

# Package ‘cOde’

May 8, 2026

**Type** Package

**Title** Automated C Code Generation for 'deSolve', 'bvptwp'

**Version** 1.1.1

**NeedsCompilation** no

**Depends** R (>= 3.0)

**Suggests** deSolve, bvptwp, testthat

**Date** 2022-02-23

**Author** Daniel Kaschek

**Maintainer** Daniel Kaschek <daniel.kaschek@gmail.com>

**Description** Generates all necessary C functions allowing the user to work with the compiled-code interface of `ode()` and `bvptwp()`. The implementation supports ``forcings" and ``events". Also provides functions to symbolically compute Jacobians, sensitivity equations and adjoint sensitivities being the basis for sensitivity analysis.

**License** GPL (>= 2)

**RoxygenNote** 7.1.1

**Repository** CRAN

**Date/Publication** 2022-02-23 22:10:02 UTC

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adjointSymb	<i>Compute adjoint equations of a function symbolically</i>
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### Description

Compute adjoint equations of a function symbolically

### Usage

```
adjointSymb(f, states = names(f), parameters = NULL, inputs = NULL)
```

### Arguments

f	Named vector of type character, the functions
states	Character vector of the ODE states for which observations are available
parameters	Character vector of the parameters
inputs	Character vector of the "variable" input states, i.e. time-dependent parameters (in contrast to the forcings).

### Details

The adjoint equations are computed with respect to the functional

$$(x, u) \mapsto \int_0^T \|x(t) - x^D(t)\|^2 + \|u(t) - u^D(t)\|^2 dt,$$

where  $x$  are the states being constrained by the ODE,  $u$  are the inputs and  $x^D$  and  $u^D$  indicate the trajectories to be best possibly approached. When the ODE is linear with respect to  $u$ , the attribute `inputs` of the returned equations can be used to replace all occurrences of  $u$  by the corresponding character in the attribute. This guarantees that the input course is optimal with respect to the above function.

### Value

Named vector of type character with the adjoint equations. The vector has attributes "chi" (integrand of the chisquare functional), "grad" (integrand of the gradient of the chisquare functional), "forcings" (character vector of the forcings necessary for integration of the adjoint equations) and "inputs" (the input expressed as a function of the adjoint variables).

**Examples**

```

## Not run:

#####
## Solve an optimal control problem:
#####

library(bvpSolve)

# O2 + O <-> O3
# O3 is removed by a variable rate u(t)
f <- c(
  O3 = " build_O3 * O2 * O - decay_O3 * O3 - u * O3",
  O2 = "-build_O3 * O2 * O + decay_O3 * O3",
  O  = "-build_O3 * O2 * O + decay_O3 * O3"
)

# Compute adjoints equations and replace u by optimal input
f_a <- adjointSymb(f, states = c("O3"), inputs = "u")
inputs <- attr(f_a, "inputs")
f_tot <- replaceSymbols("u", inputs, c(f, f_a))
forcings <- attr(f_a, "forcings")

# Initialize times, states, parameters
times <- seq(0, 15, by = .1)
boundary <- data.frame(
  name = c("O3", "O2", "O", "adjO3", "adjO2", "adjO"),
  yini = c(0.5, 2, 2.5, NA, NA, NA),
  yend = c(NA, NA, NA, 0, 0, 0))

pars <- c(build_O3 = .2, decay_O3 = .1, eps = 1)

# Generate ODE function
func <- func(f = f_tot, forcings = forcings,
            jacobian = "full", boundary = boundary,
            modelname = "example5")

# Initialize forcings (the objective)
forcData <- data.frame(time = times,
                      name = rep(forcings, each=length(times)),
                      value = rep(
                        c(0.5, 0, 1, 1), each=length(times)))
forc <- setForcings(func, forcData)

# Solve BVP
out <- bvptwPC(x = times, func = func, parms = pars, forcings = forc)

# Plot solution
par(mfcol=c(1,2))
t <- out[,1]
M1 <- out[,2:4]
M2 <- with(list(uD = 0, O3 = out[,2]),

