# Package 'ShellChron'

October 12, 2022

Title Builds Chronologies from Oxygen Isotope Profiles in Shells

Version 0.4.0

**Description** Takes as input a stable oxygen isotope (d18O) profile measured in growth direction (D) through a shell + uncertainties in both variables (d18O\_err & D\_err). It then models the seasonality

in the d18O record by fitting a combination of a growth and temperature sine wave to year-length chunks of

the data (see Judd et al., (2018) <doi:10.1016/j.palaeo.2017.09.034>). This modeling is carried out along a sliding window through the data and yields estimates of

the day of the year (Julian Day) and local growth rate for each data point. Uncertainties in both modeling

routine and the data itself are propagated and pooled to obtain a confidence envelope around the age of

each data point in the shell. The end result is a shell chronology consisting of estimated ages of shell

formation relative to the annual cycle with their uncertainties. All formulae in the package serve this

purpose, but the user can customize the model (e.g. number of days in a year and the mineralogy of the

shell carbonate) through input parameters.

```
Imports rtop (>= 0.5.14), zoo (>= 1.8.7), ggplot2 (>= 3.2.1), ggpubr (>= 0.4.0), tidyr (>= 1.1.1), scales (>= 1.1.0), dplyr, magrittr
```

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URL https://github.com/nielsjdewinter/ShellChron

BugReports https://github.com/nielsjdewinter/ShellChron/issues

Encoding UTF-8
LazyData true
RoxygenNote 7.1.1
Depends R (>= 3.5.0)

Suggests knitr, rmarkdown

NeedsCompilation no

2 age\_corr

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Repository CRAN

**Date/Publication** 2021-07-05 12:40:02 UTC

# **R** topics documented:

Index		24
	wrap_function	22
	Virtual_shell	21
	temperature_curve	20
	sinreg	20
	sd_wt	19
	run_model	17
	peakid	16
	mc_err_proj	15
	mc_err_orth	14
	mc_err_form	13
	growth_rate_curve	12
	growth_model	
	export_results	7
	data_import	6
	d18O_model	5
	cumulative_day	4
	cumdy	
	age_con	

age\_corr

Function that corrects chronologies for sudden jumps in time

# Description

Some occurrences in the model results can lead the CumDY function to detect extra year transitions, resulting in sudden jumps in the shell chronology or a start of the chronology at an age beyond 1 year. This function removes these sharp transitions and late onset by adding or subtracting whole years to the age result.

# Usage

```
age_corr(resultarray, T_per = 365, plot = TRUE, agecorrection = TRUE)
```

cumdy 3

# **Arguments**

resultarray Array containing the full results of the optimized growth model

T\_per The period length of one year (in days)

plot Should the results be plotted? (/codeTRUE/FALSE)

agecorrection Correct for jumps in age (/codeTRUE) or only for starting time (/codeFALSE)

#### Value

An updated and corrected version of resultarray

#### References

```
package dependencies: ggplot2 3.2.1
```

# **Examples**

cumdy

Function to detect year transitions and calculate cumulative age of model results

# **Description**

Takes the result of iterative growth modeling and transforms data from Julian Day (0 - 365) to cumulative day of the shell age by detecting where transitions from one year to the next occur and adding full years (365 days) to simulations in later years.

# Usage

```
cumdy(resultarray, threshold = 5, plotyearmarkers = TRUE)
```

# Arguments

resultarray Array containing the full results of the optimized growth model

threshold Artificial threshold value used to recognize peaks in occurrences of year transi-

tions (default = 5)

plotyearmarkers

Should the location of identified year transitions be plotted? TRUE/FALSE

4 cumulative\_day

#### Value

A new version of the resultarray with Julian Day model estimates replaced by estimates of cumulative age of the record in days.

#### References

```
package dependencies: zoo 1.8.7
```

# **Examples**

```
testarray <- array(NA, dim = c(20, 16, 9)) # Create empty array
# with correct third dimension
windowfill <- seq(50, 500, 50) # Create dummy simulation data
# (ages) to copy through the array
for(i in 6:length(testarray[1, , 1])){
    testarray[, i, 3] <- c(windowfill, rep(NA, length(testarray[, 1, 3]) -
        length(windowfill)))
    windowfill <- c(NA, (windowfill + 51) %% 365)
}
testarray[, 1, 3] <- seq(1, length(testarray[, 1, 3]), 1) # Add
# dummy /code{D} column.
testarray2 <- cumdy(testarray, 3, FALSE) # Apply function on array</pre>
```

cumulative\_day

Function to detect year transitions and calculate cumulative age of model results

# Description

Takes the result of iterative growth modeling and transforms data from Julian Day (0 - 365) to cumulative day of the shell age by detecting where transitions from one year to the next occur and adding full years (365 days) to simulations in later years.

