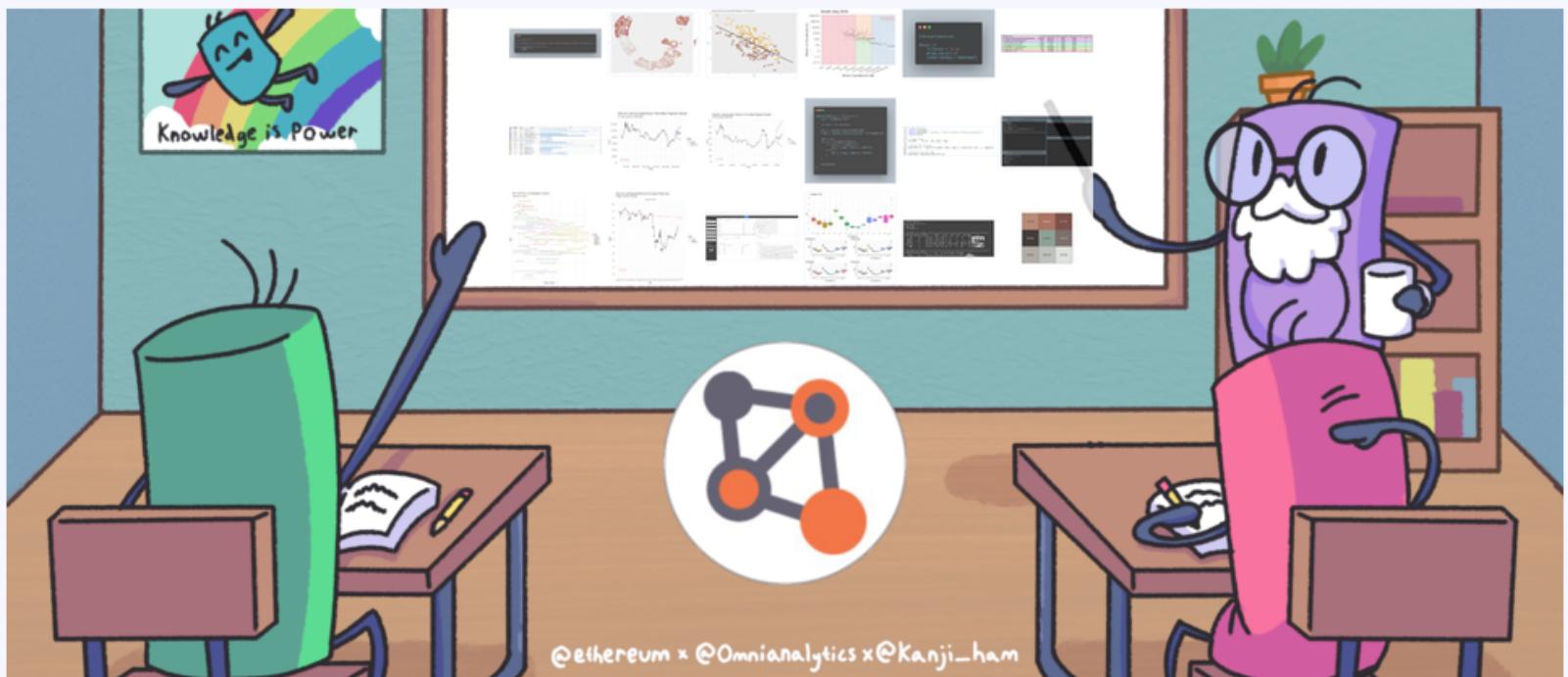


The Omniacs' Data Science Code Snippet Tweet Book Vol 1

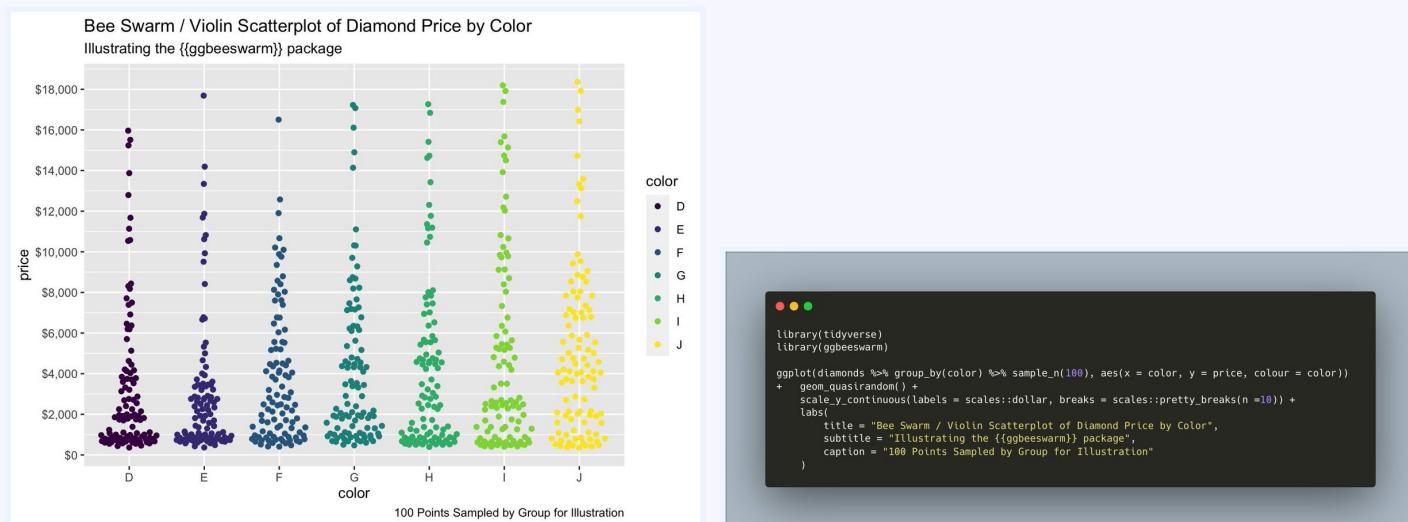


Quirkily Created by **Omni Analytics Group**

#rstats and #tidyverse



A **#beeswarm** plot, otherwise known as a Violin Scatter Plot, lets us view both distributional characteristics of the data, as well as details of the individual points. {{ ggbeeswarm }} makes producing such plots very simple! **#rstats #dataviz #tidyverse**



Tweeted on: 2022-05-17 20:06:46, Retweets: 6, Likes: 22

#rstats and #tidytuesday



Spruce up an **#RShiny** app in three easy steps: (1) Add a theme from **#shinythemes**, (2) Display loading indicators with **#shinycssloaders**, and (3) Add inline documentation with **#bsplus!** **#rstats** **#programming** **#tidytuesday**

The collage illustrates the transformation of a Shiny app's interface through three steps:

- 1. Add a theme from shinythemes**: Shows the transition from a standard light gray theme to the "cerulean" theme by shinythemes. The UI elements like the slider and histogram are styled with a blue and white color scheme.
- 2. Display Loading Indicators with shinycssloaders**: Shows the addition of a spinner loading indicator next to a plot output. The spinner is a small blue bar with dots, indicating the app is processing data.
- 3. Add inline documentation with bsplus**: Shows the addition of a tooltip with inline documentation. A blue circular icon with a question mark has a tooltip pointing to it, containing the text "About This" and "Bins in Histogram".

Each step includes a snippet of R code at the top:

- 1. Add a theme from shinythemes**:

```
ui <- fluidPage(theme = shinythemes::shinytheme("cerulean"))
```
- 2. Display Loading Indicators with shinycssloaders**:

```
shinycssloaders::withSpinner(plotOutput("distPlot"))
```
- 3. Add inline documentation with bsplus**:

```
sliderInput("bins", "Number of bins:", min = 1, max = 50, value = 30) %>%  
  shinyInput::shinyInput("distPlot",  
    shiny::icon("link", 30, 30),  
    bs::bs_embedPopover(  
      title = "About This",  
      content = "Bins in Histogram",  
      placement = "left"  
    )
```

Tweeted on: 2019-05-21 11:12:44, Retweets: 5, Likes: 17

#rstats and #tidyverse



If you want a human-readable description of the `#dplyr` data pipeline steps taken in your code, simply load the `#tidylog` package before running it! `#tidyverse #rstats #TidyTuesday`

dplyr code

```
library(tidyverse)
library(tidylog)

diamonds %>%
  mutate(ppc = price / carat) %>%
  select(cut, color, clarity, ppc) %>%
  group_by(cut) %>%
  summarise(ppc = mean(ppc))
```

tidylog output

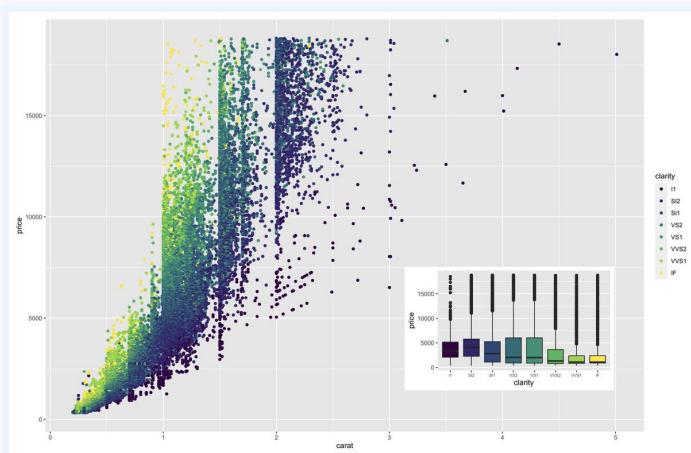
```
mutate: new variable 'ppc' with 26,693 unique values and 0% NA
select: dropped 7 variables (carat, depth, table, price, x, ...)
group_by: one grouping variable (cut)
summarise: now 5 rows and 2 columns, ungrouped
```

Tweeted on: 2020-01-14 13:52:21, Retweets: 8, Likes: 17

#rstats and #tidyverse



The newest {{ patchwork }} release on **#cran** goes beyond alignment of plots to a grid, and enables support for inset plots, allowing plots to be overlaid on top of one another at the exact location specified! **#rstats #dataviz #tidyverse**



```
library(ggplot2)
library(patchwork)

ggplot(data = diamonds, aes(x = carat, y = price, color = clarity)) +
  geom_point() +
  inset_element(
    ggplot(data = diamonds, aes(x = clarity, y = price, fill = clarity)) +
      geom_boxplot() +
      theme(legend.position = "off",
            axis.text.x = element_text(size = 6)),
    left = 0.6, bottom = 0.1, right = 0.95, top = 0.4
  )
```

Tweeted on: 2020-11-13 17:48:25, Retweets: 6, Likes: 15

#rstats and #tidyverse



For **#tidytuesday**, we're highlighting **#glue**, a **#tidyverse**-compatible alternative to the `paste()` function for string concatenation in R. Have a look at some standalone examples, and how it can be used in a **#dplyr** chain. **#rstats #datascience #programming**

String Concatenation in the Tidyverse



1. Basic Usage

Code

```
library(glue)
company_name <- "Omni Analytics Group"
glue("We are {company_name}")
```

Output

```
We are Omni Analytics Group
```

2. Advanced Usage

Code

```
library(lubridate)
company_name <- "Omni Analytics Group"
twitter_date <- ymd("2018-01-19")
glue("{company_name} created their twitter profile
on {format(twitter_date, '%b %d, %Y')}")
```

Output

```
Omni Analytics Group created their twitter profile
on January 19, 2018.
```

