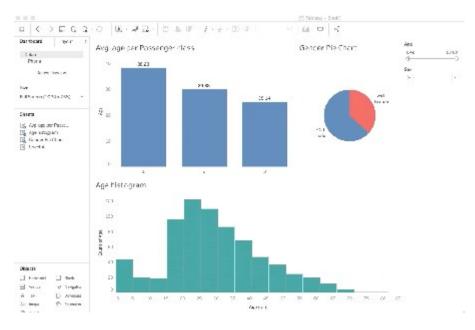
Alright, let's see who will win in *The Battle of the Dashboards*. We won't go too complex here, as dashboards charts we'll put into our dashboard:

- Bar chart average age per passenger class
- Pie chart genders
- Histogram ages

Also, all three charts should update when the user changes one of two filters:

- Age slider range slider from minimum to the maximum age of the passengers
- Gender picker option to exclude males or females from the visualizations

Here's what the final dashboard looks like in *Tableau*:



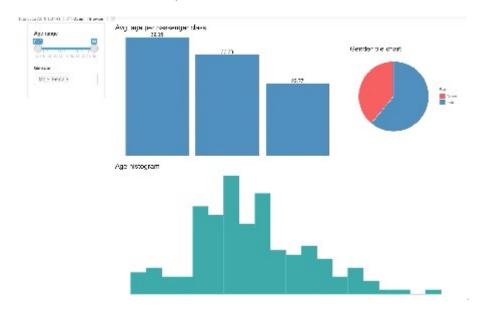
This entire dashboard required 0 lines of code. We just needed a couple of minutes of clicking in the right actual R code there. However, it's not too bad of a deal, since R is relatively easy to learn, and the custor

The following code recreates the dashboard we had in Tableau:

```
library(shiny)
library(dplyr)
library(ggplot2)
library(titanic)
titanic <- titanic::titanic train</pre>
ui <- fluidPage(</pre>
    sidebarPanel(
        width = 2,
        sliderInput(
            inputId='ageslider',
            label='Age range',
            min=min(titanic$Age, na.rm=T),
            max=max(titanic$Age, na.rm=T),
            value=c(min(titanic$Age, na.rm=T), max(titanic$Age, na.rm=T))
        ),
        selectInput(
            inputId='genderselect',
            label='Gender',
            choices=c('Male', 'Female'),
            selected=c('Male', 'Female'),
            multiple = TRUE
        )
    ),
    mainPanel(
        fluidRow(
            column(8, plotOutput('bar')),
            column(4, plotOutput('pie'))
        ),
        fluidRow(
            column(12, plotOutput('hist'))
        )
    )
)
server <- function(input, output) {</pre>
    output$bar <- renderPlot({</pre>
        pclass age data <- titanic %>%
            filter(Sex %in% tolower(input$genderselect)) %>%
            select(Pclass, Age) %>%
            filter(Age >= input$ageslider) %>%
            filter(Age <= input$ageslider) %>%
            group by(Pclass) %>%
            summarise(Age = round(mean(Age, na.rm=TRUE), 2))
        ggplot(data=pclass age data, aes(x=Pclass, y=Age)) +
            geom bar(stat='identity', fill='#4179ab') +
            geom text(aes(label=Age), vjust=-0.3, size=5) +
            ggtitle('Avg. age per passenger class') +
            theme void() +
            theme(plot.title=element text(size=20), panel.background=element :
    output$pie <- renderPlot({</pre>
```

```
pie data <- titanic %>%
            filter(Sex %in% tolower(input$genderselect)) %>%
            filter(Age >= input$ageslider) %>%
            filter(Age <= input$ageslider) %>%
            select(Sex) %>%
            group_by(Sex) %>%
            count() %>%
            mutate(Midpoint = cumsum(n) - n / 2)
        ggplot(pie data, aes(x='', y=n, fill=Sex)) +
            geom bar(width=1, stat='identity', color='white') +
            coord polar('y', start=0) +
            scale fill manual(values=c('#f34a53', '#4179ab')) +
            ggtitle('Gender pie chart') +
            theme void() +
            theme(plot.title=element text(size=20))
    })
    output$hist <- renderPlot({</pre>
        hist data <- titanic %>%
            filter(Age >= input$ageslider) %>%
            filter(Age <= input$ageslider) %>%
            filter(Sex %in% tolower(input$genderselect)) %>%
            select (Age)
            ggplot(hist data, aes(x=Age)) +
            geom histogram(bins=20, color='#62aca8', fill='#1e9a95') +
            ggtitle('Age histogram') +
            theme void() +
            theme(plot.title=element text(size=20))
    })
}
shinyApp(ui=ui, server=server)
```