```

```

        adj03 = out[,5], eps = 1, weightuD = 1),
    eval(parse(text=inputs)))

matplot(t, M1, type="l", lty=1, col=1:3,
        xlab="time", ylab="value", main="states")
abline(h = .5, lty=2)
legend("topright", legend = names(f), lty=1, col=1:3)
matplot(t, M2, type="l", lty=1, col=1,
        xlab="time", ylab="value", main="input u")
abline(h = 0, lty=2)

## End(Not run)

```

---

 bvptwpC

*Interface to bvptwp()*


---

## Description

Interface to bvptwp()

## Usage

```

bvptwpC(
  yini = NULL,
  x,
  func,
  yend = NULL,
  parms,
  xguess = NULL,
  yguess = NULL,
  ...
)

```

## Arguments

yini	named vector of type numeric. Initial values to be overwritten.
x	vector of type numeric. Integration times
func	return value from funC() with a boundary argument.
yend	named vector of type numeric. End values to be overwritten.
parms	named vector of type numeric. The dynamic parameters.
xguess	vector of type numeric, the x values
yguess	matrix with as many rows as variables and columns as x values
...	further arguments going to bvptwp()

**Details**

See `bvpSolve`-package for a full description of possible arguments

**Value**

matrix with times and states

**Examples**

```
## Not run:

#####
## Boundary value problem: Ozon formation with fixed ozon/oxygen ratio
## at final time point
#####

library(bvpSolve)

# O2 + O <-> O3
# diff = O2 - O3
# build_O3 = const.
f <- c(
  O3 = " build_O3 * O2 * 0 - decay_O3 * O3",
  O2 = "-build_O3 * O2 * 0 + decay_O3 * O3",
  O  = "-build_O3 * O2 * 0 + decay_O3 * O3",
  diff = "-2 * build_O3 * O2 * 0 + 2 * decay_O3 * O3",
  build_O3 = "0"
)

bound <- data.frame(
  name = names(f),
  yini = c(0, 3, 2, 3, NA),
  yend = c(NA, NA, NA, 0, NA)
)

# Generate ODE function
func <- func(f, jacobian="full", boundary = bound, modelname = "example4")

# Initialize times, states, parameters and forcings
times <- seq(0, 15, by = .1)
pars <- c(decay_O3 = .1)
xguess <- times
yguess <- matrix(c(1, 1, 1, 1, 1), ncol=length(times),
                 nrow = length(f))

# Solve BVP
out <- bvptwpC(x = times, func = func, parms = pars,
              xguess = xguess, yguess = yguess)

# Solve BVP for different ini values, end values and parameters
yini <- c(O3 = 2)
yend <- c(diff = 0.2)
```

```

pars <- c(decay_03 = .01)
out <- bvptwpC(yini = yini, yend = yend, x = times, func = func,
              parms = pars, xguess = xguess, yguess = yguess)

# Plot solution
par(mfcol=c(1,2))
t <- out[,1]
M1 <- out[,2:5]
M2 <- cbind(out[,6], pars)

matplot(t, M1, type="l", lty=1, col=1:4,
        xlab="time", ylab="value", main="states")
legend("topright", legend = c("03", "02", "0", "02 - 03"),
      lty=1, col=1:4)
matplot(t, M2, type="l", lty=1, col=1:2,
        xlab="time", ylab="value", main="parameters")
legend("right", legend = c("build_03", "decay_03"), lty=1, col=1:2)

## End(Not run)

```

---

compileAndLoad

*Compile and load shared object implementing the ODE system.*

---

### Description

Compile and load shared object implementing the ODE system.

### Usage

```
compileAndLoad(filename, dllname, fcontrol, verbose)
```

### Arguments

filename	Full file name of the source file.
dllname	Base name for source and dll file.
fcontrol	Interpolation method for forcings.
verbose	Print compiler output or not.