3. dplyr Chain Usage

Code

```
library(tidyverse)
mpg %>%
  group_by(manufacturer) %>%
  summarise(cty = round(mean(cty), digits = 2)) %>%
  mutate(description = glue("{manufacturer} has an",
                           "average city mpg of {cty}")) %>%
  .$description
```

Output

```
audi has an average city mpg of 17.61
chevrolet has an average city mpg of 15
dodge has an average city mpg of 13.14
ford has an average city mpg of 17.1
honda has an average city mpg of 24.44
hyundai has an average city mpg of 18.64
jeep has an average city mpg of 13.5
land rover has an average city mpg of 11.5
lincoln has an average city mpg of 11.33
mercury has an average city mpg of 12.25
mitsubishi has an average city mpg of 18.08
pontiac has an average city mpg of 17
subaru has an average city mpg of 19.29
toyota has an average city mpg of 18.53
volkswagen has an average city mpg of 20.93
```

Tweeted on: 2019-04-09 15:04:51, Retweets: 7, Likes: 13

#rstats and #tidyverse



Using paradigms from the `{glue}` package but thinking in reverse, the `{unglue}` **#rstats** package provides for some help **#regex** functionality, including the creation of a **#dataframe** automatically from semi-structured sentences that match a defined pattern.
#tidyverse #tidytuesday

```
# A set of semi-structured sentences
sentences <- c("The quick brown fox jumped over the lazy dog.",
             "The slow red hamster climbed over the frenetic lizard.",
             "The handsome mauve turtle swam over the unsuspecting anchovy.",
             "The confused rainbow pigeon flew over the happy hippo.",
             "The slick colorless rstats-package imported over the neglected base.")
```

```
> unglue_data(sentences, patterns)
   adjective1      color      animal1     verb  adjective2      animal2
1      quick      brown       fox  jumped      lazy      dog
2      slow       red      hamster  climbed  frenetic lizard
3    handsome     mauve     turtle   swam  unsuspecting anchovy
4    confused     rainbow   pigeon   flew      happy     hippo
5    slick      colorless rstats-package imported  neglected base
```

```
library(unglue)

# A set of semi-structured sentences
sentences <- c("The quick brown fox jumped over the lazy dog.",
             "The slow red hamster climbed over the frenetic lizard.",
             "The handsome mauve turtle swam over the unsuspecting anchovy.",
             "The confused rainbow pigeon flew over the happy hippo.",
             "The slick colorless rstats-package imported over the neglected base.")

# Define a pattern for the data
patterns <- c("The {adjective1} {color} {animal1} {verb} the {adjective2} {animal2}.")
```

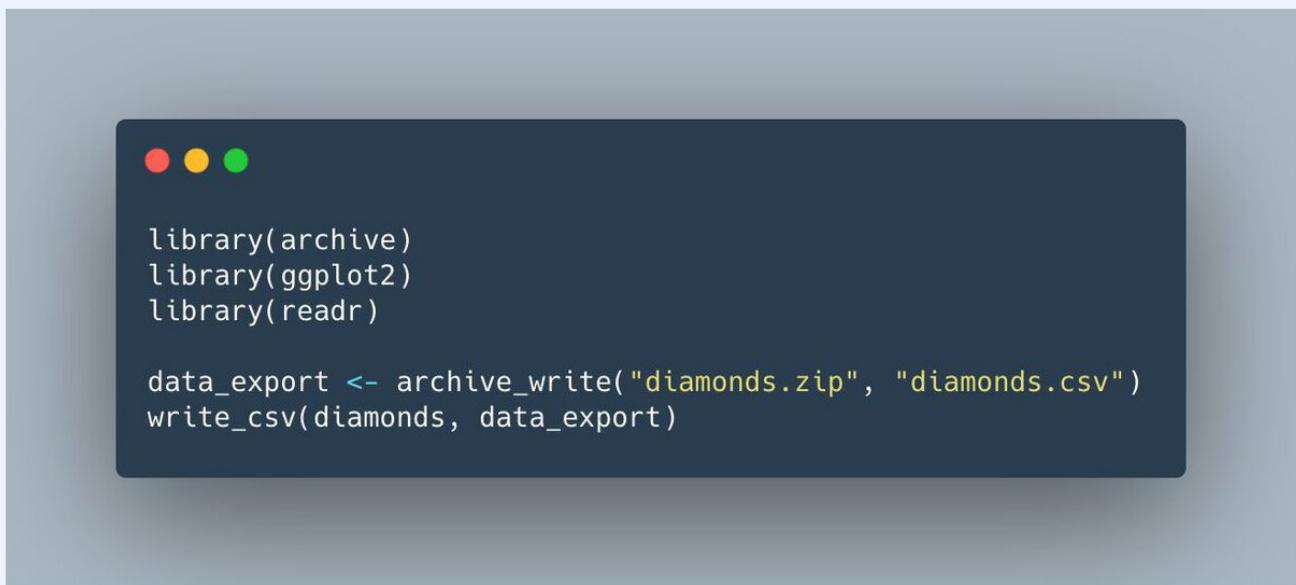
```
# Unglue it into a data frame
unglue_data(sentences, patterns)
```

Tweeted on: 2020-06-23 16:49:42, Retweets: 6, Likes: 13

#rstats and #tidyverse



Whether due to disk space constraints, or just ease and convenience, sometimes its nice to work with archives like **#zip** files directly in **#rstats** The {{ archive }} package makes this painless, and even supports writing archives natively! **#tidyverse**



A screenshot of an RStudio session window. The code in the console is:

```
library(archive)
library(ggplot2)
library(readr)

data_export <- archive_write("diamonds.zip", "diamonds.csv")
write_csv(diamonds, data_export)
```

Tweeted on: 2021-11-09 15:07:25, Retweets: 8, Likes: 12

#rstats and #tidyverse



`{ggsvg}`, another extremely useful `#ggplot2` extension, allows the plotting of arbitrary SVGs as points. Here we define an SVG by plotting a polygon in the shape of a star, and color/size it according to properties of the underlying data! `#rstats #tidyverse`



Tweeted on: 2021-12-30 18:59:52, Retweets: 5, Likes: 12

#rstats and #tidyverse



Brand new functionality in the dev version of **#tidyverse** allows for columns containing multiple variables to be easily separated with the `names_pattern` argument from the `pivot_longer()` function!
#rstats #tidyverse #tidytuesday #datascience

tidyr 0.8.3.9000 Pivoting

Preliminary Code:

```
library(tidyverse) # devtools::install_github("tidyverse/tidyr")
library(lubridate)
```

Read the Data

```
yield_raw <- read_csv("https://omnianalytics.io/data/yield.csv")
```

yield_raw

Index	Month	Day	Yield
1	1	1	100.0
1	1	2	100.0
1	1	3	100.0
1	1	4	100.0
1	1	5	100.0
1	1	6	100.0
1	1	7	100.0
1	1	8	100.0
1	1	9	100.0
1	1	10	100.0
1	1	11	100.0
1	1	12	100.0
1	1	13	100.0
1	1	14	100.0
1	1	15	100.0
1	1	16	100.0
1	1	17	100.0
1	1	18	100.0
1	1	19	100.0
1	1	20	100.0
1	1	21	100.0
1	1	22	100.0
1	1	23	100.0
1	1	24	100.0
1	1	25	100.0
1	1	26	100.0
1	1	27	100.0
1	1	28	100.0
1	1	29	100.0
1	1	30	100.0
1	1	31	100.0
1	2	1	100.0
1	2	2	100.0
1	2	3	100.0
1	2	4	100.0
1	2	5	100.0
1	2	6	100.0
1	2	7	100.0
1	2	8	100.0
1	2	9	100.0
1	2	10	100.0
1	2	11	100.0
1	2	12	100.0
1	2	13	100.0
1	2	14	100.0
1	2	15	100.0
1	2	16	100.0
1	2	17	100.0
1	2	18	100.0
1	2	19	100.0
1	2	20	100.0
1	2	21	100.0
1	2	22	100.0
1	2	23	100.0
1	2	24	100.0
1	2	25	100.0
1	2	26	100.0
1	2	27	100.0
1	2	28	100.0
1	2	29	100.0
1	2	30	100.0
1	2	31	100.0
1	3	1	100.0
1	3	2	100.0
1	3	3	100.0
1	3	4	100.0
1	3	5	100.0
1	3	6	100.0
1	3	7	100.0
1	3	8	100.0
1	3	9	100.0
1	3	10	100.0
1	3	11	100.0
1	3	12	100.0
1	3	13	100.0
1	3	14	100.0
1	3	15	100.0
1	3	16	100.0
1	3	17	100.0
1	3	18	100.0
1	3	19	100.0
1	3	20	100.0
1	3	21	100.0
1	3	22	100.0
1	3	23	100.0
1	3	24	100.0
1	3	25	100.0
1	3	26	100.0
1	3	27	100.0
1	3	28	100.0
1	3	29	100.0
1	3	30	100.0
1	3	31	100.0
1	4	1	100.0
1	4	2	100.0
1	4	3	100.0
1	4	4	100.0
1	4	5	100.0
1	4	6	100.0
1	4	7	100.0
1	4	8	100.0
1	4	9	100.0
1	4	10	100.0
1	4	11	100.0
1	4	12	100.0
1	4	13	100.0
1	4	14	100.0
1	4	15	100.0
1	4	16	100.0
1	4	17	100.0
1	4	18	100.0
1	4	19	100.0
1	4	20	100.0
1	4	21	100.0
1	4	22	100.0
1	4	23	100.0
1	4	24	100.0
1	4	25	100.0
1	4	26	100.0
1	4	27	100.0
1	4	28	100.0
1	4	29	100.0
1	4	30	100.0
1	4	31	100.0
1	5	1	100.0
1	5	2	100.0
1	5	3	100.0
1	5	4	100.0
1	5	5	100.0
1	5	6	100.0
1	5	7	100.0
1	5	8	100.0
1	5	9	100.0
1	5	10	100.0
1	5	11	100.0
1	5	12	100.0
1	5	13	100.0
1	5	14	100.0
1	5	15	100.0
1	5	16	100.0
1	5	17	100.0
1	5	18	100.0
1	5	19	100.0
1	5	20	100.0
1	5	21	100.0
1	5	22	100.0
1	5	23	100.0
1	5	24	100.0
1	5	25	100.0
1	5	26	100.0
1	5	27	100.0
1	5	28	100.0
1	5	29	100.0
1	5	30	100.0
1	5	31	100.0
1	6	1	100.0
1	6	2	100.0
1	6	3	100.0
1	6	4	100.0
1	6	5	100.0
1	6	6	100.0
1	6	7	100.0
1	6	8	100.0
1	6	9	100.0
1	6	10	100.0
1	6	11	100.0
1	6	12	100.0
1	6	13	100.0
1	6	14	100.0
1	6	15	100.0
1	6	16	100.0
1	6	17	100.0
1	6	18	100.0
1	6	19	100.0
1	6	20	100.0
1	6	21	100.0
1	6	22	100.0
1	6	23	100.0
1	6	24	100.0
1	6	25	100.0
1	6	26	100.0
1	6	27	100.0
1	6	28	100.0
1	6	29	100.0
1	6	30	100.0
1	6	31	100.0
1	7	1	100.0
1	7	2	100.0
1	7	3	100.0
1	7	4	100.0
1	7	5	100.0
1	7	6	100.0
1	7	7	100.0
1	7	8	100.0
1	7	9	100.0
1	7	10	100.0
1	7	11	100.0
1	7	12	100.0
1	7	13	100.0
1	7	14	100.0
1	7	15	100.0
1	7	16	100.0
1	7	17	100.0
1	7	18	100.0
1	7	19	100.0
1	7	20	100.0
1	7	21	100.0
1	7	22	100.0
1	7	23	100.0
1	7	24	100.0
1	7	25	100.0
1	7	26	100.0
1	7	27	100.0
1	7	28	100.0
1	7	29	100.0
1	7	30	100.0
1	7	31	100.0
1	8	1	100.0
1	8	2	100.0
1	8	3	100.0
1	8	4	100.0
1	8	5	100.0
1	8	6	100.0
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1	8	9	100.0
1	8	10	100.0
1	8	11	100.0
1	8	12	100.0
1	8	13	100.0
1	8	14	100.0
1	8	15	100.0
1	8	16	100.0
1	8	17	100.0
1	8	18	100.0
1	8	19	100.0
1	8	20	100.0
1	8	21	100.0
1	8	22	100.0
1	8	23	100.0
1	8	24	100.0
1	8	25	100.0
1	8	26	100.0
1	8	27	100.0
1	8	28	100.0
1	8	29	100.0
1	8	30	100.0
1	8	31	100.0
1	9	1	100.0
1	9	2	100.0
1	9	3	100.0
1	9	4	100.0
1	9	5	100.0
1	9	6	100.0
1	9	7	100.0
1	9	8	100.0
1	9	9	100.0
1	9	10	100.0
1	9	11	100.0
1	9	12	100.0
1	9	13	100.0
1	9	14	100.0
1	9	15	100.0
1	9	16	100.0
1	9	17	100.0
1	9	18	100.0
1	9	19	100.0
1	9	20	100.0
1	9	21	100.0
1	9	22	100.0
1	9	23	100.0
1	9	24	100.0
1	9	25	100.0
1	9	26	100.0
1	9	27	100.0
1	9	28	100.0
1	9	29	100.0
1	9	30	100.0
1	9	31	100.0
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1	10	2	100.0
1	10	3	100.0
1	10	4	100.0
1	10	5	100.0
1	10	6	100.0
1	10	7	100.0
1	10	8	100.0
1	10	9	100.0
1	10	10	100.0
1	10	11	100.0
1	10	12	100.0
1	10	13	100.0
1	10	14	100.0
1	10	15	100.0
1	10	16	100.0
1	10	17	100.0
1	10	18	100.0
1	10	19	100.0
1	10	20	100.0
1	10	21	100.0
1	10	22	100.0
1	10	23	100.0
1	10	24	100.0
1	10	25	100.0
1	10	26	100.0
1	10	27	100.0
1	10	28	100.0
1	10	29	100.0
1	10	30	100.0
1	10	31	100.0
1	11	1	100.0
1	11	2	100.0
1	11	3	100.0
1	11	4	100.0
1	11	5	100.0
1	11	6	100.0
1	11	7	100.0
1	11	8	100.0
1	11	9	100.0
1	11	10	100.0
1	11	11	100.0
1	11	12	100.0
1	11	13	100.0
1	11	14	100

