

# Package ‘ravetools’

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**Type** Package

**Title** Signal Processing Toolbox for Analyzing 'Electrophysiology' Data

**Version** 0.0.5

**Language** en-US

**Description** Implemented fast and memory-efficient 'Notch'-filter, 'Welch-periodogram', and discrete wavelet transform algorithm for hours of high-resolution signals; providing fundamental toolbox for 'iEEG' preprocess pipelines. Documentation and examples about 'RAVE' project are provided at <https://openwetware.org/wiki/RAVE>, and the paper by John F. Magnotti, Zhengjia Wang, Michael S. Beauchamp (2020) [doi:10.1016/j.neuroimage.2020.117341](https://doi.org/10.1016/j.neuroimage.2020.117341); see 'citation("ravetools")' for details.

**BugReports** <https://github.com/dipterix/ravetools/issues>

**URL** <https://dipterix.org/ravetools/>

**License** GPL-3

**Encoding** UTF-8

**RoxygenNote** 7.2.1

**Depends** R (>= 4.0.0)

**SystemRequirements** fftw3 (libfftw3-dev (deb), or fftw-devel (rpm))

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baseline_array	<i>Calculate Contrasts of Arrays in Different Methods</i>
----------------	---

---

### Description

Provides five methods to baseline an array and calculate contrast.

### Usage

```
baseline_array(x, along_dim, unit_dims = seq_along(dim(x))[-along_dim], ...)

## S3 method for class 'array'
baseline_array(
  x,
  along_dim,
  unit_dims = seq_along(dim(x))[-along_dim],
  method = c("percentage", "sqrt_percentage", "decibel", "zscore", "sqrt_zscore"),
  baseline_indexpoints = NULL,
  baseline_subarray = NULL,
  ...
)
```

**Arguments**

<code>x</code>	array (tensor) to calculate contrast
<code>along_dim</code>	integer range from 1 to the maximum dimension of <code>x</code> . baseline along this dimension, this is usually the time dimension.
<code>unit_dims</code>	integer vector, baseline unit: see Details.
<code>...</code>	passed to other methods
<code>method</code>	character, baseline method options are: "percentage", "sqrt_percentage", "decibel", "zscore", and "sqrt_zscore"
<code>baseline_indexpoints</code>	integer vector, which index points are counted into baseline window? Each index ranges from 1 to <code>dim(x)[[along_dim]]</code> . See Details.
<code>baseline_subarray</code>	sub-arrays that should be used to calculate baseline; default is NULL (automatically determined by <code>baseline_indexpoints</code> ).

**Details**

Consider a scenario where we want to baseline a bunch of signals recorded from different locations. For each location, we record  $n$  sessions. For each session, the signal is further decomposed into frequency-time domain. In this case, we have the input  $x$  in the following form:

$$session \times frequency \times time \times location$$

Now we want to calibrate signals for each session, frequency and location using the first 100 time points as baseline points, then the code will be

$$baseline\_array(x, along\_dim = 3, baseline\_window = 1 : 100, unit\_dims = c(1, 2, 4))$$

`along_dim=3` is dimension of time, in this case, it's the third dimension of  $x$ . `baseline_indexpoints=1:100`, meaning the first 100 time points are used to calculate baseline. `unit_dims` defines the unit signal. Its value `c(1, 2, 4)` means the unit signal is per session (first dimension), per frequency (second) and per location (fourth).

In some other cases, we might want to calculate baseline across frequencies then the unit signal is *frequency\*time*, i.e. signals that share the same session and location also share the same baseline. In this case, we assign `unit_dims=c(1, 4)`.

There are five baseline methods. They fit for different types of data. Denote  $z$  is a unit signal,  $z_0$  is its baseline slice. Then these baseline methods are:

"percentage"

$$\frac{z - \bar{z}_0}{\bar{z}_0} \times 100\%$$

"sqrt\_percentage"

$$\frac{\sqrt{z} - \sqrt{\bar{z}_0}}{\sqrt{\bar{z}_0}} \times 100\%$$

"decibel"

$$10 \times (\log_{10}(z) - \log_{10}(\bar{z}_0))$$

"zscore"

$$\frac{z - \bar{z}_0}{sd(\bar{z}_0)}$$

"sqrt\_zscore"

$$\frac{\sqrt{z} - \sqrt{\bar{z}_0}}{sd(\sqrt{\bar{z}_0})}$$

### Value

Contrast array with the same dimension as x.