# Usage

```
cumulative_day(
  resultarray,
  plotyearmarkers = TRUE,
  export_peakid = TRUE,
  path = tempdir()
)
```

# **Arguments**

```
resultarray Array containing the full results of the optimized growth model
plotyearmarkers
Should the location of identified year transitions be plotted? TRUE/FALSE
export_peakid Should the result of peak identification be plotted? TRUE/FALSE
path Export path (defaults to tempdir())
```

d18O\_model 5

#### Value

A new version of the Julian Day tab of the resultarray with Julian Day model estimates replaced by estimates of cumulative age of the record in days.

#### References

package dependencies: zoo 1.8.7; scales 1.1.0; graphics function dependencies: peakid

# **Examples**

```
testarray \leftarrow array(NA, dim = c(40, 36, 9)) # Create empty array
# with correct third dimension
windowfill <- seq(50, 500, 50) %% 365 # Create dummy simulation data
# (ages) to copy through the array
for(i in 6:length(testarray[1, , 1])){
    testarray[, i, 3] <- c(windowfill, rep(NA, length(testarray[, 1, 3]) -</pre>
        length(windowfill)))
    windowfill <- c(NA, (windowfill + 51) %% 365)
}
# Add dummy /code{D} column.
testarray[, 1, 3] <- seq(1, length(testarray[, 1, 3]), 1)
# Add dummy YEARMARKER column
testarray[, 3, 3] < c(0, rep(c(0, 0, 0, 0, 0, 0, 1), 5), 0, 0, 0, 0)
# Add dummy d180c column
testarray[, 2, 3] < sin((2 * pi * (testarray[, 1, 3] - 8 + 7 / 4)) / 7)
testarray2 <- suppressWarnings(cumulative_day(testarray, FALSE, FALSE, tempdir()))</pre>
# Apply function on array
```

d180\_model

Function to convert SST data to d180

# **Description**

Takes a matrix of SST data (in degrees C) against time (in days), information about the d18O value (in permille VSMOW) of the water and how it changes through the year and the transfer function used for of the record (e.g. Kim and O'Neil, 1997 or Grossman and Ku, 1986). Converts the SST data to d18O data using the supplied empirical transfer function.

# Usage

```
d180_model(SST, d180w = 0, transfer_function = "KimONeil97")
```

# **Arguments**

SST

Matrix with a time column (values in days) and an SST column (values in degrees C)

6 data\_import

d180w

Either a single value (constant d18Ow) or a vector of length equal to the period in SST data (365 days by default) containing information about seasonality in d18Ow. Defaults to constant d18Ow of 0 permille VSMOW (the modern mean ocean value)

transfer\_function

String containing the name of the transfer function (for example: "KimONei197" or "GrossmanKu86"). Defaults to Kim and O'Neil (1997).

#### Value

A vector containing d18O values for each SST value in "SST"

#### References

Grossman, E.L., Ku, T., Oxygen and carbon isotope fractionation in biogenic aragonite: temperature effects, *Chemical Geology* **1986**, *59.1*, 59-74. doi: 10.1016/01689622(86)900576 Kim, S., O'Niel, J.R., Equilibrium and nonequilibrium oxygen isotope effects in synthetic carbonates, *Geochimica et Cosmochimica Acta* **1997**, *61.16*, 3461-3475. doi: 10.1016/S00167037(97)001695 Dettman, D.L., Reische, A.K., Lohmann, K.C., Controls on the stable isotope composition of seasonal growth bands in aragonitic fresh-water bivalves (Unionidae), *Geochimica et Cosmochimica Acta* **1999**, *63.7-8*, 1049-1057. doi: 10.1016/S00167037(99)000204 Brand, W.A., Coplen, T.B., Vogl, J., Rosner, M., Prohaska, T., Assessment of international reference materials for isotope-ratio analysis (IUPAC Technical Report), *Pure and Applied Chemistry* **2014**, *86.3*, 425-467. doi: 10.1515/pac-20131023

# **Examples**

```
# Create dummy SST data
t <- seq(1, 40, 1)
T <- sin((2 * pi * (seq(1, 40, 1) - 8 + 10 / 4)) / 10)
SST <- cbind(t, T)
# Run d180 model function
d180 <- d180_model(SST, 0, "KimONeil97")</pre>
```

data\_import

Function to import d180 data and process yearmarkers and calculation windows

# Description

Takes the name of a file that is formatted according to the standard format and converts it to an object to be used later in the model. In doing so, the function also reads the user-provided yearmarkers in the file and uses them as a basis for the length of windows used throughout the model. This ensures that windows are not too short and by default contain at least one year of growth for modeling.

