

smb12dMod

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Load all important libraries

```
library(dMod)
# devtools::load_all("~/Promotion/Software/dMod/")
library(stringr)
# library(conveniencefunctions)
# library(magrittr)
#' importFrom magrittr "%>%"
# "%>%" <- magrittr::"%>%"
library(tidyverse)
library(magrittr)
library(SBMLR)
```

Load the example model “curto” from the SBMLR-Package and have a first look at it

```
curto_file <- system.file("models", "curto.xml", package = "SBMLR")
curto <- curto_file %>% readSBML()
summary(curto) %>% head
```

```
## $nSpecies
## [1] 18
##
## $sIDs
## [1] "PRPP" "IMP" "SAMP" "ATP" "SAM" "Ade" "XMP" "GTP" "dATP" "dGTP"
## [11] "RNA" "DNA" "HX" "Xa" "Gua" "UA" "R5P" "Pi"
##
## $S0
## PRPP IMP SAMP ATP SAM Ade
## 5.01742e+00 9.82634e+01 1.98189e-01 2.47535e+03 3.99187e+00 9.84730e-01
## XMP GTP dATP dGTP RNA DNA
## 2.47930e+01 4.10223e+02 6.01413e+00 3.02581e+00 2.86805e+04 5.17934e+03
## HX Xa Gua UA R5P Pi
## 9.51785e+00 5.05941e+00 5.50638e+00 1.00293e+02 1.80000e+01 1.40000e+03
##
## $BC
## PRPP IMP SAMP ATP SAM Ade XMP GTP dATP dGTP RNA DNA
## FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
## HX Xa Gua UA R5P Pi
## FALSE FALSE FALSE FALSE TRUE TRUE
##
## $nStates
## [1] 16
##
## $y0
## PRPP IMP SAMP ATP SAM Ade
## 5.01742e+00 9.82634e+01 1.98189e-01 2.47535e+03 3.99187e+00 9.84730e-01
## XMP GTP dATP dGTP RNA DNA
## 2.47930e+01 4.10223e+02 6.01413e+00 3.02581e+00 2.86805e+04 5.17934e+03
```

```
##           HX           Xa           Gua           UA
## 9.51785e+00 5.05941e+00 5.50638e+00 1.00293e+02
```

Make dMod-Model

Extract the various things needed to make a dMod object out of it. We need:

```
Parameters = c(initial_values, kinetic_parameters)
```

```
mysummary <- curto %>% summary() # SBMLR::summary() calculates some other values like the stoichiometric matrix
initial_values <- mysummary$S0

kinetic_pars <- curto[["reactions"]] %>% lapply("[", "parameters") %>% unname() %>% do.call(c,.) %>% duplicated_names <- names(kinetic_pars)[(names(kinetic_pars) %>% duplicated())]
if(length(duplicated_names)>0) warning(paste0("Parameters", paste(duplicated_names, collapse = ", "), "are duplicated"))
pars <- c(initial_values, kinetic_pars)
```

ODEs

In dMod you can supply the stoichiometric matrix “S” and the rate vector “reactions”. From this you can make an eqnlist-object which you can then use to make the eqnvec, a named vector specifying the right hand side of the ODE.

```
xdot = f(x,pars) = S %*% reactions
```

```
S <- mysummary[["incid"]] %>% # get stoichiometric matrix
t %>% # transpose
structure(.,dimnames = list(mysummary[["rIDs"]], colnames(.))) %>% # set col- and rownames
{.[.==0] <- NA # replace zeroes by NA
. # return the modified matrix
}

reactions <- data.frame(Description = mysummary[["rIDs"]], Rate = mysummary[["rLaws"]], S) # get reactions
f <- reactions %>%
as.eqnlist(volumes = curto$species %>% sapply(. %>% {.[["compartment"]]})) %>% # turn into eqnlist
as.eqnvec() # turn into eqnvec
```

Compile the model

Call the function odemodel() does two things: sensitivity equations and C-code ## Sensitivity equations dMod relies heavily on sensitivity equations for gradient computation. Unlike in finite differences, the derivatives are expressed as ODEs themselves:

$$d/dt(dx_i/dp_j) = df_i/dx \%*\% dx/dp_j + df_i/dp_j$$

However, this augments the ODE system and the sensitivity equations are calculated symbolically. This can take extremely long for large equation systems:

N states, M parameters $f = N$ equations $dx/dp = N \times M$ equations $dx/dx_0 = N$ equations ($dx_i/dx_0_j = 0$)

