

R package rodeoFABM: Basic Use and Sample Applications

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1 Main features of rodeoFABM

The package `rodeoFABM` is a collection of small tools to help create water quality models that can be coupled to physical host models using the FABM interface (Bruggeman and Bolding (2014)). As the name suggests it is heavily influenced by the R package `rodeo` (Kneis, Petzoldt, and Berendonk (2017)). The principle idea is to have a system that:

- Helps users that don't have the technical know how
- make model adaptation, communication, and maintenance easy

Therefore the water quality model is written in the standard Peterson matrix notation and stored in text files or spread sheets. The package `rodeoFABM` automatically generates FABM specific FORTRAN code from these files and can automatically compile GOTM (Burchard et al. (2006)) coupled with the newly created model, as well as .yaml control files for the water quality model.

2 Installation and requirements

In order to fully use `rodeoFABM` and run the examples some tools are needed:

- The GNU compilers

- GNU Make
- GNU CMake
- Rdevtools
- R packages: `readODS`, `gotmtools`

The package `rodeoFABM` can be installed from github using:

```
library("devtools")
install_github("JFeldbauer/rodeoFABM")
```

3 Basic use

A simple example that is extended along the way

3.1 First example (how it works)

To demonstrate the workflow we will use a very simple model that is provided in the package. The files are contained in the package and can be loaded using:

```
# copy example ods file
example_model <- system.file("extdata/simple_model.ods", package = "rodeoFABM")
file.copy(from = example_model, to = ".", recursive = TRUE)
```

Now we can read in the tables with the declaration of the state variables, model parameters, used functions and external dependencies, process rate descriptions, and the stoichiometry matrix.

```
library(readODS)

# read in example ods file
odf_file <- "simple_model.ods"
vars <- read_ods(odf_file, 1)
pars <- read_ods(odf_file, 2)
fun <- read_ods(odf_file, 3)
pros <- read_ods(odf_file, 4)
stoi <- read_ods(odf_file, 5)
```

From these we can now generate FORTRAN files

```
library(rodeoFABM)

# generate fabm code
gen_fabm_code(vars,pars,fun,pros,stoi,"simple_model.f90",diags = TRUE)
```

And compile GOTM-FABM with them. Therfore first we need to clone the lake branche of GOTM-FABM from github and prepare the build process using cmake. This needs only to be done once, using the function `clone_GOTM()`:

```
# clone github repo
clone_GOTM(build_dir = "build", src_dir = "gotm_src")
```

Now we can build GOTM-FABM with our own model using:

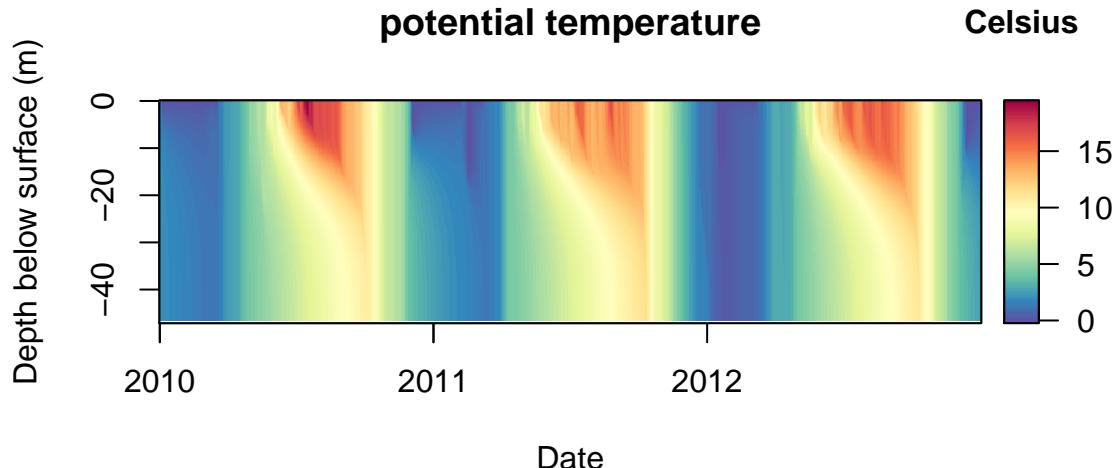
```
# build GOTM
build_GOTM(build_dir = "build",fabm_file = "simple_model.f90",
           src_dir = "gotm_src")
```

After copying the example `gotm.yaml` (the GOTM controll file), the exymple hypsograph, and the example forcing data, we finally can run GOTM-FABM with our own small model using:

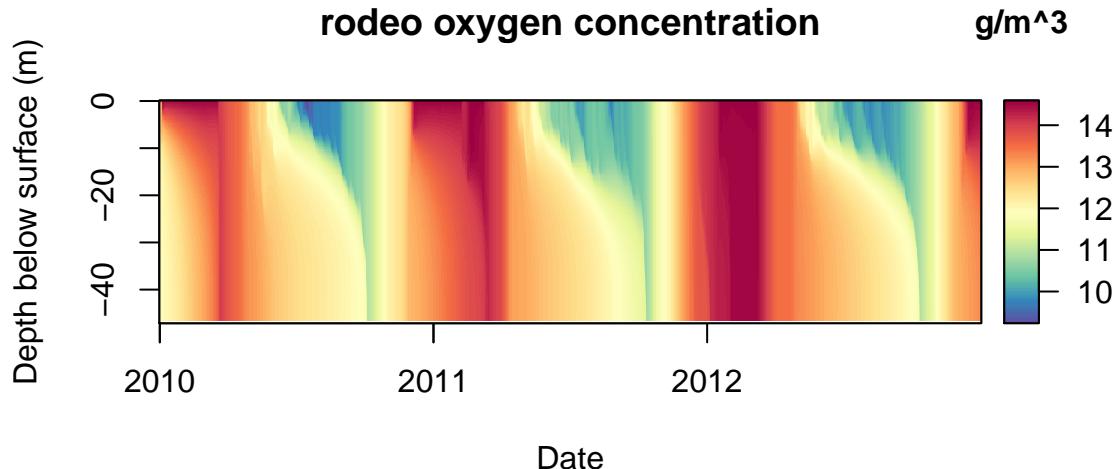
```
# copy example gotm.yaml
yaml <- system.file("extdata/gotm.yaml", package = "rodeoFABM")
file.copy(from = yaml, to = ".", recursive = TRUE)
# write hypsograph
write.table(hypsograph, "hypsograph.dat", sep = "\t", row.names = FALSE,
            quote = FALSE)
# write meteo data
write.table(meteo_file, "meteo_file.dat", sep = "\t", row.names = FALSE,
            quote = FALSE)
# run gotm
system2("./gotm")
```

We can plot the results e.g. using the `plot_var()` function using:

```
# plot temperature
plot_var("output.nc", "temp")
```



```
# plot oxygen
plot_var("output.nc", "rodeo_C_O2")
```



These are the essential steps used. In the next section we will go into the details of building a model using the `rodeoFABM` package step by step.

3.2 Create a model step by step

In order to demonstrate the necessary steps and functionality we will create a very simple phytoplankton-nutrients model and step by step add more processes.

The libre office spread sheets with the model information, GOTM controll file, and forcing data are contained in the `rodeoFABM` package. You can copy all necessary files to run GOTM using the same method as before. We will use the same meteorological forcing (`meteo_file.dat`) and hypsographic curve (`hypsograph.dat`) as in the first example. Additionally we now have one inflow and one outflow (files `inflow_m.dat`, `outflow.dat`, and `inflow_wq_m.dat` containing the nutrient concentrations of the inflow)

```
# GOTM controll file
yaml <- system.file("extdata/examples/gotm.yaml",
                     package = "rodeoFABM")
file.copy(from = yaml, to = ".", recursive = TRUE)
# inflow hydrological data
infl <- system.file("extdata/examples/inflow_m.dat",
                     package = "rodeoFABM")
file.copy(from = infl, to = ".", recursive = TRUE)
# inflow nutrient data
nut <- system.file("extdata/examples/inflow_wq_m.dat",
                     package = "rodeoFABM")
file.copy(from = nut, to = ".", recursive = TRUE)
# outflow data
out <- system.file("extdata/examples/outflow.dat",
                     package = "rodeoFABM")
file.copy(from = out, to = ".", recursive = TRUE)
```

