

# Package ‘autohrf’

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**Type** Package

**Title** Automated Generation of Data-Informed GLM Models in Task-Based fMRI Data Analysis

**Version** 1.1.0

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## Description

Analysis of task-related functional magnetic resonance imaging (fMRI) activity at the level of individual participants is commonly based on general linear modelling (GLM) that allows us to estimate to what extent the blood oxygenation level dependent (BOLD) signal can be explained by task response predictors specified in the GLM model. The predictors are constructed by convolving the hypothesised timecourse of neural activity with an assumed hemodynamic response function (HRF). To get valid and precise estimates of task response, it is important to construct a model of neural activity that best matches actual neuronal activity. The construction of models is most often driven by predefined assumptions on the components of brain activity and their duration based on the task design and specific aims of the study. However, our assumptions about the onset and duration of component processes might be wrong and can also differ across brain regions. This can result in inappropriate or suboptimal models, bad fitting of the model to the actual data and invalid estimations of brain activity. Here we present an approach in which theoretically driven models of task response are used to define constraints based on which the final model is derived computationally using the actual data. Specifically, we developed 'autohrf' — a package for the 'R' programming language that allows for data-driven estimation of HRF models. The package uses genetic algorithms to efficiently search for models that fit the underlying data well. The package uses automated parameter search to find the onset and duration of task predictors which result in the highest fitness of the resulting GLM based on the fMRI signal under predefined restrictions. We evaluate the usefulness of the 'autohrf' package on publicly available datasets of task-related fMRI activity. Our results suggest that by using 'autohrf' users can find better task related brain activity models in a quick and efficient manner.

**License** GPL (>= 3)

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**LazyData** true

**Imports** cowplot (>= 1.1.1), doParallel (>= 1.0.17), dplyr (>= 1.0.8),  
foreach (>= 1.5.2), ggplot2 (>= 3.3.5), gtools (>= 3.9.2),  
lubridate (>= 1.8.0), magrittr (>= 2.0.2), RColorBrewer (>=

1.1)

**Suggests** knitr (>= 1.38), testthat (>= 3.1.3), tidyverse (>= 1.3.0)**VignetteBuilder** knitr**RoxygenNote** 7.2.0**URL** <https://github.com/demsarjure/autohrf>**BugReports** <https://github.com/demsarjure/autohrf/issues>**NeedsCompilation** no**Author** Jure Demšar [cre, aut],  
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Grega Repovš [aut]**Depends** R (>= 3.5.0)**Repository** CRAN**Date/Publication** 2022-11-19 18:20:02 UTC

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autohrf	<i>autohrf</i>
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## Description

A function that automatically finds the parameters of model's that best match the underlying data.

## Usage

```
autohrf(
  d,
  model_constraints,
  tr,
  roi_weights = NULL,
  allow_overlap = FALSE,
  population = 100,
  iter = 100,
  mutation_rate = 0.1,
  mutation_factor = 0.05,
  elitism = 0.1,
  hrf = "spm",
  t = 32,
  p_boynton = c(2.25, 1.25, 2),
  p_spm = c(6, 16, 1, 1, 6, 0),
  f = 100,
  cores = NULL,
  autohrf = NULL,
  verbose = TRUE
)
```

## Arguments

d	A dataframe with the signal data: roi, t and y. ROI is the name of the region, t is the timestamp and y the value of the signal.
model_constraints	A list of model specifications to use for fitting. Each specification is represented as a data frame containing information about it (event, start_time, end_time, min_duration and max_duration).
tr	MRI's repetition time.
roi_weights	A data frame with ROI weights: roi, weight. ROI is the name of the region, weight a number that defines the importance of that roi, the default weight for a ROI is 1. If set to 2 for a particular ROI that ROI will be twice as important.
allow_overlap	Whether to allow overlap between events.
population	The size of the population in the genetic algorithm.
iter	Number of iterations in the genetic algorithm.



---

autohrf-datasets	<i>Datasets for autohrf examples Example datasets for use in <b>autohrf</b> examples and vignettes. The datasets were extracted from the internal Mind and Brain Lab's (MBLab, <a href="http://www.mblab.si">http://www.mblab.si</a> repository. MBLab is a research lab at the Faculty of Arts, Department of Psychology, University of Ljubljana, Slovenia.</i>
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## Description

Datasets for autohrf examples Example datasets for use in **autohrf** examples and vignettes. The datasets were extracted from the internal Mind and Brain Lab's (MBLab, <http://www.mblab.si> repository. MBLab is a research lab at the Faculty of Arts, Department of Psychology, University of Ljubljana, Slovenia.

