




UnfoldMakie.jl: EEG/ERP visualization package


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Statement of need

UnfoldMakie.jl is a Julia package for visualization of electroencephalography (EEG) data, with a focus on event-related potentials (ERPs) and regression-ERPs (rERPs). UnfoldMakie.jl fills a gap as one of the only dedicated EEG visualization libraries and offers ease of use, customization, speed, and detailed documentation. It allows for both explorative analysis (zooming/panning) and export to publication-ready vector graphics. This is achieved via multiple backends of Makie.jl ([Danisch & Krumbiegel, 2021](#)): vector graphics with CairoMakie.jl, GPU-accelerated interactive graphics with GLMakie.jl and the browser-based WGLMakie.jl.

In an earlier study ([Mikheev et al., 2024](#)), we asked novice and expert practitioners for insights into their ERP visualization practices. The results of this survey were used to develop and improve UnfoldMakie.jl. Further, UnfoldMakie.jl is agnostic (independent) of any specific analysis framework, but it nicely accompanies the rERP analysis package Unfold.jl ([Ehinger & Dimigen, 2019](#)).

The target audience of this package is anyone working with EEG, MEG, ERP, iEEG or other time-series data.

Functionality

UnfoldMakie.jl excels in various fields:

- **Focused.** UnfoldMakie.jl focuses exclusively on visualizing (r)ERPs, unlike other toolboxes such as EEGLAB, FieldTrip, Brainstorm, or MNE. This makes it easier to understand, use, and maintain the package in the future.
- **Customizable.** The majority of EEG researchers perceive the flexibility of coding as the most important feature of the EEG toolbox ([Mikheev et al., 2024](#)). Consequently, users of UnfoldMakie.jl have great flexibility in customizing most aspects of the plots, such as colors, line styles, and axis decorations.
- **Combining plots.** The layout system for subplots in Makie.jl makes it easy to combine and align various plot types.
- **Flexible mapping.** With AlgebraOfGraphics.jl, it is easy to map conditions, groups, channels, or other features, to a large variety of aesthetics like linestyle, color, marker and others. It works similarly to the popular R package ggplot2.
- **Flexible data input.** All functions support basic Arrays and tidy DataFrames.
- **Fast.** Julia and Makie.jl allows fast plotting of figures with very complex layouts. As an example, it is able to plot one figure with 50 topoplots in 1.9 seconds (1.6 sec with

DelaunayMesh interpolation), which is ~20 times faster than EEGLAB. Although, the Python-based MNE is faster by one second (Table 1).¹

- Faster updating. Makie.jl is incredibly fast at updating figures, which is beneficial for developing interactive tools and creating animations. UnfoldMakie.jl can create and save a topoplot gif file with 50 frames in 1.7 times faster than MNE (Table 2).
- Interactive. Several plots in our package have interactive features. They are supported by Observables.jl, which allows for fast data exchange and quick plot updating.
- Scientific color maps. According to our previous study (Mikheev et al., 2024), 40% of EEG researchers are not aware of the issue of perceptually non-uniform color maps. UnfoldMakie.jl uses scientific color maps throughout (Crameri et al., 2020) (Moreland, 2015).
- Documented. There is extensive documentation with many usage examples and docstrings (Hatherly et al., 2024).

Table 1: Table 1. Benchmark for a topoplot series with 50 topolots

Languages	Package	Median speed
MATLAB	EEGLAB	36 sec
Julia	UnfoldMakie	1.998 sec
Python	MNE	0.826 sec

Table 2: Table 2. Benchmark for generating (and saving) a topoplot animation with 50 timepoints. No similar functionality exists in EEGLAB.

Languages	Package	Median speed
Julia	UnfoldMakie (.gif)	5.801 sec
Python	MNE (.gif)	9.494 sec

We currently support nine general EEG plot types (Figure 1) and two Unfold-specific plots: Design matrices and Spline plots.

State of the field

There are dozens of libraries in Python and MATLAB for ERP analysis and visualization. According to a recent survey (Mikheev et al., 2024), most EEG practitioners (82%) have experience with MATLAB-based tools like EEGLAB (Delorme & Makeig, 2004), FieldTrip (Oostenveld et al., 2011), ERPLAB (Lopez-Calderon & Luck, 2014) and Brainstorm (Tadel et al., 2019). The Python-based MNE (Gramfort et al., 2013) (41%) and the commercial software Brain Vision Analyzer (22%) further showed strong popularity. None of these toolboxes focuses particularly on visualizations. Indeed, in terms of specialized EEG visualization toolboxes, we are aware of only two such libraries, both MATLAB-based and both named eegvis (Robbins, 2012) and (Ehinger, 2018).

Few EEG/ERP analysis and/or visualization libraries have been written in Julia. We are aware of NeuroAnalyzer.jl (Wysokiński, 2024), EEGToolkit.jl (Pereyra, 2024), Neuroimaging.jl (Luke, 2021). There are also [traces](#) of several abandoned projects. Worth highlighting is PyMNE.jl, a wrapper for the Python-MNE toolbox. (Alday & Arslan, 2024)

¹Be aware that results of benchmarking can vary each run and depends on a OS, package environment, other processes running on computer etc. Current measurements were done by using BenchmarkTools.jl. (Chen & Revels, 2016)

However, all these packages are focused on the analysis of EEG data, while our package is specialized on the visualization of ERPs and rERPs. This is the gap that `UnfoldMakie.jl` fills.