# Usage

```
data_import(file_name)
```

export\_results 7

# **Arguments**

file\_name

Name of the file that contains sampling distance and d18O data. Note that sampling distance should be given in micrometers, because the SCEUA model underperforms when the growth rate figures are very small (<0.1 mm/day).

# Value

A list containing an object with the original data and details on the position and length of modeling windows

# **Examples**

```
importlist <- data_import(file_name = system.file("extdata",
    "Virtual_shell.csv", package = "ShellChron")) # Run function on attached
    # dummy data

# Bad data file lacking YEARMARKER column

## Not run: importlist <- data_import(file_name = system.file("extdata",
    "Bad_data.csv", package = "ShellChron"))
## End(Not run)</pre>
```

export\_results

Function to merge and export the results of the ShellChron model

# **Description**

Takes the input data and model results and reformats them to tables of key parameters such as growth rate and shell age for each datapoint for easy plotting. This final function also combines uncertainties in the model result arising from uncertainties in input data (provided by the user) and uncertainties of the model (from overlapping modeling windows). Includes some optional plotting options.

# Usage

```
export_results(
  path = getwd(),
  dat,
  resultarray,
  parmat,
  MC = 1000,
  dynwindow,
  plot = FALSE,
  plot_export = TRUE,
  export_raw = FALSE
)
```

8 export\_results

# Arguments

path Path where result files are exported dat Matrix containing the input data

resultarray Array containing the full results of the optimized growth model

parmat Matrix listing all optimized growth rate and SST parameters used to model d18O

in each data window

MC Number of Monte Carlo simulations to apply for error propagation. Default =

1000

dynwindow Information on the position and length of modeling windows

plot Should an overview of the results of modeling be plotted? TRUE/FALSE

plot\_export Should the overview plot be exported as a PDF file? TRUE/FALSE

export\_raw Export tables containing all raw model results before being merged into tidy

tables? TRUE/FALSE

#### Value

CSV tables of model results in the current working directory + optional plots in PDF format

#### References

package dependencies: tidyverse 1.3.0; ggpubr 0.4.0; magrittr function dependencies: sd\_wt

```
# Create dummy input data column by column
dat <- as.data.frame(seq(1000, 40000, 1000))</pre>
colnames(dat) <- "D"</pre>
dat$d180c <- sin((2 * pi * (seq(1, 40, 1) - 8 + 7 / 4)) / 7)
dat = c(0, rep(c(0, 0, 0, 0, 0, 0, 1), 5), 0, 0, 0, 0)
dat$D_err <- rep(100, 40)
dat$d180c_err <- rep(0.1, 40)
testarray \leftarrow array(NA, dim = c(40, 36, 9)) # Create empty array
# with correct third dimension
windowfill <- seq(50, 500, 50) %% 365 # Create dummy simulation data
# (ages) to copy through the array
for(i in 6:length(testarray[1, , 1])){
    testarray[, i, 3] <- c(windowfill, rep(NA, length(testarray[, 1, 3]) -</pre>
        length(windowfill)))
    windowfill <- c(NA, (windowfill + 51) %% 365)</pre>
}
# Add dummy /code{D} column.
testarray[, 1, 3] <- seq(1, length(testarray[, 1, 3]), 1)
# Add dummy YEARMARKER column
testarray[, 3, 3] \leftarrow c(0, rep(c(0, 0, 0, 0, 0, 0, 1), 5), 0, 0, 0, 0)
# Add dummy d180c column
testarray[, 2, 3] < sin((2 * pi * (testarray[, 1, 3] - 8 + 7 / 4)) / 7)
# Create dummy seasonality data
```