## Format

swm fMRI dataset for a spatial working memory experiment.

Source: Internal MBLab repository.

11520 obs. of 3 variables

- roi region of interest.
- time time stamp.
- y BOLD value.

swm\_autofit Stored results from a pre-completed autohrf run.

Source: Internal MBLab repository.

swm\_autofit1 Stored results from a pre-completed autohrf run.

Source: Internal MBLab repository.

swm\_autofit2 Stored results from a pre-completed autohrf run.

Source: Internal MBLab repository.

flanker fMRI dataset for a flanker experiment.

Source: Internal MBLab repository.

192 obs. of 3 variables

- roi region of interest.
- time time stamp.
- y BOLD value.

flanker\_autofit Stored results from a pre-completed autohrf run.

Source: Internal MBLab repository.

## Examples

```
# load swm data
data_swm <- swm
```

```
# load the previously completed autofits
```

```

autofit <- swm_autofit
autofit1 <- swm_autofit1
autofit2 <- swm_autofit2

# load flanker data
data_flanker <- flanker

# load the previously completed autofits
autofit3 <- flanker_autofit

```

---

convolve\_events      *convolve\_events*

---

### Description

A helper function for convolving events of a model with a generated HRF signal.

### Usage

```

convolve_events(
  model,
  tr,
  max_duration,
  hrf = "spm",
  t = 32,
  p_boynton = c(2.25, 1.25, 2),
  p_spm = c(6, 16, 1, 1, 6, 0),
  f = 100
)

```

### Arguments

model	A data frame containing information about the model to use and its events (event, start_time and duration).
tr	MRI's repetition time.
max_duration	Maximum duration of the signal.
hrf	Method to use for HRF generation, can be "boynton" or "spm".
t	The t parameter for Boynton or SPM HRF generation.
p_boynton	Parameters for the Boynton's HRF.
p_spm	Parameters for the SPM HRF.
f	Upsampling factor.

### Value

Returns a list with the convolved signal and time series.

---

convolve_hrf	<i>convolve_hrf</i>
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---

**Description**

A helper function for convolving HRF with a signal.

**Usage**

```
convolve_hrf(y, hrf_s)
```

**Arguments**

y	The signal.
hrf_s	The HRF.

**Value**

Returns the convolution between HRF and the signal.

---

create_boynton_hrf	<i>create_boynton_hrf</i>
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---

**Description**

A helper function for creating a Boynton HRF.

**Usage**

```
create_boynton_hrf(tr, t = 32, p = c(2.25, 1.25, 2))
```

**Arguments**

tr	MRI's repetition time.
t	The t parameter for Boynton or SPM HRF generation.
p	Parameters for the Boynton's HRF.

**Value**

Returns a Boynton HRF function.

---

create_child	<i>create_child</i>
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---

**Description**

A helper function for creating a child from parents.

**Usage**

```
create_child(  
  start_time,  
  end_time,  
  n_events,  
  mutation_rate,  
  mutation_factor,  
  current_model,  
  p1,  
  p2,  
  allow_overlap  
)
```

**Arguments**

start_time	A list with model's event start times.
end_time	A list with model's event end times.
n_events	Number of events in the model.
mutation_rate	The mutation rate in the genetic algorithm.
mutation_factor	The mutation factor in the genetic algorithm.
current_model	The constraints of the current model.
p1	The first selected parent.
p2	The second selected parent.
allow_overlap	Whether to allow overlap between events.

**Value**

A child model created from two parents.



---

`create_first_generation`  
*create\_first\_generation*

---

### **Description**

A helper function for creating the first generation.

### **Usage**

```
create_first_generation(current_model, n_events, population, allow_overlap)
```

### **Arguments**

`current_model` The constraints of the current model.  
`n_events` Number of events in the model.  
`population` The size of the population in the genetic algorithm.  
`allow_overlap` Whether to allow overlap between events.

### **Value**

Returns the first generation of models.

---

`create_new_generation` *create\_new\_generation*

---

### **Description**

A helper function for creating a new generation of possible solutions.

### **Usage**

```
create_new_generation(  
    elitism,  
    population,  
    start_time,  
    end_time,  
    fitness,  
    n_events,  
    mutation_factor,  
    mutation_rate,  
    current_model,  
    allow_overlap  
)
```

**Arguments**

elitism	The degree of elitism (promote a percentage of the best solutions) in the genetic algorithm.
population	The size of the population in the genetic algorithm.
start_time	A list with model's event start times.
end_time	A list with model's event end times.
fitness	A fitness score of all candidate models.
n_events	Number of events in the model.
mutation_factor	The mutation factor in the genetic algorithm.
mutation_rate	The mutation rate in the genetic algorithm.
current_model	The constraints of the current model.
allow_overlap	Whether to allow overlap between events.

