

statsExpressions: R Package for Tidy Dataframes and Expressions with Statistical Details

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Summary

The statsExpressions package is designed to facilitate producing dataframes with rich statistical details for the most common types of statistical approaches and tests: parametric, nonparametric, robust, and Bayesian t-test, one-way ANOVA, correlation analyses, contingency table analyses, and meta-analyses. The functions are pipe-friendly and provide a consistent syntax to work with tidy data. These dataframes additionally contain expressions with statistical details, and can be used in graphing packages to display these details

Statement of need

The aim of this package is to provide an approachable and intuitive syntax to carry out common statistical tests across diverse statistical approaches.

Comparison to Other Packages

Behind the scenes, statsExpressions uses stats package for parametric and non-parametric (R Core Team, 2021), WRS2 package for robust (Mair & Wilcox, 2020), and BayesFactor package for Bayesian statistics (Morey & Rouder, 2020). Additionally, random-effects meta-analysis is carried out using metafor (parametric) (Viechtbauer, 2010), metaplus (robust) (Beath, 2016), and metaBMA (Bayesian) (Heck et al., 2019) packages. So one can naturally ask why there needs to be another package that wraps around these packages.

There is a lot of diversity among these packages in terms of their syntax and expected input type that can make it difficult to switch from one statistical approach to another. For example, some functions expect vectors as inputs, while others expect dataframes. Depending on whether it is a repeated measures design or not, different functions might expect data to be in wide or long format. Some functions can internally omit missing values, while other functions error in their presence. Furthermore, if someone wishes to utilize the objects returned by these packages downstream in their workflow, this is not straightforward either because even functions from the same package can return a list, a matrix, an array, a dataframe, etc., depending on the function. So on and so forth.

The result of sustained exposure to such inconsistencies is that data exploration can become a cognitively demanding task and discourage users to explore different statistical approaches. In the long run, this might even solidify into a habit of sticking to the defaults without giving much thought to the alternative approaches (e.g., exploring if Bayesian hypothesis testing is to be preferred over null hypothesis significance testing in context of the problem).

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This is where statsExpressions comes in: It can be thought of as a unified portal through which most of the functionality in these underlying packages can be accessed, with a little to no cognitive overhead. The package offers just six primary functions that let users choose a statistical approach without changing the syntax. The users are always expected to provide a dataframe in tidy format (Wickham et al., 2019) to functions, all functions work with missing data, and they always return a dataframe that can be further utilized downstream in the pipeline (for a visualization, e.g.).

Table 1: A summary table listing the primary functions in the package and the statistical approaches they support. For a more detailed description of the tests and outputs from these functions, the readers are encouraged to read vignettes on the package website: https://indrajeetpatil.github.io/statsExpressions/articles/.

Function	Parametric	Non-parametric	Robust	Bayesian
one_sample_test	√	✓	✓	√
two_sample_test	\checkmark	\checkmark	\checkmark	\checkmark
oneway_anova	\checkmark	\checkmark	\checkmark	\checkmark
corr_test	\checkmark	\checkmark	\checkmark	\checkmark
contingency_table	\checkmark	\checkmark	-	\checkmark
meta_analysis	\checkmark	-	\checkmark	\checkmark

Note that, unlike broom (Robinson, Hayes, & Couch, 2021) or parameters (Lüdecke, Ben-Shachar, Patil, & Makowski, 2020), the goal of statsExpressions is not to convert model objects into tidy dataframes, but to provide a consistent and easy syntax to carry out statistical tests.

Tidy Dataframes from Statistical Analysis

All functions return dataframes containing exhaustive details from inferential statistics, and appropriate effect size/posterior estimates and their confidence/credible intervals. The package internally relies on easystats ecosystem of packages to achieve this (Ben-Shachar, Lüdecke, & Makowski, 2020; Lüdecke, Ben-Shachar, Patil, & Makowski, 2020; Lüdecke, Ben-Shachar, Patil, Waggoner, & Makowski, 2021; Lüdecke, Waggoner, & Makowski, 2019; Makowski, Ben-Shachar, & Lüdecke, 2019; Makowski, Ben-Shachar, Patil, & Lüdecke, 2020).

To illustrate the simplicity of this syntax, let's say we want to compare equality of a measure among two independent groups. We can use the two_sample_test function here.

If we first run a parametric t-test:

```
set.seed(123) # for reproducibility
library(statsExpressions) # loading needed package
# Welch's t-test
mtcars %>% two_sample_test(am, wt, type = "parametric")
#> # A tibble: 1 x 14
     term group mean.group1 mean.group2 statistic df.error
                                                                  p.value
#>
     <chr> <chr>
                        <d.b 1.>
                                    <d.b 1.>
                                               <d.b1.>
                                                         <d.b1.>
                                                                     <d.b1.>
                         3.77
#> 1 wt
                                     2.41
                                                5.49
                                                          29.2 0.00000627
#>
     method
                              estimate conf.level conf.low conf.high effectsize
     <chr>
                                  <db1>
                                             <db1>
                                                       <db1>
                                                                 <dbl> <chr>
#> 1 Welch Two Sample t-test
                                   1.84
                                              0.95
                                                        1.00
                                                                  2.66 Hedges' g
```



```
#> expression
#> t>
#> 1 <language>
```

