

Some numerical issues in PBPK modelling for efficient pharmacotherapy

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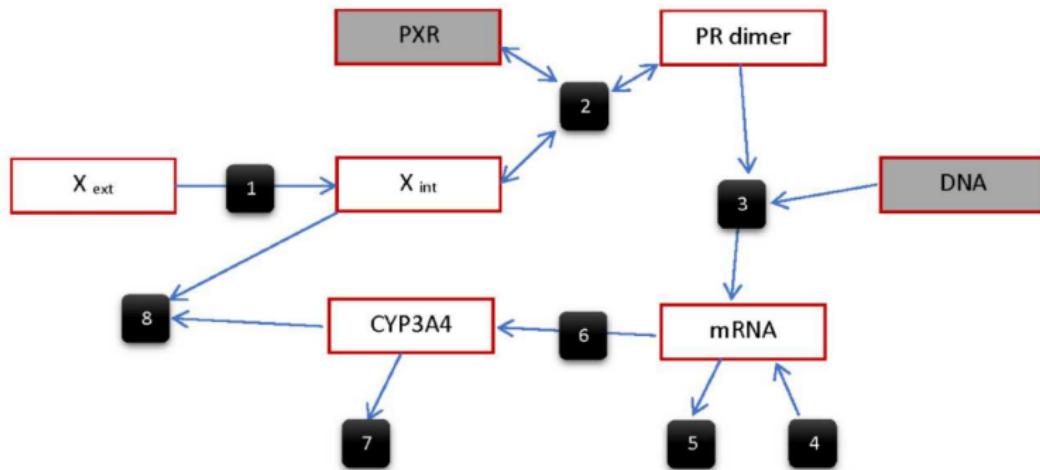
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Jurjen

From in vitro to in vivo model

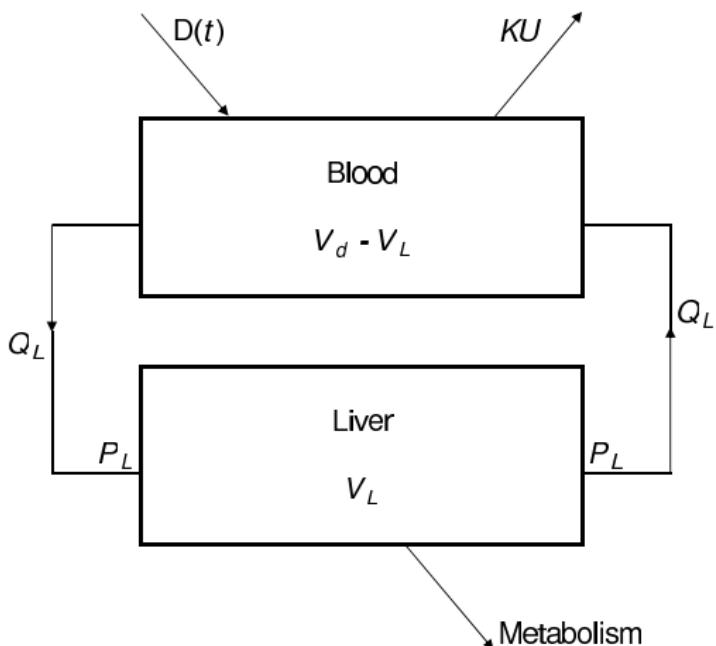
A physiologically based pharmacokinetic (PBPK) model for the action of rifampicin:



$$X_{\text{ext}}(t) \rightarrow X_B(t), \quad X_{\text{int}}(t) \rightarrow X_L(t)$$

From in vitro to in vivo model

Schematic of the two compartment pharmacokinetic model used to determine distribution of rifampicin in vivo:



In vivo model

$$\frac{dX_B(t)}{dt} = \frac{D(t) + Q_L \cdot \frac{X_L(t)}{P_L} - Q_L \cdot X_B(t) - KU \cdot X_B(t)}{V_d - V_L}$$

$$\begin{aligned}\frac{dX_L(t)}{dt} &= \frac{Q_L}{V_L} \cdot \left(X_B(t) - \frac{X_L(t)}{P_L} \right) - k_{assoc} \cdot X_L(t) \cdot (PXR_{tot} - PR(t)) \\ &\quad - k_{met} \cdot CYP3A4(t) \cdot X_L(t) + k_{dis} \cdot PR(t)\end{aligned}$$

$$\frac{dPR(t)}{dt} = k_{assoc} \cdot X_L(t) \cdot (PXR_{tot} - PR(t)) - k_{dis} \cdot PR(t)$$

$$\frac{dmRNA(t)}{dt} = k_3 \cdot PR(t) - mRNAdeg \cdot mRNA(t) + mRNAbck$$

$$\frac{dCYP3A4(t)}{dt} = k_4 \cdot mRNA(t) - cypdeg \cdot CYP3A4(t)$$

In vivo model

Functions dependent on time t [min]:

- $X_B(t)$ [μM] – concentration of rifampicin in the blood
- $X_L(t)$ [μM] – concentration of rifampicin in the liver
- $\text{PR}(t)$ [μM] – concentration of PXR-RXR heterodimer
- $\text{mRNA}(t)$ [μM] – rate of mRNA production;
fold induction $\text{mRNA}_{\text{fold}}(t) = \text{mRNA}(t)/\text{mRNA}(0)$
- $\text{CYP3A4}(t)$ [μM] – change in CYP3A4 protein levels;
fold induction $\text{CYP3A4}_{\text{fold}}(t) = \text{CYP3A4}(t)/\text{CYP3A4}(0)$
- $D(t)$ [mg/min] – function describing intravenous rifampicin dosing

In vivo model

Known parameters obtained from literature:

- $Q_L = 1.3 \text{ [L/min]}$ – blood flow rate
- $V_L = 1.815 \text{ [L]}$ – volume of the liver
- $V_d = 15.7 \text{ [L]}$ – overall volume of distribution
- $K_d = 8.36 \text{ [\mu M]}$ – binding affinity constant
- $k_3 = 39.3 \text{ [min}^{-1}]$ – transcription rate constant
- $k_4 = 2.5 \text{ [min}^{-1}]$ – translation rate constant
- $mRNAddeg = 0.04 \text{ [min}^{-1}]$ – degradation coefficient for mRNA
- $cypdeg = 2.7 \cdot 10^{-4} \text{ [min}^{-1}]$ – CYP3A4 degradation rate parameter
- $k_{assoc} = \frac{k_{dis}}{K_d} \text{ [(\mu M min)}^{-1}]$ – association rate constant for the formation of the PXR-RXR heterodimer

In vivo model

Unknown parameters to be optimized:

- PXR_{tot} [μM] – total concentration of PXR in the system
- k_{dis} [min^{-1}] – first order dissociation constant
- $mRNAbck$ [$\mu\text{M}/\text{min}$] – background mRNA production rate
- k_{met} [$(\mu\text{M min})^{-1}$] – second order metabolic constant
- KU [L/min] – urinary excretion rate
- P_L [unitless] – liver/blood partition coefficient

In vivo model

Initial conditions:

- $D(t)$ – a dose of 600 mg of rifampicin administered intravenously over 60 minutes, repeatedly every day:
$$D(t) = D_t \text{ for } t \in [0, 60] \cup [1440, 1500] \cup \dots \text{ and } D(t) = 0 \text{ otherwise, where } D_t = 10$$
- $X_B(0) = D_t$ – initial concentration of rifampicin in the blood is equal to the dose of rifampicin being administered
- $X_L(0) = 0$ – concentration within the liver has an initial value of zero
- $PR(0) = 0$ – there is no concentration of PXR-RXR heterodimer initially
- $mRNA(0) = mRNA_{ss} = \frac{mRNAbck}{mRNAdeg}$ – initial value is assumed to be the steady state value
- $CYP3A4(0) = CYP3A4_{ss} = \frac{k_4}{cypdeg} \cdot mRNA_{ss}$ – initial value is assumed to be the steady state value