### Author(s)

Daniel Kaschek, <daniel.kaschek@physik.uni-freiburg.de>  
Wolfgang Mader, <Wolfgang.Mader@fdm.uni-freiburg.de>

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forcData	<i>Forcings data.frame</i>
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**Description**

Forcings data.frame

---

funC	<i>Generate C code for a function and compile it</i>
------	--

---

**Description**

Generate C code for a function and compile it

**Usage**

```
funC(
  f,
  forcings = NULL,
  events = NULL,
  fixed = NULL,
  outputs = NULL,
  jacobian = c("none", "full", "inz.lsodes", "jacvec.lsodes"),
  rootfunc = NULL,
  boundary = NULL,
  compile = TRUE,
  fcontrol = c("nospline", "einspline"),
  nGridpoints = -1,
  includeTimeZero = TRUE,
  precision = 1e-05,
  modelname = NULL,
  verbose = FALSE,
  solver = c("deSolve", "Sundials")
)
```

**Arguments**

f	Named character vector containing the right-hand sides of the ODE. You may use the key word <code>time</code> in your equations for non-autonomous ODEs.
forcings	Character vector with the names of the forcings
events	data.frame of events with columns "var" (character, the name of the state to be affected), "time" (numeric or character, time point), "value" (numeric or character, value), "method" (character, either "replace" or "add"). See <a href="#">events</a> . If "var" and "time" are characters, their values need to be specified in the parameter vector when calling <code>odeC</code> . An event function is generated and compiled with the ODE.

fixed	character vector with the names of parameters (initial values and dynamic) for which no sensitivities are required (will speed up the integration).
outputs	Named character vector for additional output variables, see arguments <code>nout</code> and <code>outnames</code> of <code>lsode</code>
jacobian	Character, either "none" (no jacobian is computed), "full" (full jacobian is computed and written as a function into the C file) or "inz.lsodes" (only the non-zero elements of the jacobian are determined, see <code>lsodes</code> )
rootfunc	Named character vector. The root function (see <code>lsoda</code> ). Besides the variable names ( <code>names(f)</code> ) also other symbols are allowed that are treated like new parameters.
boundary	data.frame with columns <code>name</code> , <code>yini</code> , <code>yend</code> specifying the boundary condition set-up. NULL if not a boundary value problem
compile	Logical. If FALSE, only the C file is written
fcontrol	Character, either "nospline" (default, forcings are handled by <code>deSolve</code> ) or "einspline" (forcings are handled as splines within the C code based on the <code>einspline</code> library).
nGridpoints	Integer, defining for which time points the ODE is evaluated or the solution is returned: Set -1 to return only the explicitly requested time points (default). If additional time points are introduced through events, they will not be returned. Set $\geq 0$ to introduce additional time points between <code>tmin</code> and <code>tmax</code> where the ODE is evaluated in any case. Additional time points that might be introduced by events will be returned. If splines are used with <code>fcontrol = "einspline"</code> , <code>nGridpoints</code> also indicates the number of spline nodes.
includeTimeZero	Logical. Include $t = 0$ in the integration time points if TRUE (default). Consequently, integration starts at $t = 0$ if only positive time points are provided by the user and at <code>tmin</code> , if also negative time points are provided.
precision	Numeric. Only used when <code>fcontrol = "einspline"</code> .
modelName	Character. The C file is generated in the working directory and is named <code>&lt;modelName&gt;.c</code> . If NULL, a random name starting with ".f" is chosen, i.e. the file is hidden on a UNIX system.
verbose	Print compiler output to R command line.
solver	Select the solver suite as either <code>deSolve</code> or <code>Sundials</code> (not available any more). Defaults to <code>deSolve</code> .

### Details

The function replaces variables by arrays `y[i]`, etc. and replaces " $\wedge$ " by `pow()` in order to have the correct C syntax. The file name of the C-File is derived from `f`. I.e. `funC(abc, ...)` will generate a file `abc.c` in the current directory. Currently, only explicit ODE specification is supported, i.e. you need to have the right-hand sides of the ODE.

### Value

the name of the generated shared object file together with a number of attributes

**Examples**

```
## Not run:
# Exponential decay plus constant supply
f <- c(x = "-k*x + supply")
func <- funcC(f, forcings = "supply")

# Example 2: root function
f <- c(A = "-k1*A + k2*B", B = "k1*A - k2*B")
rootfunc <- c(steadyState = "-k1*A + k2*B - tol")

func <- funcC(f, rootfunc = rootfunc, modelname = "test")

yini <- c(A = 1, B = 2)
parms <- c(k1 = 1, k2 = 5, tol = 0.1)
times <- seq(0, 10, len = 100)

odeC(yini, times, func, parms)

## End(Not run)
```