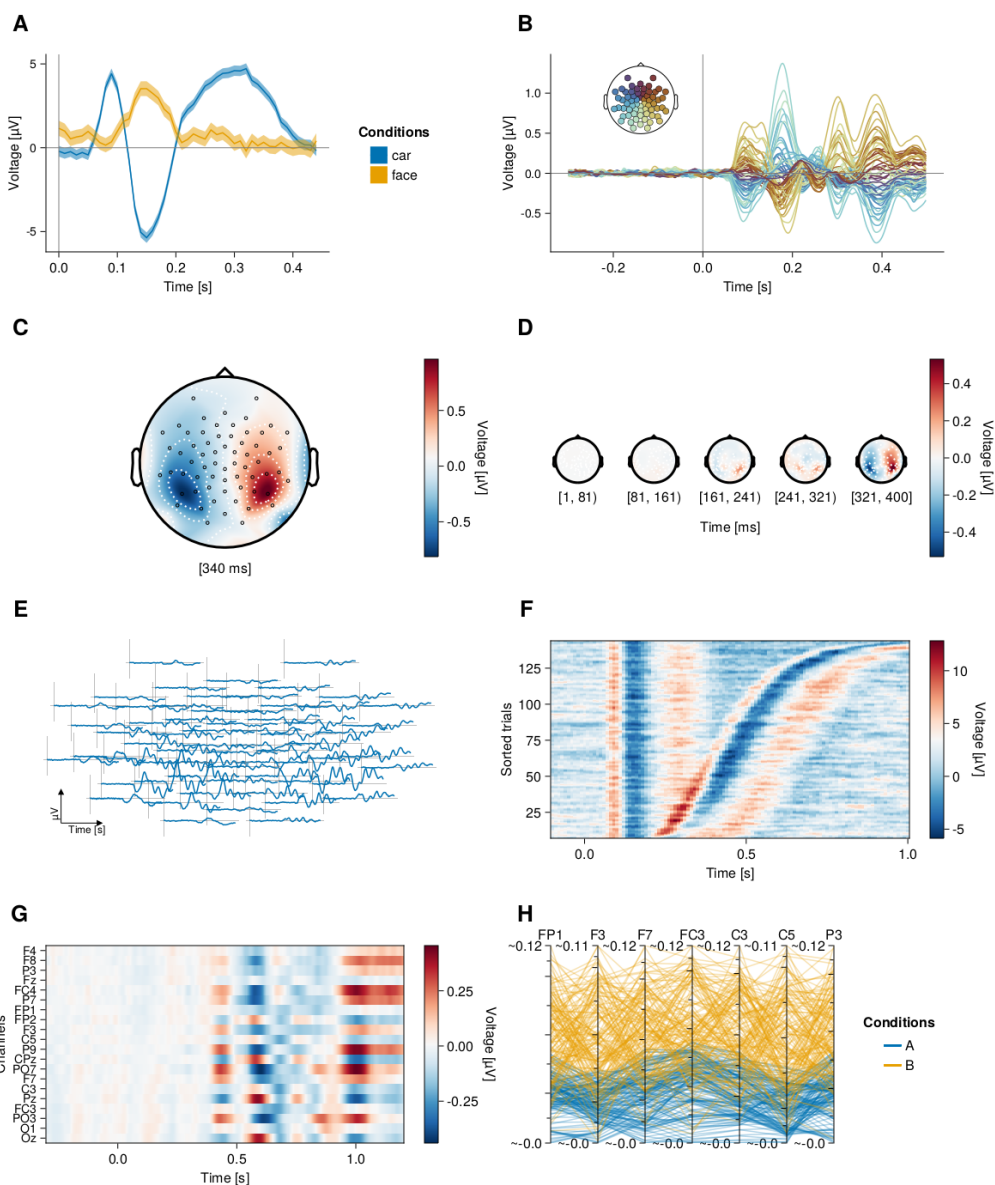


Figure 1: 8 plots generated by `UnfoldMakie.jl`. A) ERP plot, B) Butterfly plot, C) Topoplot, D) Topoplot timeseries, E) ERP image (Jung et al., 1998), G) Channel image, H) Parallel coordinate plot (Ten Caat et al., 2006).

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Toolbox dependencies

Only the main dependencies are listed here. The dependencies of the dependencies can be found in the respective Manifest.toml files. Furthermore, please note that we only list rather than cite the packages for which we could not find any citation instructions.

AlgebraOfGraphics.jl (Krumbiegel & Vertechi, 2024), BenchmarkTools.jl, BSplineKit.jl, CSV.jl, CairoMakie.jl, CategoricalArrays.jl, ColorSchemes.jl, ColorTypes.jl, Colors.jl, CoordinateTransformations.jl, DataFrames.jl (Bouchet-Valat & Kamiński, 2023), DataFramesMeta.jl, DataStructures.jl, DocStringExtensions.jl, Documenter.jl (Hatherly et al., 2024), DSP.jl (Kornblith et al., 2023), FFMPEG_jll.jl, GeometryBasics.jl, Glob.jl, GridLayoutBase.jl, ImageFiltering.jl, Interpolations.jl, Literate.jl, LiveServer.jl, LinearAlgebra.jl, Makie.jl (Danisch & Krumbiegel, 2021), MakieThemes.jl, Observables.jl, PyMNE.jl (Alday & Arslan, 2024), PythonCall.jl, PythonPlot.jl, Random.jl, Revise.jl, SparseArrays.jl, StaticArrays.jl, Statistics.jl, StatsModels.jl, Test.jl, TopoPlots.jl (Ehinger et al., 2024), Unfold.jl (Ehinger & Dimigen, 2019), UnfoldSim.jl (Schepers et al., 2024), XML2_jll.jl

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