**Value**

A new generation of candidate models.

---

create_spm_hrf	<i>create_boynton_hrf</i>
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---

**Description**

A helper function for creating a SPM HRF.

**Usage**

```
create_spm_hrf(tr, t = 32, p = c(6, 16, 1, 1, 6, 0))
```

**Arguments**

tr	MRI's repetition time.
t	The t parameter for Boynton or SPM HRF generation.
p	Parameters for the SPM HRF.

**Value**

Returns a SPM HRF function.

---

downsample	<i>downsample</i>
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---

**Description**

A helper function for downsampling a given signal.

**Usage**

```
downsample(y, f = 100)
```

**Arguments**

y	The signal.
f	Upsampling factor.

**Value**

Returns the downsampled signal.

---

evaluate_model	<i>evaluate_model</i>
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---

**Description**

A function for evaluating the model against the data.

**Usage**

```
evaluate_model(  
  d,  
  model,  
  tr,  
  roi_weights = NULL,  
  hrf = "spm",  
  t = 32,  
  p_boynton = c(2.25, 1.25, 2),  
  p_spm = c(6, 16, 1, 1, 6, 0),  
  f = 100,  
  verbose = TRUE  
)
```

**Arguments**

d	A dataframe with the signal data: roi, t and y. ROI is the name of the region, t is the timestamp and y the value of the signal.
model	A data frame containing information about the model to use and its events (event, start_time and duration).
tr	MRI's repetition time.
roi_weights	A data frame with ROI weights: roi, weight. ROI is the name of the region, weight a number that defines the importance of that roi, the default weight for a ROI is 1. If set to 2 for a particular ROI that ROI will be twice as important.
hrf	Method to use for HRF generation, can be "boynton" or "spm".
t	The t parameter for Boynton or SPM HRF generation.
p_boynton	Parameters for the Boynton's HRF.
p_spm	Parameters for the SPM HRF.
f	Upsampling factor.
verbose	Whether to print a report of the evaluation results.

**Value**

Returns a list that contains the model, fits of events for each ROI, convolved events, TR and evaluation scores for each ROI.

**Examples**

```
# create the model
m <- data.frame(event = c("encoding", "delay", "response"),
start_time = c(0, 2.5, 12.5), duration = c(2.5, 10, 5))

# evaluate
df <- flanker
res <- evaluate_model(df, m, tr = 2.5)
```

---

fit\_to\_constraints      *fit\_to\_constraints*

---

**Description**

A helper function for fitting a model to constraints.

**Usage**

```

fit_to_constraints(
  model_id,
  d,
  model_constraints,
  tr,
  roi_weights,
  allow_overlap,
  population,
  iter,
  mutation_rate,
  mutation_factor,
  elitism,
  hrf,
  t,
  p_boynton,
  p_spm,
  f,
  autohrf = NULL,
  verbose = TRUE
)

```

**Arguments**

model_id	ID of the model.
d	A dataframe with the signal data: roi, t and y. ROI is the name of the region, t is the timestamp and y the value of the signal.
model_constraints	A list of model specifications to use for fitting. Each specification is represented as a data frame containing information about it (event, start_time, end_time, min_duration and max_duration).
tr	MRI's repetition time.
roi_weights	A data frame with ROI weights: roi, weight. ROI is the name of the region, weight a number that defines the importance of that roi, the default weight for a ROI is 1. If set to 2 for a particular ROI that ROI will be twice as important.
allow_overlap	Whether to allow overlap between events.
population	The size of the population in the genetic algorithm.
iter	Number of iterations in the genetic algorithm.
mutation_rate	The mutation rate in the genetic algorithm.
mutation_factor	The mutation factor in the genetic algorithm.
elitism	The degree of elitism (promote a percentage of the best solutions) in the genetic algorithm.
hrf	Method to use for HRF generation.
t	The t parameter for Boynton or SPM HRF generation.

<code>p_boynton</code>	Parameters for the Boynton's HRF.
<code>p_spm</code>	Parameters for the SPM HRF.
<code>f</code>	Upsampling factor.
<code>autohrf</code>	Results of a previous autohrf run to continue.
<code>verbose</code>	Whether to print progress of the fitting process.

**Value**

Returns the best model given provided constraints.