Parameter estimation

Collecting of experimental data from donors and subsequent estimation of missing parameters: the estimates are obtained through curve fitting, i.e. minimization of a sum of squares based on comparing observed and computed concentrations:

$$J(q_*) = \min_q J(q), \quad \text{where} \quad J(q) = \sum_{i=1}^N (f(t_i, q) - R_i)^2$$

and

- q_* is the optimal parameter set that minimizes the ordinary least squares (OLS) cost function J
- R represents the rifampicin levels in blood (values $X_B(t)$) presented in Loos, see the next slide
- $f(t_i, q)$ is the time course model result (solution of ODEs)
- N is a number of observations

Experimental data for $X_B(t)$

Time values of rifampicin concentration in blood ($X_B(t)$),
repeated dosing 600 mg during 60 minutes every day

Time	0	20	40	60	80	100
day 1	10.0	26.1	33.7	38.3	27.8	23.0
day 8	8.80	23.7	27.8	33.7	23.7	20.2
day 22	6.81	19.6	24.5	30.0	19.6	14.7

Time	120	240	360	480	600	720
day 1	21.5	10.7	5.62	2.45	*	0.64
day 8	19.6	9.38	4.35	1.38	0.45	0.12
day 22	13.8	7.26	1.67	0.32	*	*

* indicates missing data



Loos U., Musch E., Jensen J.C., Mikus G., Schwabe H.K., Eichelbaum M.
Pharmacokinetics of oral and intravenous rifampicin during chronic administration.

Klin. Wochenschr. 63 (1985), 1205–1211.

Computation of the ordinary least squares cost function J for optimal parameters: two-level problem:

- outer problem – minimization of J subject to q

software used: UFO (Ladislav Lukšan et al.)

<http://www.cs.cas.cz/luksan/ufo.html>

method used: line-search variable metric method for direction determination

- inner problem – solution of a system of ODEs ($f(t_i, q)$)

software used: ODEPACK (Alan Hindmarsh)

<https://people.sc.fsu.edu/~jburkardt/>

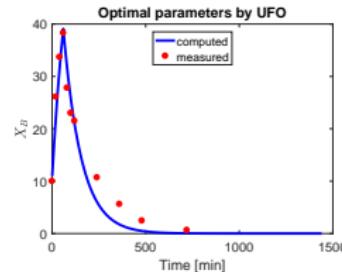
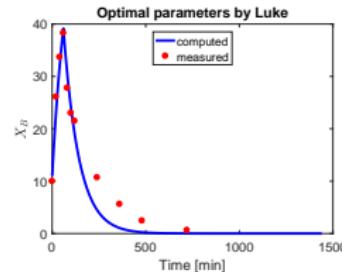
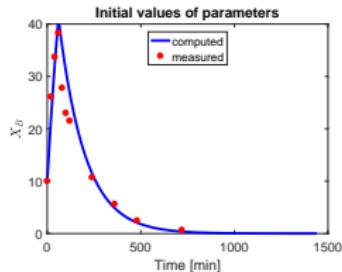
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method used: method based on backward differentiation formulas (stiff), chord iteration with a user-supplied full Jacobian