export\_results 9

```
seas <- as.data.frame(seq(1, 365, 1))</pre>
colnames(seas) <- "t"</pre>
seasSST \leftarrow 15 + 10 * sin((2 * pi * (seq(1, 365, 1) - 182.5 +
    365 / 4)) / 365)
seas$GR <- 10 + 10 * sin((2 * pi * (seq(1, 365, 1) - 100 + 365 / 4)) / 365)
seas$d180 <- (exp((18.03 * 1000 / (seas$SST + 273.15) - 32.42) / 1000) - 1) *
    1000 + (0.97002 * 0 - 29.98)
# Apply dummy seasonality data to generate other tabs of testarray
testarray[, , 1] <- seas$d180[match(testarray[, , 3], seas$t)] # d180 values</pre>
tab <- testarray[, , 1]</pre>
tab[which(!is.na(tab))] <- 0.1</pre>
testarray[, , 2] <- tab # dummy d180 residuals</pre>
testarray[, , 4] <- seas$GR[match(testarray[, , 3], seas$t)] # growth rates</pre>
testarray[, , 5] <- seas$SST[match(testarray[, , 3], seas$t)] # temperature</pre>
tab[which(!is.na(tab))] <- 0.1
testarray[, , 6] <- tab # dummy d180 SD
tab[which(!is.na(tab))] <- 20</pre>
testarray[, , 7] <- tab # dummy time SD</pre>
tab[which(!is.na(tab))] <- 3</pre>
testarray[, , 8] <- tab # dummy GR SD
tab[which(!is.na(tab))] <- 1</pre>
testarray[, , 9] <- tab # dummy temperature SD</pre>
darray \leftarrow array(rep(as.matrix(dat), 9), dim = c(40, 5, 9))
testarray[, 1:5, ] <- darray</pre>
# Create dummy dynwindow data
dynwindow <- as.data.frame(seq(1, 31, 1))</pre>
colnames(dynwindow) <- "x"</pre>
dynwindow$y <- rep(10, 31)
dimnames(testarray) <- list(</pre>
    paste("sample", 1:length(testarray[, 1, 3])),
    c(colnames(dat), paste("window", 1:length(dynwindow$x))),
    c("Modeled_d180",
        "d180_residuals",
         "Time_of_year",
         "Instantaneous_growth_rate",
         "Modeled temperature",
         "Modeled_d180_SD",
         "Time_of_Year_SD",
         "Instantaneous_growth_rate_SD",
         "Modeled_temperature_SD")
)
# Set parameters
G_amp <- 20
G_per <- 365
G_pha <- 100
G_av <- 15
G_skw <- 70
T_amp <- 20
T_per <- 365
T_pha <- 150
```

10 growth\_model

```
pars <- c(T_amp, T_pha, T_av, G_amp, G_pha, G_av, G_skw)
parsSD <- c(3, 10, 3, 5, 10, 3, 5) # Artificial variability in parameters
parmat <- matrix(rnorm(length(pars) * length(dynwindow$x)), nrow =</pre>
    length(pars)) * parsSD + matrix(rep(pars, length(dynwindow$x)),
    nrow = length(pars))
rownames(parmat) <- c("T_amp", "T_pha", "T_av", "G_amp", "G_pha", "G_av",</pre>
    "G_skw")
# Run export function
test <- export_results(path = tempdir(),</pre>
    dat,
    testarray,
    parmat,
    MC = 1000,
    dynwindow,
    plot = FALSE,
    plot_export = FALSE,
    export_raw = FALSE)
```

growth\_model

Function that models a d180 curve through SST and GR sinusoids

# **Description**

The core function of the ShellChron growth model. Uses growth rate and SST (Sea Surface Temperature) sinusoids to model d18O data to be matched with the input. In the ShellChron modeling routine, this function is optimized using the SCEUA algorithm and applied on sliding windows through the dataset to estimate the age of each datapoint

# Usage

```
growth_model(
  pars,
  T_per = 365,
  G_per = 365,
  years = 1,
  t_{int} = 1,
  transfer_function = "KimONeil97",
  d180w = "default",
  Dsam,
  Osam,
  t_maxtemp = 182.5,
  plot = FALSE,
 MC = 1000,
  D_err = NULL,
  0_err = NULL,
  return = "SSR"
)
```

growth\_model 11

#### **Arguments**

pars List of parameters for temperature and growth rate sinusoids pars <- c(T\_amp,

T\_pha, T\_av, G\_amp, G\_pha, G\_av, G\_skw)

T\_per Period of SST sinusoid (in days; default = 365)

G\_per Period of growth rate sinusoid (in days; default = 365)

years Number of years to be modeled (default = 1)

t\_int Time interval (in days; default = 1)

transfer\_function

Transfer function used to convert d18Oc to temperature data.

d180w Either a single value (constant d180w) or a vector of length equal to the period

in SST data (365 days by default) containing information about seasonality in d18Ow. Defaults to constant d18Ow of 0 permille VSMOW (the modern mean

ocean value)

Dsam Vector of D values serving as input (keep unit consistent throughout model)

