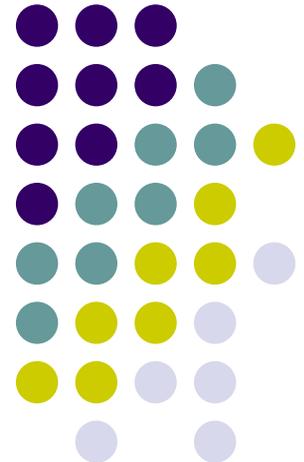


A real time optimal design for model discrimination and parameter estimation for itraconazole

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Introduction



- Approximately 10% of patients who have cystic fibrosis (CF) develop allergic bronchopulmonary aspergillosis – an allergic reaction to antigens on the surface of the pathogen
- Itraconazole is one of the drugs of choice for the treatment of Aspergillosis spp infections.
- At present very little is known about the pharmacokinetics of itraconazole in patients with CF – but limited information support the use of higher doses than other patient groups
- Itraconazole has an active metabolite (hydroxyitraconazole)

Capsule or solution?



- Both a capsule and solution are available – standard treatment involves the capsule only.
- Limited data suggest that the solution provides a considerably greater rate and extent of absorption than the capsule
- Also... the solution does not appear to be affected by the presence of food (cf the capsule)
- The most appropriate dosing of the solution in patients with CF has not been determined...

Aim



- To choose an experimental design (set of elementary designs) for model selection and efficient estimation of relevant parameters for itraconazole and hydroxyitraconazole



Prior information

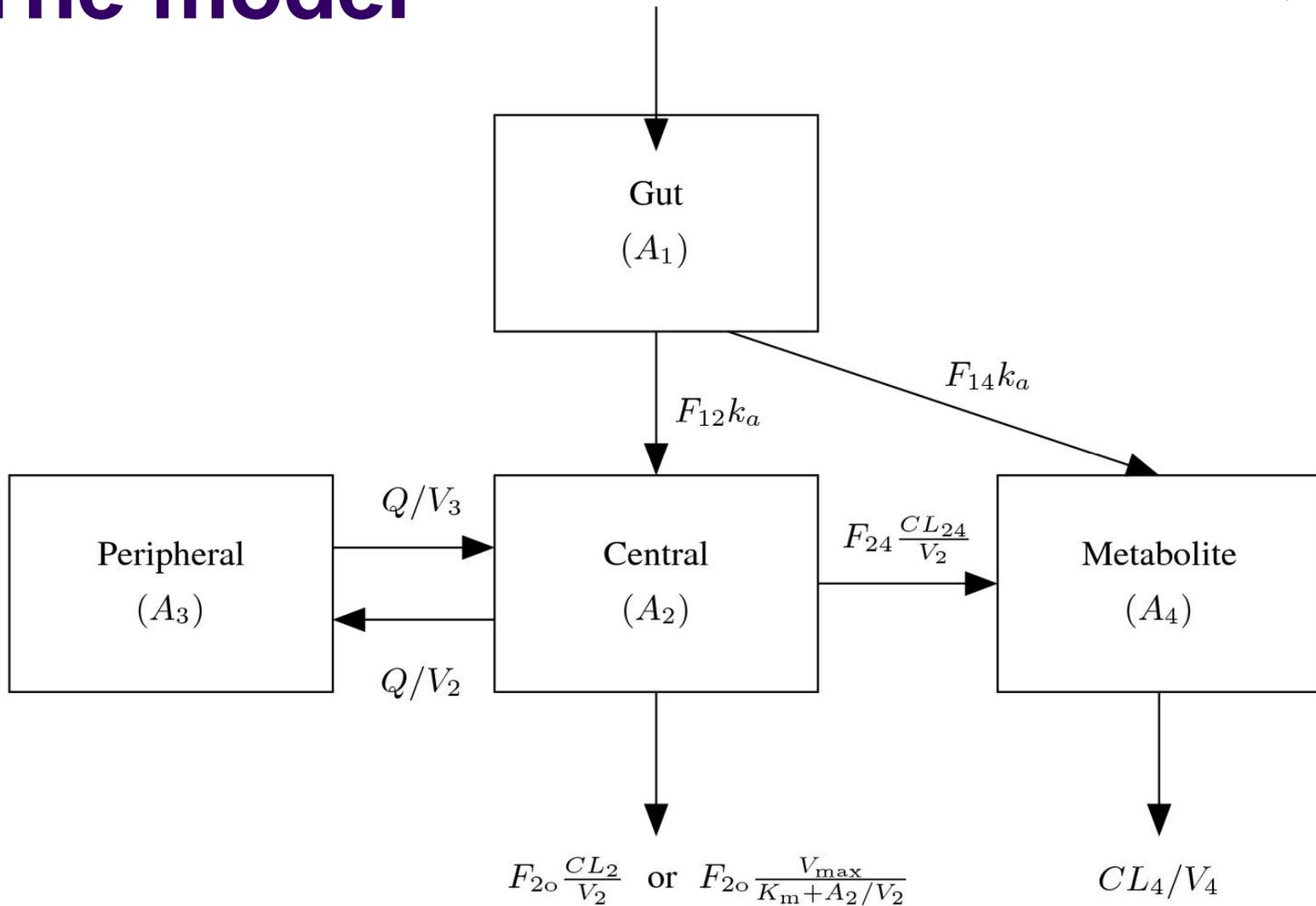
- Data were available from 4 studies of itraconazole and 1 described both itraconazole and hydroxyitraconazole
 - Koks et al. Ther Drug Monit 2003;25:229-233
 - Barone et al. Antimicrob Agents Chemother. 1993;778-784
 - FDA web site (x 2 documents)
- The model for itraconazole and hydroxyitraconazole were described by 2 linked compartmental models
- Data from Barone et al., suggested that the elimination of itraconazole from the central compartment was non-linear (this has been supported in other studies)