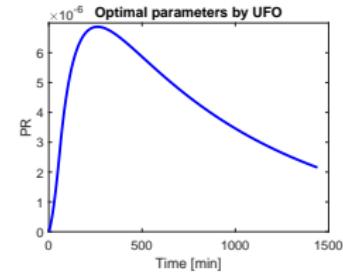
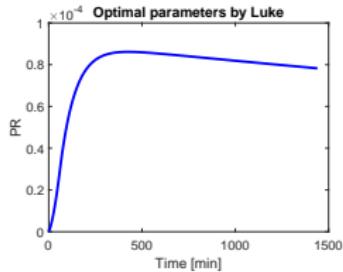
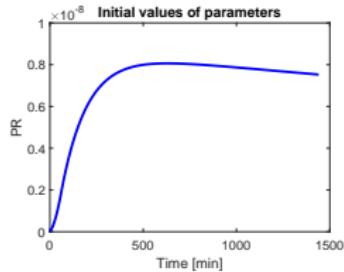
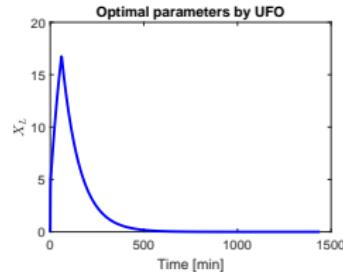
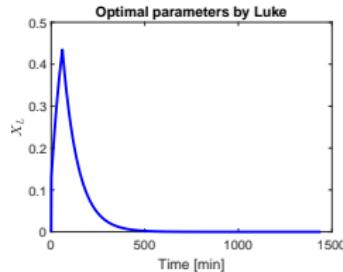
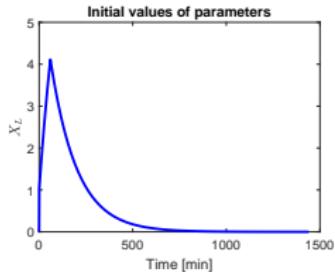
Optimal parameters for 1 day

The ordinary least squares cost function J for optimal parameters computed by Luke and UFO using the data for 1 day only.

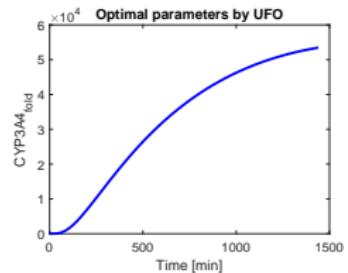
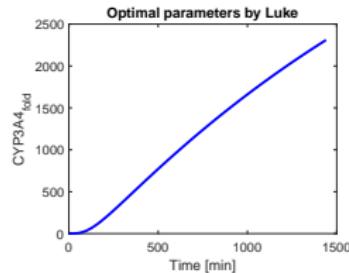
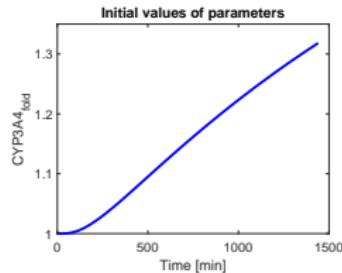
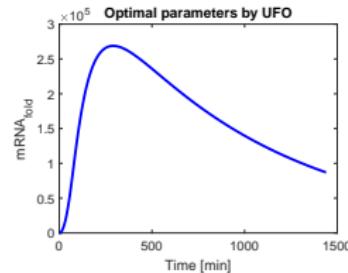
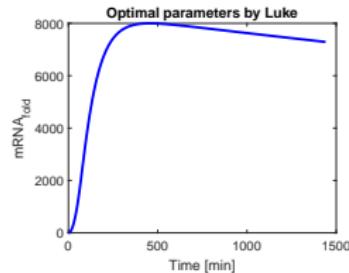
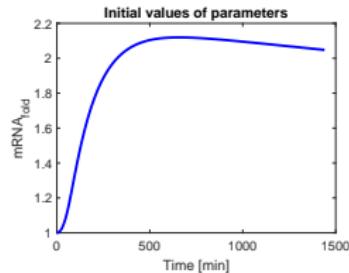
parameter	initial value	optimal value by Luke	optimal value by UFO
PXR_{tot}	9.47D-7	1.35D-1	3.43D-5
k_{dis}	1.03D-4	1.03D-4	1.08D-3
$mRNAbck$	2.83D-7	4.23D-7	1.00D-8
k_{met}	2.47D-5	2.47D-5	1.00D-8
KU	1.00D-1	1.67D-1	1.52D-1
P_L	1.00D-1	1.11D-2	4.31D-1
J	185.2735	97.3111	87.0709