Osam Vector of d180c values serving as input (in permille VPDB)

t\_maxtemp Timing of the warmest day of the year (in julian day; default = 182.5, or May

26th halfway through the year)

plot Should results of modeling be plotted? TRUE/FALSE

MC Number of Monte Carlo simulations to apply for error propagation Default =

1000

D\_err OPTIONAL: Vector containing errors on Dsam
O\_err OPTIONAL: Vector containing errors on Osam

return String indicating whether to return just the Sum of Squared Residuals ("SSR")

or a matrix containing the results of the model and the propagated uncertainties

(if applicable)

# Value

Depending on the value of the "return" parameter either a single value representing the Sum of Squared Residuals ("SSR") as a measure for the closeness of the match between modeled d18O and input values, or a matrix containing the full result of the modeling including propagated uncertainties if applicable.

#### References

package dependencies: ggplot2 3.2.1 function dependencies: temperature\_curve, d18O\_model, growth\_rate\_curve, mc\_err\_orth

doi: 10.1016/j.palaeo.2017.09.034

```
# Set parameters
G_amp <- 20
G_per <- 365</pre>
```

12 growth\_rate\_curve

```
G_pha <- 100
G_av <- 15
G_skw <- 70
T_amp <- 20
T_per <- 365
T_pha <- 150
T_av <- 15
pars <- c(T_amp, T_pha, T_av, G_amp, G_pha, G_av, G_skw)</pre>
d180w <- 0
# Create dummy data
Dsam <- seq(1, 40, 1)
Osam \leftarrow sin((2 * pi * (seq(1, 40, 1) - 8 + 30 / 4)) / 30)
# Test returning residual sum of squares for optimization
SSR <- growth_model(pars, T_per, G_per, Dsam = Dsam, Osam = Osam,
    return = "SSR")
# Test returning full model result
resmat <- growth_model(pars, T_per, G_per, Dsam = Dsam, Osam = Osam,</pre>
    return = "result")
```

growth\_rate\_curve

Function that creates a skewed sinusoidal growth rate (GR) curve from a list of parameters

# **Description**

Takes the specified parameters for amplitude, period, phase, average value and skewness factor as well as the number of years specified and the time interval. It then creates a skewed sinusoid based on the boundary conditions. The skewness factor (G\_skw) determines whether the sinusoid is skewed towards the front (G\_skw < 50) or the back of the annual peak in growth rate (G\_skw > 50). Used as intermediate step during iterative modeling.

# Usage

```
growth_rate_curve(G_par, years = 1, t_int = 1)
```

# **Arguments**

G_par	List of four parameters describing (in order) amplitude (G_amp; in micrometer/day), period (G_per; in days), phase (G_pha in day of the year), average growth rate (G_av; in micrometer/day) and the skewness factor (G_skw between 0 and 100)
years	Length of the preferred sinusoid in number of years (defaults to 1)
t_int	Time interval of sinusoidal record (in days)

# Value

A matrix containing columns for time (in days) and GR (in micrometer/day)

mc\_err\_form 13

# References

doi: 10.1016/j.palaeo.2017.09.034

# **Examples**

```
# Set parameters
G_amp <- 20
G_per <- 365
G_pha <- 100
G_av <- 15
G_skw <- 70
G_par <- c(G_amp, G_per, G_pha, G_av, G_skw)
# Run GR model function
GR <- growth_rate_curve(G_par, 1, 1)</pre>
```

mc\_err\_form

Function that propagates measurement uncertainty through model results

# **Description**

Function to propagate combined errors on x (= Dsam) and y (= Osam) on the modeled X (= D) and Y (= d180c) values by means of projection of uncertainties through the modeled X-Y relationship

# Usage

```
mc_{err_form}(x, x_{err}, y, y_{err}, X, Y, MC = 1000)
```

# Arguments

X	Vector of x values of input data
x_err	Vector of uncertainties on x values
у	Vector of y values of input data
y_err	Vector of uncertainties on y values
Χ	Vector of modeled X values on which the uncertainty is to be projected
Υ	Matrix of modeled x and Y values
MC	Number of Monte Carlo simulations to apply for error propagation Default = 1000

#### **Details**

Note: projection leads to large uncertainties on shallow parts of the XY curve

# Value

A vector listing the standard deviations of propagated errors propagated on all X values.

mc\_err\_orth

# **Examples**

mc\_err\_orth

Function that propagates measurement uncertainty through model results

# **Description**

Function to propagate combined errors on x (= Dsam) and y (= Osam) on the modeled X (= D) and Y (= d180c) values by means of orthogonal projection of uncertainty on x and y onto the model curve