---

<code>get_best_models</code>	<i>get_best_models</i>
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---

**Description**

Returns and prints the best fitted model for each of the specs used in autohrf.

**Usage**

```
get_best_models(autofit, return_fitness = FALSE, verbose = TRUE)
```

**Arguments**

<code>autofit</code>	Output of the autohrf function.
<code>return_fitness</code>	Whether to return models or fitness.
<code>verbose</code>	Whether to print information or only return the result.

**Value**

Returns a list containing the best models for each of the provided constraints.

**Examples**

```
# prepare model specs
model3 <- data.frame(
  event      = c("encoding", "delay", "response"),
  start_time = c(0,          2.65,   12.5),
  end_time   = c(3,          12.5,   16)
)

model4 <- data.frame(
  event      = c("fixation", "target", "delay", "response"),
  start_time = c(0,          2.5,    2.65,  12.5),
  end_time   = c(2.5,        3,      12.5,  15.5)
)

model_constraints <- list(model3, model4)
```

```
# run autohrf
df <- flanker
autofit <- autohrf(df, model_constraints, tr = 2.5,
                  population = 2, iter = 2, cores = 1)

# print best models
get_best_models(autofit)
```

---

get_parents	<i>get_parents</i>
-------------	--------------------

---

**Description**

A helper function for getting parents for the child model.

**Usage**

```
get_parents(fitness)
```

**Arguments**

fitness            A fitness score of all candidate models.

**Value**

Parents for the child model.

---

plot_best_models	<i>plot_best_models</i>
------------------	-------------------------

---

**Description**

Plots the best fitted model for each of the specs in autohrf.

**Usage**

```
plot_best_models(autofit, ncol = NULL, nrow = NULL)
```

**Arguments**

autofit            Output of the autohrf function.  
ncol                Number of columns in the plot.  
nrow                Number of rows in the plot.

**Value**

Plots the grid containing a visualization of the best models for each of the provided constraints.

**Examples**

```
# prepare model specs
model3 <- data.frame(
  event      = c("encoding", "delay", "response"),
  start_time = c(0,          2.65,  12.5),
  end_time   = c(3,          12.5,  16)
)

model4 <- data.frame(
  event      = c("fixation", "target", "delay", "response"),
  start_time = c(0,          2.5,    2.65,  12.5),
  end_time   = c(2.5,        3,      12.5,  15.5)
)

model_constraints <- list(model3, model4)

# run autohrf
df <- flanker
autofit <- autohrf(df, model_constraints, tr = 2.5,
                  population = 2, iter = 2, cores = 1)

# plot best models
plot_best_models(autofit)
```

---

plot\_events

*plot\_events*

---

**Description**

A helper function for plotting events of a fitted model.

**Usage**

```
plot_events(af, i = NULL)
```

**Arguments**

**af**                    The output from the autohrf function.  
**i**                     Model index.

**Value**

Returns a plot of the events.



---

plot_fitness	<i>plot_fitness</i>
--------------	---------------------

---

## Description

Plots how fitness changed through iterations of autohrf. Use this to investigate whether your solution converged.

## Usage

```
plot_fitness(autofit)
```

## Arguments

autofit            Output of the autohrf function.

## Value

A ggplot visualization of fitness through time.

## Examples

```
# prepare model specs
model3 <- data.frame(
  event      = c("encoding", "delay", "response"),
  start_time = c(0,          2.65,   12.5),
  end_time   = c(3,          12.5,   16)
)

model4 <- data.frame(
  event      = c("fixation", "target", "delay", "response"),
  start_time = c(0,          2.5,    2.65,  12.5),
  end_time   = c(2.5,        3,      12.5,  15.5)
)

model_constraints <- list(model3, model4)

# run autohrf
df <- flanker
autofit <- autohrf(df, model_constraints, tr = 2.5,
                  population = 2, iter = 2, cores = 1)

# plot fitness
plot_fitness(autofit)
```



---

run_model	<i>run_model</i>
-----------	------------------

---

**Description**

A helper function for evaluating a model.

**Usage**

```
run_model(d, ce, model, roi_weights = NULL)
```

**Arguments**

d	A dataframe with the signal data: roi, t and y. ROI is the name of the region, t is the timestamp and y the value of the signal.
ce	Result of the convolve_events function.
model	A data frame containing information about the model to use and its events (event, start_time and duration).
roi_weights	A data frame with ROI weights: roi, weight. ROI is the name of the region, weight a number that defines the importance of that roi, the default weight for a ROI is 1. If set to 2 for a particular ROI that ROI will be twice as important.

**Value**

Returns the model's evaluation.

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