#### And here's what the final product looks like:



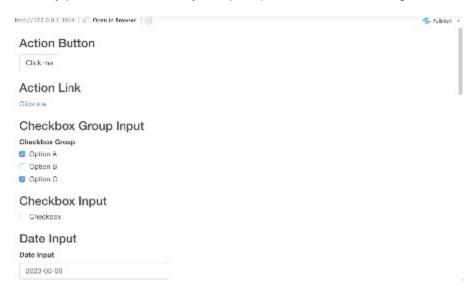
As you can see, this dashboard is quite unfinished. That's what we'll discuss in the last section of the articoloid to work in. Tableau's dashboard is way too sensitive to accidental clicks, and the dashboards look p much faster with Tableau than with R Shiny. For this reason, we'll declare Tableau the winner for simple c

Winner (Simple Dashboards): Tak

### **User Input and Interactivity**

Here's where things get especially interesting in our Tableau vs. R Shiny comparison. There's simply not for such things. However, inputs are essential for a well-built and interactive dashboard as they provide a Tableau.

R Shiny provides a wide array of input options. Instead of listing them, it's better to show some of them:



Yeah, that's a lot, and you can certainly do any task with them. One question that might immediately pop

Well, interactivity makes any dataset much easier to analyze. For example, the *file input* component can further filtered by date (*date range input*) and column selectors (*var select*). This is contrasted with read-c

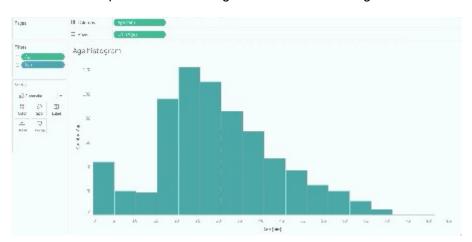
With R Shiny dashboards, you can also build complete interactive forms with password and text field inpulateractivity is a large part of what separates simple dashboards from powerful, enterprise-level business

Winner (User Input): R 5

### Visual Styling

Unfortunately, there isn't much you can do to tweak the visuals in Tableau. It's designed to look good outthings. The amount of tweakable aspects depends on the type of plot you're tweaking, but in general, Tak

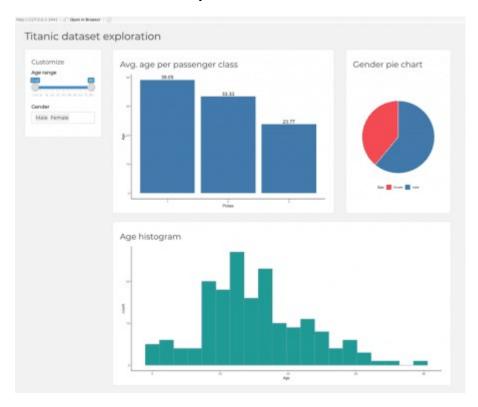
Here's an example of how to change the colors of a histogram:



R Shiny is a whole other animal. You can customize everything. To start, create a folder www right where can use everything CSS has to offer when it comes to styling R Shiny dashboards.

Learn More: How to Use CSS to Style Your R Shiny Dashboards

To connect the Shiny app to a CSS file, we can put theme=main.css right after the fluidPage in the F minutes with CSS and R Shiny:



Here's the source code for the final styled Shiny dashboard. Give this CSS optionality to a qualified desig visuals. One of our engineers even created an entire functional card-swiping video game in R Shiny by ta



Shiny Decisions is a video game made in R Shiny b

The winner for the visual styling section is once again pretty obvious. R Shiny blows Tableau out of the w decent out of the box.

#### Winner (Visual Styling): F

Looking for some attractive Shiny dashboard examples? Find inspiration here in our demo gallery.

### Conclusion

Here are the final results, according to our calculations:

- R Shiny 3 points
- Tableau 2 points
- Tie 1 point

So, is Shiny is a clear winner for all situations? Well, no. It depends on the type of work you need to do. It dashboard every now and then, Shiny may be overkill. Tableau was designed for exactly these sorts of ta the actual code, then Tableau is a great choice for simple charts and dashboards.