#rstats and #tidyverse



Parallel computing in the `#tidyverse` - `multidplyr` 0.1 is now on CRAN and provides a quick and easy framework for distributing long-running computations to the different processes in a cluster, then collecting the results! `#rstats`

Tweeted on: 2021-02-25 16:08:45, Retweets: 5, Likes: 11

#rstats and #programming



12 34 Simple steps to convert your **#forloop** to a parallel {{ foreach }} loop:

- 1 Change “for” to “foreach”
- 2 Change “in” to “=”
- 3 Assign the result to a variable
- 4 Add “%do%” after the iterator statement

#rstats #programming

```
● ● ●  
for (i in 1:10) {  
  j = i^2 - 5  
  
  print(j)  
}
```

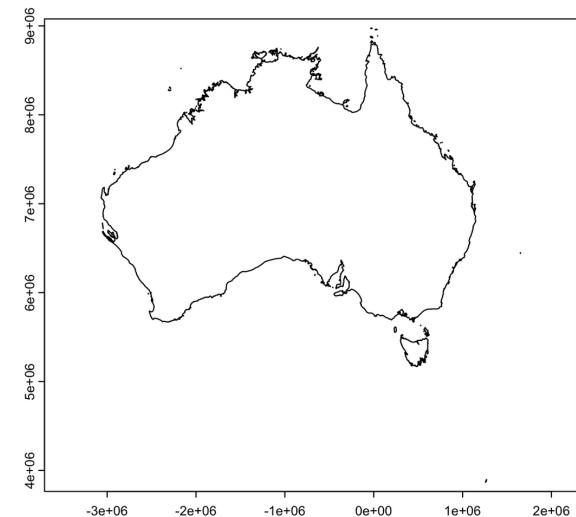
```
● ● ●  
library(foreach)  
  
res <- foreach (i = 1:10) %do% {  
  j = i^2 - 5  
  
  print(j)  
}
```

Tweeted on: 2022-03-31 15:41:24, Retweets: 8, Likes: 10

#rstats and #tidyverse



For working with **#rasters**, we love the {{ terra }} package in **#rstats**
- Combine that with {{ geodata }} and in just a few lines of code,
we can manipulate and project geographic data from around the
world! **#tidyverse**



```
library(terra)
library(geodata)

w <- world(path = ".", resolution =
australia <- w[w$GID_0=="AUS", ]

prj <- "epsg:28355"

p_aus <- project(australia, prj)
plot(p_aus)
```

Tweeted on: 2022-05-25 14:24:24, Retweets: 6, Likes: 10

#rstats and #tidytuesday



Adding `#testthat` tests to your `#rstats` package helps with proper regression testing. But [@_lionelhenrys](#) `#vdiffrr` package extends this functionality even further, allowing you to test for regressions in visuals such as `#ggplot2` plots! `#rstats #tidytuesday`

The screenshot illustrates the `vdiffrr` package's workflow for visual regression testing:

- 1. Create Your Tests**
Write `testthat-compatible` tests
Code

```
context("OMGGraphicsTests")  
library(ggplot2)  
mpg_hist <- ggplot(data = mpg, aes(x = cty)) +  
  geom_histogram()  
mpg_scatter <- ggplot(data = mpg, aes(x = cty, y = hwy)) +  
  geom_point()  
expect_doppelganger("MPG Histogram", mpg_hist)  
expect_doppelganger("MPG Scatter Plot", mpg_scatter)
```
- 2. Add to your R Package**
Code

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

A screenshot of an RStudio session showing the addition of the test code to the `OMGGraphicsTests` file.
- 3. Manage the Test Cases**
Visually confirm the test plot appearance
Code

```
vdiffrr::manage_cases()
```

A screenshot of the `vdiffrr` interface showing a histogram plot used for visual comparison.
- 4. Run the Tests!**
Code

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

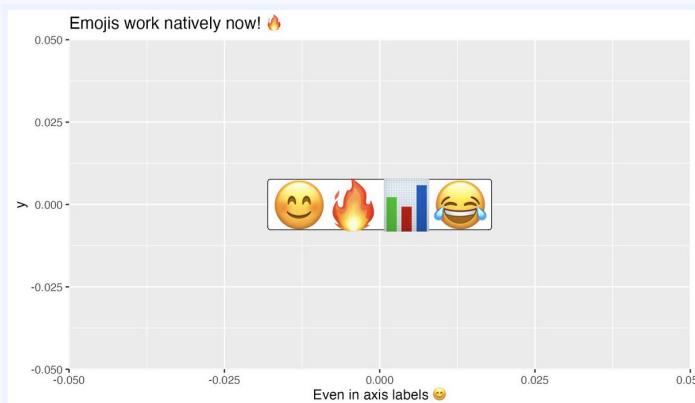
A screenshot of an RStudio session showing the execution of the test code and the resulting output indicating no failures.

Tweeted on: 2019-04-30 14:41:33, Retweets: 5, Likes: 9

#rstats and #tidyverse



Configure **@RStudio** 1.4 to use the AGG backend, install the latest `{ragg}` package, and like magic, emojis are natively rendered in **#ggplot** in the RStudio plot pane without any configuration needed! **#rstats #tidyverse** Read more here: <https://www.tidyverse.org/blog/2021/02/modern-text-features/>



```
library(ragg)
library(ggplot2)

emojis <- "😊🔥🇮🇹😂"

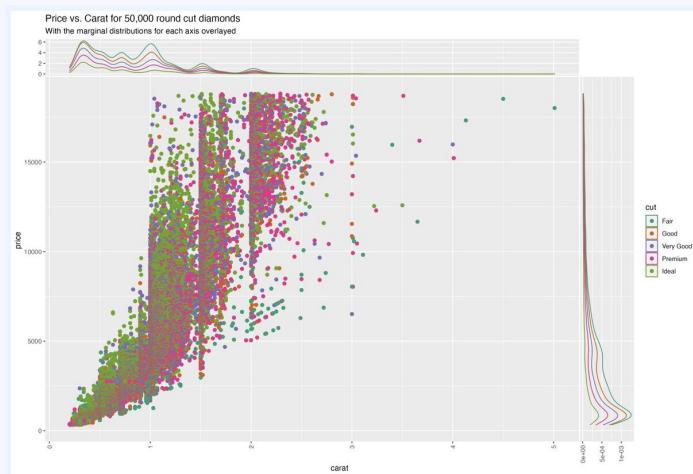
p <- ggplot() +
  geom_label(
    aes(x = 0, y = 0, label = emojis),
    family = "Apple Color Emoji",
    size = 15
  ) +
  labs(
    title = "Emojis work natively now! 🔥",
    x = "Even in axis labels 😊"
  )
```

Tweeted on: 2021-02-17 20:19:17, Retweets: 6, Likes: 9

#rstats and #dataviz



Another very useful {{ ggplot2 }} extension is {{ ggsid }} , which allows for a quick and easy **#visualization** of the marginal distributions of your variables side by side with your plot. **#rstats #dataviz #TidyTuesday**



```
library(tidyverse)
library(ggsid)

ggplot(diamonds, aes(carat, price, colour = cut)) +
  geom_point(size = 2) +
  scale_colour_brewer(palette = "Dark2") +
  geom_xsidedensity(aes(y = after_stat(density)), position = "stack") +
  geom_ysidedensity(aes(x = after_stat(density)), position = "stack") +
  theme(axis.text.x = element_text(angle = 90, vjust = .5)) +
  labs(
    title = "Price vs. Carat for 50,000 round cut diamonds",
    subtitle = "With the marginal distributions for each axis overlaid"
  )
```

