

Group project report

A template for a more structured writing of the group project report

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Abstract

You can include a short abstract here. It should summarize the overall aim and key findings of your project.

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1 Introduction

The report starts with an introduction, which should explain the broader context of the project and define project aims. Ultimately, it should serve as a *motivation* for the work presented in the subsequent sections.

2 Methods

The Methods section describes the “how”. It should explain

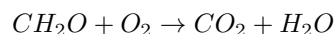
- the theoretical underpinning of the model (state variables, including units, conceptual diagram, chemical equations, if relevant, mass balance equations);

- key assumptions (rate expressions, boundary conditions);
- model implementation in R (model parameters & state variables, model function).

2.1 Equations

When writing chemical or mathematical equations, you can use L^AT_EX commands within the Rmd file. For example:

- Displayed chemical equation (without a label):



- Displayed chemical equation (with a label):



- Displayed mathematical equation (with a label):

$$\frac{\partial O_2}{\partial t} = Transport(O_2) + R_{aeration} - R_{miner} \tag{2}$$

The label can then be used to reference Eq. 2 and Eq. 1 even if you change the order of equations later on.

2.2 Model implementation in R

In this section you will show, and explain, how you implemented the model in R. As an example, here we include the implementation of the BOD model (available in the RTM_1D template).

```
require(ReacTran)

# units: time=days, space=meters, amount=moles, concentration=mol/m3

# model grid
Length <- 1000 # [m]
N <- 100 # [-] number of boxes
Grid <- setup.grid.1D(L = Length, N = N) # grid of N equally-sized boxes

# initial conditions - state variables are defined in the middle of grid cells
O2 <- rep(0.1, times = N) # [mol/m3]
BOD <- rep(0.001, times = N) # [mol/m3]

# initial state of the system: a vector with all state variables (2*N)
state.ini <- c(O2, BOD)

# names of the modeled state variables
SVnames <- c("O2", "BOD")

# model parameters
pars <- c(
```

```

D      = 100,  # [m2/d]      dispersion coefficient (tidal mixing)
v      = 10,   # [m/d]      advection velocity
kDecay = 0.05 , # [/d]      rate constant of BOD decay (first-order process)
K.O2   = 0.001, # [mol/m3]   half-saturation O2 concentration for BOD decay
inputBOD = 10,  # [mol/m2/d] BOD input rate upstream
BODdown = 0.1,  # [mol/m3]   BOD concentration downstream
O2up    = 0.25, # [mol/m3]   O2 concentration upstream
satO2    = 0.3, # [mol/m3]   saturation concentration of O2 (i.e., solubility)
kAeration = 0.1 # [/d]      rate constant for air-water O2 exchange
)

# Model function
BOD1D <- function(t, state, parms) { # state is a long vector, at time t
  with (as.list(parms),{

    # The vectors of the state variables O2 and BOD are
    # "extracted" from the LONG vector state passed to the function as input.
    O2 <- state[ (0*N+1) : (1*N) ] # first N elements for O2
    BOD <- state[ (1*N+1) : (2*N) ] # second N elements for BOD

    # Transport - tran.1D approximates the spatial derivatives
    # note: for O2: zero-gradient boundary downstream (default)
    tranO2 <- tran.1D(C      = O2,
                      C.up = O2up,      # imposed conc upstream,
                      D = D, v = v,      # dispersion, advection
                      dx = Grid)         # Grid

    tranBOD <- tran.1D(C      = BOD,
                      flux.up = inputBOD, # imposed flux upstream
                      C.down = BODdown,   # imposed conc downstream
                      D = D, v = v,       # dispersion, advection
                      dx = Grid)           # Grid

    # rate expressions [mol/m3/d] - values in the middle of grid cells
    Decay <- kDecay * BOD * O2/(O2+K.O2) # BOD decay, limited by O2
    Aeration <- kAeration * (satO2-O2)   # air-water exchange of O2

    # Time-derivatives: dC/dt = transport + production - consumption [mol/m3/d]
    dO2.dt <- tranO2$dC + Aeration - Decay
    dBOD.dt <- tranBOD$dC - Decay

    # return vector of time-derivatives and ordinary variables as a list
    return(list(c(dO2.dt, dBOD.dt), # time-derivatives
               # (the same order as state variables!!)

    # additional output:

    # process rates along the domain (1D vector)
    Decay = Decay, # mol/m3/d