The inflows and especially the inflow of state variables to a *FABM* model are defined in the `stream` section of the GOTM control file (`gotm.yaml`). The section looks like this:

```
streams:
  inflow:                                     # stream configuration
```

```

method: 4                                # inflow method, default=1
zu: 0.0                                    # upper limit m
z1: 0.0                                    # lower limit m
flow:
  method: 2                                # 0=constant, 2=from file, default = 0
  constant_value: 1.0                      # constant value( m^3/s)
  file: inflow_m.dat                      # path to file with time series
  column: 1                                 # index of column to read from
temp:
  method: 2                                # 0=constant, 2=from file; default=0
  constant_value: 10.0                     # constant value (°C)
  file: inflow_m.dat                      # path to file with time series
  column: 2                                 # index of column to read from
salt:
  method: 0                                # 0=constant, 2=from file; default=0
  constant_value: -1.0                     # constant value (PSU)
  file: inflow.dat                         # path to file with time series
  column: 3                                 # index of column to read from
rodeo_HPO4:
  method: 0                                # 0=constant, 2=from file; default=0
  constant_value: 0.5                       # constant value (gP/m^3)
  file: inflow_wq_m.dat                   # path to file with time series
  column: 4                                 # index of column to read from

```

Within the `streams` section several in- and outflows can be defined with any desired name (here `inflow`). The inflow/outflow depth is defined by `streams/method`, whereas 1 means surface, 2 means bottom, 3 mean a specified range of depths defined by `streams/zu` (upper) and `streams/z1` (l), and 4 means inflow to the depth with same temperature as the inflow temperature. Every in- or outflow needs the `streams/flow` section defining the flow rate in m^3/s and can have additional entries like `streams/temp` for temperature or inflowing state variables of the *FABM* model (like `streams/rodeo_HPO4`). The *FABM* state variables need to start with `rodeo_` followed by the defined state variable name. The values can either be constant (`streams/rodeo_HPO4/method = 0`) or a time series given by a tab separated file (`streams/rodeo_HPO4/method = 2`) with first column datetime (as `YYYY-mm-dd HH:MM:ss`). The name of the file is supplied by `streams/rodeo_HPO4/file` and the column the variable is in by `streams/rodeo_HPO4/column`, take care: the first column with datetime is not counted and if the columns have a header it needs to start with an exclamation mark “!”.

We can creat the firs phytoplankton nutrients model similar to how we created the first example:

```

# copy the spread sheet
ods <- system.file("extdata/examples/simple_alg.ods",
                    package = "rodeoFABM")
file.copy(from = ods, to = ".", recursive = TRUE)

# declare data frames for vars, pars, funs, pros, and stoi
vars <- read_ods("simple_alg.ods", sheet = "vars")
pars <- read_ods("simple_alg.ods", sheet = "pars")
funs <- NULL
pros <- read_ods("simple_alg.ods", sheet = "pros")
stoi <- read_ods("simple_alg.ods", sheet = "stoi")

```

This first model is a very simple model with two state variables, which are declared in the `vars` data frame. The table needs to have at least three columns *name* giving the identifier of the state variable, *unit* giving the used unit, and *description* giving a short description of the state variable. If additionally the column *default* is supplied the initial value will be included in the *FABM* control file (`fabm.yaml`), which is automatically generated by `gen_fabm_code()`.

Table 1: Data set `vars`: Declaration of state variables.

name	unit	description	default
C	gDM/m ³	algae concentration	0.0
$HPO4$	gP/m ³	phosphorus concentration	0.1

The models parameters are defined in the `pars` data frame in a similar fashion. They need the same three columns *name*, *unit*, and *description* and can have the additional column *default* as well. Take care that *FABM* requires all parameters with relation to time to be in units of second.

Table 2: Data set `pars`: Declaration of model parameters.

name	unit	description	default
μ_{max}	1/s	maximum growth rate	1e-05
K_P	W/m ²	half saturation concentration of HPO4 limitation	2e-02
k_{death}	1/s	death rate	2e-06
a_P	gP/gDM	phosphorus content of phytoplankton	5e-02

External functions, or forcing data that needs to be obtained from the physical host model (e.g. water temperature) are defined in `funs`. As the first simple model has no such things this is explained in the later steps. As in this example the data frame is not needed it has to be set to `NULL`.

The declaration of the processes and process rates is done in the `pros` data frame. It has four required columns: *name* giving the name of the process, *unit* giving the unit of the process rate (again in seconds!), *description* giving a short description of the process, and *expression* giving the mathematical expression of the process. There can be additional columns to define the spatial domain of the process, or to declare sinking processes, but they will be explained later.

Table 3: Data set `pros`: Declaration of processes.

name	unit	description	expression
growth	g/m ³ /d	growth of algae	$C \cdot \mu_{max} \cdot HPO4 / (HPO4 + K_P)$
death	g/m ³ /d	death of algae	$C \cdot k_{death}$

The phytoplankton have a simple linear growth term with a Monod like limitation for the limiting nutrient Phosphorus and a linear decay/death term.

The last data frame `stoi` gives the stoichiometry table (in long format) connecting the process rates with the state variables. It has three required columns: *variable* giving the variable affected by the *process*, and *expression* giving a factor to multiply the process rate by:

Table 4: Data set `stoi`: Declaration of stoichiometry matrix in long format.

variable	process	expression
C	growth	1
C	death	-1
$HPO4$	growth	$-1 \cdot a_P$

The growth of phytoplankton is increasing its concentration C and decreasing the nutrient $HPO4$ by the

fraction of a_P , which is the Phosphorus content of the phytoplankton. Decay/death is just decreasing phytoplankton concentration C .

Having declared all five data frames we can now generate the fortran code using `gen_fabm_code()`. This will also perform some automated checks e.g. if all used parameters and state variables are also declared, and will issue a warning if the used units are not using seconds for time. It will also create the *FABM* control file `fabm.yaml` and insert the default values for parameters and initial values (if declared). If the argument `diags` is set to `TRUE` the process rates are stored as diagnostic variables in the output netcdf file.

```
# create the fabm code
gen_fabm_code(vars, pars, funs, pros, stoi, "model_1.f90", diags = TRUE)
```

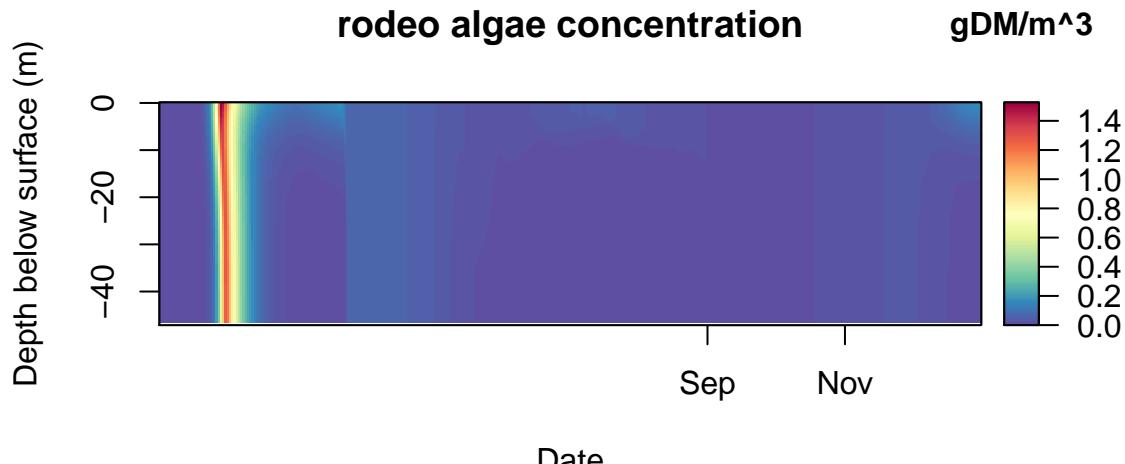
After creating the fortran source code, *GOTM-FABM* can be automatically compiled using the function `build_GOTM()` (assuming the source code was already fetched and prepared for compilation using `clone_GOTM()`), this will also copy the compiled executable to the current working directory, which then can be ran using e.g. `system2("./gotm")`.

```
# build GOTM with the model
build_GOTM(build_dir = "build", src_dir = "gotm_src",
            fabm_file = "model_1.f90")

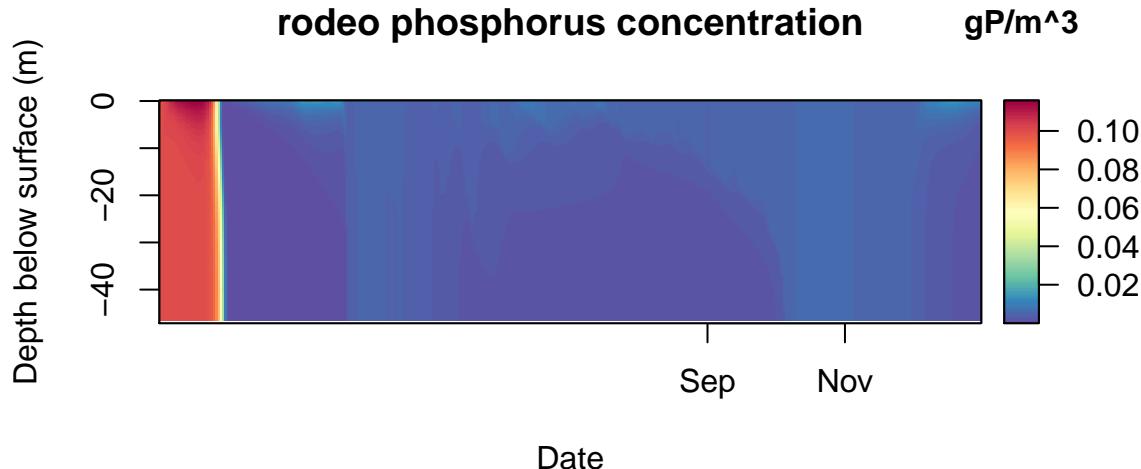
# run the model
system2("./gotm")
```