Functions $X_L(t)$, $PR(t)$ – 1 day



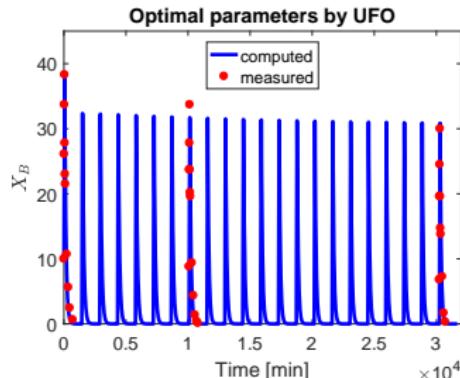
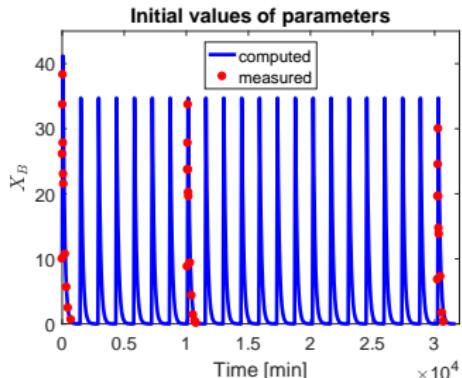
Functions mRNA(t), CYP3A4(t) – 1 day



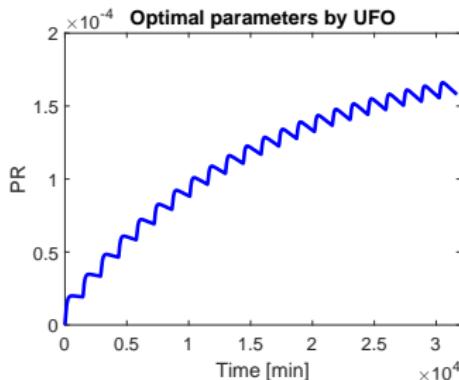
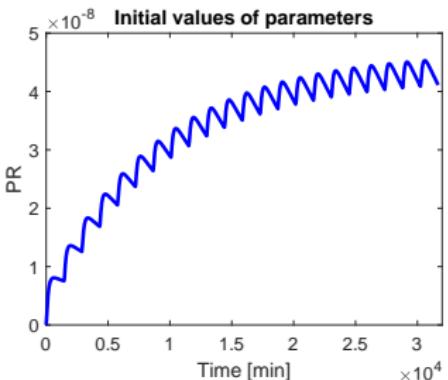
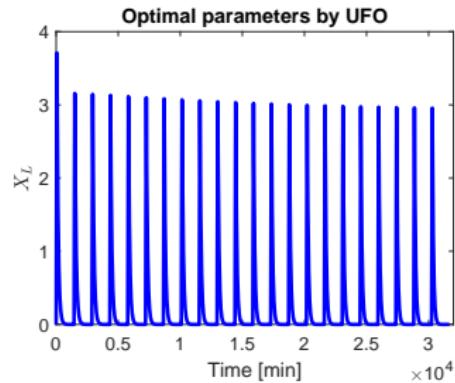
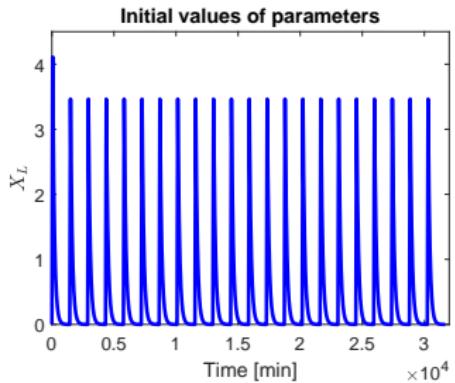
Optimal parameters for 22 days

The ordinary least squares cost function J for optimal parameters computed by UFO using the data for 22 days.