# Usage

```
mc_err_orth(x, x_err, y, y_err, X, Y, MC = 1000)
```

# Arguments

Х	Vector of x values of input data
x_err	Vector of uncertainties on x values
у	Vector of y values of input data
y_err	Vector of uncertainties on y values
Χ	Vector of modeled X values on which the uncertainty is to be projected
Υ	Matrix of modeled X and Y values
MC	Number of Monte Carlo simulations to apply for error propagation Default = 1000

# Value

A vector listing the standard deviations of propagated errors propagated on all X values.

mc\_err\_proj 15

# **Examples**

mc\_err\_proj

Function that propagates measurement uncertainty through model results

# Description

Function to propagate combined errors on x (= Dsam) and y (= Osam) on the modeled X (= D) and Y (= d180c) values by means of direct projection of y-uncertainty on x and then combine the errors on both in the x domain

# Usage

```
mc_err_proj(x, x_err, y, y_err, X, Y, MC = 1000)
```

# **Arguments**

x	Vector of x values of input data
x_err	Vector of uncertainties on x values
у	Vector of y values of input data
y_err	Vector of uncertainties on y values
Χ	Vector of modeled X values on which the uncertainty is to be projected
Υ	Matrix of modeled x and Y values
MC	Number of Monte Carlo simulations to apply for error propagation Default = 1000

# **Details**

Note: projection y\_err on x\_err leads to large X errors on shallow slopes due to numerical calculation of fist derivative.

# Value

A vector listing the standard deviations of propagated errors propagated on all X values.

16 peakid

# **Examples**

peakid

Function that identifies peaks in a dataset

# **Description**

Developed by William A. Huber

# Usage

```
peakid(x, y, w = 1, ...)
```

# **Arguments**

X	Vector of x values of input data
у	Vector of y values of input data
W	Window size for smoothing data
	Additional arguments to be passed into LOESS function

# Value

A vector listing the standard deviations of propagated errors propagated on all X values.

# References

```
package dependencies: zoo 1.8.7
Huber, W.A., Data Smoothing and Peak Detection, Rpubs, Last accessed: December 8th, 2020. 
https://rpubs.com/mengxu/peak_detection
```

#### See Also

https://rpubs.com/mengxu/peak\_detection

run\_model 17

# **Examples**

```
# Create dummy periodic data x \leftarrow seq(1, 100, 1) y \leftarrow sin((2 * pi * (seq(1, 100, 1) - 8 + 20 / 4)) / 20) # Run peakid function result <- peakid(x, y, w = 20)
```

run\_model

Function that optimizes sinusoid parameters to fit d180 data

# Description

The second core function of the ShellChron growth model. Loops through all data windows and uses the growth\_model function to create d18O series that match the input data. This step is iterated and optimized (minimizing the Sum of Squared Residuals) through the SCEUA algorithm (by Duan et al., 1992) which finds the optimal input parameters to the growth rate and Sea Surface Temperature (SST) sinusoids to simulate d18O data.

# Usage

```
run_model(
    dat,
    dynwindow,
    transfer_function = "KimONeil97",
    d180w = 0,
    T_per = 365,
    G_per = 365,
    t_int = 1,
    t_maxtemp = 182.5,
    SCEUApar = c(1, 25, 10000, 5, 0.01, 0.01),
    sinfit = TRUE,
    MC = 1000,
    plot = FALSE
)
```

# **Arguments**

dat Matrix containing the input data

dynwindow Information on the position and length of modeling windows

transfer\_function

Transfer function used to convert d18Oc to temperature data.

d180w Either a single value (constant d180w) or a vector of length equal to the period

in SST data (365 days by default) containing information about seasonality in d18Ow. Defaults to constant d18Ow of 0 permille VSMOW (the modern mean

ocean value)

T\_per Period of SST sinusoid (in days; default = 365)

run\_model

G_per	Period of growth rate sinusoid (in days; default = 365)
t_int	Time interval (in days; default = 1)
t_maxtemp	Timing of the warmest day of the year (in julian day; default = 182.5, or May 26th halfway through the year)
SCEUApar	Parameters for SCEUA optimization (iniflg, ngs, maxn, kstop pcento, peps). For details, refer to Duan et al. (1992) in references
sinfit	Apply sinusoidal fitting to guess initial parameters for SCEUA optimization? $\ensuremath{TRUE/FALSE}$
MC	Number of Monte Carlo simulations to apply for error propagation Default = 1000
plot	Should results of modeling be plotted? TRUE/FALSE