---

getSymbols

*Get symbols from a character*

---

**Description**

Get symbols from a character

**Usage**

```
getSymbols(char, exclude = NULL)
```

**Arguments**

char	Character vector (e.g. equation)
exclude	Character vector, the symbols to be excluded from the return value

**Value**

character vector with the symbols

**Examples**

```
getSymbols(c("A*AB+B^2"))
```

---

jacobianSymb	<i>Compute Jacobian of a function symbolically</i>
--------------	--

---

**Description**

Compute Jacobian of a function symbolically

**Usage**

```
jacobianSymb(f, variables = NULL)
```

**Arguments**

f	named vector of type character, the functions
variables	other variables, e.g. paramters, f depends on. If variables is given, f is derived with respect to variables instead of names(f)

**Value**

named vector of type character with the symbolic derivatives

**Examples**

```
jacobianSymb(c(A="A*B", B="A+B"))
jacobianSymb(c(x="A*B", y="A+B"), c("A", "B"))
```

---

odeC	<i>Interface to ode()</i>
------	---------------------------

---

**Description**

Interface to ode()

**Usage**

```
odeC(y, times, func, parms, ...)
```

**Arguments**

y	named vector of type numeric. Initial values for the integration
times	vector of type numeric. Integration times. If includeTimeZero is TRUE (see <a href="#">funC</a> ), the times vector is augmented by t = 0. If nGridpoints (see <a href="#">funC</a> ) was set >= 0, uniformly distributed time points between the first and last time point are introduced and the solution is returned for these time points, too. Any additional time points that are introduced during integration (e.g. event time points) are returned unless nGridpoints = -1 (the default).

```

func          return value from funC()
parms        named vector of type numeric.
...          further arguments going to ode()

```

### Details

See deSolve-package for a full description of possible arguments

### Value

matrix with times and states

### Examples

```

## Not run:

#####
## Ozone formation and decay, modified by external forcings
#####

library(deSolve)
data(forcData)
forcData$value <- forcData$value + 1

# O2 + O <-> O3
f <- c(
  O3 = " (build_O3 + u_build) * O2 * O - (decay_O3 + u_degrade) * O3",
  O2 = "-(build_O3 + u_build) * O2 * O + (decay_O3 + u_degrade) * O3",
  O  = "-(build_O3 + u_build) * O2 * O + (decay_O3 + u_degrade) * O3"
)

# Generate ODE function
forcings <- c("u_build", "u_degrade")
func <- funC(f, forcings = forcings, modelname = "test",
            fcontrol = "nospline", nGridpoints = 10)

# Initialize times, states, parameters and forcings
times <- seq(0, 8, by = .1)
yini <- c(O3 = 0, O2 = 3, O = 2)
pars <- c(build_O3 = 1/6, decay_O3 = 1)

forc <- setForcings(func, forcData)

# Solve ODE
out <- odeC(y = yini, times = times, func = func, parms = pars,
           forcings = forc)

# Plot solution

par(mfcol=c(1,2))
t1 <- unique(forcData[,2])
M1 <- matrix(forcData[,3], ncol=2)

```

```

t2 <- out[,1]
M2 <- out[,2:4]
M3 <- out[,5:6]

matplot(t1, M1, type="l", lty=1, col=1:2, xlab="time", ylab="value",
main="forcings", ylim=c(0, 4))
matplot(t2, M3, type="l", lty=2, col=1:2, xlab="time", ylab="value",
main="forcings", add=TRUE)

legend("topleft", legend = c("u_build", "u_degrade"), lty=1, col=1:2)
matplot(t2, M2, type="l", lty=1, col=1:3, xlab="time", ylab="value",
main="response")
legend("topright", legend = c("O3", "O2", "O"), lty=1, col=1:3)

#####
## Ozone formation and decay, modified by events
#####

f <- c(
  O3 = " (build_O3 + u_build) * O2 * 0 -
        (decay_O3 + u_degrade) * O3",
  O2 = "-(build_O3 + u_build) * O2 * 0 +
        (decay_O3 + u_degrade) * O3",
  O  = "-(build_O3 + u_build) * O2 * 0 +
        (decay_O3 + u_degrade) * O3",
  u_build = "0", # piecewise constant
  u_degrade = "0" # piecewise constant
)

# Define parametric events
events.pars <- data.frame(
  var = c("u_degrade", "u_degrade", "u_build"),
  time = c("t_on", "t_off", "2"),
  value = c("plus", "minus", "2"),
  method = "replace"
)

# Declar parametric events when generating funC object
func <- funC(f, forcings = NULL, events = events.pars, modelname = "test",
            fcontrol = "nospline", nGridpoints = -1)

# Set Parameters
yini <- c(O3 = 0, O2 = 3, O = 2, u_build = 1, u_degrade = 1)
times <- seq(0, 8, by = .1)
pars <- c(build_O3 = 1/6, decay_O3 = 1, t_on = exp(rnorm(1, 0)), t_off = 6, plus = 3, minus = 1)

# Solve ODE with additional fixed-value events
out <- odeC(y = yini, times = times, func = func, parms = pars)