### Examples

```
# Set ncores = 2 to comply to CRAN policy. Please don't run this line
ravetools_threads(n_threads = 2L)

library(ravetools)
set.seed(1)

# Generate sample data
dims = c(10,20,30,2)
x = array(rnorm(prod(dims))^2, dims)

# Set baseline window to be arbitrary 10 timepoints
baseline_window = sample(30, 10)

# ----- baseline percentage change -----

# Using base functions
re1 <- aperm(apply(x, c(1,2,4), function(y){
  m <- mean(y[baseline_window])
  (y/m - 1) * 100
}), c(2,3,1,4))

# Using ravetools
re2 <- baseline_array(x, 3, c(1,2,4),
  baseline_indexpoints = baseline_window,
  method = 'percentage')

# Check different, should be very tiny (double precisions)
range(re2 - re1)

# Check speed for large dataset
if(interactive()){
```

```

ravetools_threads(n_threads = -1)

dims <- c(200,20,300,2)
x <- array(rnorm(prod(dims))^2, dims)
# Set baseline window to be arbitrary 10 timepoints
baseline_window <- seq_len(100)
f1 <- function(){
  aperm(apply(x, c(1,2,4), function(y){
    m <- mean(y[baseline_window])
    (y/m - 1) * 100
  }), c(2,3,1,4))
}
f2 <- function(){
  # equivalent as bl = x[, ,baseline_window, _]
  #
  baseline_array(x, along_dim = 3,
                 baseline_indexpoints = baseline_window,
                 unit_dims = c(1,2,4), method = 'percentage')
}
range(f1() - f2())
microbenchmark::microbenchmark(f1(), f2(), times = 10L)

}

```

---

collapse

*Collapse array*


---

## Description

Collapse array

## Usage

```

collapse(x, keep, ...)

## S3 method for class 'array'
collapse(
  x,
  keep,
  average = TRUE,
  transform = c("asis", "10log10", "square", "sqrt"),
  ...
)

```

**Arguments**

<code>x</code>	A numeric multi-mode tensor (array), without NA
<code>keep</code>	Which dimension to keep
<code>...</code>	passed to other methods
<code>average</code>	collapse to sum or mean
<code>transform</code>	transform on the data before applying collapsing; choices are 'asis' (no change), '10log10' (used to calculate decibel), 'square' (sum-squared), 'sqrt' (square-root and collapse)

**Value**

a collapsed array with values to be mean or summation along collapsing dimensions

**Examples**

```
# Set ncores = 2 to comply to CRAN policy. Please don't run this line
ravetools_threads(n_threads = 2L)

# Example 1
x = matrix(1:16, 4)

# Keep the first dimension and calculate sums along the rest
collapse(x, keep = 1)
rowMeans(x) # Should yield the same result

# Example 2
x = array(1:120, dim = c(2,3,4,5))
result = collapse(x, keep = c(3,2))
compare = apply(x, c(3,2), mean)
sum(abs(result - compare)) # The same, yield 0 or very small number (1e-10)

if(interactive()){
  ravetools_threads(n_threads = -1)
}

# Example 3 (performance)

# Small data, no big difference
x = array(rnorm(240), dim = c(4,5,6,2))
microbenchmark::microbenchmark(
  result = collapse(x, keep = c(3,2)),
  compare = apply(x, c(3,2), mean),
  times = 1L, check = function(v){
    max(abs(range(do.call('-', v)))) < 1e-10
  }
)

# large data big difference
x = array(rnorm(prod(300,200,105)), c(300,200,105,1))
```

```

microbenchmark::microbenchmark(
  result = collapse(x, keep = c(3,2)),
  compare = apply(x, c(3,2), mean),
  times = 1L, check = function(v){
    max(abs(range(do.call('-', v)))) < 1e-10
  })
}

```

---

decimate

*Decimate with 'FIR' or 'IIR' filter*


---

### Description

Decimate with 'FIR' or 'IIR' filter

### Usage

```
decimate(x, q, n = if (ftype == "iir") 8 else 30, ftype = "fir")
```

### Arguments

x	signal to be decimated
q	integer factor to down-sample by
n	filter order used in the down-sampling; default is 30 if ftype='fir', or 8 if ftype='iir'
ftype	filter type, choices are 'fir' (default) and 'iir'

### Details

This function is migrated from signal package, but with bugs fixed on 'FIR' filters. The result agrees with 'Matlab' decimate function with 'FIR' filters. Under 'IIR' filters, the function is identical with signal::decimate, and is slightly different with 'Matlab' version.