C-Code

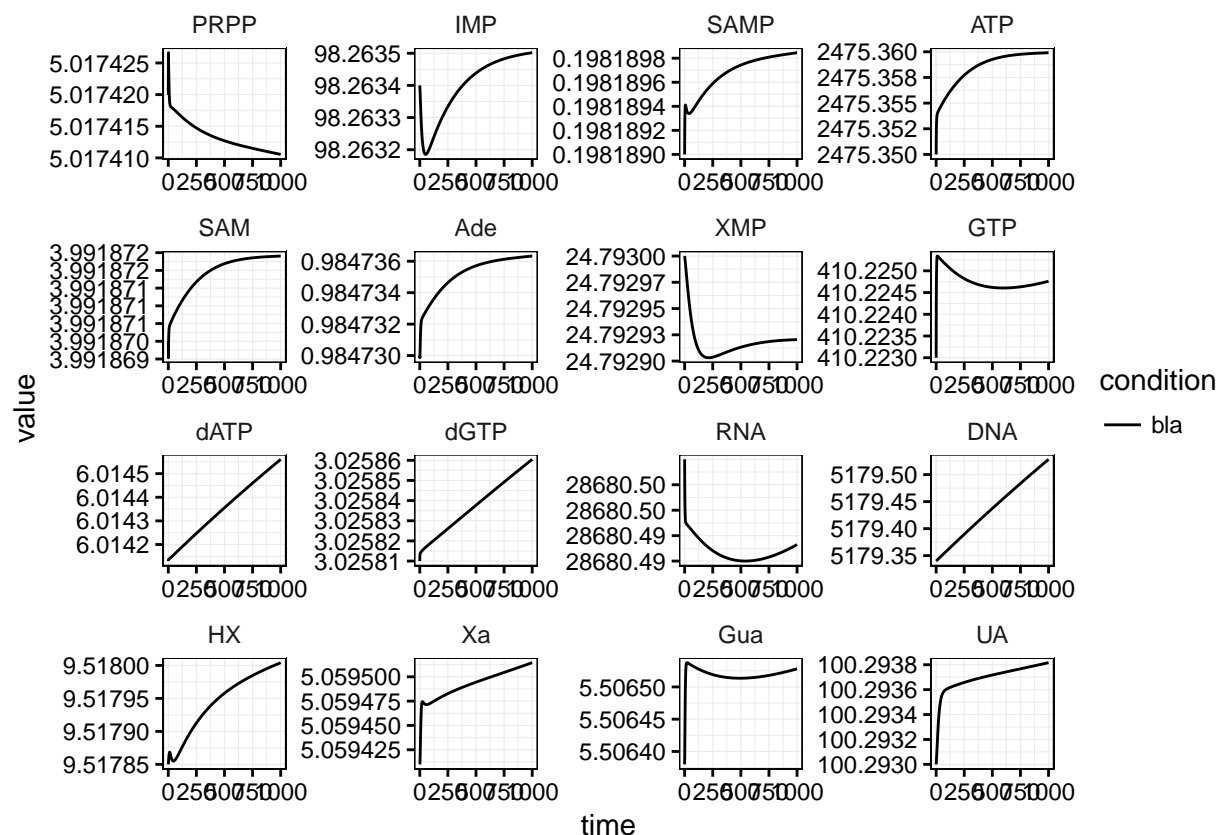
The symbolic equations are then written into C-code and compiled. This can also take quite long for large systems.

These two things greatly restrict the number of states and parameters that can be used with our methods.

```
myodemodel <- odemodel(f, modelname = "sbml2r") # calculate sensitivity equations and compile C-code
# save(myodemodel, file = "sbml2r.rda")
# load("sbml2r.rda")
```

Compute predictions

```
x <- Xs(myodemodel, condition = "bla") # make prediction function
mypred <- x(seq(0,1000,1), pars, deriv = F) #compute the prediction. If deriv == T, also the derivative.
plotPrediction(mypred)
```



simulate data

I did this yesterday in the train, too lazy to delete it, but also too lazy to complete it. For all the estimation procedure it's better to read the vignette and the paper.

```

pars[names(initial_values)] <- initial_values*runif(length(initial_values), 0.5,2) # change some pars to

mydata <- x(times = c(1,5,10,50,1000), pars, deriv = F) %>%
  lapply(function(pred_cond) {
    pred_cond[,-1] <- pred_cond[,-1]*exp(rnorm(length(pred_cond[,-1]), sd = 0.05))
    return(pred_cond)
  }) %>% wide2long() %>%
  mutate(sigma = 0.05*value) %>%
  as.datalist()

plotCombined(x(times = seq(1,2000, 10), pars, deriv = F), data = mydata)

```