Tweeted on: 2021-05-25 14:31:19, Retweets: 4, Likes: 9

#rstats and #tidyverse



The `#python` package {{ polars }}

is an extremely efficient library for `#dataframe` manipulation. Better yet, it has a syntax for data summaries that should be quite familiar to most `#rstats` and `#tidyverse` users!

```
In [8]: df
Out[8]: shape: (5, 8)
          cut      carat_sum    depth_sum    table_sum    price_sum    x_sum    y_sum    z_sum
          ---      f64        f64        f64        f64        f64        f64        f64
Good   695.209999  5.48361e00  5.4841699  3194268  5241.49  5255.0799  3261.8299
       9999986  0000001e4  99999999e4  99999999e4  99999999e4  99999999e4  99999999e4  999999977
Very Good 1623.15999  1.48152599  9999992e5  999999993  7715165  1.3923292  1.3111710
       99999958  99999958  99999958  99999958  99999958  99999958  99999958  99999958
Ideal   2257.49999  2.40763099  9999997e5  999999993  10130238  2.0357128  2.0421330
       99999513  99999513  99999513  99999513  99999513  99999513  99999513  99999513
Fair    191.079099  1.41836000  1.32977e4  824838  1323.6380  1312.2399  833.76000
       9999999  00000002e4  0000000003  0000000005  0000000006  0000000016  999999998  000000001
Premium 1677.36999  1.42958099  1.373676e5  8270443  1.3057090  1.298987e4  7966.5399
       99999856  999999856  999999856  999999856  999999856  999999856  999999856  999999856
```

```
In [2]: df
Out[2]: shape: (53940, 10)
          carat      cut      color      clarity      ...      price      x      y      z
          ---      ---      ---      ---      ...      ---      ---      ---      ---
0.23  Ideal      E      SI2      ...      326      3.95      3.98      2.43
0.21  Premium    E      SI1      ...      326      3.89      3.84      2.31
0.23  Good       E      VS1      ...      327      4.05      4.07      2.31
0.29  Premium    I      VS2      ...      334      4.2      4.23      2.63
...
...
...
...
...
0.72  Ideal      D      SI1      ...      2757      5.75      5.76      3.5
0.72  Good       D      SI1      ...      2757      5.69      5.75      3.61
0.7  Very Good  D      SI1      ...      2757      5.66      5.68      3.56
0.86  Premium    H      SI2      ...      2757      6.15      6.12      3.74
0.75  Ideal      D      SI2      ...      2757      5.83      5.87      3.64
```

```
● ● ●

import polars as pl

df = pl.read_csv("https://raw.githubusercontent.com/tidyverse/ggplot2/main/data-raw/diamonds.csv")

df = (df.filter(pl.col("color") == "E")
      .groupby("cut")
      .sum())
```

Tweeted on: 2021-11-23 20:13:03, Retweets: 3, Likes: 9

#rstats and #tidyverse



An essential package in our **#rstats** analysis toolkit, `tidyverse` gets a new 1.2.0 release today. Were still unpacking the new features, but one cool one - the ability to expand the names in pivot operations, to retain missing factor levels automatically! **#tidyverse**

```
> months <- c("January", "February", "March", "April",
+   "May", "June", "July", "August",
+   "September", "October", "November", "December")
>
> m_series <- tibble(
+   month = factor(months[-c(2, 10)], levels = months),
+   value = sample(1:10)
+ )
>
> pivot_wider(m_series, names_from = month,
+   values_from = value, names_expand = TRUE)
# A tibble: 1 × 12
  January February March April May June July August September October November December
    <int>     <int> <int> <int> <int> <int> <int> <int> <int> <int> <int> <int>
1      6       NA     3    10     8     5     9     7     2     NA     1     4
```

```
library(tidyr)

months <- c("January", "February", "March", "April",
          "May", "June", "July", "August",
          "September", "October", "November", "December")

m_series <- tibble(
  month = factor(months[-c(2, 10)], levels = months),
  value = sample(1:10)
)

pivot_wider(m_series, names_from = month,
            values_from = value, names_expand = TRUE)
```

Tweeted on: 2022-02-03 01:21:42, Retweets: 7, Likes: 9

#rstats and #tidyverse



The latest release of **#santoku** 🍴 for **#rstats** adds even more functionality for chopping vectors, including support for dates! Here we easily chop the airquality data into Early, Middle, and Late categories based on the date of observation. **#tidyverse**
#tidytuesday

```
library(santoku)
library(lubridate)
library(tidyverse)

airquality %>%
  mutate(Date = ymd(paste0("1973-", Month, "-", Day))) %>%
  mutate(Cut = chop_quantiles(Date, c(0, .33, .66, 1), labels = c("Early", "Middle", "Late"))) %>%
  slice(1:5)
```

	Ozone	Solar.R	Wind	Temp	Month	Day	Date	Cut
	<int>	<int>	<dbl>	<int>	<int>	<int>	<date>	<fct>
1	41	190	7.4	67	5	1	1973-05-01	Early
2	36	118	8	72	5	2	1973-05-02	Early
3	12	149	12.6	74	5	3	1973-05-03	Early
4	18	313	11.5	62	5	4	1973-05-04	Early
5	NA	NA	14.3	56	5	5	1973-05-05	Early
6	NA	150	6.3	77	6	21	1973-06-21	Middle
7	NA	59	1.7	76	6	22	1973-06-22	Middle
8	NA	91	4.6	76	6	23	1973-06-23	Middle
9	NA	250	6.3	76	6	24	1973-06-24	Middle
10	NA	135	8	75	6	25	1973-06-25	Middle
11	NA	222	8.6	92	8	10	1973-08-10	Late
12	NA	137	11.5	86	8	11	1973-08-11	Late
13	44	192	11.5	86	8	12	1973-08-12	Late
14	28	273	11.5	82	8	13	1973-08-13	Late
15	65	157	9.7	80	8	14	1973-08-14	Late

Tweeted on: 2020-06-09 22:42:56, Retweets: 10, Likes: 8

#rstats and #tidyverse



The latest {{ dbplyr }} 2.0 release adds compatibility with {{ dplyr }}'s 1.0 release features, and drastically improves the `#sql` translation. Heres some `#rstats #tidyverse` code and the corresponding `#sql` automatically generated!

```
library(dbplyr)
library(tidyverse)

lf <- tbl_lazy(mpg)

lf %>%
  mutate(model = toupper(model)) %>%
  group_by(model) %>%
  summarise(across(c(cty, hwy), mean, na.rm = TRUE)) %>%
  arrange(desc(cty))
```

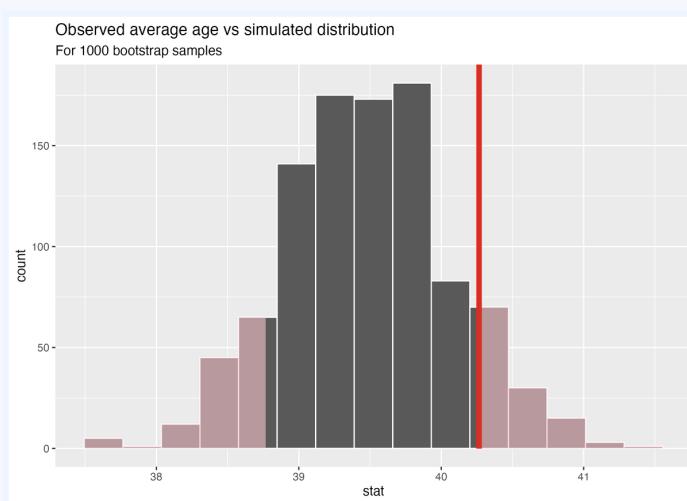
```
SELECT `model`, AVG(`cty`) AS `cty`, AVG(`hwy`) AS `hwy`
FROM (SELECT `manufacturer`, UPPER(`model`) AS `model`, `displ`, `year`, `cyl`, `trans`, `drv`, `cty`,
`hwy`, `class`
FROM `mpg` ) `df`
GROUP BY `model`
ORDER BY `cty` DESC
```

Tweeted on: 2020-11-05 19:41:09, Retweets: 4, Likes: 8

#rstats and #tidyverse



Powerful and easy-to-use new **#rstats** package for **#tidyverse**-compliant statistical inference! The `{infer}` package makes it dead simple to perform a number of statistical inference tasks, including simulation from a null-distribution
<https://www.tidyverse.org/blog/2021/08/infer-1-0-0/>



```
library(infer)
library(ggplot2)

point_estimate <- gss %>%
  specify(response = age) %>%
  calculate(stat = "mean")

gss %>%
  specify(response = age) %>%
  hypothesize(null = "point", mu = 39.5) %>%
  generate(reps = 1000, type = "bootstrap") %>%
  calculate(stat = "mean") %>%
  visualize() +
  shade_p_value(obs_stat = point_estimate, direction = "two-sided") +
  labs(
    title = "Observed average age vs simulated distribution",
    subtitle = "For 1000 bootstrap samples"
  )
```

Tweeted on: 2021-08-17 22:16:16, Retweets: 5, Likes: 8

#rstats and #python



Its now even easier to import **#rstats #dataframes** into **#python**. With {{ pyreadr }}, RData files can be natively imported and converted into a **#pandas** dataframe like so!

```
[In [6]: mpg_df
Out[6]:
   manufacturer  model  displ  year  cyl      trans  drv  cty  hwy  fl  class
0       audi     a4    1.8  1999    4  auto(l5)   f  18  29  p  compact
1       audi     a4    1.8  1999    4  manual(m5)   f  21  29  p  compact
2       audi     a4    2.0  2008    4  manual(m6)   f  20  31  p  compact
3       audi     a4    2.0  2008    4  auto(av)    f  21  30  p  compact
4       audi     a4    2.8  1999    6  auto(l5)    f  16  26  p  compact
..      ...  ...
229  volkswagen  passat  2.0  2008    4  auto(s6)   f  19  28  p  midsized
230  volkswagen  passat  2.0  2008    4  manual(m6)   f  21  29  p  midsized
231  volkswagen  passat  2.8  1999    6  auto(l5)    f  16  26  p  midsized
232  volkswagen  passat  2.8  1999    6  manual(m5)   f  18  26  p  midsized
233  volkswagen  passat  3.6  2008    6  auto(s6)   f  17  26  p  midsized
[234 rows x 11 columns]
```

```
● ● ●
import pyreadr

mpg = pyreadr.read_r('Downloads/Snapshot/mpg.RData')
mpg_df = mpg["mpg"]
```

Tweeted on: 2021-12-08 16:35:54, Retweets: 5, Likes: 8

#rstats and #programming



Our **#rstats #programming #tipoftheday** - Did you know you can use the `enframe()` function from the **#tibble** package to easily convert named vectors and lists to a two column **#dataframe**? Heres some sample code!

The slide has a header 'Data Transformation' and an 'OMNI analytics group' logo. It shows three stages: 'BEFORE' (a screenshot of the mtcars dataset), 'CODE' (the R code), and 'AFTER' (the resulting tibble).

BEFORE:

car	mpg
Mazda RX4	21.0
Hornet 4 Drive	21.0
Duster 360	14.3
Merc 240D	22.8
Merc 230	22.8
Merc 280	19.2
Merc 450SL	15.8
Lincoln Continental	18.7
Honda Civic	30.4
Dodge Challenger	12.5
Pontiac Firebird	19.2
Lotus Europa	21.4
Maserati Bora	15.0
Mazda RX4 Wag	21.0
Hornet Sportabout	18.7
Merc 280C	16.4
Merc 450SLC	16.4
Cadillac Fleetwood	10.4
Chrysler Imperial	14.3
Toyota Corolla	33.9
AMC Javelin	15.2
Fiat X1-9	26.0
Ford Pantera L	15.8
Volvo 144E	21.4
Datsun 710	22.8
Valiant	18.1
Merc 223C	22.8
Ferrari Dino	19.7

CODE:

```
myvector <- mtcars$mpg  
names(myvector) <- rownames(mtcars)  
  
tibble::enframe(myvector,  
                name = "car",  
                value = "mpg")
```

AFTER:

