

easystats: A Unified Framework for Statistical Analysis in R

Daniel Lüdtke¹, Mattan S. Ben-Shachar², Indrajeet Patil³, Brenton M. Wiernik⁴, Etienne Bacher⁵, Rémi Thériault⁶, Philip D. Waggoner^{7, 8}, and Dominique Makowski^{9, 10}

1 University Medical Center Hamburg-Eppendorf, Germany **2** Independent Researcher, Ramat Gan, Israel **3** Carl Zeiss AG, Germany **4** Independent Researcher, Tampa, FL, USA **5** Luxembourg Institute of Socio-Economic Research (LISER), Luxembourg **6** Department of Psychology, New York University, New York, NY, USA **7** Department of Applied Mathematics and Statistics, Colorado School of Mines, Golden, CO, USA **8** School of Medicine, Stanford University, Stanford, CA, USA **9** School of Psychology, University of Sussex, Brighton, UK **10** Sussex Centre for Consciousness Science, University of Sussex, Brighton, UK

DOI:

Software

- [Review](#) ↗
- [Repository](#) ↗
- [Archive](#) ↗

Submitted:

Published:

License

Authors of papers retain copyright and release the work under a Creative Commons Attribution 4.0 International License ([CC-BY](#)).

Summary

The **easystats** project is a collection of R packages that provides a unified and intuitive framework for data wrangling and statistical analysis. The ecosystem is built around a philosophy of consistency, user-friendliness, and interoperability between packages, similar to the Tidyverse (Wickham et al., 2019). It aims to simplify the process of preparing, conducting, interpreting, and reporting statistical analyses by offering tools for a wide range of common tasks. These tasks include, for instance, data wrangling (`{datawizard}`, Patil et al., 2022), model assessment (`{performance}`, Lüdtke, Ben-Shachar, Patil, Waggoner, & Makowski, 2021), understanding and describing model parameters (`{parameters}`, Lüdtke, Ben-Shachar, Patil, & Makowski, 2020), including Bayesian models (`{bayestestR}`, Makowski, Ben-Shachar, & Lüdtke, 2019), computation of effect sizes (`{effectsize}`, Ben-Shachar, Lüdtke, & Makowski, 2020), calculating and visualizing marginal effects (`{modelbased}`, Makowski et al., 2025), and generating publication-ready figures (`{see}`, Lüdtke, Patil, et al., 2021) or reports of statistical models (`{report}`, Makowski et al., 2023). The **easystats** ecosystem is designed to be accessible to both novice and experienced R users, promoting reproducible research and a more seamless workflow from data exploration to result communication.

Statement of Need

R is a powerful language for statistical computing, but its capabilities are scattered across a fragmented ecosystem of packages. Conducting a full analysis consisting of modeling, diagnostics, interpretation, and visualization, often requires juggling multiple tools with different syntax, design principles, outputs, and classes. This creates barriers for newcomers and inefficiencies even for experienced users.

The **easystats** ecosystem addresses this challenge by providing a unified framework for wrangling, exploration, modeling, visualization, and reporting in R. Its lightweight, interoperable packages focus on different stages of the statistical workflow, and share consistent syntax and integrate seamlessly, making robust analysis more accessible while reducing cognitive load.

Other software toolkits often serve different purposes. The `{tidyverse}` (Wickham et al., 2019), for example, provides a world-class environment for data manipulation and general-purpose plotting but does not focus on the intricacies of statistical model interpretation and reporting. Specialist packages like `{lme4}` (Bates, Mächler, Bolker, & Walker, 2015) for mixed-effects models or `{marginaleffects}` (Arel-Bundock, Greifer, & Heiss, 2024) for marginal effects and predictions are essential tools, but **easystats** serves as a complementary meta-layer that provides a single, easy-to-learn interface for interacting with the outputs from these and many other modeling packages. This allows analysts and researchers to focus on scientific questions rather than the technical idiosyncrasies of software implementations. **easystats** meets a critical need when doing statistics in R by delivering a coherent, intuitive suite of tools that span the statistical modeling pipeline.

The modularity of the **easystats** packages enables developers to select and use only the necessary components. For example, `{insight}`, a dependency-free package for retrieving model information, is utilized by 45 other CRAN packages, and `{parameters}` is used by 22. In contrast, the `{easystats}` meta-package provides users with a cohesive experience, granting access to the entire ecosystem and its consistent design principles without needing to know the specific package of each function.

A Harmonized and Integrated Workflow

A key design principle of the **easystats** ecosystem is the harmonization and integration of different packages into a simple, sequential workflow. The typical workflow for a statistical analysis using `{easystats}` starts with importing data and bringing the data into shape for the next step - fitting a model - and then sequentially using different functions to obtain a comprehensive understanding of the model.