We can now plot the model results e.g. using `gotmtools::plot_vari()`.

```
# plot the variables
plot_var("output.nc", "rodeo_C")
```



```
plot_var("output.nc", "rodeo_HP04")
```



3.2.1 Getting dependencies from the host model

As many biogeochemical processes depend on external forcing, such as temperature or available irradiation, these values can be obtained from the physical host model. In the next step we want to add the dependency of phytoplankton growth on available irradiation. We first copy the prepared spread sheet and declare the data frames:

```
ods <- system.file("extdata/examples/simple_alg_par.ods",
                   package = "rodeoFABM")
file.copy(from = ods, to = ".", recursive = TRUE)

# declare data frames for vars, pars, funs, pros, and stoi
vars <- read_ods("simple_alg_par.ods", sheet = "vars")
pars <- read_ods("simple_alg_par.ods", sheet = "pars")
funs <- read_ods("simple_alg_par.ods", sheet = "funs")
pros <- read_ods("simple_alg_par.ods", sheet = "pros")
stoi <- read_ods("simple_alg_par.ods", sheet = "stoi")
```

Now we need to get values for the photosynthetic active radiation (PAR) from GOTM. *FABM* has so called “standard-variables” with defined names (stored in the `std_names_FABM` data). If you want to access these variables you need to define them as a function in the `funs` data frame and add the additional column *dependency* which contains the full standard-variable name. The data frame `funs` has three required columns that are the same as in `vars`, and `pars`: *name*, *unit*, and *description*, additionally the column *dependency*. If you declare several functions of whom some are not dependencies the corresponding entry in column *dependency* needs to be `NA` for these and the corresponding standard-name for the ones that are dependencies.

Table 5: Data set `funs`: Declaration of model functions and dependencies from the host model.

name	unit	description	dependency
<code>par</code>	W/m^2	Downwelling photosynthetic radiative flux	<code>downwelling_photosynthetic_radiative_flux</code>

The declared functions/dependencies can now be used in the process expression, same as parameters and state variables. We added a Monod Term for light limitation in the *growth* process:

Table 6: Data set `pros`: Declaration of processes.

name	unit	description	expression
growth	g/m ³ /d	growth of algae	$C \cdot \mu_{max} \cdot HPO_4 / (HPO_4 + K_P) \cdot par / (par + K_{par})$
death	g/m ³ /d	death of algae	$C \cdot k_{death}$

For this we need to declare the additional parameter for the half-saturation irradiation in the `pars` data frame:

Table 7: Data set `pars`: Declaration of model parameters.

name	unit	description	default
μ_{max}	1/s	maximum growth rate	1.0e-05
K_P	W/m ²	half saturation of photosyntetic flux	2.0e-02
k_{death}	1/s	death rate	2.0e-06
a_P	gP/gDM	phosphorus content of phytoplankton	5.0e-02
K_{par}	W/m ²	half saturation of photosyntetic flux	2.7e+01

Now we can generate the fortran code, compile *GOTM-FABM*, and run the adapted model.

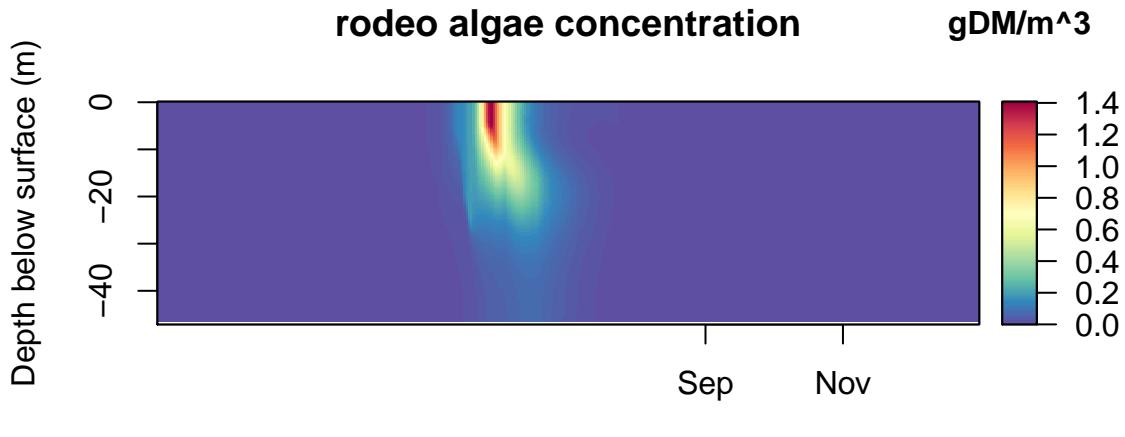
```
# create the fabm code
gen_fabm_code(vars, pars, funs, pros, stoi, "model_2.f90", diags = TRUE)

# build GOTM with the model
build_GOTM(build_dir = "build", src_dir = "gotm_src",
            fabm_file = "model_2.f90")

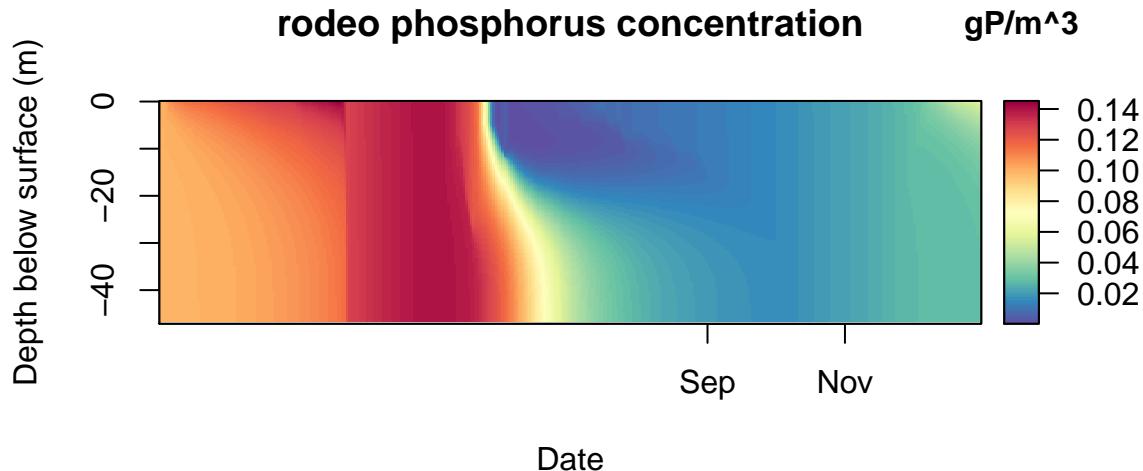
# run the model
system2("./gotm")
```

And plot some of the simulated state variables:

```
# plot the variables
plot_var("output.nc", "rodeo_C")
```