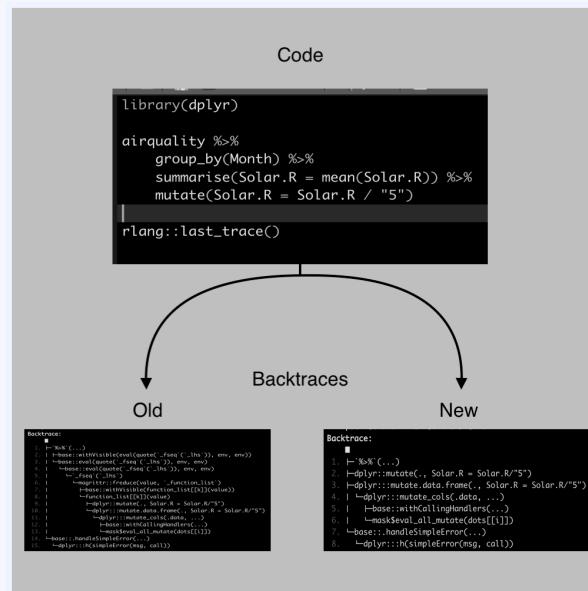
car	mpg
Mazda RX4	21
Mazda RX4 Wag	21
Datsun 710	22.8
Hornet 4 Drive	21.4
Hornet Sportabout	18.7
Valiant	18.1
Duster 360	14.3
Merc 240D	22.8
Merc 230	22.8
Merc 280	19.2
Merc 450SL	15.8
Lincoln Continental	18.7
Honda Civic	30.4
Dodge Challenger	12.5
Pontiac Firebird	19.2
Lotus Europa	21.4
Maserati Bora	15.0
Mazda RX4 Wag	21
Hornet Sportabout	18.7
Merc 280C	16.4
Merc 450SLC	16.4
Cadillac Fleetwood	10.4
Chrysler Imperial	14.3
Toyota Corolla	33.9
AMC Javelin	15.2
Fiat X1-9	26.0
Ford Pantera L	15.8
Volvo 144E	21.4
Datsun 710	22.8
Valiant	18.1
Duster 360	14.3
Merc 223C	22.8
Merc 280	19.2
Merc 450SL	15.8
Lincoln Continental	18.7
Honda Civic	30.4
Dodge Challenger	12.5
Pontiac Firebird	19.2
Lotus Europa	21.4
Maserati Bora	15.0

Tweeted on: 2019-05-23 13:26:59, Retweets: 6, Likes: 7

#rstats and #debugging



We're looking forward to the latest {{ magrittr }} 2.0 release for **#rstats** - In addition to evaluation changes and speed improvements, the new version cleans up the error backtrace to ease the process of **#debugging** (Left: Old backtrace, Right: New)



Tweeted on: 2020-09-03 16:19:42, Retweets: 2, Likes: 7

#rstats and #tidyverse



Hot off the presses, {{ dplyr }} 1.0.4 implements two new functions which perform column-wise filtering across a data frame - if_all() checks that all columns satisfy the given condition, while if_any() checks that any one of them does. **#rstats #tidyverse**

```
> mtcars %>%
+   filter(if_all(c(gear, carb), `>`, 4))
      mpg cyl disp hp drat wt qsec vs am gear carb
Ferrari Dino 19.7 6 145 175 3.62 2.77 15.5 0 1 5 6
Maserati Bora 15.0 8 301 335 3.54 3.57 14.6 0 1 5 8
> mtcars %>%
+   filter(if_any(c(gear, carb), `>`, 4))
      mpg cyl disp hp drat wt qsec vs am gear carb
Porsche 914-2 26.0 4 120.3 91 4.43 2.140 16.7 0 1 5 2
Lotus Europa 30.4 4 95.1 113 3.77 1.513 16.9 1 1 5 2
Ford Pantera L 15.8 8 351.0 264 4.22 3.170 14.5 0 1 5 4
Ferrari Dino 19.7 6 145.0 175 3.62 2.770 15.5 0 1 5 6
Maserati Bora 15.0 8 301.0 335 3.54 3.570 14.6 0 1 5 8
> |
```

```
library(dplyr)

mtcars %>%
  filter(if_all(c(gear, carb), `>`, 4))

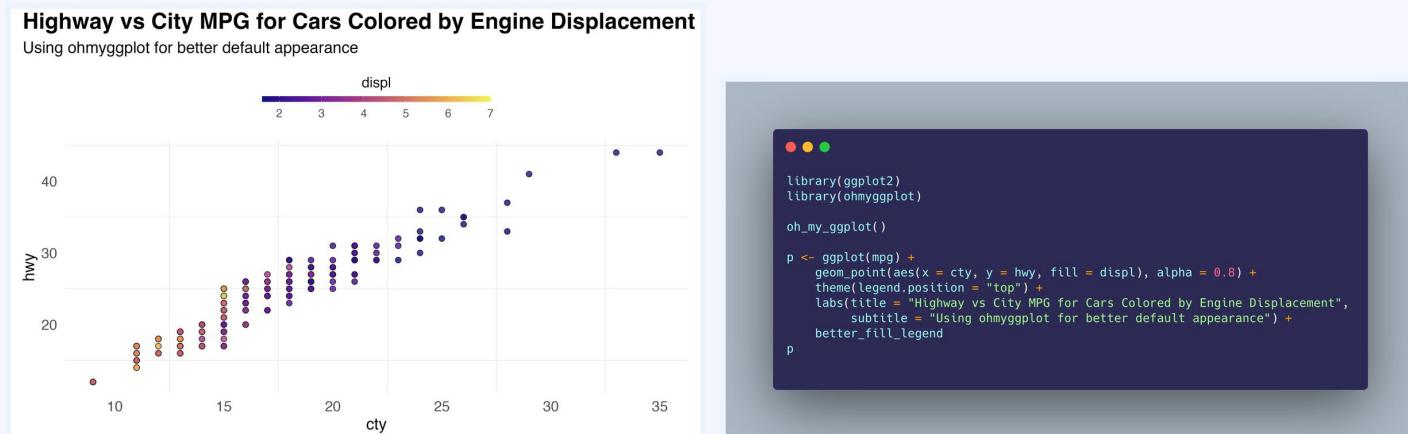
mtcars %>%
  filter(if_any(c(gear, carb), `>`, 4))
```

Tweeted on: 2021-02-03 16:31:53, Retweets: 5, Likes: 7

#rstats and #dataviz



Came across [@KejunYings](#) package `{ohmyggplot}`, which sets certain aesthetic and stylistic defaults for `#ggplot2` that make the default plot appearance much nicer with only one single initialization line. Check it out! `#rstats #dataviz`
<https://github.com/albert-ying/ohmyggplot>



Tweeted on: 2021-11-11 18:47:20, Retweets: 5, Likes: 7

#rstats and #dataframes



The {{ openxlsx }} package allows easy creation of #excel file from #rstats #dataframes. Compared to existing solutions, it maintains the same relative performance, adds many styling/formatting options, while also removing a dependency on #Java.

The image shows a Microsoft Excel spreadsheet titled "writeXLSXTable3.xlsx" and a block of R code. The Excel spreadsheet contains a table with columns: Date, USD Value, Percentage, SmallNumber, and stringsAsFactors. The R code uses the tidyverse and openxlsx packages to create this table, applying specific styles like border colors and date formats.

Date	USD Value	Percentage	SmallNumber	stringsAsFactors
11/22/2022	\$ 1	0.00%	1.19E-10	FALSE
11/21/2022	\$ 2	5.26%	7.00E-10	FALSE
11/20/2022	\$ 3	10.53%	6.96E-10	FALSE
11/19/2022	\$ 4	15.79%	2.83E-10	FALSE
11/18/2022	\$ 5	21.05%	5.41E-10	FALSE
11/17/2022	\$ 6	26.32%	6.38E-10	FALSE
11/16/2022	\$ 7	31.59%	8.05E-10	FALSE
11/15/2022	\$ 8	36.84%	4.98E-10	FALSE
11/14/2022	\$ 9	42.11%	5.78E-10	FALSE
11/13/2022	\$ 10	47.37%	6.22E-10	FALSE
11/12/2022	\$ 11	52.63%	5.98E-10	FALSE
11/11/2022	\$ 12	57.89%	2.15E-10	FALSE
11/10/2022	\$ 13	63.16%	2.40E-10	FALSE
11/09/2022	\$ 14	68.42%	2.37E-10	FALSE
11/08/2022	\$ 15	73.68%	9.11E-10	FALSE
11/07/2022	\$ 16	78.95%	3.24E-10	FALSE
11/06/2022	\$ 17	84.21%	4.60E-10	FALSE
11/05/2022	\$ 18	89.47%	4.52E-10	FALSE
11/04/2022	\$ 19	94.74%	9.33E-10	FALSE
11/03/2022	\$ 20	100.00%	2.78E-10	FALSE

```
library(tidyverse)
library(openxlsx)

options("openxlsx.borderColour" = "#F07649")
options("openxlsx.dateFormat" = "mm/dd/yyyy")
options("openxlsx.numFmt" = NULL) ## For default style rounding of numeric columns

df <- tibble("Date" = Sys.Date() - 0:19,
            "USD Value" = paste("$", 1:20),
            "Percentage" = seq(0, 1, length.out=20),
            "SmallNumber" = runif(20) / 1E9, stringsAsFactors = FALSE)

class(df$Percentage) <- "percentage"
class(df$SmallNumber) <- "scientific"

write.xlsx(df, file = "writeXLSXTable3.xlsx", asTable = TRUE)
```

Tweeted on: 2021-12-20 18:10:52, Retweets: 3, Likes: 7

#rstats and #modeldown



A brand new member of the *down series of **#rstats** packages, **#modeldown** quickly and easily creates a **#bootstrap**-based **#HTML** page to showcase model fit and performance results!

The screenshot shows the modelDown web application. At the top, there's a navigation bar with tabs for 'modelDown', 'Auditor', 'Model Performance', and 'Variable Importance'. Below the navigation bar, a purple header bar says 'modelDown' and 'Explore your model!'. Underneath, there are two main sections: 'Basic data information' (with a table showing 234 observations and 2 columns) and 'Explainers' (with a link to 'lm (download)'). Below these are 'Summaries for numerical variables' (a table for 'cty' and 'hwy') and a 'Model Performance' section showing a 'Distribution of |residual|' plot for the 'lm' model.

```
1 # Load libraries
2 library(tidyverse)
3 library(modelDown) # devtools::install_github("MI2DataLab/modelDown")
4 library(DALEX)
5
6 # Create model
7 lm_model <- lm(cty ~ hwy, data = mpg)
8
9 # Define the explainer
10 explainer_lm <- explain(lm_model, data = mpg %>% select(cty, hwy), y = mpg$cty)
11
12 # Generate the HTML Page
13 modelDown::modelDown(explainer_lm)
```

Tweeted on: 2019-06-19 12:24:38, Retweets: 2, Likes: 6

#rstats and #tidyverse



An **#rstats** pop-quiz: Provide a succinct human readable description of what this **#dplyr** chain is computing. **#tidyverse**

```
library(tidyverse)
library(nycflights13)
library(lubridate)

flights %>%
  mutate(Date = ymd(paste(year, month, day, sep = "-"))) %>%
  group_by(Date, carrier) %>%
  arrange(desc(dep_delay)) %>%
  slice(1) %>%
  group_by(carrier) %>%
  summarise(distance = mean(distance))
```

Tweeted on: 2020-07-30 15:26:57, Retweets: 4, Likes: 6

#rstats and #tidyverse



Our go-to tool for **#webscraping**, the brand new release of {{ rvest }} 1.0 improves the compatibility and performance of both table and text extraction from **#html** pages! **#rstats #tidyverse**

```
> library(rvest)
> 
<- "https://www.w3schools.com/html/html_tables.asp" %>%
+   read_html() %>% html_to_text()
[[1]]
# A tibble: 6 x 3
  Company          Contact      Country
  <chr>            <chr>        <chr>
1 Alfreds Futterkiste  Maria Andress Germany
2 Centro comercial Móstolesno Francisco Chang Mexico
3 Ernst Handel         Roland Mendel Austria
4 Italia Gusto          Helen Bennett UK
5 Laughing Bacchus Winesellers York Tennerup Canada
6 Magazzini Alimentari Riuniti Giovanni Rovelli Italy
```