### Value

Decimated signal

### Examples

```

x <- 1:100
y <- decimate(x, 2, ftype = "fir")
y

# compare with signal package
z <- signal::decimate(x, 2, ftype = "fir")

```

```
# Compare decimated results
plot(x, type = 'l')
points(seq(1,100, 2), y, col = "green")
points(seq(1,100, 2), z, col = "red")
```

---

detrend

*Remove the trend for one or more signals*

---

## Description

'Detrending' is often used before the signal power calculation.

## Usage

```
detrend(x, trend = c("constant", "linear"), break_points = NULL)
```

## Arguments

x	numerical or complex, a vector or a matrix
trend	the trend of the signal; choices are 'constant' and 'linear'
break_points	integer vector, or NULL; only used when trend is 'linear' to remove piecewise linear trend; will throw warnings if trend is 'constant'

## Value

The signals with trend removed in matrix form; the number of columns is the number of signals, and number of rows is length of the signals

## Examples

```
x <- rnorm(100, mean = 1) + c(
  seq(0, 5, length.out = 50),
  seq(5, 3, length.out = 50))
plot(x)

plot(detrend(x, 'constant'))
plot(detrend(x, 'linear'))
plot(detrend(x, 'linear', 50))
```



---

diagnose_channel	<i>Show channel signals with diagnostic plots</i>
------------------	---

---

### Description

The diagnostic plots include 'Welch Periodogram' ([pwelch](#)) and histogram ([hist](#))

### Usage

```
diagnose_channel(
  s1,
  s2 = NULL,
  sc = NULL,
  srate,
  name = "",
  try_compress = TRUE,
  max_freq = 300,
  window = ceiling(srate * 2),
  noverlap = window/2,
  std = 3,
  cex = 1.5,
  lwd = 0.5,
  plim = NULL,
  nclass = 100,
  main = "Channel Inspection",
  col = c("black", "red"),
  which = NULL,
  start_time = 0,
  boundary = NULL,
  mar = c(5.2, 5.1, 4.1, 2.1),
  mai = c(0.6, 0.54, 0.4, 0.1),
  ...
)
```

### Arguments

s1	the main signal to draw
s2	the comparing signal to draw; usually s1 after some filters; must be in the same sampling rate with s1; can be NULL
sc	decimated s1 to show if srate is too high; will be automatically generated if NULL
srate	sampling rate
name	name of s1, or a vector of two names of s1 and s2 if s2 is provided
try_compress	whether try to compress (decimate) s1 if srate is too high for performance concerns

max\_freq the maximum frequency to display in 'Welch Periodograms'  
 window, noverlap see [pwelch](#)  
 std the standard deviation of the channel signals used to determine boundary; default is plus-minus 3 standard deviation  
 cex, lwd, mar, mai, ... graphical parameters; see [par](#)  
 plim the y-axis limit to draw in 'Welch Periodograms'  
 nclass number of classes to show in histogram ([hist](#))  
 main the title of the signal plot  
 col colors of s1 and s2  
 which NULL or integer from 1 to 4; if NULL, all plots will be displayed; otherwise only the subplot will be displayed  
 start\_time the starting time of channel (will only be used to draw signals)  
 boundary a red boundary to show in channel plot; default is to be automatically determined by std

**Value**

A list of boundary and y-axis limit used to draw the channel

**Examples**

```

library(ravetools)

# Generate 20 second data at 2000 Hz
time <- seq(0, 20, by = 1 / 2000)
signal <- sin( 120 * pi * time) +
  sin(time * 20*pi) +
  exp(-time^2) *
  cos(time * 10*pi) +
  rnorm(length(time))

signal2 <- notch_filter(signal, 2000)

diagnose_channel(signal, signal2, srate = 2000,
  name = c("Raw", "Filtered"), cex = 1)

```

---

fast\_cov

*Calculate massive covariance matrix in parallel*


---

**Description**

Speed up covariance calculation for large matrices. The default behavior is the same as [cov](#) ('pearson', no NA handling).