```

```

# Plot solution

par(mfcol=c(1,2))
t2 <- out[,1]
M2 <- out[,2:4]
M3 <- out[,5:6]

matplot(t2, M3, type="l", lty=2, col=1:2, xlab="time", ylab="value",
        main="events")
legend("topleft", legend = c("u_build", "u_degrade"), lty=1, col=1:2)
matplot(t2, M2, type="l", lty=1, col=1:3, xlab="time", ylab="value",
        main="response")
legend("topright", legend = c("O3", "O2", "O"), lty=1, col=1:3)

#####
## Ozone formation and decay, modified by events triggered by root
#####

f <- c(
  O3 = " (build_O3 + u_build) * O2 * 0 -
        (decay_O3 + u_degrade) * O3",
  O2 = "-(build_O3 + u_build) * O2 * 0 +
        (decay_O3 + u_degrade) * O3",
  O  = "-(build_O3 + u_build) * O2 * 0 +
        (decay_O3 + u_degrade) * O3",
  u_build = "0", # piecewise constant
  u_degrade = "0" # piecewise constant
)

# Define parametric events
events.pars <- data.frame(
  var = c("u_degrade", "u_degrade", "u_build", "O3"),
  time = c("t_on", "t_off", "2", "t_thres_O3"),
  value = c("plus", "minus", "2", "0"),
  root = c(NA, NA, NA, "O3 - thres_O3"),
  method = "replace"
)

# Declar parametric events when generating funC object
func <- funC(f, forcings = NULL, events = events.pars, modelname = "test",
            fcontrol = "nospline", nGridpoints = -1)

# Set Parameters
yini <- c(O3 = 0, O2 = 3, O = 2, u_build = 1, u_degrade = 1)
times <- seq(0, 8, by = .01)
pars <- c(build_O3 = 1/6, decay_O3 = 1,
          t_on = exp(rnorm(1, 0)), t_off = 6, plus = 3, minus = 1,
          thres_O3 = 0.5, t_thres_O3 = 1)

# Solve ODE with additional fixed-value events

```

```

out <- odeC(y = yini, times = times, func = func, parms = pars, method = "lsode")

class(out) <- c("deSolve")
plot(out)
# Plot solution

par(mfcol=c(1,2))
t2 <- out[,1]
M2 <- out[,2:4]
M3 <- out[,5:6]

matplot(t2, M3, type="l", lty=2, col=1:2, xlab="time", ylab="value",
        main="events")
legend("topleft", legend = c("u_build", "u_degrade"), lty=1, col=1:2)
matplot(t2, M2, type="l", lty=1, col=1:3, xlab="time", ylab="value",
        main="response")
legend("topright", legend = c("O3", "O2", "O"), lty=1, col=1:3)

## End(Not run)

```

---

oxygenData

*Time-course data of O, O2 and O3*

---

### Description

Forcings data.frame

---

prodSymb

*Compute matrix product symbolically*

---

### Description

Compute matrix product symbolically

### Usage

prodSymb(M, N)

### Arguments

M                   matrix of type character  
N                   matrix of type character

**Value**

Matrix of type character, the matrix product of M and N

---

reduceSensitivities    *reduceSensitivities*

---

**Description**

reduceSensitivities

**Usage**

reduceSensitivities(sens, vanishing)

**Arguments**

sens	Named character, the sensitivity equations
vanishing	Character, names of the vanishing sensitivities

**Details**

Given the set vanishing of vanishing sensitivities, the algorithm determines sensitivities that vanish as a consequence of the first set.