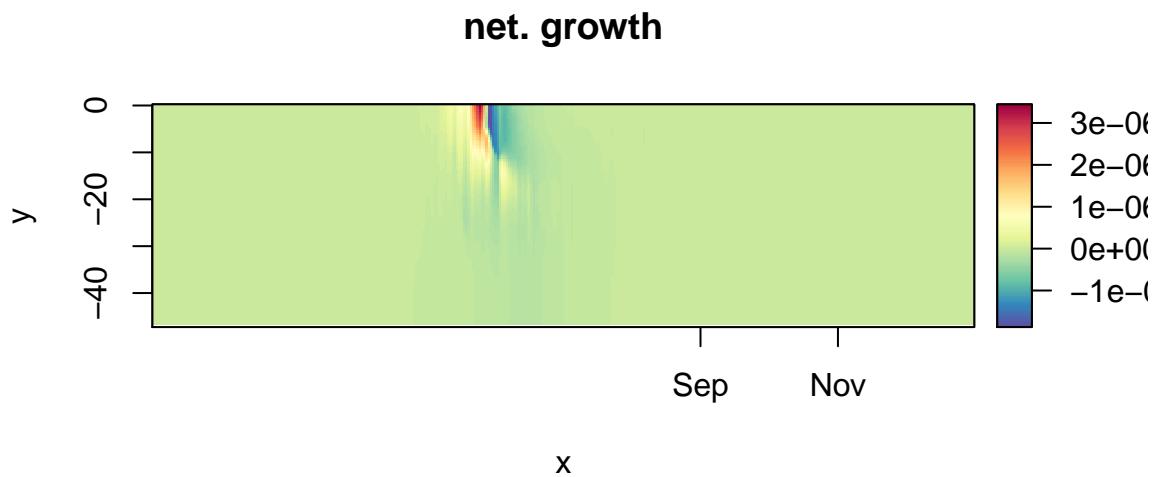
```
plot_var("output.nc", "rodeo_HP04")
```



From the saved process rates (the diagnostic variables) we can plot e.g. the net. growth rate. We can access the values stored in the netcdf file e.g. using the function `gotmtools::get_var()` and plot them using `gotmtools::long_heatmap()`:

```
library(plot3D)

# also plot net. growth
growth <- get_var("output.nc", "rodeo_growth")
death <- get_var("output.nc", "rodeo_death")
net_growth <- growth$var - death$var
# nice colors
mycol <- colorRampPalette(rev(RColorBrewer::brewer.pal(11, 'Spectral')))
# plot the net. growth
image2D(net_growth, growth$time, growth$z, main = "net. growth", col = mycol(100))
```



3.3 Sedimentation

Often in biogeochemical models some state variables are sinking in the water body (e.g. phytoplankton or particular organic matter). In the next adaptation of the model we want to include a constant sinking velocity for the phytoplankton. Therefore, again we first copy the spread sheets from the package data and declare the data frames:

```
ods <- system.file("extdata/examples/simple_alg_par_sed.ods",
                    package = "rodeoFABM")
file.copy(from = ods, to = ".", recursive = TRUE)

# declare data frames for vars, pars, funs, pros, and stoi
vars <- read_ods("simple_alg_par_sed.ods", sheet = "vars")
pars <- read_ods("simple_alg_par_sed.ods", sheet = "pars")
funs <- read_ods("simple_alg_par_sed.ods", sheet = "funs")
pros <- read_ods("simple_alg_par_sed.ods", sheet = "pros")
stoi <- read_ods("simple_alg_par_sed.ods", sheet = "stoi")
```

FABM allows for time varying sinking of state variables. This is implemented in *rodeoFABM* as a process declared in the *pros* data frame that has a logical flag set in an additional column called *sedi*. The expression for this can also be a function of external dependencies (e.g. water density) or internal state variables (e.g. nutrient concentration), in this simple case we choose a constant sinking velocity:

Table 8: Data set *pros*: Declaration of processes.

name	unit	description	expression	sedi
growth	g/m ³ /d	growth of algae	$C \cdot mu_max \cdot HPO4 / (HPO4 + K_P) \cdot par / (par + K_par)$	NA
death	g/m ³ /d	death of algae	$C \cdot k_death$	NA
sed	g/m ³ /s	sinking	v_sed	TRUE

For this to work we need to declare the additional parameter for the sinking velocity:

Table 9: Data set *pars*: Declaration of model parameters.

name	unit	description	default
<i>mu_max</i>	1/s	maximum growth rate	1.0e-05
<i>K_P</i>	W/m ²	half saturation of photosynthetic flux	2.0e-02
<i>k_death</i>	1/s	death rate	2.0e-06
<i>a_P</i>	gP/gDM	phosphorus content of phytoplankton	5.0e-02
<i>K_par</i>	W/m ²	half saturation of photosynthetic flux	2.7e+01
<i>v_sed</i>	m/s	sedimentation velocity	1.0e-06

And add the process to the stoichiometry table:

Table 10: Data set *stoi*: Declaration of stoichiometry matrix in long format.

variable	process	expression
<i>C</i>	growth	1
<i>C</i>	death	-1
<i>C</i>	sed	-1

variable	process	expression
<i>HPO4</i>	growth	$-1 \cdot a_P$

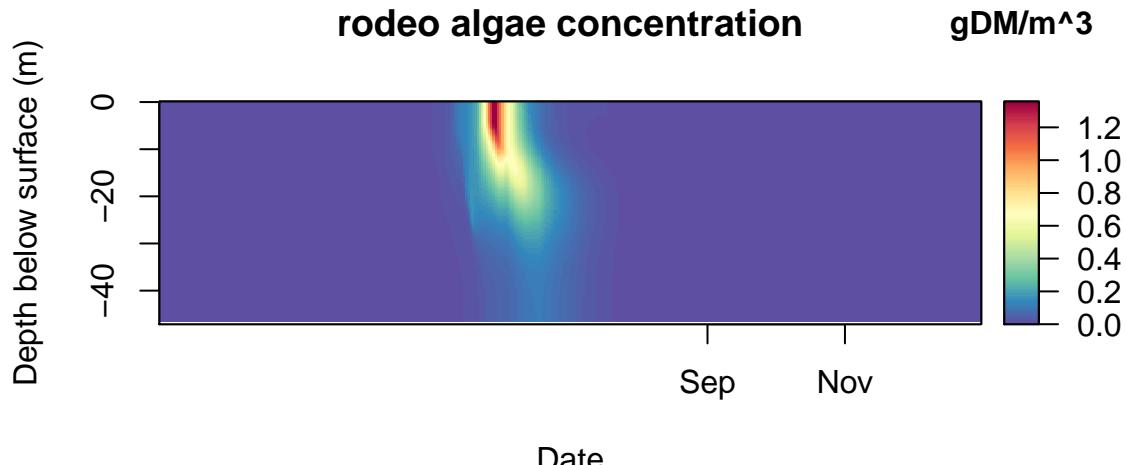
Now we can create the fortran source file, compile *GOTM-FABM*, run the model and plot some of the results:

```
# create the fabm code
gen_fabm_code(vars, pars, funs, pros, stoi, "model_3.f90")

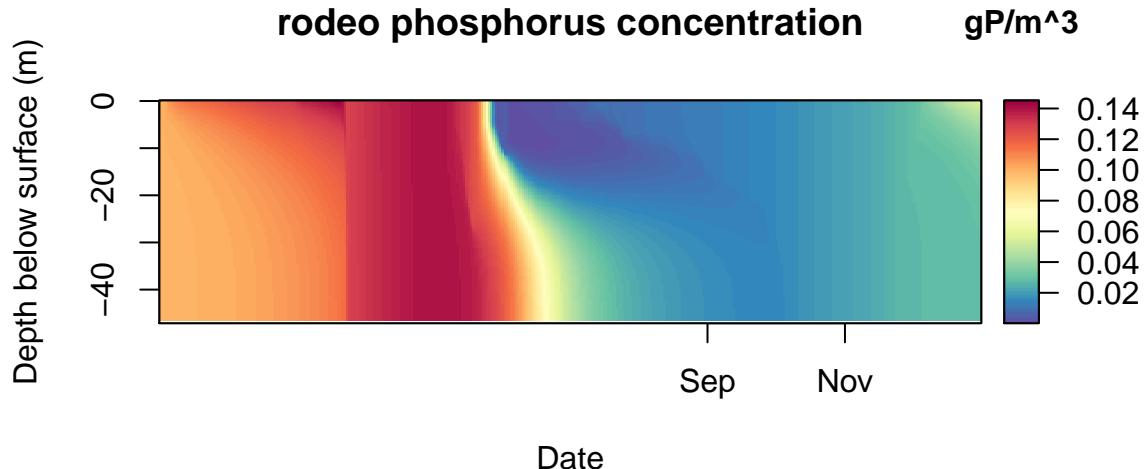
# build GOTM with the model
build_GOTM(build_dir = "build", src_dir = "gotm_src",
            fabm_file = "model_3.f90")

# run the model
system2("./gotm")

# plot the variables
plot_var("output.nc", "rodeo_C")
```