**Usage**

```
fast_cov(x, y = NULL, col_x = NULL, col_y = NULL, df = NA)
```

**Arguments**

x	a numeric vector, matrix or data frame; a matrix is highly recommended to maximize the performance
y	NULL (default) or a vector, matrix or data frame with compatible dimensions to x; the default is equivalent to y = x
col_x	integers indicating the subset indices (columns) of x to calculate the covariance, or NULL to include all the columns; default is NULL
col_y	integers indicating the subset indices (columns) of y to calculate the covariance, or NULL to include all the columns; default is NULL
df	a scalar indicating the degrees of freedom; default is nrow(x)-1

**Value**

A covariance matrix of x and y. Note that there is no NA handling. Any missing values will lead to NA in the resulting covariance matrices.

**Examples**

```
# Set ncores = 2 to comply to CRAN policy. Please don't run this line
ravetools_threads(n_threads = 2L)

x <- matrix(rnorm(400), nrow = 100)

# Call `cov(x)` to compare
fast_cov(x)

# Calculate covariance of subsets
fast_cov(x, col_x = 1, col_y = 1:2)

if(interactive()){

# Speed comparison, better to use multiple cores (4, 8, or more)
# to show the differences.

ravetools_threads(n_threads = -1)
x <- matrix(rnorm(100000), nrow = 1000)
microbenchmark::microbenchmark(
  fast_cov = {
    fast_cov(x, col_x = 1:50, col_y = 51:100)
  },
  cov = {
    cov(x[,1:50], x[,51:100])
  },
  unit = 'ms', times = 10
)
```

```
}

```

---

matlab_palette	<i>'Matlab' heat-map plot palette</i>
----------------	---------------------------------------

---

**Description**

'Matlab' heat-map plot palette

**Usage**

```
matlab_palette()
```

**Value**

vector of 64 colors

---

multitaper	<i>Compute 'multitaper' spectral densities of time-series data</i>
------------	--

---

**Description**

Compute 'multitaper' spectral densities of time-series data

**Usage**

```
multitaper_config(
  data_length,
  fs,
  frequency_range = NULL,
  time_bandwidth = 5,
  num_tapers = NULL,
  window_params = c(5, 1),
  nfft = NA,
  detrend_opt = "linear"
)
```

```
multitaper(
  data,
  fs,
  frequency_range = NULL,
  time_bandwidth = 5,
  num_tapers = NULL,
```

```

    window_params = c(5, 1),
    nfft = NA,
    detrend_opt = "linear"
  )

```

### Arguments

<code>data_length</code>	length of data
<code>fs</code>	sampling frequency in 'Hz'
<code>frequency_range</code>	frequency range to look at; length of two
<code>time_bandwidth</code>	a number indicating time-half bandwidth product; i.e. the window duration times the half bandwidth of main lobe; default is 5
<code>num_tapers</code>	number of 'DPSS' tapers to use; default is NULL and will be automatically computed from <code>floor(2*time_bandwidth - 1)</code>
<code>window_params</code>	vector of two numbers; the first number is the window size in seconds; the second number is the step size; default is <code>c(5, 1)</code>
<code>nfft</code>	'NFFT' size, positive; see 'Details'
<code>detrend_opt</code>	how you want to remove the trend from data window; options are 'linear' (default), 'constant', and 'off'
<code>data</code>	numerical vector, signal traces

### Details

The original source code comes from 'Prerau' Lab (see 'Github' repository 'multitaper\_toolbox' under user 'preraulab'). The results tend to agree with their 'Python' implementation with precision on the order of at  $1E-7$  with standard deviation at most  $1E-5$ . The original copy was licensed under a Creative Commons Attribution 'NC'-'SA' 4.0 International License (<https://creativecommons.org/licenses/by-nc-sa/4.0/>).

This package ('ravetools') redistributes the multitaper function under minor modifications on `nfft`. In the original copy there is no parameter to control the exact numbers of `nfft`, and the `nfft` is always the power of 2. While choosing `nfft` to be the power of 2 is always recommended, the modified code allows other choices.

### Value

`multitaper_config` returns a list of configuration parameters for the filters; `multitaper` also returns the time, frequency and corresponding spectral power.