**Value**

Named character, the sensitivity equations with zero entries for vanishing sensitivities.

---

replaceNumbers    *Replace integer number in a character vector by other double*

---

**Description**

Replace integer number in a character vector by other double

**Usage**

replaceNumbers(x)

**Arguments**

x	vector of type character, the object where the replacement should take place
---	--

**Value**

vector of type character, conserves the names of x.

replaceOperation      *Replace a binary operator in a string by a function*

---

**Description**

Replace a binary operator in a string by a function

**Usage**

```
replaceOperation(what, by, x)
```

**Arguments**

what	character, the operator symbol, e.g. "^"
by	character, the function string, e.g. "pow"
x	vector of type character, the object where the replacement should take place

**Value**

vector of type character

**Examples**

```
replaceOperation("^", "pow", "(x^2 + y^2)^.5")
```

---

replaceSymbols      *Replace symbols in a character vector by other symbols*

---

**Description**

Replace symbols in a character vector by other symbols

**Usage**

```
replaceSymbols(what, by, x)
```

**Arguments**

what	vector of type character, the symbols to be replaced, e.g. c("A", "B")
by	vector of type character, the replacement, e.g. c("x[0]", "x[1]")
x	vector of type character, the object where the replacement should take place

**Value**

vector of type character, conserves the names of x.

**Examples**

```
replaceSymbols(c("A", "B"), c("x[0]", "x[1]"), c("A*B", "A+B+C"))
```

---

sensitivitiesSymb	<i>Compute sensitivity equations of a function symbolically</i>
-------------------	---

---

**Description**

Compute sensitivity equations of a function symbolically

**Usage**

```
sensitivitiesSymb(
  f,
  states = names(f),
  parameters = NULL,
  inputs = NULL,
  events = NULL,
  reduce = FALSE
)
```

**Arguments**

f	named vector of type character, the functions
states	Character vector. Sensitivities are computed with respect to initial values of these states
parameters	Character vector. Sensitivities are computed with respect to initial values of these parameters
inputs	Character vector. Input functions or forcings. They are excluded from the computation of sensitivities.
events	data.frame of events with columns "var" (character, the name of the state to be affected), "time" (numeric or character, time point), "value" (numeric or character, value), "method" (character, either "replace" or "add"). See <a href="#">events</a> . Within <code>sensitivitiesSymb()</code> a data.frame of additional events is generated to reset the sensitivities appropriately, depending on the event method.
reduce	Logical. Attempts to determine vanishing sensitivities, removes their equations and replaces their right-hand side occurrences by 0.

**Details**

The sensitivity equations are ODEs that are derived from the original ODE `f`. They describe the sensitivity of the solution curve with respect to parameters like initial values and other parameters contained in `f`. These equations are also useful for parameter estimation by the maximum-likelihood method. For consistency with the time-continuous setting provided by [adjointSymb](#), the returned equations contain attributes for the chisquare functional and its gradient.

**Value**

Named vector of type character with the sensitivity equations. Furthermore, attributes "chi" (the integrand of the chisquare functional), "grad" (the integrand of the gradient of the chisquare functional), "forcings" (Character vector of the additional forcings being necessary to compute chi and grad) and "yini" (The initial values of the sensitivity equations) are returned.