Company	Contact	Country
Alfreds Futterkiste	Maria Andress	Germany
Centro comercial Móstolesno	Francisco Chang	Mexico
Ernst Handel	Roland Mendel	Austria
Italia Gusto	Helen Bennett	UK
Lauing Bacchus Winesellers	Yoshi Tannamuri	Canada
Magazzini Alimentari Riuniti	Giovanni Rovelli	Italy

Tag	Description
table	Defines a table
thead	Defines a header cell in a table
tbody	Defines a row in a table
tr	Defines a cell in a table
caption	Defines a table caption
colgroup	Specifies a group of one or more columns in a table for formating
col	Specifies column properties for each column within a <colgroup> element
thead	Groups the header content in a table
tbody	Groups the body content in a table
tfoot	Groups the footer content in a table

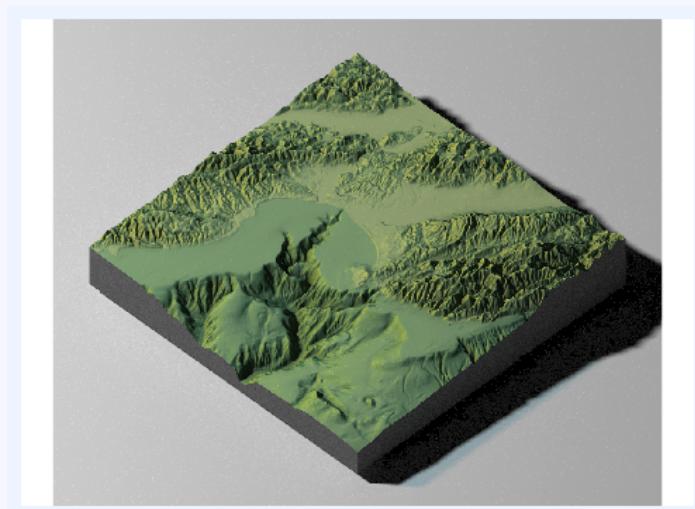
```
library(rvest)
"https://www.w3schools.com/html/html_tables.asp" %>%
  read_html() %>% html_table()
```

Tweeted on: 2021-03-12 15:26:31, Retweets: 5, Likes: 6

#rstats and #GitHub



We're following the **#GitHub** README for the `{rayshader}` package to learn how to render complex scenes in 3D, as well as export high quality snapshots from the rendered scene! More here: <https://github.com/tylermorganwall/rayshader>



```
library(rayshader)

montshadow = ray_shade(montereybay, zscale = 50, lambert = FALSE)
montamb = ambient_shade(montereybay, zscale = 50)
montereypay %>%
  sphere_shade(zscale = 10, texture = "imhof1") %>%
  add_shadow(montamb, 0.5) %>%
  add_shadow(montshadow, 0.5) %>%
  plot_3d(montereybay, zscale = 50, fov = 0, theta = -45, phi = 45,
          windowsize = c(1000, 800), zoom = 0.75,
          water = TRUE, waterdepth = 0, wateralpha = 0.5, watercolor = "lightblue",
          waterlinecolor = "white", waterlinealpha = 0.5)
render_highquality(clamp_value=10, samples=256)
```

Tweeted on: 2021-07-27 18:50:51, Retweets: 2, Likes: 6

#rstats and #python



A common data manipulation task in **#python** is to quickly take two list objects and construct a **#dictionary**. The `zip()` function makes this seamless. For our **#rstats** followers, how would you produce a named list from two distinct lists/vectors in a similar fashion? 🤔

```
my_list = ["apple", "orange", "cherry"]
my_properties = [{"color": "red", "size": "medium"}, {"color": "orange", "size": "medium"}, {"color": "red", "size": "small"}]

my_dict = dict(zip(my_list, my_properties))

Out[45]:
{'apple': {'color': 'red', 'size': 'medium'},
 'orange': {'color': 'orange', 'size': 'medium'},
 'cherry': {'color': 'red', 'size': 'small'}}
```

Tweeted on: 2021-11-19 18:16:27, Retweets: 3, Likes: 6

#rstats and #tidyverse



A major new release of **#rstats**, version 4.1, is now available. Among other features, it introduces a native pipe operator that behaves much the same way as the `{ magrittr }` pipe in the **#tidyverse**. (Left: Old, Right: New)

```
library(tidyverse)

mtcars %>%
  filter(wt > 3) %>%
  group_by(cyl) %>%
  summarise(mpg = mean(mpg))
```

```
library(tidyverse)

mtcars |>
  filter(wt > 3) |>
  group_by(cyl) |>
  summarise(mpg = mean(mpg))
```

Tweeted on: 2021-05-20 21:54:00, Retweets: 4, Likes: 5

#rstats and #tidyverse



The first public release of the new {{ clock }} package provides a set of utilities for manipulating **#datetimes** in **#rstats**, based on the **#date** library in C++. Here we group a set of airquality observations by month and compute summaries over the numeric variables. **#tidyverse**

```
# A tibble: 5 x 7
  Month      Ozone Solar.R  Wind   Temp   Day Year
  <date>     <dbl>    <dbl> <dbl>  <dbl> <dbl> <dbl>
1 1973-05-01  23.6    181. 11.6   65.5   16    1973
2 1973-06-01  29.4    190. 10.3   79.1   15.5  1973
3 1973-07-01  59.1    216.  8.94  83.9   16    1973
4 1973-08-01  60.0    172.  8.79  84.0   16    1973
5 1973-09-01  31.4    167. 10.2   76.9   15.5  1973
```

```
library(tidyverse)
library(clock)

airquality %>%
  mutate(Year = 1973,
        Date = date_build(Year, Month, Day),
        Month = date_group(Date, "month")) %>%
  group_by(Month) %>%
  summarise(across(where(is.numeric), mean, na.rm = TRUE))
```

Tweeted on: 2021-04-05 16:34:15, Retweets: 5, Likes: 4

#rstats and #dataframe



As its name suggests, the {{ arsenal }} package provides, well, an arsenal of tools for large scale summaries, such as variable summary tables, #dataframe comparison, and exporting results to html! #rstats

	4 (N=77)	5 (N=4)	6 (N=78)	8 (N=70)	Total (N=229)	p value
drv						< 0.001
4	23 (29.9%)	0 (0.0%)	32 (41.0%)	48 (68.6%)	103 (45.0%)	
f	54 (70.1%)	4 (100.0%)	42 (53.8%)	1 (1.4%)	101 (44.1%)	
r	0 (0.0%)	0 (0.0%)	4 (5.1%)	21 (30.0%)	25 (10.9%)	
trans						< 0.001
auto(av)	1 (1.3%)	0 (0.0%)	3 (3.8%)	0 (0.0%)	4 (1.7%)	
auto(d3)	2 (2.6%)	0 (0.0%)	0 (0.0%)	0 (0.0%)	2 (0.9%)	
auto(d4)	24 (31.2%)	0 (0.0%)	29 (37.2%)	30 (42.9%)	83 (36.2%)	
auto(l5)	5 (6.5%)	0 (0.0%)	15 (19.2%)	17 (24.3%)	37 (16.2%)	
auto(l6)	0 (0.0%)	0 (0.0%)	2 (2.6%)	4 (5.7%)	6 (2.6%)	
auto(m4)	2 (2.6%)	0 (0.0%)	0 (0.0%)	1 (1.4%)	3 (1.3%)	
auto(s5)	1 (1.3%)	0 (0.0%)	1 (1.3%)	1 (1.4%)	3 (1.3%)	
auto(s6)	4 (5.2%)	2 (50.0%)	5 (6.4%)	5 (7.1%)	16 (7.0%)	
manual(m5)	32 (41.6%)	2 (50.0%)	18 (23.1%)	5 (7.1%)	57 (24.9%)	
manual(m6)	6 (7.8%)	0 (0.0%)	5 (6.4%)	7 (10.0%)	18 (7.9%)	

```
library(arsenal)
library(tidyverse)

mpg_copy <- mpg %>%
  select(-cty) %>%
  mutate(cty_other = hwy * 2) %>%
  slice(-c(1:5))

comparedf(mpg, mpg_copy)

tbl <- tableby(cyl ~ drv + trans, data=mpg_copy)
write2html(tbl, "output.html")
```

```
Compare Object

Function Call:
comparedf(x = mpg, y = mpg_copy)

Shared: 10 non-by variables and 229 observations.
Not shared: 2 variables and 5 observations.

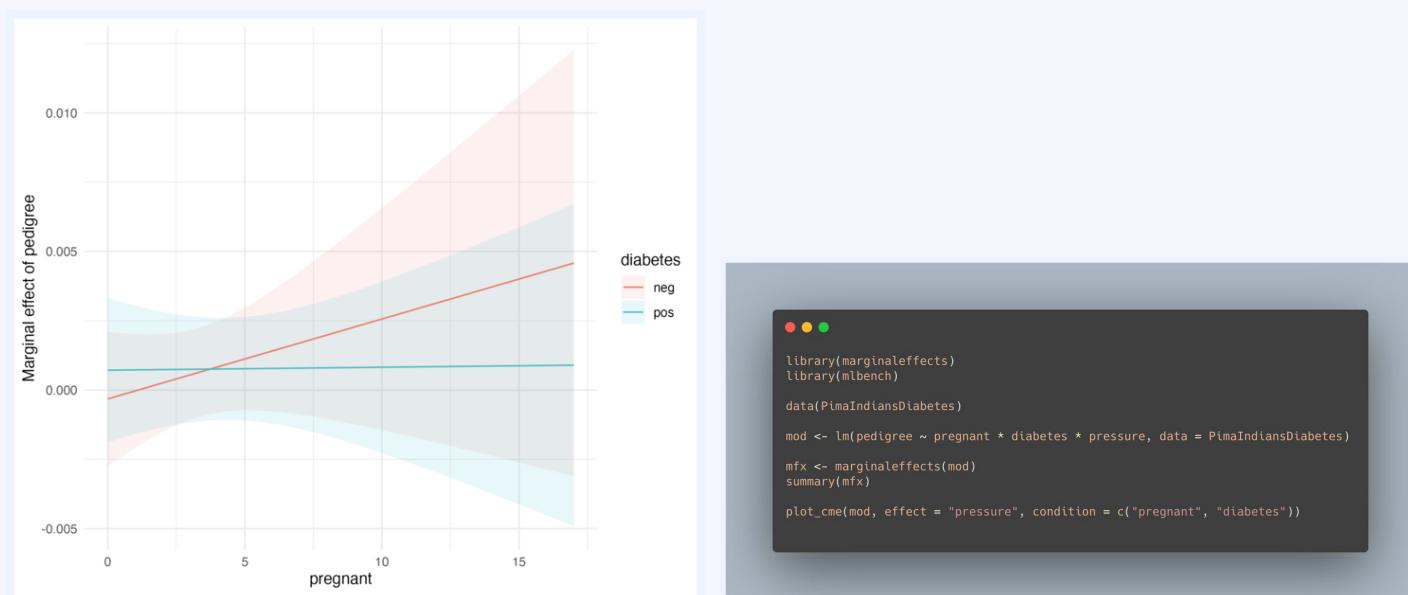
Differences found in 10/10 variables compared.
0 variables compared have non-identical attributes.
> |
```

Tweeted on: 2021-06-15 16:26:11, Retweets: 3, Likes: 4

#rstats and #dataviz



Now on **#CRAN**, `{marginaleffects}` allows **#rstats** users to extract estimates of the marginal effect of a variable conditioned on the values of other variables, and visualize them quickly and seamlessly! **#dataviz**