Let's demonstrate this with an example, where the user starts by preparing some data and then fits a simple linear model:

```
# we don't load each package individually,  
# but rather the entire ecosystem  
library(easystats)  
data(coffee_data, package = "modelbased")  
  
# dichotomize outcome variable  
coffee_data$alertness <- categorize(coffee_data$alertness, lowest = 0)  
  
# rename variable  
coffee_data <- data_rename(coffee_data, select = c(treatment = "coffee"))  
  
# fit model  
model <- glm(  
  alertness ~ treatment,  
  data = coffee_data,  
  family = binomial()  
)
```

The `model` object can then be passed to functions from different **easystats** packages. For instance, the user can get a summary of the model parameters using the `{parameters}` package:

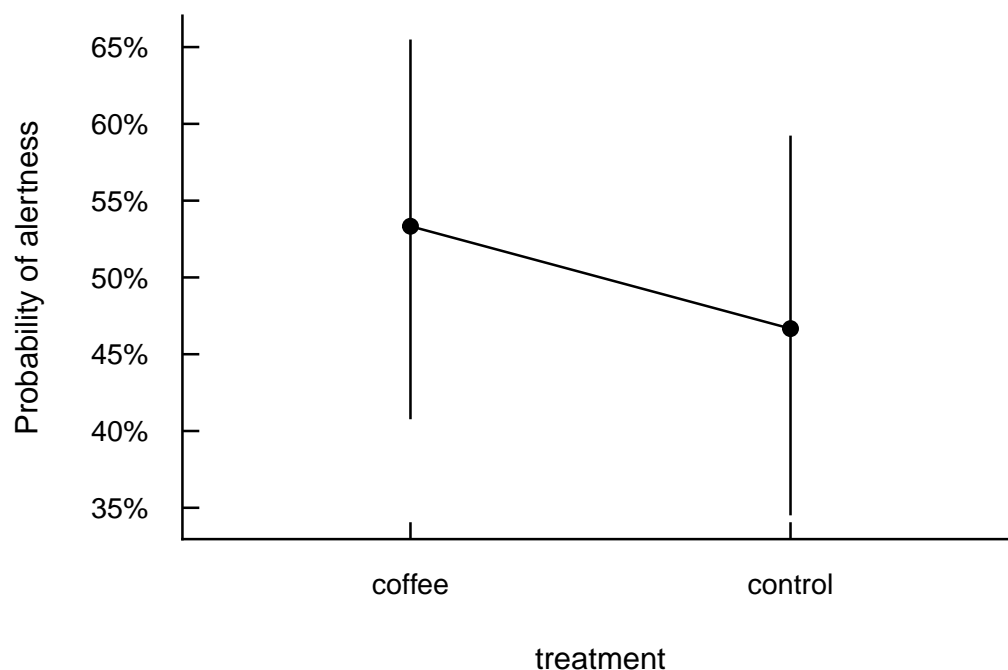


Figure 1: Predicted probability of alertness by treatment group.

```
model_parameters(model)
#> Parameter          / Log-Odds /   SE /          95% CI /    z /    p
#> -----
#> (Intercept)          /    0.13 / 0.26 / [-0.37, 0.65] /  0.52 / 0.606
#> treatment [control] /   -0.27 / 0.37 / [-0.99, 0.45] / -0.73 / 0.466
```

Then, the performance of the model can be assessed with the `{performance}` package:

```
model_performance(
  model,
  metrics = c("AIC", "BIC", "R2", "PCP")
)
#> # Indices of model performance
#>
#> AIC   /   BIC / Tjur's R2 /   PCP
#> -----
#> 169.8 / 175.4 /    0.004 / 0.502
```

The results can be visualized using the `{see}` package by, for example, plotting the model's predictions from `{modelbased}`:

```
predictions <- estimate_means(model, "treatment")
plot(predictions) + theme_modern(show.ticks = TRUE) # add nice theme
```

Finally, a full report of the analysis can be generated with the `{report}` package:

```
report(model)
#> We fitted a logistic model (estimated using ML) to predict alertness with
#> treatment (formula: alertness ~ treatment). The model's explanatory power is
#> very weak (Tjur's R2 = 4.44e-03). The model's intercept, corresponding to
#> treatment = coffee, is at 0.13 (95% CI [-0.37, 0.65], p = 0.606). Within this
#> model:
#>
#> - The effect of treatment [control] is statistically non-significant and
#> negative (beta = -0.27, 95% CI [-0.99, 0.45], p = 0.466; Std. beta = -0.27, 95%
#> CI [-0.99, 0.45])
#>
#> Standardized parameters were obtained by fitting the model on a standardized
#> version of the dataset. 95% Confidence Intervals (CIs) and p-values were
#> computed using a Wald z-distribution approximation.
```

This seamless integration between **easystats** packages allows users to move from model fitting and interpretation to visualization and reporting in a fluid and intuitive way, without having to learn different syntaxes, data structures, or idiosyncratic software design choices.