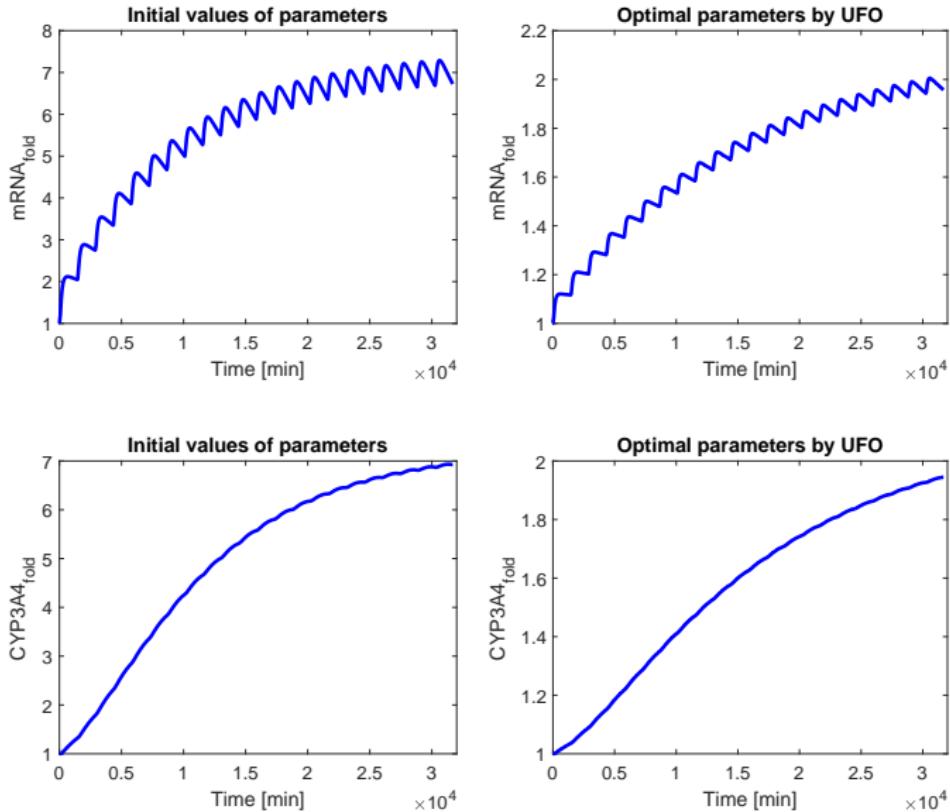
parameter	initial value	optimal value by UFO
PXR_{tot}	9.47D-7	6.7385D-3
k_{dis}	1.03D-4	4.7560D-5
$mRNAbck$	2.83D-7	6.4921D-3
k_{met}	2.47D-5	1.0790D-4
KU	1.00D-1	1.0639D-1
P_L	1.00D-1	1.0006D-1
J	911.3656	510.3724



Functions $X_L(t)$, $PR(t)$ – 22 days



Functions mRNA(t), CYP3A4(t) – 22 days



Discussion

Discussion:

- Important: Need more experimental data! (There exist a lot of different local minima.)
- What parameters most influence the results? (The question of sensitivity analysis – preliminary done by Luke.)
- The obtained optimal parameters and corresponding functions must be physically correct – adding constraints or bounds to the optimization problem.
- How to choose local parameters in UFO and ODEPACK (method, initial approximation, time step in solving ODEs, step-length for direction determination,...)? They influence the solution obtained...
- Better software for solving stiff ODEs?
- Robust optimization method (quasi-Newton, Gauss-Newton)?
- ...

Štěpán