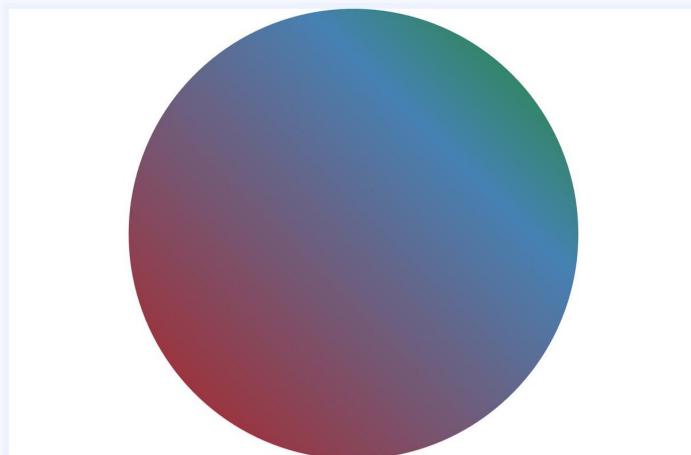


Tweeted on: 2021-10-07 14:24:06, Retweets: 3, Likes: 4

#rstats and #tidyverse



#DidYouKnow that you can draw elegant #color gradients using the #rstats graphic engine? If you update to the latest version of R, plus install companion packages {{ ragg }} and {{ svglite }}, you can produce this! See more in the #tidyverse blog:
<https://www.tidyverse.org/blog/2022/02/new-graphic-features/>



```
library(grid)
grid.circle(
  gp = gpar(
    fill = linearGradient(
      c("firebrick", "steelblue", "forestgreen"),
      stops = c(0, 0.7, 1)
    ),
    col = NA
  )
)
```

Tweeted on: 2022-03-08 21:38:37, Retweets: 3, Likes: 4

#rstats and #programming



#rstats #programming #tipoftheday: If you need to dynamically grow a #vector but know its maximum size in advance, pre-initialize it! Compare the #microbenchmark runtimes of the following two code samples - a 100x speedup!

Dynamically Growing Vectors

```
slow_expr <- function() {
  result <- NULL
  for (i in 1:100000) {
    result <- c(result, i)
  }
}

fast_expr <- function() {
  result <- integer(100000)
  for (i in 1:100000) {
    result[i] <- i
  }
}

microbenchmark::microbenchmark(slow_expr(), unit = "us")
Unit: microseconds
expr      min       lq     mean   median      uq     max neval
slow_expr() 31677.34 38995.9 42249.1 40620.51 44659.66 86773.35    100

microbenchmark::microbenchmark(fast_expr(), unit = "us")
Unit: microseconds
expr      min       lq     mean   median      uq     max neval
fast_expr() 319.639 351.5635 390.7448 376.057 405.8195 711.525    100
```

Tweeted on: 2019-04-17 14:53:37, Retweets: 4, Likes: 3

#rstats and #python



#didyouknow that you dont need #python in order to interact with @Googles #BERT? Using the #Keras-bert pip package and the reticulate #rstats package, we can load, train, and predict with BERT directly in R!

The screenshot shows an RStudio interface. On the left, a script editor window titled "BERT_from_R.r" contains R code for loading the Keras-BERT model and tokenizing text. On the right, a terminal window shows the execution of this code and its output. The terminal output includes the R code, followed by a command to tokenize some input, and the resulting tokens.

```
library(reticulate)
# Choose the version of python to use
use_python("/usr/local/bin/python3", required = TRUE)
# Set the path to your downloaded BERT model
pretrained_path = 'uncased_L-2_H-128_A-2'
config_path = file.path(pretrained_path, 'bert_config.json')
checkpoint_path = file.path(pretrained_path, 'bert_model.ckpt')
vocab_path = file.path(pretrained_path, 'vocab.txt')

# Load the keras bert module, and load the vocabulary
k_bert = import('keras_bert')
token_dict = k_bert$load_vocabulary(vocab_path)
tokenizer = k_bert$Tokenizer(token_dict)

# Tokenize some input!
tokenizer$tokenize("Google's BERT can be used directly within R!")

# Tokenize some input!
> tokenizer$tokenize("This is a test")
[1] "[CLS]" "this" "is" "a" "test" "[SEP]"
> # Tokenize some input!
> tokenizer$tokenize("Google's BERT can be used directly within R!")
[1] "[CLS]" "google" "" "bert" "can" "be"
[8] "used" "directly" "within" "r" "!" "[SEP]"
> |
```

```
library(reticulate)
# Choose the version of python to use
use_python("/usr/local/bin/python3", required = TRUE)
# Set the path to your downloaded BERT model
pretrained_path = 'uncased_L-2_H-128_A-2'
config_path = file.path(pretrained_path, 'bert_config.json')
checkpoint_path = file.path(pretrained_path, 'bert_model.ckpt')
vocab_path = file.path(pretrained_path, 'vocab.txt')

# Load the keras bert module, and load the vocabulary
k_bert = import('keras_bert')
token_dict = k_bert$load_vocabulary(vocab_path)
tokenizer = k_bert$Tokenizer(token_dict)

# Tokenize some input!
tokenizer$tokenize("Google's BERT can be used directly within R!")
```

Tweeted on: 2020-05-26 19:59:02, Retweets: 2, Likes: 3

#rstats and #data



The **@Socrata** API platform is used by non-profits and government organizations all over the world to provide programmatic access to open **#data**. You can use the {{RSocrata}} **#rstats** package to seamlessly access this data (example shown: Mesa, AZ police reports)

```
library(RSocrata)
mesa <- read.socrata("https://data.mesaaz.gov/resource/39rt-2rfj.json")
View(mesa)
```

crime_id	crime_type	report_date	report_month	report_year	other
20160010880	WARRANT ARREST-FOJ ARIZ JURISD	2016-01-01	Jan	2016	2016
20160057001	PROPERTY-LOST	2016-01-01	Jan	2016	2016
20160010024	PUBLIC SEXUAL INDEC-CONTACT	2016-01-01	Jan	2016	2016
20160010034	DUI - LIQUOR - DRUGS	2016-01-01	Jan	2016	2016
20160010095	MISSING JUVENILE	2016-01-01	Jan	2016	2016
20160010182	CRIMINAL DAMAGE-DEFACE - DV	2016-01-01	Jan	2016	2016
20160010262	INFORMATION-ALCOHOL RELATED	2016-01-01	Jan	2016	2016
20160010274	DEATH-NATURAL	2016-01-01	Jan	2016	2016
20160010275	WARRANT ARREST-FOJ ARIZ JURISD	2016-01-01	Jan	2016	2016
20160010287	PROPERTY-FOUND	2016-01-01	Jan	2016	2016
20160010303	DANGEROUS DRUG-POSS-USE	2016-01-01	Jan	2016	2016
20160010402	DISORDERLY CONDUCT-FIGHTING - DV	2016-01-01	Jan	2016	2016
20160010403	WARRANT ARREST-MESA CITY COURT	2016-01-01	Jan	2016	2016
20160010493	FAIL TO STOP/UNATTENDED VEHICLE - COLLISION	2016-01-01	Jan	2016	2016
20160010500	WARRANT ARREST-MESA CITY COURT	2016-01-01	Jan	2016	2016

Tweeted on: 2020-07-27 23:19:10, Retweets: 5, Likes: 3

#rstats and #tidyverse



With a new `#svg` rendering engine and a streamlined codebase extending `{testthat}` 3, the 1.0 release of `{vdiffr}` makes testing for regressions in graphical output even easier, helping ensure our `#ggplot` visualizations are as we expect. `#rstats #tidyverse`

The screenshot shows a dark-themed RStudio interface. On the left, an R script window contains the following code:

```
library(testthat)
library(vdiffr)
library(ggplot2)

test_that("Check histogram appearance", {
  p <- ggplot(mpg) + geom_histogram(aes(x = cty))
  expect_doppelganger("CTY Histogram", p)
})
```

On the right, the terminal window shows the execution results:

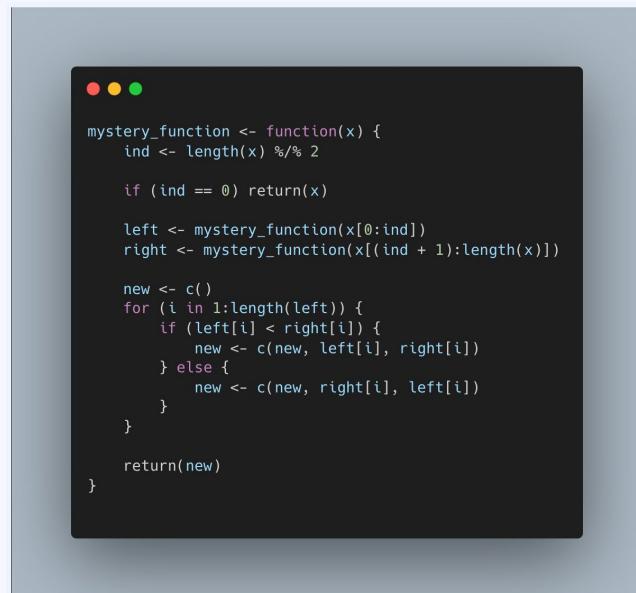
```
> test_that("Check histogram appearance", {
+   p <- ggplot(mpg) + geom_histogram(aes(x = cty))
+   expect_doppelganger("CTY Histogram", p)
+ })
`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
Can't compare snapshot to reference when testing interactively
  Run `devtools::test()` or `testthat::test_file()` to see changes
  New path: /var/folders/x6/13byt17d36g2zh_dx587dbzm0000gn/T//RtmpfzpjX8/cty-histogram176a949a
43223.svg
Test passed ✅
```

Tweeted on: 2021-06-25 21:53:01, Retweets: 3, Likes: 3

#rstats and #compsci



An **#rstats** pop-quiz: This function solves a popular **#compsci** algorithm, for vectors of length 2^n . What comp sci algorithm is it? Can you generalize the code to handle any arbitrary length vector of numeric values? 🤔



```
mystery_function <- function(x) {
  ind <- length(x) %/% 2

  if (ind == 0) return(x)

  left <- mystery_function(x[0:ind])
  right <- mystery_function(x[(ind + 1):length(x)])

  new <- c()
  for (i in 1:length(left)) {
    if (left[i] < right[i]) {
      new <- c(new, left[i], right[i])
    } else {
      new <- c(new, right[i], left[i])
    }
  }

  return(new)
}
```

Tweeted on: 2021-11-29 22:11:41, Retweets: 4, Likes: 3

#rstats and #tidyverse



New release of {{ dtplyr }}, detailed in the **#tidyverse** blog. The package, which allows **#dplyr** verbs to be used with **#datatable** objects, gains support for a wider variety of verbs, combining **#datatables** speed with **#dplyrs** elegant syntax. **#rstats**

```
# A tibble: 53,940 x 11
  carat cut     color clarity depth table price     x     y     z   PPC
  <dbl> <ord> <ord> <ord> <dbl> <dbl> <int> <dbl> <dbl> <dbl> <dbl>
1 0.23 Ideal     E     SI2    61.5    55    326  3.95  3.98  2.43 1417.
2 0.21 Premium   E     SI1    59.8    61    326  3.89  3.84  2.31 1552.
3 0.23 Good      E     VS1    56.9    65    327  4.05  4.07  2.31 1422.
4 0.29 Premium   I     VS2    62.4    58    334  4.2   4.23  2.63 1152.
5 0.31 Good      J     SI2    63.3    58    335  4.34  4.35  2.75 1081.
6 0.24 Very Good J     VVS2   62.8    57    336  3.94  3.96  2.48 1400.
7 0.24 Very Good I     VVS1   62.3    57    336  3.95  3.98  2.47 1400.
8 0.26 Very Good H     SI1    61.9    55    337  4.07  4.11  2.53 1296.
9 0.22 Fair       E     VS2    65.1    61    337  3.87  3.78  2.49 1532.
10 0.23 Very Good H     VS1    59.4    61    338  4     4.05  2.39 1470.
# ... with 53,930 more rows
```

```
library(dtplyr)
library(tidyverse)

dt <- data.table::data.table(diamonds)

dt %>%
  mutate(PPC = price / carat) %>%
  replace_na(list(PPC = 0)) %>%
  as_tibble()
```

