Torsten: A PKPD Model Library for Stan

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Introduction

Metrum Research Group, LLC has developed a prototype PKPD model library for Stan. The current version includes:

- One compartment model with first order absortpion
- Two compartment model with elimination from and first order absorption into the central compartment
- General compartmental model described by a system of first order ODEs

The models and data format are based on NONMEM^{®1}/NMTRAN/PREDPP conventions including:

- Recursive calculation of model predictions, which permits piecewise constant covariate values
- Bolus or constant rate inputs into any compartment
- Handling of single dose, multiple doses and steady-state dosing histories
- Implemented NMTRAN data items: TIME, EVID, CMT, AMT, RATE, ADDL, II, SS

The library provides Stan language functions that calculate amounts in each compartment.

Implementation Details

- Stan (http://mc-stan.org/)
- An extensible object-oriented programing approach is used to facilitate development of additional models
- The core functions are programmed in C++
- One and two compartment models: analytical calculations programmed in C++
- General compartment models with numerical solutions to ODE using built-in ODE integrators in Stan (Runge-Kutta 4th/5th and CVODES). NOTE: currently, general compartment models do not handle steady state approximations

Warning: The current version of Torsten is a *prototype*. It is being released for review and comment. It is still being developed and should not be used for critical applications without further testing or cross-checking by comparison with other methods.

Development

This project is open-source and forsters collaboration. We have recieved help and advice from researchers at various institutions, notably the Stan Group. See https://groups.google.com/forum/#!forum/stan-dev..

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Install Stan and Torsten

Examples and installation files are available on GitHub: https://github.com/charlesm93/example-models/tree/feature/issue-70-PKPDexamples-torsten/PKPD/torsten

Torsten has been tested with cmdStan-2.10, the command line interface of Stan. I recommend using cmdStan-dev and downloading Torsten from gitHub. The easiest way to do so is to run the bash file setupTorsten.sh posted on example-models (see link above). You can also do this step by step on terminal.

Download cmdstan-dev:

```
git clone https://github.com/stan-dev/cmdstan.git
```

Add the version of Stan that exposes the Torsten functions to the language, and switch to the branch feature/issue-1953-exposing-torsten-pharmacometrics:

```
cd cmdstan
git clone https://github.com/charlesm93/stan.git
cd stan
git checkout feature/issue-1953-exposing-torsten-pharmacometrics
```

Next, install the version of stan-math that contains the Torsten library, and switch to the branch feature/issue-314-torsten-pharmacometrics:

```
cd lib
rm -r stan_math
git clone https://github.com/charlesm93/math.git
cd math
git checkout feature/issue-314-torsten-pharmacometrics
cd ..
mv math stan_math
```

Using the Torsten library

Linear One and Two compartment model

The one and two compartment model functions have the form:

```
<modelName>(theta, time, amt, rate, ii, evid, cmt, addl, ss)
```

where time, amt, rate, ii are arrays of real and evid, cmt, addl, and ss arrays of integers. All arrays have the same length, which corresponds to the number of events. theta, also known as the *parameter matrix*, is an array of vectors. If the parameters are constant for all events, theta should be an array of length 1, else it should have the same length as time, which amounts to having one vector of parameters per event. In the latter case, the *ith* row contains a vector of parameters for the time interval (time[i-1], time[i]).

The options for *modelName* are:

- PKModelOneCpt
- PKModelTwoCpt

which respectively correspond to the one and two compartment model with first order absorption (see figure 1).

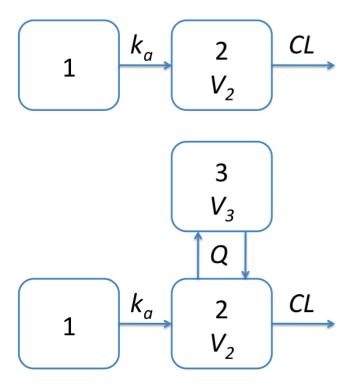


Figure 1: One and Two Compartment Model with First Order Absorption

General compartment model

The general compartment model functions have the form:

where ODE_System is a System of first order ODEs defined in the function block of Stan (see section 19.2 of the Stan reference manual for instructions on how to code such a system) and nCmt the number of compartments in the model. rel_tol, abs_tol, and max_step are the tuning parameters of the ODE integrator, respectively the relative tolerance, the absolute tolerance, and the maximum number of steps.