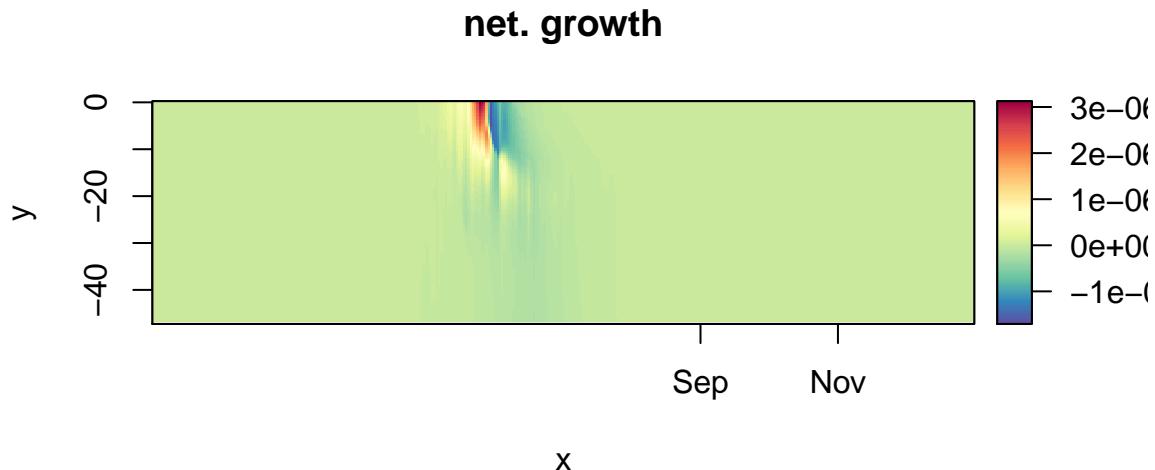


```
plot_var("output.nc", "rodeo_HPO4")
```



```
# also plot net. growth
growth <- get_var("output.nc", "rodeo_growth")
death <- get_var("output.nc", "rodeo_death")
net_growth <- growth$var - death$var

image2D(net_growth, growth$time, growth$z, main = "net. growth", col = mycol(100))
```



3.3.1 Processes at the upper and lower boundaries (surface and sediment)

There are some processes that only take place at the surface or bottom (sediment) of lakes. *FABM* knows three spatial domains: open water (pelagial), surface, and bottom (sediment) and processes can be declared to only take place at one of these domains. To demonstrate this we add oxygen along with surface exchange and a constant oxygen consumption in the sediment to the model. We again start by copying the spread sheet from the package data:

```

ods <- system.file("extdata/examples/simple_alg_02.ods",
  package = "rodeoFABM")
file.copy(from = ods, to = ".", recursive = TRUE)
# read in first simple model
vars <- read_ods("simple_alg_02.ods", sheet = "vars")
pars <- read_ods("simple_alg_02.ods", sheet = "pars")
funs <- read_ods("simple_alg_02.ods", sheet = "funs")
pros <- read_ods("simple_alg_02.ods", sheet = "pros")
stoi <- read_ods("simple_alg_02.ods", sheet = "stoi")

```

Now we need to add the new state variable $O2$ to the `vars` data frame:

Table 11: Data set `vars`: Declaration of state variables.

name	unit	description	default
C	gDM/m^3	algae concentration	0.0
$HPO4$	gP/m^3	phosphorus concentration	0.1
$O2$	gO/m^3	oxygen concentration	10.0

If processes occur only at the surface or bottom interface we can declare this by setting a logical flag in additional columns in the `pros` data frame called `bot` and `surf`. We add two new processes $O2_exch$, and $O2_cons$ and set the flag in the corresponding columns to TRUE:

Table 12: Data set ‘pros’: Declaration of processes.

name	unit	description	expression	surf	bot	sed
growth	$\text{g}/\text{m}^3/\text{d}$	growth of algae	$C \cdot mu_max \cdot HPO4 / (HPO4 + K_P) \cdot par / (par + K_par)$			
death	$\text{g}/\text{m}^3/\text{d}$	death of algae	$C \cdot k_death$			
sed	$\text{g}/\text{m}^3/\text{s}$	sinking	v_sed			TRUE
$O2_exch$	$\text{g}/\text{m}^3/\text{d}$	exchange of Oxygen at the surface	$v_O2 \cdot (exp(7.7117 - 1.31403 \cdot log(Temp + 45.93)) \cdot (p/101325) - O2)$	TRUE		
$O2_cons$	$\text{g}/\text{m}^3/\text{d}$	consumption of Oxygen in the pelagic	$O2 / (O2 + K_O2) \cdot k_O2_cons$		TRUE	

We need to declare the additional parameter for the oxygen exchange velocity, the constant consumption in the sediment, and the half-saturation concentration of oxygen limiting the oxygen consumption in the sediment:

Table 13: Data set `pars`: Declaration of model parameters.

name	unit	description	default
mu_max	1/s	maximum growth rate	1.0e-05
K_P	W/m^2	half saturation of photosynthetic flux	2.0e-02
k_death	1/s	death rate	2.0e-06
a_P	gP/gDM	phosphorus content of phytoplankton	5.0e-02
K_par	W/m^2	half saturation of photosynthetic flux	2.7e+01
v_sed	m/s	sedimentation velocity	1.0e-06
v_O2	1/s	speed of oxygen transfer	1.0e-05
k_O2_cons	$1/\text{s}/\text{m}^2$	Oxygen consumption rate in sediment	5.0e-07
K_O2	gO/m^3	half saturation concentration of oxygen consumption	5.0e+00
a_O	gO/gDM	oxygen production per growth of algae	1.0e+00

We need to declare the used functions \log , and \exp , as well as the external dependencies p (the barometric pressure at the surface), and $Temp$ (water temperature) which are needed to calculate the oxygen saturation concentration:

Table 14: Data set `funs`: Declaration of model functions and dependencies from the host model.

name	unit	description	dependency
<i>par</i>	W/m ²	Downwelling photosynthetic radiative flux	downwelling_photosynthetic_radiative_flux
<i>p</i>	Pa	Atmospheric Pressure	surface_air_pressure
<i>Temp</i>	celsius	Water temperature	temperature
<i>exp</i>	-	exponential function	NA
<i>log</i>	-	logarithmic function	NA

After adding the new processes to the stoichiometry table:

Table 15: Data set `stoi`: Declaration of stoichiometry matrix in long format.

variable	process	expression
<i>C</i>	growth	1
<i>C</i>	death	-1
<i>C</i>	sed	-1
<i>HPO4</i>	growth	-1 · <i>a_P</i>
<i>O2</i>	<i>O2_exch</i>	1
<i>O2</i>	<i>O2_cons</i>	-1
<i>O2</i>	growth	<i>a_O</i>