Clinical application

- A maximum of 30 patients are to be recruited
- A maximum of 4 samples per patient per occasion
- The study would be a single dose cross-over with capsule 200 mg followed by solution 200 mg
- A minimum washout period of 72 hours would be observed

The model



The ODE's



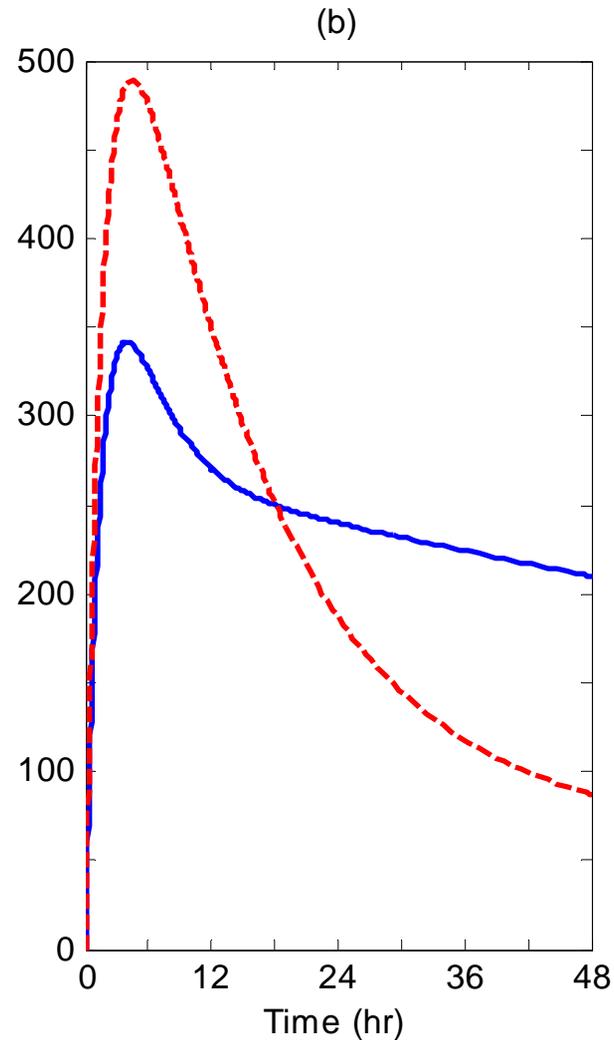
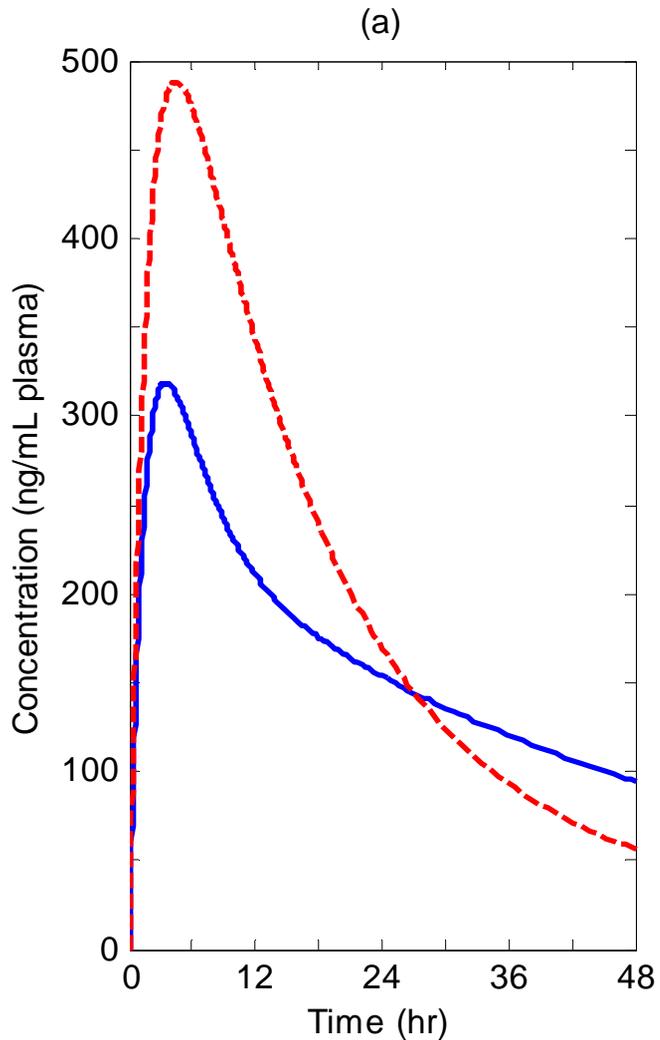
$$\frac{dA_1(t)}{dt} = -F_{12}k_a A_1(t) - F_{14}k_a A_1(t) \quad (1)$$

$$\begin{aligned} \frac{dA_2(t)}{dt} = & F_{12}k_a A_1(t) + \frac{Q}{V_3} A_3(t) - \frac{Q}{V_2} A_2(t) - F_{24} \frac{CL_{24}}{V_2} A_2(t) \\ & + \begin{cases} -F_{20} \frac{CL_2}{V_2} A_2(t) & \text{Model 1} \\ -F_{20} \frac{V_{\max}}{K_m + A_2(t)/V_2} A_2(t) & \text{Model 2} \end{cases} \end{aligned} \quad (2)$$

$$\frac{dA_3(t)}{dt} = \frac{Q}{V_2} A_2(t) - \frac{Q}{V_3} A_3(t) \quad (3)$$

$$\frac{dA_4(t)}{dt} = F_{14}k_a A_1(t) + F_{24} \frac{CL_{24}}{V_2} A_2(t) - \frac{CL_4}{V_4} A_4(t) \quad (4)$$

PK profiles



— itraconazole

- - - hydroxy-
itraconazole



Design – the information matrix

- The information matrix was the same as that provided by Retout et al and was assumed to take the form:

$$M_F(\boldsymbol{\psi}, \boldsymbol{\xi}_q) \approx \frac{1}{2} \begin{bmatrix} A(E, V) & C(E, V) \\ C^T(E, V) & B(E, V) \end{bmatrix}$$

where $\boldsymbol{\psi}$ is a vector of population parameters (fixed and variance of the random effects), $\boldsymbol{\xi}_q$ is the q^{th} elementary design and $C(E, V)$ is a matrix of zeros

Retout et al., Stat Med 2002;21:2623-2639

Design – multiple responses (parent & metabolite)