The options for modelName are:

- \bullet generalCptModel_bdf
- generalCptModel rk45

These respectively call the built-in CVODES and Runge-Kutta 4th/5th ODE integrators.

Example 1: Two Compartment Model

To illustrate use of the model library, let's start with a fairly simple example. We modeled drug absorption in a single patient and simulated plasma drug concentrations:

• Single Dose: 5000 mg

- PK: plasma concentrations of parent drug
- PK measured at 0, 0.25, 0.50, 0.75, 1, 1.25, 1.50, 1.75, 2, 4, 6, 8, 10, 12, 14, 16, 18, and 20 hours after

The plasma drug concentration (c) is simulated according to the following equations:

$$log(c) \sim N(log(\hat{c}), \sigma)$$

$$\hat{c} = f_{2cpt}(t, CL, Q, V_1, V_2, ka)$$

$$(CL, Q, V_1, V_2, ka) = (5 \text{ L/h}, 8 \text{ L/h}, 20 \text{ L}, 70 \text{ L}, 1.2 \text{ h}^{-1}), \sigma = 0.05$$

The simulations are generated using the R package *mrgsolve*. Four MCMC chains of 2000 iterations were simulated. The first 1000 iteration of each chain were discarded. Thus 1000 MCMC samples were used for the subsequent analyses. To fit this model, we used TwoCptModelExample.stan and the data and initial estimates simulated by TwoCptModelSimulation.R, respectively saved in TwoCptModelExample.data.R and TwoCptModelExample.init.R.

This example uses the function PKModelTwoCpt, which analytically computes the solution for a two compartment model. Alternatively, GenTwoCptModelExample.stan calls genCptModel_bdf and makes a numerical approximation of the solution. Since we are looking at a relatively simple model, the fitting only takes a few seconds in both cases.

Additionally, we coded some priors in the model block:

```
CL ~ lognormal(log(10), 0.25);

Q ~ lognormal(log(15), 0.5);

V1 ~ lognormal(log(35), 0.25);

V2 ~ lognormal(log(105), 0.5);

ka ~ lognormal(log(2.5), 1);

sigma ~ cauchy(0, 1);
```

These priors are not essential to fit this particular model, but we included them to make the example more complete.

Result

The MCMC history plots (Figure 2) and the posterior marginal densities of the model parameters (figure 3) suggest that the 4 chains have converged to a common distribution for all of the key model parameters. The predicted data agrees with the simulated data (not surprisingly so since the fitted model is identical to the one used to simulate data; figure 4). Similarly the parameter estimates summarized in table 1 are consistent with the values used for the simulation.

Table 1: Summary of the MCMC simulations of the marginal posterior distributions of the model parameters

	mean	se_mean	sd	2.5%	25%	50%	75%	97.5%	n_eff	Rhat
CL	5.21	0.01	0.25	4.65	5.06	5.22	5.37	5.65	1460.46	1.00
Q	7.93	0.01	0.48	6.97	7.62	7.93	8.23	8.89	1737.66	1.00
V1	20.13	0.06	2.20	15.99	18.67	20.13	21.55	24.55	1226.46	1.00
V2	63.72	0.19	6.98	51.92	59.20	62.96	67.39	79.63	1337.96	1.00
ka	1.24	0.00	0.17	0.95	1.13	1.23	1.34	1.58	1299.38	1.00