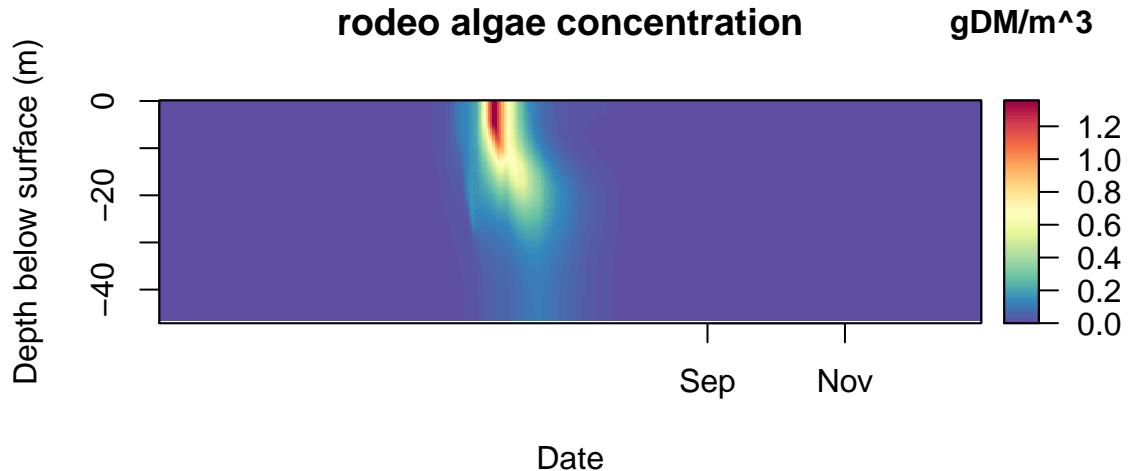
We can generate the source code, compile *GOTM-FABM*, run the model, and plot some of the results:

```
# create the fabm code
gen_fabm_code(vars, pars, funs, pros, stoi, "model_4.f90")

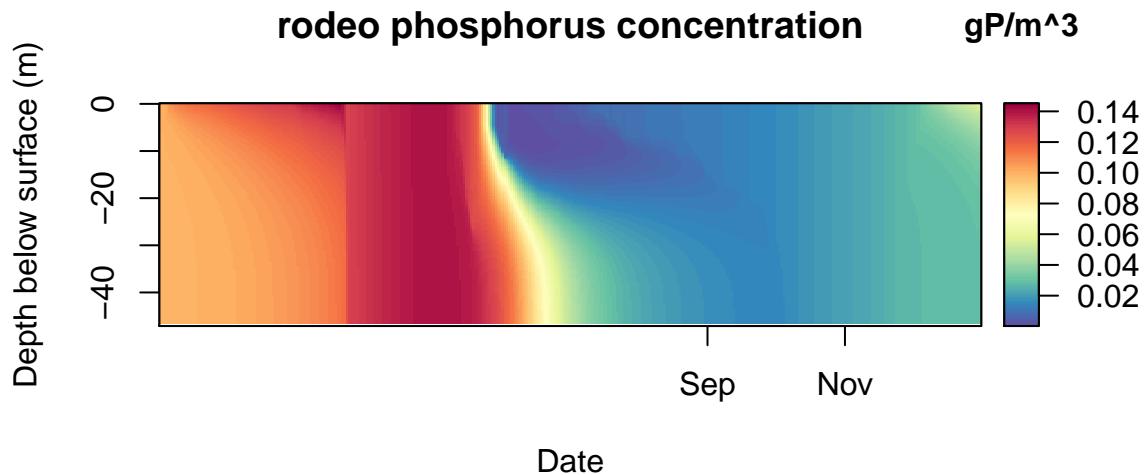
# build GOTM with the model
build_GOTM(build_dir = "build", src_dir = "gotm_src",
            fabm_file = "model_4.f90")

# run the model
system2("./gotm")

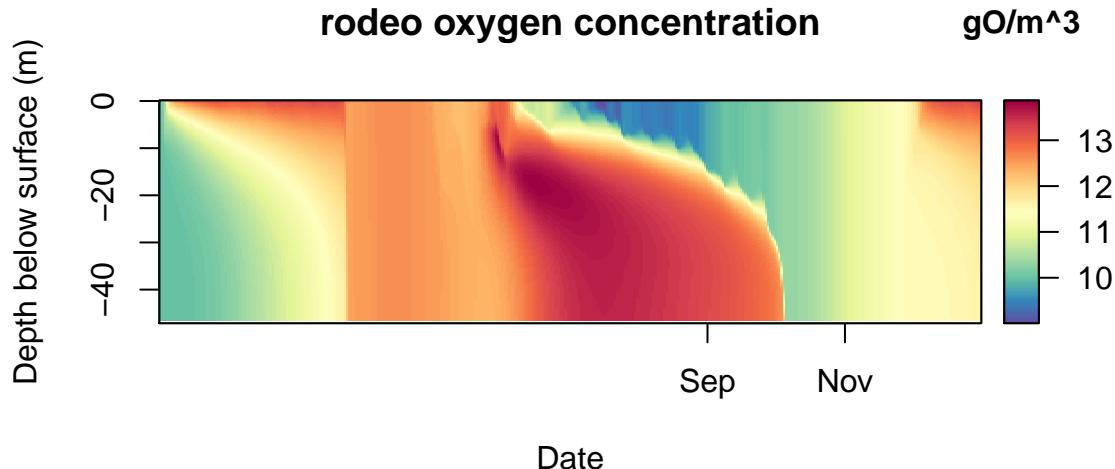
# plot the variables
plot_var("output.nc", "rodeo_C")
```



```
plot_var("output.nc", "rodeo_HP04")
```



```
plot_var("output.nc", "rodeo_02")
```



3.3.2 Sediment or surface attached state variables

As mentioned before *FABM* recognizes three spatial domains: open water, surface, and sediment. Like Processes, also state variables can be attaced to one of these domaines (e.g. sedimented particulated organic matter). To demonstrate this feature we will include two more state variables in our model; particulated organic matter (*POM*) and sedimented particulated organic matter (*SPOM*). Again we first need to copy the spread sheet from the package:

```
ods <- system.file("extdata/examples/simple_alg_02_POM.ods",
                    package = "rodeoFABM")
file.copy(from = ods, to = ".", recursive = TRUE)

# read in first simple model
vars <- read_ods("simple_alg_02_POM.ods", sheet = "vars")
pars <- read_ods("simple_alg_02_POM.ods", sheet = "pars")
funs <- read_ods("simple_alg_02_POM.ods", sheet = "funs")
pros <- read_ods("simple_alg_02_POM.ods", sheet = "pros")
stoi <- read_ods("simple_alg_02_POM.ods", sheet = "stoi")
```

Then we add the two new state variables *POM* and *SPOM*. We declare *SPOM* as bottom bound state variable by adding another column to the *vars* data frame called *bot* and set it to TRUE for all bottom bound state variables and to NA or FALSE for all others. Surface bound state variables can be declared in a column named *surf* in the same manner.

Table 16: Data set *vars*: Declaration of state variables.

name	unit	description	default	bot
<i>C</i>	gDM/m ³	algae concentration	0.0	NA
<i>HPO4</i>	gP/m ³	phosphorus concentration	0.1	NA
<i>O2</i>	gO/m ³	oxygen concentration	10.0	NA
<i>POM</i>	gDM/m ³	particulated organic matter	0.0	NA
<i>SPOM</i>	gDM/m ²	sedimented particulated organic matter	0.0	TRUE

The death of algae generates *POM*, which settles down, and sediments to the ground to become *SPOM*.

Both POM and $SPOM$ are mineralized, releasing HPO_4 but the mineralization is faster in the sediment. We add the new sinking, sedimentation, and mineralization processes to the `pros` data frame:

Table 17: Data set ‘`pros`’: Declaration of processes.

name	unit	description	expression	surf	bot	sed
growth	gDW/m^3/d	growth of algae	$C \cdot mu_max \cdot (HPO_4) / (HPO_4 + K_P) \cdot (par) / (par + K_par)$			
death	gDW/m^3/d	death of algae	$C \cdot k_death$			
sed_ALG	gDW/m^3/s	sinking of algae	v_sed_ALG			TRUE
O2_exch	gO/m^3/d	exchange of Oxygen at the surface	$v_O2 \cdot (\exp(7.7117 - 1.31403 \cdot \log(Temp + 45.93)) \cdot (p) / (101325) - O2)$	TRUE		
O2_cons	gO/m^3/d	consumption of Oxygen in the pelagic	$(O2) / (O2 + K_O2) \cdot k_O2_cons$			
sed_POM	gDW/m^3/s	sinking of POM	v_sed_POM			TRUE
miner_POM	gDW/m^3/s	mineralization of POM	$POM \cdot k_miner_POM \cdot (O2) / (O2 + K_miner_O2)$			TRUE
miner_SPOM	gDW/m^3/s	mineralization of SPOM	$SPOM \cdot k_miner_SPOM \cdot (O2) / (O2 + K_miner_O2)$			
set_POM	gDW/m^3/s	settling of POM	$v_sed_POM \cdot POM$			TRUE

Then we need to adapt the stoichiometry table:

Table 18: Data set `stoi`: Declaration of stoichiometry matrix in long format.

variable	process	expression
C	growth	1
C	death	-1
C	sed_ALG	-1
HPO_4	growth	$-1 \cdot a_P$
HPO_4	miner_POM	a_P
HPO_4	miner_SPOM	a_P
$O2$	O2_exch	1
$O2$	O2_cons	-1
$O2$	growth	a_O
$O2$	miner_POM	$-1 \cdot a_miner$
$O2$	miner_SPOM	$-1 \cdot a_miner$
POM	sed_POM	-1
POM	set_POM	-1
POM	death	1
POM	miner_POM	-1
$SPOM$	set_POM	1
$SPOM$	miner_SPOM	-1

And declare the new parameters for the sinking velocity, the mineralization kinetic, and the half-saturation concentration limiting the mineralization:

Table 19: Data set `pars`: Declaration of model parameters.

name	unit	description	default
mu_max	1/s	maximum growth rate	1.0e-05
K_P	W/m^2	half saturation of photosynthetic flux	2.0e-02
k_death	1/s	death rate	2.0e-06
a_P	gP/gDM	phosphorus content of phytoplankton	5.0e-02
K_par	W/m^2	half saturation of photosynthetic flux	2.7e+01
v_sed_ALG	m/s	sedimentation velocity of algae	1.0e-06
v_O2	1/s	speed of oxygen transfer	1.0e-04
k_O2_cons	1/s/m^2	Oxygen consumption rate in sediment	5.0e-07
K_O2	gO/m^3	half saturation concentration of oxygen consumption	5.0e+00
a_O	gO/gDM	oxygen production per growth of algae	1.0e+00
v_sed_POM	m/s	sedimentation velocity of POM	2.0e-06
K_miner_O2	gO/m^3	half saturation concentration of oxygen for mineralization	3.0e+00

name	unit	description	default
k_{miner_POM}	1/s	maximum mineralization rate of POM	0.0e+00
k_{miner_SPOM}	1/s	maximum mineralization rate of SPOM	3.0e-07
a_{miner}	gO/gDM	oxygen consumption per oxygenation of POM/SPOM	1.0e+00