#### Value

A list containing the resultarray which contains the full result of all simulations on each data window and the parmat listing all optimized growth rate and SST parameters used to model d18O in each data window

# References

package dependencies: ggplot2 3.2.1; rtop 0.5.14 Function dependencies: sinreg, d18O\_model, growth\_model

doi: 10.1029/91WR02985

#### See Also

Duan, Qingyun, Soroosh Sorooshian, and Vijai Gupta. "Effective and efficient global optimization for conceptual rainfall runoff models." Water resources research 28.4 (1992): 1015-1031. https://doi.org/10.1029/91WR02985

```
# Create dummy input data column by column
dat <- as.data.frame(seq(1000, 40000, 1000))</pre>
colnames(dat) <- "D"</pre>
dat$d180c <- sin((2 * pi * (seq(1, 40, 1) - 8 + 7 / 4)) / 7)
datYEARMARKER \leftarrow c(0, rep(c(0, 0, 0, 0, 0, 0, 1), 5), 0, 0, 0, 0)
dat$D_err <- rep(100, 40)
dat$d180c_err <- rep(0.1, 40)
# Create dummy dynwindow data
dynwindow <- as.data.frame(seq(1, 29, 2))</pre>
colnames(dynwindow) <- "x"</pre>
dynwindow$y <- rep(12, 15)
# Run model function
resultlist <- run_model(dat = dat,
    dynwindow = dynwindow,
    transfer_function = "KimONeil97",
    d180w = 0,
    T_{per} = 365,
```

*sd\_wt* 19

```
G_per = 365,
t_int = 1,
t_maxtemp = 182.5,
SCEUApar = c(1, 25, 10000, 5, 0.01, 0.01),
sinfit = TRUE,
MC = 1000,
plot = FALSE)
```

sd\_wt

Function to calculate weighted standard deviation

# Description

Calculates the standard deviation of a weighted sample set while propagating sample weights through the calculation.

# Usage

```
sd_wt(x, w, na.rm = FALSE)
```

# **Arguments**

Χ	Vector containing the values in the set
W	Vector containing the weights to each value (in the same order as x the optimized growth model
na.rm	Should NA values be removed from the set prior to calculation? TRUE/FALSE

# Value

The standard deviation of the weighted set of x values

```
# Create dummy data x \leftarrow seq(1, 10, 0.5) w \leftarrow c(seq(0.1, 1, 0.1), seq(0.9, 0.1, -0.1)) SDw \sim sd_wt(x, w, na.rm = TRUE) # Run the function
```

20 temperature\_curve

sinreg Function that carries out a sinusoidal regression
--

# **Description**

Fits a sinusoid through data provided as an x and y vector and returns a list containing both the fitted curve and the parameters of that curve. Used to produce initial values for modeling data windows and later to find peaks in modeled julian day values to align the result to a cumulative age timeline.

# Usage

```
sinreg(x, y, fixed_period = NA, plot = FALSE)
```

#### **Arguments**

Vector of x values of input data
 Vector of y values of input data
 fixed\_period Optional variable for fixing the period of the sinusoid in the depth domain. Defaults to NA, period is not fixed. Supply a single value to fix the period.

plot Should the fitting result be plotted? TRUE/FALSE

#### Value

A list containing a vector of parameters of the fitted sinusoid and the fitted values belonging to each x value. Fitting parameters: I = the mean annual value of the sinusoid (height) A = the amplitude of the sinusoid Dper = the period of the sinusoid in x domain peak = the location of the peak in the sinusoid R2adj = the adjusted R^2 value of the fit p = the p-value of the fit

# **Examples**

```
# Create dummy data x \leftarrow seq(1000, 11000, 1000) y \leftarrow sin((2 * pi * (seq(1, 11, 1) - 8 + 7 / 4)) / 7) sinlist \leftarrow sinreg(x, y, plot = FALSE) # Run the function
```

temperature\_curve Function that creates a sinusoidal Sea Surface Temperature (SST) curve from a list of parameters

# **Description**

Takes the specified parameters for amplitude, period, phase and average value as well as the number of years specified and the time interval. It then creates a sinusoid based on the boundary conditions. Used as intermediate step during iterative modeling.