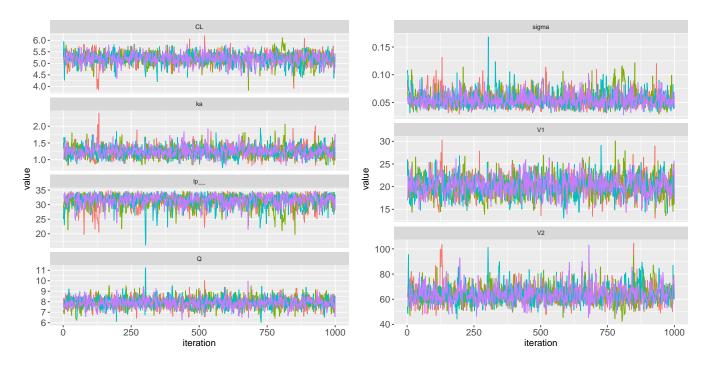


Figure 2: MCMC history plots for the parameters of a two compartment model with first order absorption (each color corresponds to a different chain)

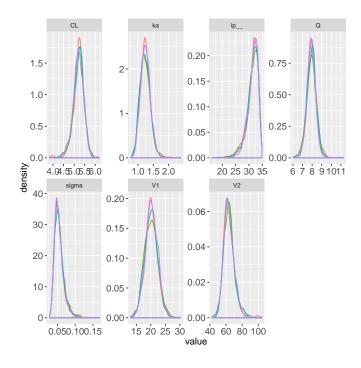


Figure 3: Posterior Marginal Densities of the Model Parameters (each color corresponds to a different chain)

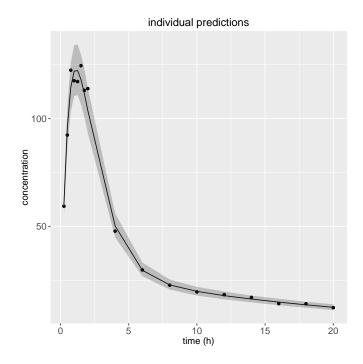


Figure 4: Predicted (posterior median and 90 % credible intervals) and observed plasma drug concentrations

Example 2: Population PK Model

The above example can be extended to a population model, in which we administer the drug to multiple patients, and where the PK parameters vary from one individual to the other:

$$\begin{split} log(c_i) \sim N(log(\hat{c_i}), \sigma) \\ \hat{c_i} &= f_{2cpt}(t_i, CL_i, Q_i, V_{1i}, V_{2i}, ka_i) \\ log(CL_i, Q_i, V_{1i}, V_{2i}, k_i) \sim N(log(\hat{CL}, \hat{Q}, \hat{V_1}, \hat{V_2}, \hat{ka}, \Omega) \\ (\hat{CL}, \hat{Q}, \hat{V_1}, \hat{V_2}, \hat{ka}) &= (5 \text{ L/h}, 8 \text{ L/h}, 20 \text{ L}, 70 \text{ L}, 1.2 \text{ h}^{-1}), \sigma = 0.05 \\ \Omega &= \begin{pmatrix} 0.05^2 & 0 & 0 & 0 \\ 0 & 0.05^2 & 0 & 0 \\ 0 & 0 & 0.05^2 & 0 \\ 0 & 0 & 0.05^2 \end{pmatrix} \end{split}$$

This model can be fitted using TwoCptModelPopulation.stan and the corresponding .data.R and .init.R files.

Example 3: Population PK of ME-2

This example was presented at the *Getting Started with Bayesian PKPD Modeling Using Stan* workshop at the PAGE 2016 conference. Relevant pages from the workshop slides that describe the model (see MetrumBayesianStanPAGE_example3.pdf) are available on gitHub. This model is more elaborate, and has the merit of demonstrating how Stan and Torsten can be used to tackle more difficult problems. Just a heads up: it took me 7.5 hours to run the model with 2000 MCMC iterations, but the result was rather good (MCMC chains converge, and data predicted by fitted model agrees with real data).

Further Examples

TwoCptModelSimulation.R and TwoCptModelSimulationPopulation.R can be used as templates to simulate PKPD data. Users are welcomed to experiment with different parameter values and different events (such as dosing at regular intervals or steady state approximation). Examples on how to simulate PKPD data sets with <code>mrgsolve</code> can be found on https://github.com/metrumresearchgroup/mrgsolve/wiki/gallery. Changing the model itself is also a possibility, though it may result in model misspecification.