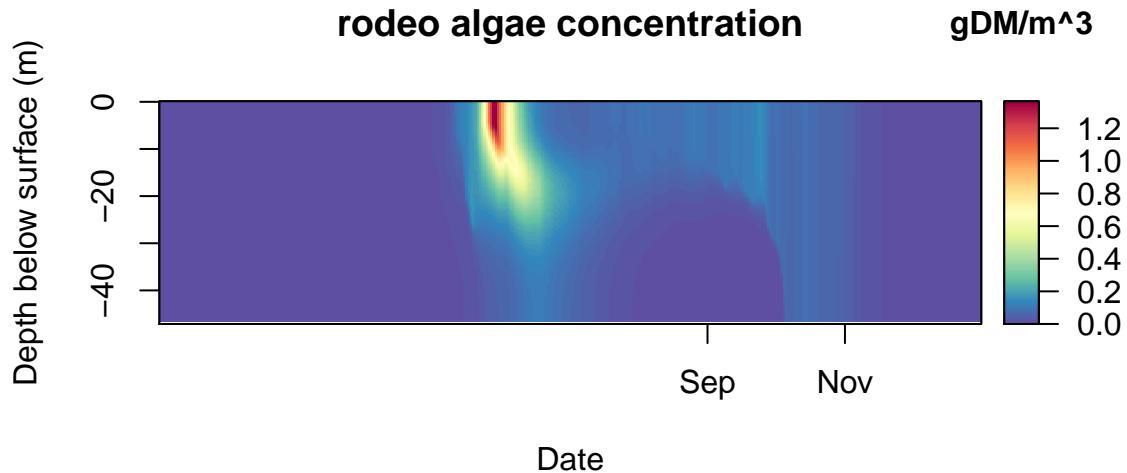
Now we can create fortran source code, compile *GOTM-FABM*, run the model, and plot the results:

```
# create the fabm code
gen_fabm_code(vars, pars, funs, pros, stoi, "model_5.f90")

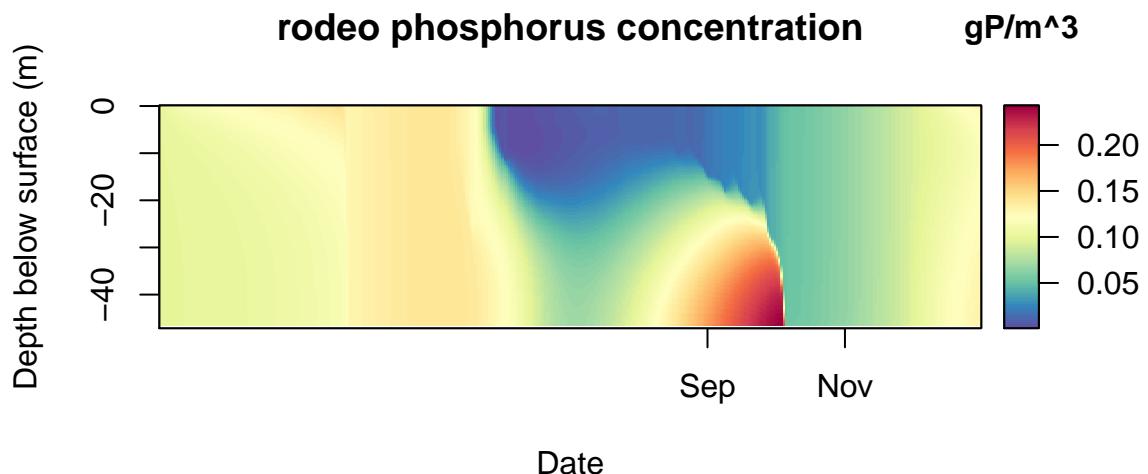
# build GOTM with the model
build_GOTM(build_dir = "build", src_dir = "gotm_src",
            fabm_file = "model_5.f90")

# run the model
system2("./gotm")

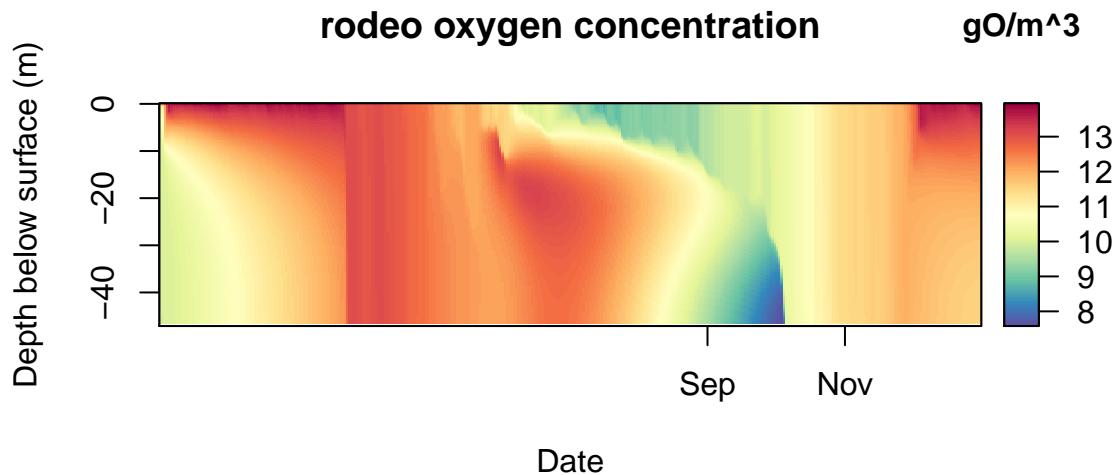
# plot the variables
plot_var("output.nc", "rodeo_C")
```



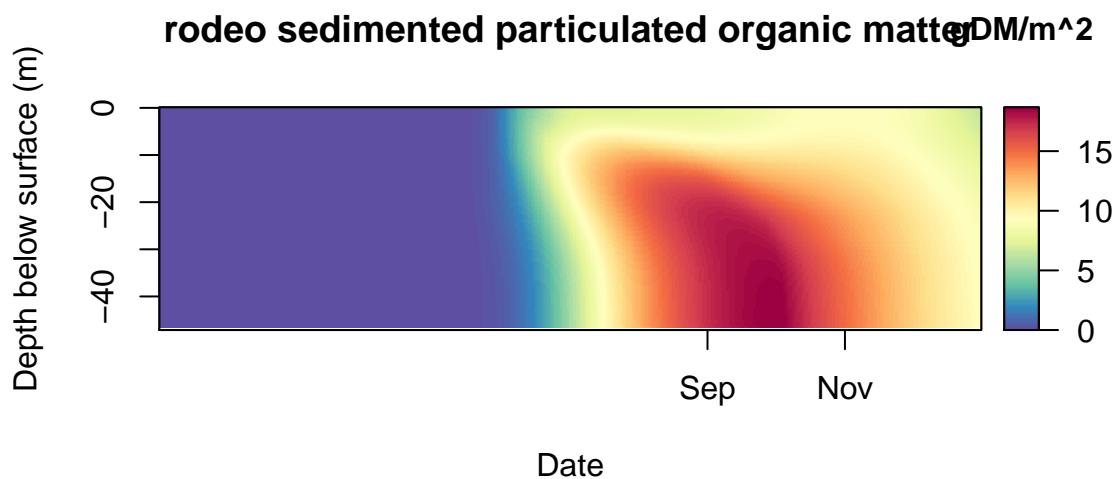
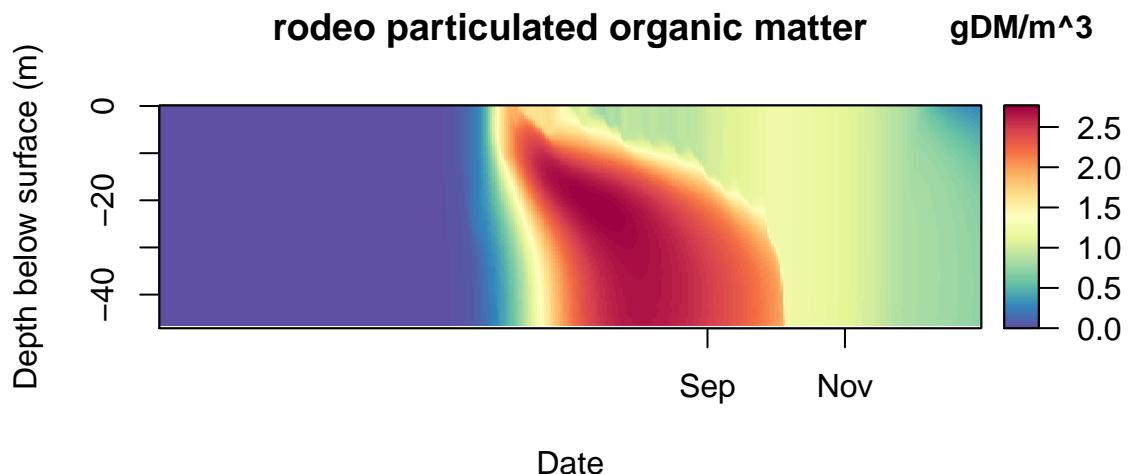
```
plot_var("output.nc", "rodeo_HP04")
```



```
plot_var("output.nc", "rodeo_02")
```



```
plot_var("output.nc", "rodeo_POM")
```



3.4 Additional features

3.4.1 Additional state variable arguments

There are a few additional arguments for state variables that can be defined in *FABM*. In order to use them a new column in the state variable data frame needs to be added with exactly the name.