And then decide to run, instead, a robust t-test. The syntax remains the same:

```
set.seed(123) # for reproducibility
# Yuen's t-test
mtcars %>% two_sample_test(am, wt, type = "robust")
#> # A tibble: 1 x 10
                           p.value
#>
     statistic df.error
#>
          <dbl>
                  <dbl>
                              <dbl>
#> 1
          5.84
                   13.6 0.0000485
#>
     method
                                                               estimate conf.low
#>
     <chr>
                                                                  <dbl>
                                                                           <db1>
#> 1 Yuen's test on trimmed means for independent samples
                                                                  0.915
                                                                           0.754
     conf.high conf.level effectsize
                                                                  expression
#>
                     <dbl> <chr>
          <d.b 1.>
                                                                  \langle l.i.s.t. \rangle
          0.977
#> 1
                      0.95 Explanatory measure of effect size <language>
```

These functions also play nicely with other popular data manipulation packages. For example, we can use dplyr to repeat the same analysis across *all* levels of a certain grouping variable:

```
set.seed(123) # for reproducibility
library(dplyr) # loading needed package for groupwise analysis
# running one-sample proportion test for all levels of `cyl`
mtcars %>%
  group_by(cyl) %>%
  group_modify(~ contingency_table(.x, am), .keep = TRUE) %>%
 ungroup()
#> # A tibble: 3 x 11
#>
       cyl statistic
                        df p.value method
#>
     <dbl>
               <dbl> <dbl>
                             <dbl> <chr>
#> 1
               2.27
                         1 0.132
                                   Chi-squared test for given probabilities
         4
#> 2
         6
               0.143
                         1 0.705
                                   Chi-squared test for given probabilities
#> 3
                         1 0.00753 Chi-squared test for given probabilities
         8
               7.14
#>
     estimate conf.level conf.low conf.high effectsize
                                                               expression
#>
        <db1>
                   <dbl>
                            <dbl>
                                      <dbl> <chr>
                                                               t>
#> 1
                                      0.917 Cramer's V (adj.) <language>
        0.344
                    0.95
                            0
                                      0
                                             Cramer's V (adj.) <language>
#> 2
        0
                    0.95
                            0
#> 3
                    0.95
                            0.127
                                      1.18 Cramer's V (adj.) <language>
        0.685
```

Expressions for Plots

In addition to other details contained in the dataframe, there is also a column titled expression, which contains expression with statistical details and can be displayed in a plot (Figure 1). Displaying statistical results in the context of a visualization is indeed a philosophy adopted by the ggstatsplot package (Patil, 2021), and statsExpressions functions as its statistical processing backend.

```
# loading needed packages
set.seed(123) # for reproducibility
```



Kruskal-Wallis Rank Sum Test

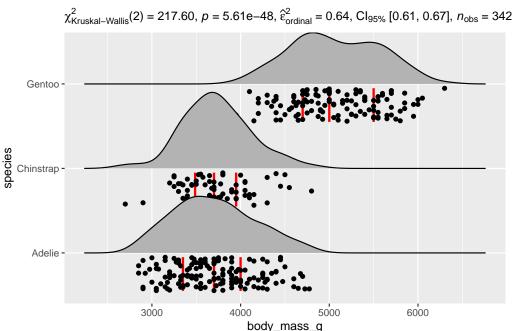


Figure 1: Example illustrating how 'statsExpressions' functions can be used to display results from a statistical test in a plot.

```
library(ggplot2)
library(palmerpenguins) # for data
library(ggridges) # for creating a ridgeplot
# creating a dataframe with results and expression
res <- oneway_anova(penguins, species, body_mass_g, type = "nonparametric")
# create a ridgeplot using `ggridges` package
ggplot(penguins, aes(x = body_mass_g, y = species)) +
  geom_density_ridges(
    jittered_points = TRUE, quantile_lines = TRUE,
    scale = 0.9, vline_size = 1, vline_color = "red",
    position = position_raincloud(adjust_vlines = TRUE)
  ) + \# use 'expression' column to display results in the subtitle
  labs(
    title = "Kruskal-Wallis Rank Sum Test",
    subtitle = res$expression[[1]]
  )
```

The details contained in these expressions (Figure 2) attempt to follow the gold standard in statistical reporting for both Bayesian (Doorn et al., 2020) and non-Bayesian (Association & others, 1985) framework tests.

Licensing and Availability

statsExpressions is licensed under the GNU General Public License (v3.0), with all source code stored at GitHub. In the spirit of honest and open science, requests and



Template for Frequentist analysis



Template for Bayesian analysis

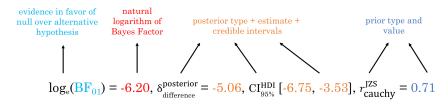


Figure 2: The templates used in 'statsExpressions' to display statistical details in a plot.

suggestions for fixes, feature updates, as well as general questions and concerns are encouraged via direct interaction with contributors and developers by filing an issue while respecting *Contribution Guidelines*.

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