Licensing and Availability

{**easystats**} is licensed under the MIT-License, with all source code stored at GitHub (<https://github.com/easystats/easystats>), and with a corresponding issue tracker for bug reporting and feature enhancements. In the spirit of honest and open science, we encourage requests, tips for fixes, feature updates, as well as general questions and concerns via direct interaction with contributors and developers.

Acknowledgments

{**easystats**} is part of the collaborative **easystats ecosystem**. Thus, we thank all **members of easystats**, contributors, and users alike.

References

- Arel-Bundock, V., Greifer, N., & Heiss, A. (2024). How to Interpret Statistical Models Using **marginaleffects** for *r* and *python*. *Journal of Statistical Software*, 111(9). doi:[10.18637/jss.v111.i09](https://doi.org/10.18637/jss.v111.i09)
- Bates, D., Mächler, M., Bolker, B., & Walker, S. (2015). Fitting Linear Mixed-Effects Models Using *lme4*. *Journal of Statistical Software*, 67(1), 1–48. doi:[10.18637/jss.v067.i01](https://doi.org/10.18637/jss.v067.i01)
- Ben-Shachar, M. S., Lüdtke, D., & Makowski, D. (2020). effectsize: Estimation of effect size indices and standardized parameters. *Journal of Open Source Software*, 5(56), 2815. doi:[10.21105/joss.02815](https://doi.org/10.21105/joss.02815)
- Lüdtke, D., Ben-Shachar, M. S., Patil, I., & Makowski, D. (2020). parameters: Extracting, computing and exploring the parameters of statistical models using R. *Journal of Open Source Software*, 5(53), 2445. doi:[10.21105/joss.02445](https://doi.org/10.21105/joss.02445)

- Lüdecke, D., Ben-Shachar, M. S., Patil, I., Waggoner, P., & Makowski, D. (2021). performance: An R package for assessment, comparison and testing of statistical models. *Journal of Open Source Software*, 6(60), 3139. doi:[10.21105/joss.03139](https://doi.org/10.21105/joss.03139)
- Lüdecke, D., Patil, I., Ben-Shachar, M., Wiernik, B., Waggoner, P., & Makowski, D. (2021). See: An R Package for Visualizing Statistical Models. *Journal of Open Source Software*, 6(64), 3393. doi:[10.21105/joss.03393](https://doi.org/10.21105/joss.03393)
- Makowski, D., Ben-Shachar, M. S., & Lüdecke, D. (2019). bayestestR: Describing effects and their uncertainty, existence and significance within the bayesian framework. *Journal of Open Source Software*, 4(40), 1541. doi:[10.21105/joss.01541](https://doi.org/10.21105/joss.01541)
- Makowski, D., Ben-Shachar, M. S., Wiernik, B. M., Patil, I., Thériault, R., & Lüdecke, D. (2025). modelbased: An R package to make the most out of your statistical models through marginal means, marginal effects, and model predictions. *Journal of Open Source Software*, 10(109), 7969. doi:[10.21105/joss.07969](https://doi.org/10.21105/joss.07969)
- Makowski, D., Lüdecke, D., Patil, I., Thériault, R., Ben-Shachar, M. S., & Wiernik, B. M. (2023). Automated results reporting as a practical tool to improve reproducibility and methodological best practices adoption. *CRAN*. doi:[10.32614/CRAN.package.report](https://doi.org/10.32614/CRAN.package.report)
- Patil, I., Makowski, D., Ben-Shachar, M. S., Wiernik, B. M., Bacher, E., & Lüdecke, D. (2022). datawizard: An R package for easy data preparation and statistical transformations. *Journal of Open Source Software*, 7(78), 4684. doi:[10.21105/joss.04684](https://doi.org/10.21105/joss.04684)
- Wickham, H., Averick, M., Bryan, J., Chang, W., McGowan, L. D., François, R., Grolemond, G., et al. (2019). Welcome to the tidyverse. *Journal of Open Source Software*, 4(43), 1686. doi:[10.21105/joss.01686](https://doi.org/10.21105/joss.01686)