### Examples

```

time <- seq(0, 3, by = 0.001)
x <- sin(time * 20*pi) + exp(-time^2) * cos(time * 10*pi)

res <- multitaper(
  x, 1000, frequency_range = c(0,15),

```

```

time_bandwidth=1.5,
window_params=c(2,0.01)
)

image(
  x = res$time,
  y = res$frequency,
  z = 10 * log10(res$spec),
  xlab = "Time (s)",
  ylab = 'Frequency (Hz)',
  col = matlab_palette()
)

```

---

notch\_filter

*Apply 'Notch' filter*


---

### Description

Apply 'Notch' filter

### Usage

```

notch_filter(
  s,
  sample_rate,
  lb = c(59, 118, 178),
  ub = c(61, 122, 182),
  domain = 1
)

```

### Arguments

s	numerical vector if domain=1 (voltage signals), or complex vector if domain=0
sample_rate	sample rate
lb	filter lower bound of the frequencies to remove
ub	filter upper bound of the frequencies to remove; shares the same length as lb
domain	1 if the input signal is in the time domain, 0 if it is in the frequency domain

### Details

Mainly used to remove electrical line frequencies at 60, 120, and 180 Hz.

### Value

filtered signal in time domain (real numerical vector)

## Examples

```
time <- seq(0, 3, 0.005)
s <- sin(120 * pi * time) + rnorm(length(time))

# Welch periodogram shows a peak at 60Hz
pwelch(s, 200, plot = 1, log = "y")

# notch filter to remove 60Hz
s1 <- notch_filter(s, 200, lb = 59, ub = 61)
pwelch(s1, 200, plot = 2, log = "y", col = "red")
```

---

parallel-options	<i>Set or get thread options</i>
------------------	----------------------------------

---

## Description

Set or get thread options

## Usage

```
detect_threads()

ravetools_threads(n_threads = "auto", stack_size = "auto")
```

## Arguments

n_threads	number of threads to set
stack_size	Stack size (in bytes) to use for worker threads. The default used for "auto" is 2MB on 32-bit systems and 4MB on 64-bit systems.

## Value

detect\_threads returns an integer of default threads that is determined by the number of CPU cores; ravetools\_threads returns nothing.

## Examples

```
if(interactive()){
  detect_threads()
  ravetools_threads(n_threads = 2)
}
```

---

pwelch

*Calculate 'Welch Periodogram'*

---

### Description

pwelch is for single signal trace only; mv\_pwelch is for multiple traces. Currently mv\_pwelch is experimental and should not be called directly.

### Usage

```
pwelch(  
  x,  
  fs,  
  window = 64,  
  noverlap = 8,  
  nfft = 256,  
  col = "black",  
  xlim = NULL,  
  ylim = NULL,  
  main = "Welch periodogram",  
  plot = 0,  
  log = c("xy", "", "x", "y"),  
  ...  
)  
  
## S3 method for class 'pwelch'  
print(x, ...)  
  
## S3 method for class 'pwelch'  
plot(  
  x,  
  log = c("xy", "x", "y", ""),  
  type = "l",  
  add = FALSE,  
  col = 1,  
  cex = 1,  
  cex.main = cex,  
  cex.sub = cex,  
  cex.lab = cex * 0.8,  
  cex.axis = cex * 0.7,  
  las = 1,  
  main = "Welch periodogram",  
  xlab,  
  ylab,  
  xlim = NULL,  
  ylim = NULL,  
  ...  
)
```



```
)
mv_pwelch(x, margin, fs, nfft)
```

### Arguments

<code>x</code>	'pwelch' instance returned by pwelch function
<code>fs</code>	sample rate, average number of time points per second
<code>window</code>	window length in time points, default size is 64
<code>noverlap</code>	overlap between two adjacent windows, measured in time points; default is 8
<code>nfft</code>	number of basis functions to apply
<code>col, xlim, ylim, main, type, cex, cex.main, cex.sub, cex.lab, cex.axis, las, xlab, ylab</code>	parameters passed to <a href="#">plot.default</a>
<code>plot</code>	integer, whether to plot the result or not; choices are 0, no plot; 1 plot on a new canvas; 2 add to existing canvas
<code>log</code>	indicates which axis should be log <sub>10</sub> -transformed, used by the plot function. For 'x' axis, it's log <sub>10</sub> -transform; for 'y' axis, it's 10log <sub>10</sub> -transform (decibel unit). Choices are "xy", "x", "y", and "".
<code>...</code>	will be passed to <code>plot.pwelch</code> or ignored
<code>add</code>	logical, whether the plot should be added to existing canvas
<code>margin</code>	the margin in which pwelch should be applied to

### Value

A list with class 'ravetools-pwelch' that contains the following items:

<code>freq</code>	frequencies used to calculate the 'periodogram'
<code>spec</code>	resulting spectral power for each frequency
<code>window</code>	window function (in numerical vector) used
<code>noverlap</code>	number of overlapping time-points between two adjacent windows
<code>nfft</code>	number of basis functions
<code>fs</code>	sample rate
<code>x_len</code>	input signal length
<code>method</code>	a character string 'Welch'

### Examples

```
x <- rnorm(1000)
pwel <- pwelch(x, 100)
pwel

plot(pwel, log = "xy")
```

---

raw-to-sexp	<i>Convert raw vectors to R vectors</i>
-------------	---

---

**Description**

Convert raw vectors to R vectors

**Usage**

raw\_to\_uint8(x)

raw\_to\_uint16(x)

raw\_to\_uint32(x)

raw\_to\_int8(x)

raw\_to\_int16(x)

raw\_to\_int32(x)

raw\_to\_int64(x)

raw\_to\_float(x)

raw\_to\_string(x)

**Arguments**

x                      raw vector of bytes

**Details**

For numeric conversions, the function names are straightforward. For example, `raw_to_uintN` converts raw vectors to unsigned integers, and `raw_to_intN` converts raw vectors to signed integers. The number 'N' stands for the number of bits used to store the integer. For example `raw_to_uint8` uses 8 bits (1 byte) to store an integer, hence the value range is 0-255.

The input data length must be multiple of the element size represented by the underlying data. For example `uint16` integer uses 16 bits, and one raw number uses 8 bits, hence two raw vectors can form one unsigned integer-16. That is, `raw_to_uint16` requires the length of input to be multiple of two. An easy calculation is: the length of x times 8, must be divided by 'N' (see last paragraph for definition).

The returned data uses the closest available R native data type that can fully represent the data. For example, R does not have single float type, hence `raw_to_float` returns double type, which can represent all possible values in float. For `raw_to_uint32`, the potential value range is  $0 - (2^{32}-1)$ . This exceeds the limit of R integer type  $(-2^{31}) - (2^{31}-1)$ . Therefore, the returned values will be real (double float) data type.

There is no native data type that can store integer-64 data in R, package `bit64` provides `integer64` type, which will be used by `raw_to_int64`. Currently there is no solution to convert raw to unsigned integer-64 type.

`raw_to_string` converts raw to character string. This function respects null character, hence is slightly different than the native `rawToChar`, which translates raw byte-by-byte. If each raw byte represents a valid character, then the above two functions returns the same result. However, when the characters represented by raw bytes are invalid, `raw_to_string` will stop parsing and returns only the valid characters, while `rawToChar` will still try to parse, and most likely to result in errors. Please see Examples for comparisons.

## Value

Numeric vectors, except for `raw_to_string`, which returns a string.

## Examples

```
# 0x00, 0x7f, 0x80, 0xFF
x <- as.raw(c(0, 127, 128, 255))

raw_to_uint8(x)

# The first bit becomes the integer sign
# 128 -> -128, 255 -> -1
raw_to_int8(x)

## Comments based on little endian system

# 0x7f00 (32512), 0xFF80 (65408 unsigned, or -128 signed)
raw_to_uint16(x)
raw_to_int16(x)

# 0xFF807F00 (4286611200 unsigned, -8356096 signed)
raw_to_uint32(x)
raw_to_int32(x)

# ----- String -----

# ASCII case: all valid
x <- charToRaw("This is an ASCII string")

raw_to_string(x)
rawToChar(x)

x <- c(charToRaw("This is the end."),
      as.raw(0),
      charToRaw("*** is invalid"))

# rawToChar will raise error
raw_to_string(x)

# ----- Integer64 -----
```

```

# Runs on little endian system
x <- as.raw(c(0x80, 0x00, 0x7f, 0x80, 0xFF, 0x50, 0x7f, 0x00))

# Calculate bitstring, which concatenates the followings
# 10000000 (0x80), 00000000 (0x00), 01111111 (0x7f), 10000000 (0x80),
# 11111111 (0xFF), 01010000 (0x50), 01111111 (0x7f), 00000000 (0x00)

if(.Platform$endian == "little") {
  bitstring <- paste0(
    "00000000011111110101000011111111",
    "10000000011111110000000010000000"
  )
} else {
  bitstring <- paste0(
    "00000001000000001111111000000001",
    "11111111000010101111111000000000"
  )
}

# This is expected value
bit64::as.integer64(structure(
  bitstring,
  class = "bitstring"
))

# This is actual value
raw_to_int64(x)

```

---

 shift\_array

*Shift array by index*


---

## Description

Re-arrange arrays in parallel

## Usage

```
shift_array(x, along_margin, unit_margin, shift_amount)
```

## Arguments

x	array, must have at least matrix
along_margin	which index is to be shifted
unit_margin	which dimension decides shift_amount
shift_amount	shift amount along along_margin

## Details

A simple use-case for this function is to think of a matrix where each row is a signal and columns stand for time. The objective is to align (time-lock) each signal according to certain events. For each signal, we want to shift the time points by certain amount.