- Multiple responses (that conform to the same likelihood) for fixed effects model has been described by summing the information matrices over the responses (Draper and Hunter):

$$M_F(\boldsymbol{\theta}, \boldsymbol{\xi}) = \sum_{a=1}^r \sum_{b=1}^r \sigma^{ab} F_a^T F_b$$

- Where F is a vector of first partial derivatives of the response to the fixed effects parameters ($\boldsymbol{\theta}$)
- We extended this form to include fixed and random effects and simplified its computation by using only a single common residual variance

Draper and Hunter. Biometrika 1966;53:525-533

Design – competing models (linear, non-linear, capsule, solution)



- Estimation
 - A compound criterion has been described for estimating parameters over a number of competing models (Atkinson & Cox)
$$P(\Psi^1, \Psi^2, \Xi) = \left| M_F^1(\Psi^1, \Xi) \right|^{1/p1} \left| M_F^2(\Psi^2, \Xi) \right|^{1/p2}$$
- Discrimination
 - T-optimality has been described by Atkinson & Fedorov – but is highly computationally intensive and difficult to implement
 - Waterhouse et al showed that the compound design worked well for discrimination for some models – this was used here

Atkinson & Cox. J R Stat Soc Ser B 1974;36:321-348

Atkinson & Fedorov. Biometrika 1975; 62:57-70

Waterhouse et al. Research Report 106: Centre for Statistics, UQ

Execution of optimization



- The design was simplified by assuming that:
 - capsules and solution only differed in terms of the rate and extent of absorption,
 - the non-linear and linear models differed only in terms of elimination from the central compartment and all other parameters were common.
- Imposed design constraints meant that CL_{24} , V_3 and Q were not able to be estimated (and were fixed)
- The design was optimized using `POPT.m` with a simulated annealing algorithm (taking 7 days)

Design



Group (q)	N_q	Capsule		Solution	
		Elementary design ξ_q^c (hrs:mins)	Sampling window (hrs)	Elementary design ξ_q^s (hrs:mins)	Sampling window (hrs)
1	10	1:14	0.1 → 3.0	0:17	0.1 → 1.0
		8:56	7.0 → 10.0	3:55	3.0 → 3.5
		25:49	24.0 → 27.0	3:56	3.5 → 4.0
		51:45	50.0 → 53.0	3:56	4.0 → 4.5
2	10	6:13	5.0 → 8.0	0:18	0.1 → 1.0
		9:50	8.0 → 11.0	4:06	3.0 → 4.0
		29:29	28.0 → 29.5	4:06	4.0 → 5.0
		29:29	29.5 → 31.0	72:00	69.0 → 72.0
3	10	8:08	7.0 → 10.0	0:17	0.1 → 1.0
		28:00	26.5 → 29.5	4:22	3.0 → 6.0
		72:00	69.0 → 70.5	27:08	26.0 → 29.0
		72:00	70.5 → 72.0	72:00	69.0 → 72.0

Design evaluation



- The compound optimal design was 96.4 and 95.9% as efficient as the optimal design for either linear or non-linear model alone
- Simulation - 100 data sets were simulated under the optimal design from each model and both models fitted to the data using NONMEM (FOCEI)
 - In 100% of data sets the NONMEM run converged successfully
 - In 74% of the data sets simulated under the linear model, the correct model was preferred
 - In 100% of the data sets simulated under the nonlinear model, the correct model was preferred
 - The standard error estimates from `POPT.m` were close to the empirical standard deviation of parameter estimates from the NONMEM estimation runs, except for k_a and ω_{k_a} which differed by 2-fold



Discussion

- Optimal design procedures can be undertaken under the constraints imposed by time and clinical considerations
- Designs may be optimized over relatively complex model features including:
 - Multiple competing models
 - Models that can be defined only as ODEs
 - Models that involve multiple response types
- The compound design criterion appears to perform well for model discrimination and is relatively efficient for estimation
- Joint design windows can be computed efficiently – to produce clinically meaningful designs
- Optimality is a real alternative to simulation for designing popPK and popPKPD studies

We await the results of the actual study...