- *minimum*: minimum allowed value for the state variable
- *maximum*: maximum allowed value for the state variable
- *specific_light_extinction*: specific light extinction coefficient of this variable
- *no_precipitation_dilution*: the variable is not diluted by precipitation
- *no_river_dilution*: the variable is not diluted by river inflows

3.4.2 Profile initial values for the state variables

By default the initial values for the *FABM* state variables are constant throughout the whole profile. With some tinkering we can profile profiles of initial values. Therefore, you need to have (a) text file(s) with (the) profiles that you want to initialise and install the R packages `ncdf4`, `data.table`, `gotmtools` from the AEMON-J github. The approach is to first run GOTM with 0 time steps (stop date = start date). After a run with GOTM, a `restart.nc` file is created, that can be used to restart a simulation with the same settings that ended the previous simulation. By running with 0 time steps, this file contains the “standard” initial values, including the ones in your biogeochemical model. Then you need to replace the homogeneous initial profiles by your specified profiles in the `restart.nc` file. Lastly, you need to set the `restart` option to “true”. If you run `gotm` now, it will run with the initial profiles for your biogeochemical model (note that you’ll have to rerun this approach every time before you want to run GOTM, because every new GOTM run overwrites the `restart.nc` file). Alternatively you can save the created `restart.nc` file e.g. as `restart_init_profiles.nc` and use this file to override the `restart.nc` before you run GOTM.

3.4.3 Defining own functions

3.4.3.1 Defining functions in the *fun*s data.frame

It is possible to define own functions and call them. If the function is simple i.e. can be calculated in one line of code (e.g. limiting functions for algae growth) it can be defined from the *expressions* column of the *pros* data.frame. In order to define own functions two additional columns are needed in the *fun*s data.frame: `expression` and `arguments`. Similar to the *pros* data.frame the `expression` column gives the mathematical expression to calculate. The `arguments` column gives all input arguments that the functions uses, in the same order they are supplied during function calls (e.g. in the `expression` column of the *pros* data.frame), separated by commas (“,”).

3.4.3.2 Defining functions in external fortran files

More complex functions can be supplied as external fortran code. Two additional columns are needed in the *fun*s data.frame: `file` and `module`. They give the name of the source code file and the name of the module, which is then loaded in the main source code. This feature is still very experimental and might lead to errors!

3.4.4 Automatic model documentation

If wanted `rodeoFABM` can automatically generate LaTeX documentation of the state variables, parameter, processes and stoichiometry. To do so the function `document_model()` can be used. Lets create a documentation of the final phytoplankton nutrients model from our example. In order to work we need an additional column named `tex` (you can also use another name for this column and supply the name to `document_model()` using the `tex` argument) in the data frames *vars*, *pars*, *fun*s, and *pros* giving the corresponding LaTeX symbols to be used. The documentation function automatically generates LaTeX fraction, but in order for this to work all used fractions in the `expression` column of the *pros* data frame need to be in a specified format. The numerator and denominator need to be in brackets, even if they are just one single variable, number, or parameter: e.g. $(O_2)/(O_2 + K_{O_2})$. In the example spread sheet file they are already added:

```
# see column "tex"
head(vars)

##   name     unit          description default  bot    tex
## 1   C gDM/m^3      algae concentration    0.0   NA     C
## 2 HP04 gP/m^3      phosphorus concentration  0.1   NA HP0_4
## 3   O2 g0/m^3       oxygen concentration   10.0   NA    O_2
## 4   POM gDM/m^3     particulated organic matter  0.0   NA    POM
## 5 SPOM gDM/m^2     sedimented particulated organic matter  0.0 TRUE  SPOM
```

```
# create LaTeX documentation for our model
document_model(vars, pars, pros, funs, stoi, landscape = FALSE)
```

```
##  
##  finished  
## [1] TRUE
```

We can see that now there are seven additional file in our working directory:

```
grep(".*\\.tex", list.files(), value = TRUE)
```

```
## [1] "document_model.tex" "preamble-latex.tex" "pros_expr.tex"      "tab_funs.tex"      "tab_pars.tex"  
## [6] "tab_pros.tex"       "tab_stoi.tex"        "tab_vars.tex"
```

They are LaTeX tables of the models state variables (*tab_vars.tex*), used model parameters (*tab_pars.tex*), used functions (*tab_funs.tex*), declaration of the models processes (*tab_pros.tex*), description of the process equations (*pros_expr.tex*), the stoichiometry table (*tab_stoi.tex*), and a very simple latex document that can be used to compile all of the before (*document_model.tex*).

The created expressions of the processes now look like this:

```
head(readLines("pros_expr.tex"))
```

```
## [1] "\\begin{align}"  
## [2] " \\rho_{growth} =&~ C \\cdot \\mu_{max} \\cdot \\frac{HPO_4}{HPO_4 + K_P} \\cdot \\frac{par}{par + K_{par}}"  
## [3] " \\rho_{death} =&~ C \\cdot k_{death}\\\\"  
## [4] " \\rho_{Sed,ALG} =&~ v_{sed,ALG}\\\\"  
## [5] " \\rho_{O2,exch} =&~ v_{exch,O2} \\cdot \\left( \\exp(7.7117 - 1.31403 \\cdot \\log(\\vartheta_z + 45.93)) \\cdot \\frac{p}{101325} - O_2 \\right)"  
## [6] " \\rho_{O2,cons} =&~ \\frac{O_2}{O_2 + K_{O2}} \\cdot k_{O2,cons}\\\\"
```

and compiled they look like this:

$$\rho_{growth} = C \cdot \mu_{max} \cdot \frac{HPO_4}{HPO_4 + K_P} \cdot \frac{par}{par + K_{par}} \quad (1)$$

$$\rho_{death} = C \cdot k_{death} \quad (2)$$

$$\rho_{Sed,ALG} = v_{sed,ALG} \quad (3)$$

$$\rho_{O2,exch} = v_{exch,O2} \cdot \left(\exp(7.7117 - 1.31403 \cdot \log(\vartheta_z + 45.93)) \cdot \frac{p}{101325} - O_2 \right) \quad (4)$$

$$\rho_{O2,cons} = \frac{O_2}{O_2 + K_{O2}} \cdot k_{O2,cons} \quad (5)$$

$$\rho_{Sed,POM} = v_{sed,POM} \quad (6)$$

$$\rho_{Miner,POM} = POM \cdot k_{miner,POM} \cdot \frac{O_2}{O_2 + K_{miner,O2}} \quad (7)$$

$$\rho_{Miner,SPOM} = SPOM \cdot k_{miner,SPOM} \cdot \frac{O_2}{O_2 + K_{miner,O2}} \quad (8)$$

$$\rho_{Set,POM} = v_{sed,POM} \cdot POM \quad (9)$$

References

Bruggeman, Jorn, and Karsten Bolding. 2014. “A General Framework for Aquatic Biogeochemical Models.” *Environmental Modelling & Software* 61 (November): 249–65. <https://doi.org/10.1016/j.envsoft.2014.04.002>.

Burchard, Hans, Karsten Bolding, Wilfried Kühn, Andreas Meister, Thomas Neumann, and Lars Umlauf. 2006. "Description of a Flexible and Extendable Physical–Biogeochemical Model System for the Water Column." *Journal of Marine Systems* 61 (3): 180–211. <https://doi.org/https://doi.org/10.1016/j.jmarsys.2005.04.011>.

Kneis, David, Thomas Petzoldt, and Thomas U. Berendonk. 2017. "An R-Package to Boost Fitness and Life Expectancy of Environmental Models." *Environmental Modelling & Software* 96 (October): 123–27. <https://doi.org/10.1016/j.envsoft.2017.06.036>.