**Examples**

```
## Not run:

#####
## Sensitivity analysis of ozone formation
#####

library(deSolve)

# O2 + O <-> O3
f <- c(
  O3 = " build_03 * O2 * O - decay_03 * O3",
  O2 = "-build_03 * O2 * O + decay_03 * O3",
  O  = "-build_03 * O2 * O + decay_03 * O3"
)

# Compute sensitivity equations
f_s <- sensitivitiesSymb(f)

# Generate ODE function
func <- func(c(f, f_s))

# Initialize times, states, parameters and forcings
times <- seq(0, 15, by = .1)
yini <- c(O3 = 0, O2 = 3, O = 2, attr(f_s, "yini"))
pars <- c(build_03 = .1, decay_03 = .01)

# Solve ODE
out <- odeC(y = yini, times = times, func = func, parms = pars)

# Plot solution
par(mfcol=c(2,3))
t <- out[,1]
M1 <- out[,2:4]
M2 <- out[,5:7]
M3 <- out[,8:10]
M4 <- out[,11:13]
M5 <- out[,14:16]
M6 <- out[,17:19]

matplot(t, M1, type="l", lty=1, col=1:3,
        xlab="time", ylab="value", main="solution")
legend("topright", legend = c("O3", "O2", "O"), lty=1, col=1:3)
matplot(t, M2, type="l", lty=1, col=1:3,
        xlab="time", ylab="value", main="d/(d O3)")
```



```

forc <- setForcings(func, forcData)

# Solve ODE
out <- odeC(y = c(yini, yini_s, yini_chi, yini_grad),
            times = times, func = func, parms = pars, forcings = forc,
            method = "lsodes")

# Plot solution
par(mfcol=c(1,2))
t <- out[,1]
M1 <- out[,2:4]
M2 <- out[,names(grad)]
tD <- oxygenData[,1]
M1D <- oxygenData[,2:4]

matplot(t, M1, type="l", lty=1, col=1:3,
        xlab="time", ylab="value", main="states")
matplot(tD, M1D, type="b", lty=2, col=1:3, pch=4, add=TRUE)
legend("topright", legend = names(f), lty=1, col=1:3)
matplot(t, M2, type="l", lty=1, col=1:5,
        xlab="time", ylab="value", main="gradient")
legend("topleft", legend = names(grad), lty=1, col=1:5)

# Define objective function
obj <- function(p) {
  out <- odeC(y = c(p[names(f)], yini_s, yini_chi, yini_grad),
            times = times, func = func, parms = p[names(pars)],
            forcings = forc, method="lsodes")

  value <- as.vector(tail(out, 1)[,"chi"])
  gradient <- as.vector(
    tail(out, 1)[,paste("chi", names(p), sep=".")])
  hessian <- gradient%*%t(gradient)

  return(list(value = value, gradient = gradient, hessian = hessian))
}

# Fit the data
myfit <- optim(par = c(yini, pars),
             fn = function(p) obj(p)$value,
             gr = function(p) obj(p)$gradient,
             method = "L-BFGS-B",
             lower=0,
             upper=5)

# Model prediction for fit parameters
prediction <- odeC(y = c(myfit$par[1:3], yini_s, yini_chi, yini_grad),
                 times = times, func = func, parms = myfit$par[4:5],
                 forcings = forc, method = "lsodes")

# Plot solution
par(mfcol=c(1,2))
t <- prediction[,1]

```

```

M1 <- prediction[,2:4]
M2 <- prediction[,names(grad)]
tD <- oxygenData[,1]
M1D <- oxygenData[,2:4]

matplot(t, M1, type="l", lty=1, col=1:3,
        xlab="time", ylab="value", main="states")
matplot(tD, M1D, type="b", lty=2, col=1:3, pch=4, add=TRUE)
legend("topright", legend = names(f), lty=1, col=1:3)
matplot(t, M2, type="l", lty=1, col=1:5,
        xlab="time", ylab="value", main="gradient")
legend("topleft", legend = names(grad), lty=1, col=1:5)

## End(Not run)

```

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setForcings	<i>Generate interpolation spline for the forcings and write into list of matrices</i>
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### Description

Generate interpolation spline for the forcings and write into list of matrices

### Usage

```
setForcings(func, forcings)
```

### Arguments

func	result from funC()
forcings	data.frame with columns name (factor), time (numeric) and value (numeric)

### Details

Splines are generated for each name in forcings and both, function value and first derivative are evaluated at the time points of the data frame.

### Value

list of matrices with time points and values assigned to the forcings interface of deSolve

### Examples

```

## Not run:
f <- c(x = "-k*x + a - b")
func <- funC(f, forcings = c("a", "b"))
forcData <- rbind(
  data.frame(name = "a", time = c(0, 1, 10), value = c(0, 5, 2)),

```

```
data.frame(name = "b", time = c(0, 5, 10), value = c(1, 3, 6))
forc <- setForcings(func, forcData)

## End(Not run)
```

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sumSymb	<i>Compute matrix sumSymbolically</i>
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**Description**

Compute matrix sumSymbolically

**Usage**

```
sumSymb(M, N)
```

**Arguments**

M	matrix of type character
N	matrix of type character

**Value**

Matrix of type character, the matrix sum of M and N

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