Tweeted on: 2021-12-07 15:48:09, Retweets: 4, Likes: 3

#rstats and #python



While we do the majority of our random `#sampling` in `#rstats`, its useful to know that `#python` has its own module called random, where you can quickly and easily sample without (`random.sample`) and with (`random.choices`) replacement.

```
In [12]: import random
...:
...: L = [1, 2, 3, 4, 5]

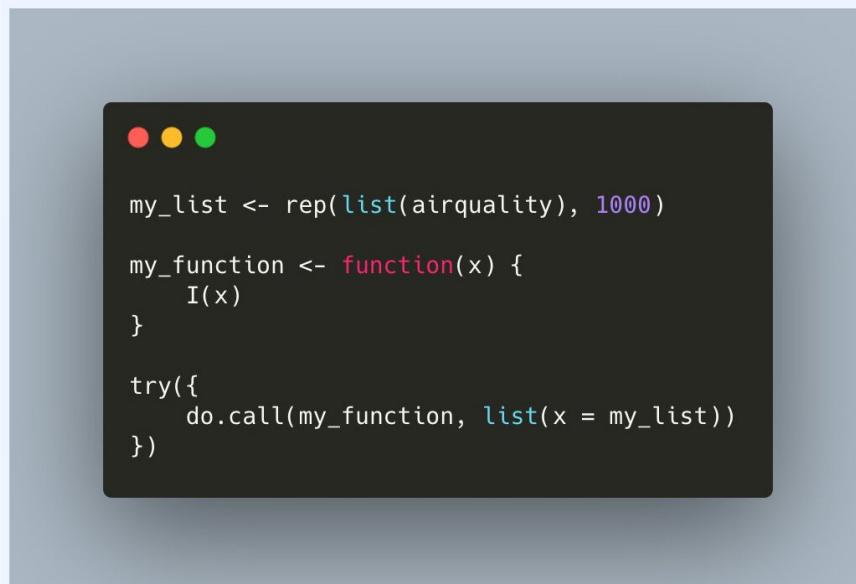
In [13]: random.sample(L, 5)
Out[13]: [4, 1, 2, 3, 5]

In [14]: random.choices(L, k=5)
Out[14]: [3, 1, 5, 1, 3]
```

Tweeted on: 2022-04-20 16:37:33, Retweets: 5, Likes: 3



Minor but important update in the release of #rstats 4.1.2 that not only fixes several segfaults, but also includes performance improvements: In particular, a try() statement in some cases is now much faster, like in this block. Technical details here: <https://stat.ethz.ch/pipermail/r-devel/2021-September/081101.html>



A screenshot of a terminal window showing R code. The code defines a list named 'my_list' and a function named 'my_function'. The 'my_function' calls 'I(x)'. A 'try' block is used to call 'do.call' with 'my_function' and 'list(x = my_list)'.

```
my_list <- rep(list(airquality), 1000)

my_function <- function(x) {
  I(x)
}

try({
  do.call(my_function, list(x = my_list))
})
```

Tweeted on: 2021-11-10 18:38:51, Retweets: 3, Likes: 2

#rstats and #tidyverse



The `{ maybe }` package implements a `#monad` which allows for functions to be written which gracefully handle invalid inputs, like this `safe_log()` function! `{ maybe }` this will be useful to us in our `#rstats` analysis! `#tidyverse`

```
library(maybe)

safe_log <- function(x, base = exp(1)) {
  if (x < 0) nothing() else just(log(x, base = base))
}

safe_log(5)
# Just
# [1] 1.609438

safe_log(-5)
# Nothing>
```

Tweeted on: 2022-02-01 17:46:27, Retweets: 5, Likes: 2

#rstats and #TalkLikeAPirateDay



Want to talk like a pirate today for **#TalkLikeAPirateDay**? Just add “ARRRRRrrrrrrrr!” to all your text strings with **#rstats**

```
> regular_text<- "Make me talk like a pirate!"  
> pirate_text<- paste0(c(regular_text, "ARRRRrrrrr!"), collapse = " ")  
> pirate_text  
[1] "Make me talk like a pirate! ARRRRrrrrr!"
```

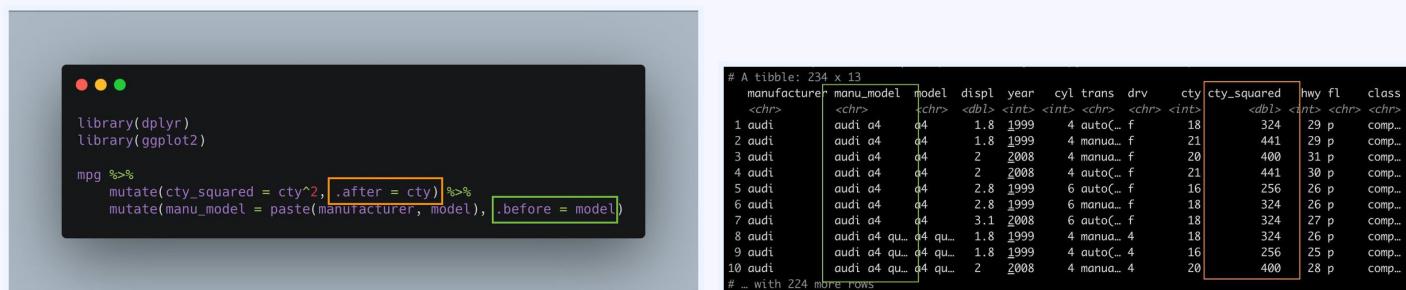
Tweeted on: 2019-09-19 14:15:30, Retweets: 3, Likes: 1

#rstats and #tidyverse



(3) The position of new columns generated by mutate() can now be specified upon creation, with a simple .before and .after argument.

#rstats #tidyverse #TidyTuesday



The screenshot shows an RStudio session. On the left, the code is written in R:

```
library(dplyr)
library(ggplot2)

mpg %>%
  mutate(cty_squared = cty^2, .after = cty) %>%
  mutate(manu_model = paste(manufacturer, model), .before = model)
```

On the right, the resulting tibble is displayed:

manufacturer	model	displ	year	cyl	trans	drv	cty	cty_squared	hwy	fl	class	
audi	a4	0.4	1.8	1999	4	auto_Lf	18	324	29	p	comp...	
audi	a4	0.4	1.8	1999	4	manua_Lf	21	441	29	p	comp...	
audi	a4	0.4	2	2008	4	manua_Lf	20	400	31	p	comp...	
audi	a4	0.4	2	2008	4	auto_Lf	21	441	30	p	comp...	
audi	a4	0.4	2.8	1999	6	auto_Lf	16	256	26	p	comp...	
audi	a4	0.4	2.8	1999	6	manua_Lf	18	324	26	p	comp...	
audi	a4	0.4	3.1	2008	6	auto_Lf	18	324	27	p	comp...	
audi	a4 qu...	0.4 qu...	1.8	1999	4	manua_Lf	4	18	324	26	p	comp...
audi	a4 qu...	0.4 qu...	1.8	1999	4	auto_Lf	4	16	256	25	p	comp...
audi	a4 qu...	0.4 qu...	2	2008	4	manua_Lf	4	20	400	28	p	comp...

... with 224 more rows

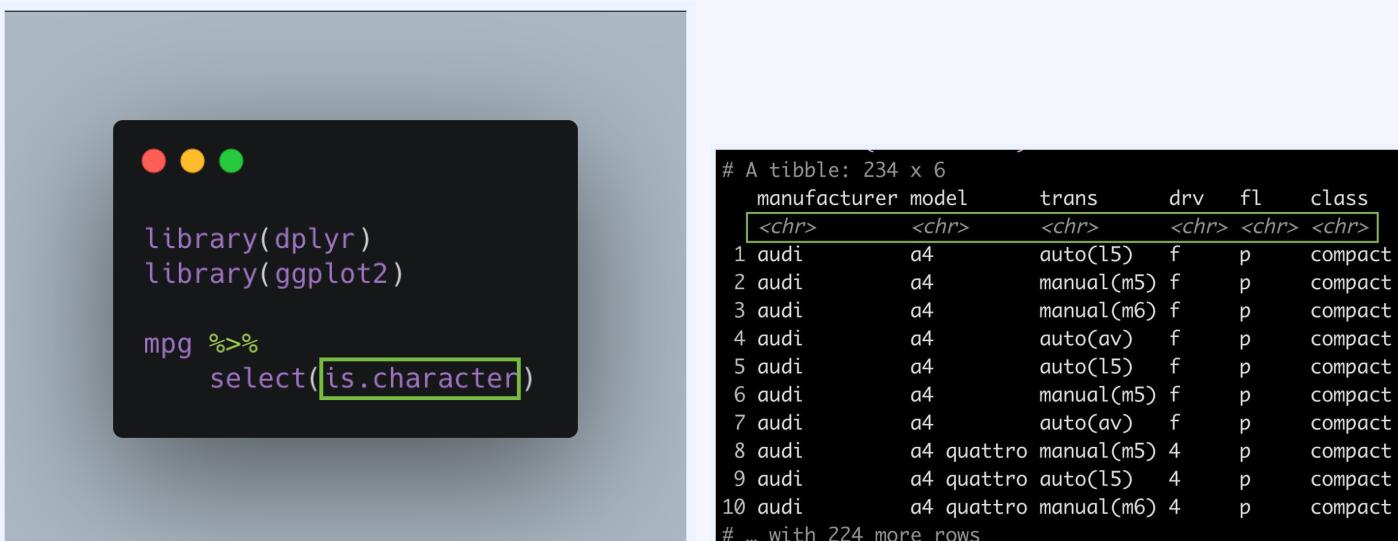
Tweeted on: 2020-03-17 17:55:40, Retweets: 4, Likes: 6

#rstats and #tidyverse



(2) Eliminating the need for the `select_if()` function, `select()` now allows direct selection of columns based on their type, like here where we select all the character columns from the mpg data

#rstats #tidyverse #tidytuesday



The screenshot shows an RStudio session. On the left, there's a dark panel with three colored dots (red, yellow, green). Below it, the R code is displayed:

```
library(dplyr)
library(ggplot2)

mpg %>%
  select(is.character)
```

On the right, the output is shown in a light-colored panel. It starts with a header: "# A tibble: 234 x 6". Below the header is a table with the following data:

	manufacturer	model	trans	drv	f1	class
	<chr>	<chr>	<chr>	<chr>	<chr>	<chr>
1	audi	a4	auto(l5)	f	p	compact
2	audi	a4	manual(m5)	f	p	compact
3	audi	a4	manual(m6)	f	p	compact
4	audi	a4	auto(av)	f	p	compact
5	audi	a4	auto(l5)	f	p	compact
6	audi	a4	manual(m5)	f	p	compact
7	audi	a4	auto(av)	f	p	compact
8	audi	a4	quattro	manual(m5)	4	p
9	audi	a4	quattro	auto(l5)	4	p
10	audi	a4	quattro	manual(m6)	4	p
# ... with 224 more rows						

Tweeted on: 2020-03-17 17:55:40, Retweets: 2, Likes: 6