```

```

Aeration          = Aeration,                # mol/m3/d

# mean process rates (a number)
MeanDecay          = mean(Decay),             # mol/m3/d
MeanAeration       = mean(Aeration),          # mol/m3/d

# rates integrated along the domain (for budgetting)
TotalDecay         = sum(Decay*Grid$dx),      # mol/m2/d
TotalAeration      = sum(Aeration*Grid$dx),   # mol/m2/d

# fluxes at domain boundaries (for budgetting)
BODinflux = tranBOD$flux.up,                 # BOD flux INTO the system upstream,      mol/m2/d
BODefflux  = tranBOD$flux.down,               # BOD flux OUT of the system downstream, mol/m2/d
O2influx   = tranO2$flux.up,                 # O2 flux INTO the system upstream,      mol/m2/d
O2efflux   = tranO2$flux.down))              # O2 flux OUT of the system downstream, mol/m2/d
})
}

```

2.3 Modeled scenarios

In this section you can also include the *calculation* (not yet plotting) of model outputs for specific model scenarios.

For example, here we find a steady-state solution for the default model parameters.

```

std <- steady.1D(y = state.ini, parms = pars, func = BOD1D,
               positive = TRUE, names = SVnames, nspec = length(SVnames), dims = N,
               atol = 1e-10, rtol = 1e-10)

```

In contrast, here we solve the model assuming no aeration.

```

p2 <- pars
p2["kAeration"] <- 0
std2 <- steady.1D(y = state.ini, parms = p2, func = BOD1D,
               positive = TRUE, names = SVnames, nspec = length(SVnames), dims = N,
               atol = 1e-10, rtol = 1e-10)

```

We also generate the content of tables. We do not display them just yet, but we wait until the Results section.

```

toselect <- c("TotalDecay", "TotalAeration", "BODinflux", "BODefflux",
             "O2influx", "O2efflux")
BUDGET <- data.frame(default      = unlist(std[toselect]),
                    no_aeration = unlist(std2[toselect]))

```

This can go on until you clarify all methodological aspects covered in your project.

3 Results

The Results section should summarize your key findings supported by graphs and/or tables.

3.1 Including variables generated in Methods

When knitting this file independently of the `methods.Rmd` file, you need to ensure that the values of the variables generated by the R-code in the `methods.Rmd` file are known here. You can do this by including the following R-chunk (enclosed between the opening and closing “‘”) somewhere at the beginning of the `results.Rmd` file:

```
{r, message=FALSE, echo=FALSE, eval=TRUE}
res <- knitr::knit_child('methods.Rmd', quiet = TRUE)
```

Once you have finished editing the `results.Rmd` file and are ready to knit the “master” file of your project, you can omit knitting of the `methods.Rmd` child here by using `eval=FALSE` in the above R-chunk:

```
{r, message=FALSE, echo=FALSE, eval=FALSE}
res <- knitr::knit_child('methods.Rmd', quiet = TRUE)
```

3.2 Plotting graphs

It is recommended to use *one* R-chunk for plotting one figure (plot). In this way, you can individually optimize the size of the figure and the corresponding figure caption.

Note that if you specify `fig.cap`, the figure will become a *floating* object and will have a reference number. How this floating object is placed within your final document can be prioritized in the yaml header of your Rmd file. Here, we use

```
\usepackage{float}
\floatplacement{figure}{ht}
```

which means that the first preference is to place the figure in the same place where the figure is generated (**h**, meaning *here*), while the second preference is to place it at the top of the next page (**t**, meaning *top*). These are L^AT_EX-specific tricks.

Note that there is no need to display the R-code that is used to generate the graphs, as this code is typically quite long but not overly informative. This is achieved by using the `echo=FALSE` flag in the corresponding R-chunk.

3.3 Reporting values in tables

Tables generated in the Methods section can be nicely typeset using the `kable` function from the `knitr` package.

```
knitr::kable(BUDGET, digits = 2, caption = "Budgets for the different model scenarios.")
```

Table 1: Budgets for the different model scenarios.

	default	no_aeration
TotalDecay	9.96	3.08
TotalAeration	9.87	0.00
BODinflux	10.00	10.00
BODefflux	0.04	6.92
O2influx	2.95	3.08
O2efflux	2.86	0.00