In this case, the shift amount is defined by `shift_amount`, whose length equals to number of signals. `along_margin=2` as we want to shift time points (column, the second dimension) for each signal. `unit_margin=1` because the shift amount is depend on the signal number.

## Value

An array with same dimensions as the input `x`, but with index shifted. The missing elements will be filled with NA.

## Examples

```
# Set ncores = 2 to comply to CRAN policy. Please don't run this line
ravetools_threads(n_threads = 2L)

x <- matrix(1:10, nrow = 2, byrow = TRUE)
z <- shift_array(x, 2, 1, c(1,2))

y <- NA * x
y[1,1:4] = x[1,2:5]
y[2,1:3] = x[2,3:5]

# Check if z and y are the same
z - y

# array case
# x is Trial x Frequency x Time
x <- array(1:27, c(3,3,3))

# Shift time for each trial, amount is 1, -1, 0
shift_amount <- c(1,-1,0)
z <- shift_array(x, 3, 1, shift_amount)

if(interactive()){
  par(mfrow = c(3, 2), mai = c(0.8, 0.6, 0.4, 0.1))
  for( ii in 1:3 ){
    image(t(x[ii, ,]), ylab = 'Frequency', xlab = 'Time',
          main = paste('Trial', ii))
    image(t(z[ii, ,]), ylab = 'Frequency', xlab = 'Time',
          main = paste('Shifted amount:', shift_amount[ii]))
  }
}
```

---

wavelet *'Morlet' wavelet transform (Discrete)*

---

### Description

Transform analog voltage signals with 'Morlet' wavelets: complex wavelet kernels with  $\pi/2$  phase differences.

### Usage

```

wavelet_kernels(freqs, srate, wave_num)

morlet_wavelet(
  data,
  freqs,
  srate,
  wave_num,
  precision = c("float", "double"),
  trend = c("constant", "linear", "none"),
  ...
)

```

### Arguments

freqs	frequency in which data will be projected on
srate	sample rate, number of time points per second
wave_num	desired number of cycles in wavelet kernels to balance the precision in time and amplitude (control the smoothness); positive integers are strongly suggested
data	numerical vector such as analog voltage signals
precision	the precision of computation; choices are 'float' (default) and 'double'.
trend	choices are 'constant': center the signal at zero; 'linear': remove the linear trend; 'none' do nothing
...	further passed to <a href="#">detrend</a> ;

### Value

wavelet\_kernels returns wavelet kernels to be used for wavelet function; morlet\_wavelet returns a file-based array if precision is 'float', or a list of real and imaginary arrays if precision is 'double'

### Examples

```

if(interactive()){
  # generate sine waves

```

```
time <- seq(0, 3, by = 0.01)
x <- sin(time * 20*pi) + exp(-time^2) * cos(time * 10*pi)

plot(time, x, type = 'l')

# freq from 1 - 15 Hz; wavelet using float precision
freq <- seq(1, 15, 0.2)
coef <- morlet_wavelet(x, freq, 100, c(2,3))

# to get coefficients in complex number from 1-10 time points
coef[1:10, ]

# power
power <- Mod(coef[])^2

# Power peaks at 5Hz and 10Hz at early stages
# After 1.0 second, 5Hz component fade away
image(power, x = time, y = freq, ylab = "frequency")

# wavelet using double precision
coef2 <- morlet_wavelet(x, freq, 100, c(2,3), precision = "double")
power2 <- (coef2$real[])^2 + (coef2$imag[])^2

image(power2, x = time, y = freq, ylab = "frequency")

# The maximum relative change of power with different precisions
max(abs(power/power2 - 1))

# display kernels
freq <- seq(1, 15, 1)
kern <- wavelet_kernels(freq, 100, c(2,3))
print(kern)

plot(kern)

}
```

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