Virtual\_shell 21

# Usage

```
temperature_curve(T_par, years = 1, t_int = 1)
```

# **Arguments**

T_par	List of four parameters describing (in order) amplitude (T_amp; in degrees C), period (T_per; in days), phase (T_pha in day of the year) and average temperature (T_av; in degrees C)
years	Length of the preferred sinusoid in number of years (defaults to 1)

t\_int Time interval of sinusoidal record (in days)

# Value

A matrix containing columns for time (in days) and SST (in degrees C)

# **Examples**

```
# Set parameters
T_amp <- 20
T_per <- 365
T_pha <- 150
T_av <- 15
T_par <- c(T_amp, T_per, T_pha, T_av)
SST <- temperature_curve(T_par, 1, 1) # Run the function</pre>
```

Virtual\_shell

Virtual input data for ShellChron

# **Description**

A dataset containing data used to test the ShellChron functions. Generated using the code in "Generate\_Virtual\_shell.r" in data-raw

# Usage

```
Virtual_shell
```

# **Format**

A data frame with 80 rows and 5 variables:

**D** Depth, in  $\mu$ m along the virtual record

d18Oc stable oxygen isotope value, in permille VPDB

**D\_err** Depth uncertainty, in  $\mu$ m

d18Oc\_err stable oxygen isotope value uncertainty, in permille VPDB

YEARMARKER "1" marking year transitions ...

22 wrap\_function

# Source

See code to generate data in data-raw Modified after virtual data described in de Winter et al., 2021 https://doi.org/gk98

wrap\_function

Full ShellChron workflow wrapped in a single function

# Description

Takes starting parameters and names of input files and directory and runs through all the steps of the ShellChron model. Function includes options for plotting and exporting raw data, which are parsed into underlying formulae.

# Usage

```
wrap_function(
  path = getwd(),
  file_name,
  transfer_function = "KimONeil97",
  t_{int} = 1,
  T_{per} = 365,
 d180w = 0,
  t_{maxtemp} = 182.5,
  SCEUApar = c(1, 25, 10000, 5, 0.01, 0.01),
  sinfit = TRUE,
 MC = 1000,
 plot = TRUE,
 plot_export = TRUE,
  export_raw = FALSE,
  export_path = getwd()
)
```

### **Arguments**

path String containing the path to the directory that contains the input data.

file\_name Name of the file that contains d18O data

transfer\_function

String containing the name of the transfer function. Defaults to Kim and O'Neil,

1997.

t\_int Time interval (in days; default = 1)

T\_per Period of SST sinusoid (in days; default = 365)

d180w Either a single value (constant d180w) or a vector of length equal to the period

in SST data (365 days by default) containing information about seasonality in d18Ow. Defaults to constant d18Ow of 0 permille VSMOW (the modern mean

ocean value)

wrap\_function 23

t_maxtemp	Timing of the warmest day of the year (in julian day; default = 182.5, or May 26th halfway through the year)
SCEUApar	Parameters for SCEUA optimization (iniflg, ngs, maxn, kstop pcento, peps) For details, refer to Duan et al. (1992) in references
sinfit	Apply sinusoidal fitting to guess initial parameters for SCEUA optimization? TRUE/FALSE
MC	Number of Monte Carlo simulations to apply for error propagation. Default = 1000
plot	Should an overview of the results of modeling be plotted? TRUE/FALSE
plot_export	Should the overview plot be exported as a PDF file? TRUE/FALSE
export_raw	Export tables containing all raw model results before being merged into tidy tables? TRUE/FALSE
export_path	Path where result files are exported

# Value

CSV tables of model results in the current working directory, optional plots in PDF format and list object of model results for further processing in the R workspace.

# References

function dependencies: data\_import, run\_model, cumulative\_day, export\_results

```
# find attached dummy data
example <- wrap_function(path = getwd(),</pre>
   file_name = system.file("extdata", "Virtual_shell.csv",
   package = "ShellChron"),
   transfer_function = "KimONeil97",
   t_int = 1,
   T_{per} = 365,
   d180w = 0,
   t_maxtemp = 182.5,
   SCEUApar = c(1, 25, 10000, 5, 0.01, 0.01),
   sinfit = TRUE,
   MC = 1000,
   plot = FALSE,
   plot_export = FALSE,
   export_raw = FALSE,
   export_path = tempdir()) # Run function
```

# **Index**

```
\ast datasets
     {\tt Virtual\_shell}, {\color{red} {21}}
age_corr, 2
cumdy, 3
\verb|cumulative_day|, 4
d180_model, 5
data_import, 6
export_results, 7
growth_model, 10
growth_rate_curve, 12
mc_err_form, 13
mc_err_orth, 14
mc_err_proj, 15
peakid, 16
run_model, 17
sd_wt, 19
{\tt sinreg}, \textcolor{red}{20}
temperature_curve, 20
Virtual_shell, 21
wrap\_function, \textcolor{red}{22}
```