The BOD-O2 model

A template markdown file for a simple dynamic model in 0D

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Contents

Introduction	
Model definition	2
Model solution	2
Dynamic solution	
Steady-state solution	
Sensitivity analysis	4
References	ŗ

Introduction

This template file contains a simple box model describing the dynamics of molecular oxygen (O₂, in $mol\ m^{-3}$) and biochemical oxygen demand (BOD, in $mol\ m^{-3}$) in a lake. It is assumed that

- O₂ and BOD are removed due to BOD decay (O₂:BOD stoichiometry of 1:1),
- BOD decay is a first-order process with respect to BOD $(r = 0.05 d^{-1})$ and depends on O₂ according to the Michaelis-Menten kinetics $(kO_2 = 0.001 \ mol \ m^{-3})$,

 • O_2 is added by air-water exchange $(k = 0.1 \ d^{-1}, O_{2,sat} = 0.3 \ mol \ m^{-3})$,

 • DOB is added at a constant rate $(0.001 \ mol \ m^{-3} \ d^{-1})$,

 • initial concentrations are $O_{2,ini} = 0.25 \ mol \ m^{-3}$ and $O_{ini} = 0.5 \ mol \ m^{-3}$.

Model definition

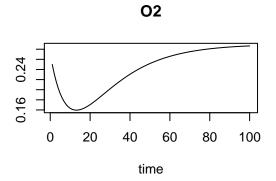
```
# Initial conditions of the state variables
yini <-c(02 = 0.25, BOD = 0.5) # both in [mol/m3]
# Model parameters
pars <- c(
 rDecay = 0.05 , # [/d]
                               first-order rate constant
         = 0.001, # [mol/m3] half-saturation 02 concentration
 sat02 = 0.3 , # [mol/m3] 02 solubility
          = 0.1 , # [/d]
                           reaeration rate constant
 inputBOD = 0.001 # [mol/m3/d] constant BOD input rate
# Model function: calculates time-derivatives and other output
BODmodel <-function(t, state, pars) {</pre>
  # t: time, state: state variables, pars: model parameters
  with (as.list(c(state, pars)),{
  # rate expressions [mol/m3/d]
          <- rDecay * BOD * 02/(02+k02) # BOD decay</pre>
   Reaeration <- k * (sat02-02)
                                            # air-water exchange
  # Time-derivatives: dC/dt = production - consumption [mol/m3/d]
            <- Reaeration - Decay</pre>
   dBODdt
              <- inputBOD - Decay</pre>
  # return time-derivatives and ordinary variables as a list
   list(c(d02dt, dB0Ddt), # vector with derivatives
                          # (the same order as state variables!)
     Decay = Decay, # other output
      Reaeration = Reaeration)
 })
```

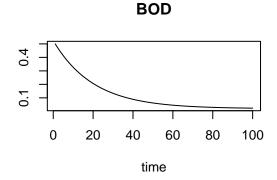
Model solution

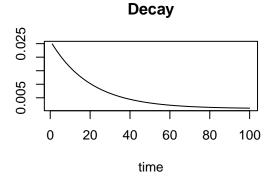
Dynamic solution

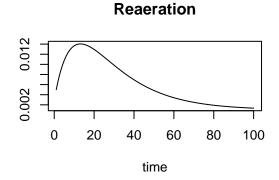
We run the model dynamically over 100 days, using two different values of the reaeration rate constant:

```
require(deSolve) # package with integration methods
# vector of output times
outtimes <- seq(from = 1, to = 100, length.out = 100)
# ode integrates the model
out <- ode(y=yini, parms=pars, func=BODmodel, times=outtimes)
# plot the model output
plot(out)</pre>
```









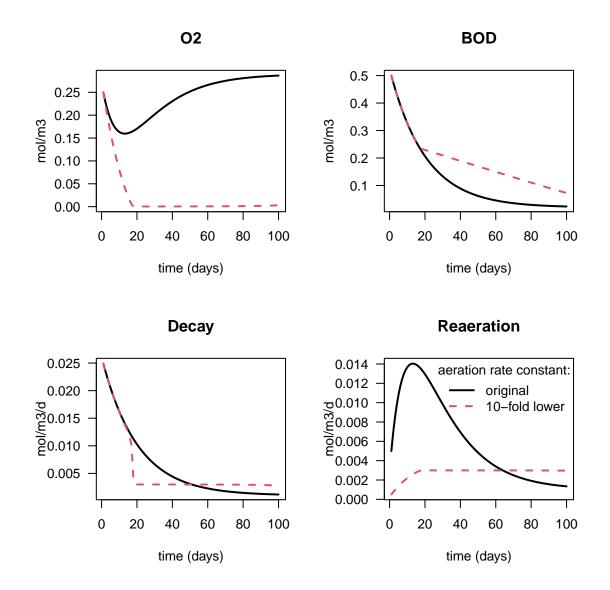
```
# change the value of the reaeration rate constant
pars2 <- pars  # copy the original parameter vector
pars2["k"] <- pars["k"]/10 # 10-fold lower k

# integrate the model with the new parameters
out2 <- ode(y = yini, parms = pars2, func = BODmodel, times = outtimes)

# print summary of the solution</pre>
```

We plot both solutions in one graph:

summary(out)



Steady-state solution

We find the steady-state solution:

Sensitivity analysis

We perform a sensitivity analysis to find how the steady-state depends on the reaeration rate constant:

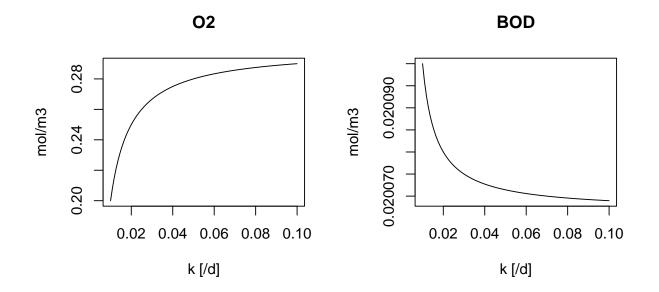
```
k.seq <- seq(from = 0.01, to = 0.1, length.out = 100)
BOD.seq <- vector() # will contain the results

O2.seq <- vector()

for (i in 1:length(k.seq)){
    # parameter values for this run
    p <- pars
    p["k"] <- k.seq[i] # reaeration rate constant based on the sequence
    # steady-state with new parameter values
    std <- steady(y = yini, parms = p, func = BODmodel, positive = TRUE)
    BOD.seq[i] <- std$y["BOD"]
    O2.seq[i] <- std$y["O2"]
}</pre>
```

Finally, we plot the results of the sensitivity analysis:

```
par(mfrow=c(1,2)) # figures in 1 row, 2 columns
plot(k.seq, 02.seq, type="l", xlab="k [/d]", main="02", ylab="mol/m3")
plot(k.seq, BOD.seq, type="l", xlab="k [/d]", main="BOD", ylab="mol/m3")
```



References

R Core Team (2020). R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria. 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, Thomas Petzoldt, R. Woodrow Setzer (2010). Solving Differential Equations in R: Package deSolve. Journal of Statistical Software, 33(9), 1–25. http://www.jstatsoft.org/v33/i09/ DOI: 10.18637/jss.v033.i09