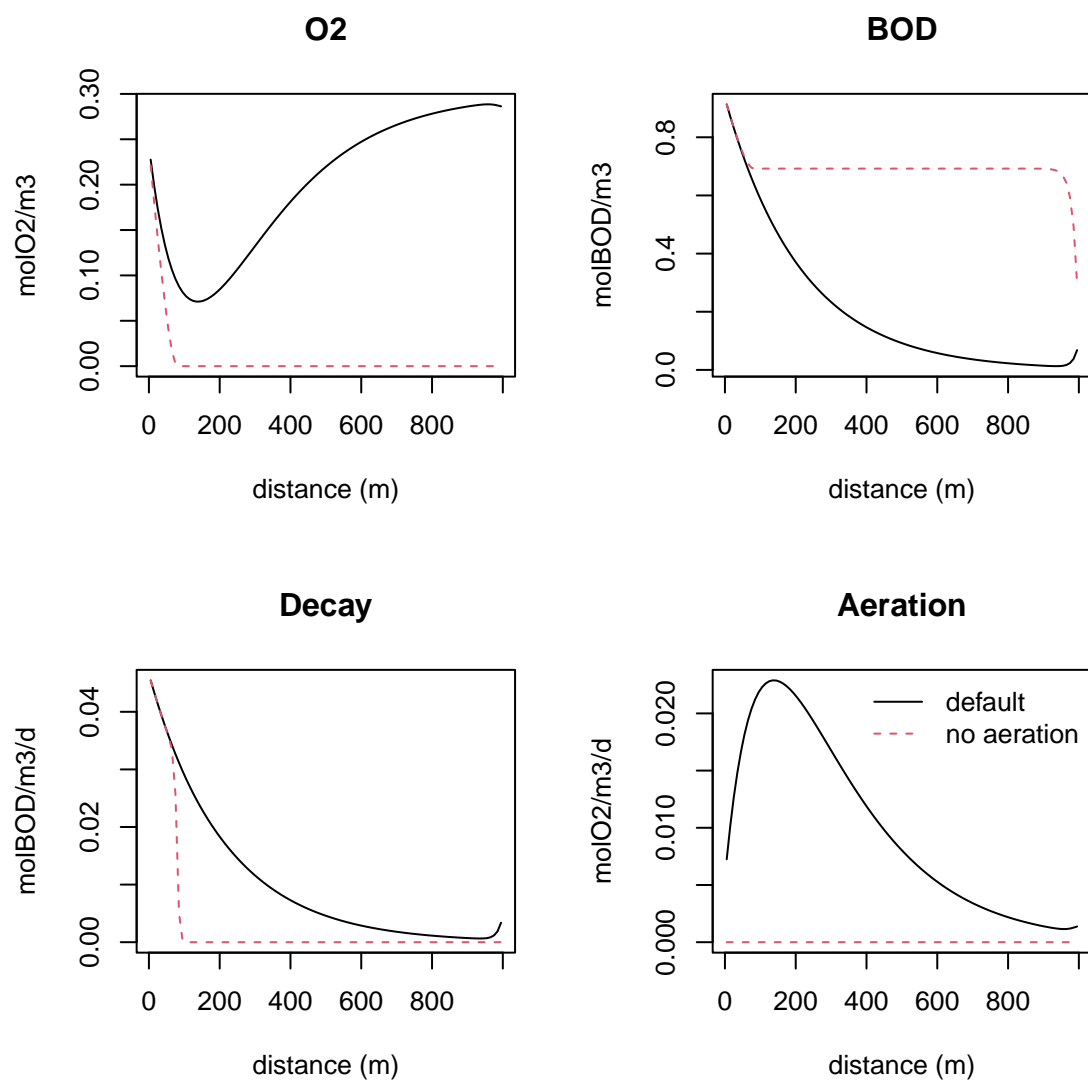


Figure 1: A comparison of model outputs for a default scenario and for a scenario without aeration.

4 Discussion

The Discussion section should tie the results into a coherent story. Specifically, it should explain what the results *mean*, clarify to what extent the project aims were reached, provide ideas for future work, etc.

5 Acknowledgements

It is always a good idea to acknowledge help of others.

6 Author's contributions

We also require that you list specific contributions of each author.

7 References

Lubos Polerecky, Dries Bonte, and Karline Soetaert (2021). RTM: learning environment for reaction-transport modelling in R. <https://github.com/dynamic-R/RTM>

R Core Team (2020). R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria. URL <https://www.R-project.org/>.

Soetaert Karline (2009). rootSolve: Nonlinear root finding, equilibrium and steady-state analysis of ordinary differential equations. R-package version 1.6

Soetaert, Karline and Meysman, Filip, (2012). Reactive transport in aquatic ecosystems: Rapid model prototyping in the open source software R Environmental Modelling & Software, 32, 49-60.