

Drug-disease modeling approach to describe the relationship between long- acting methylphenidate exposure and clinical response

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Disclosure

Roberto Gomeni, PhD was consultant for: Ironshore Pharmaceuticals & Development, Inc., Sunovion Pharmaceuticals Inc., Supernus Pharmaceuticals, Inc., Teva Branded Pharmaceutical Products R&D, Inc., Biomedical Science Institutes, Singapore, Nanomi BV, The Netherlands, Laboratorios Liconsa, S.A., Spain, General Hospital Corporation, Boston, Massachusetts, and UCB Biopharma S.P.R.L.

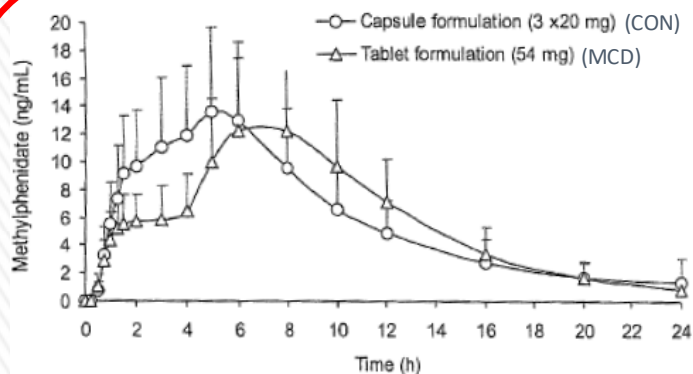
Outline

- Methylphenidate (MPH) is currently used to treat children with attention deficit hyperactivity disorder (ADHD).
- Several modified release formulations characterized by complex in-vivo drug release process have been developed in the attempt to improve the treatment efficacy.
- Model-based approach is recognized as a tool to make drug development more productive and to better support regulatory and therapeutic decisions.
- The objective of this presentation is to develop a model-based framework (i.e., a drug-disease model, and a response surface analysis) to identify the relevant factors affecting performances of MPH treatments and to use this modeling framework to evaluate the ideal properties of a MPH formulation appropriate for maximizing the clinical benefit of a treatment.

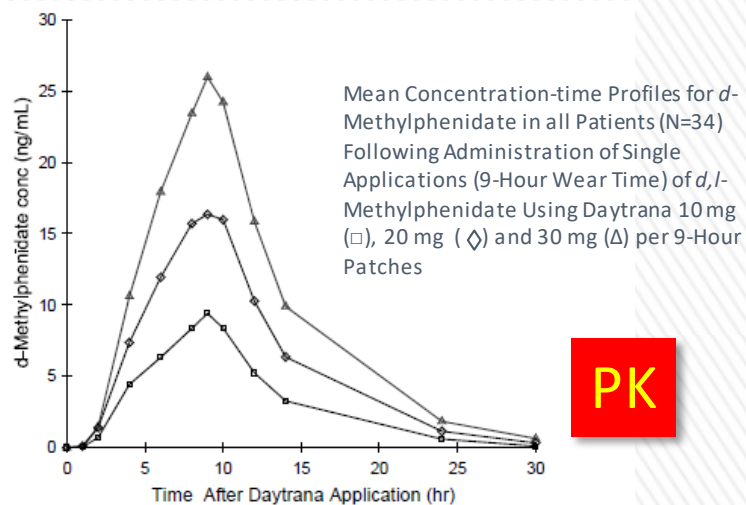
Question

Is the rate of absorption and the extent of exposure a determinant of the clinical response of a MPH treatment?

Exposure-Response

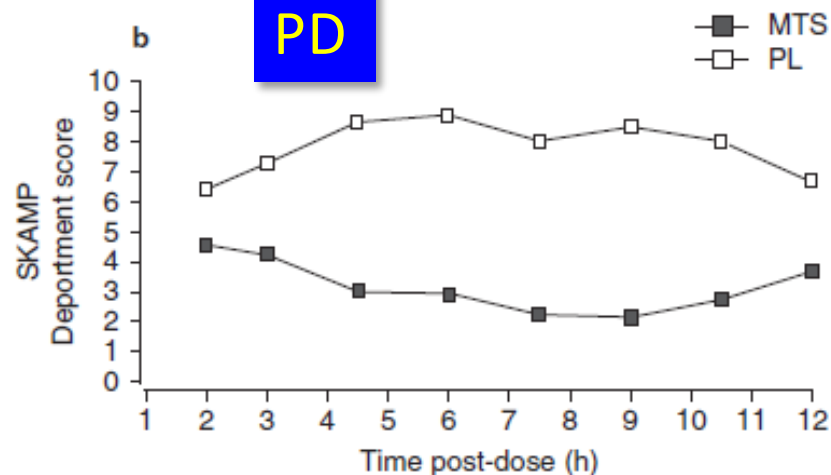
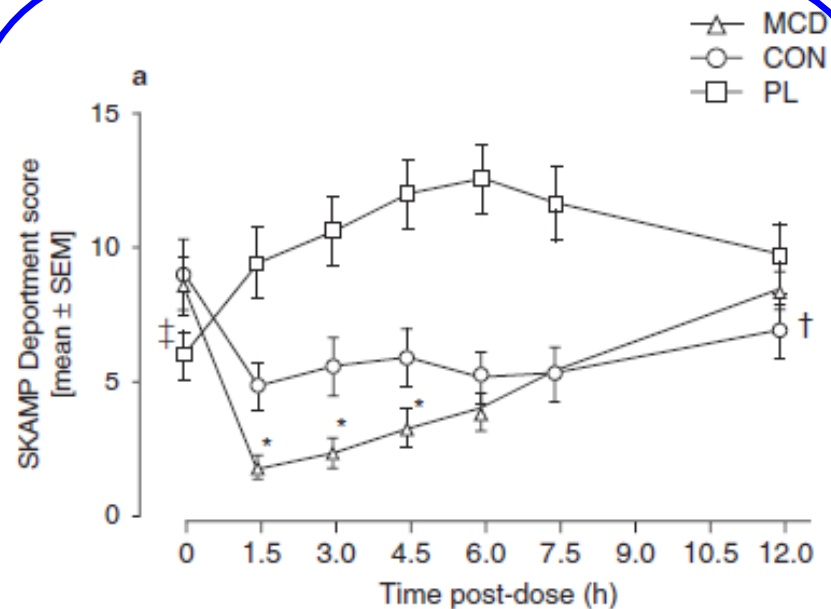


González, M.A., et al. Methylphenidate bioavailability from two extended-release formulations. *Int. J. Clin. Pharmacol. Ther.* 40, 175-84 (2002).



PK

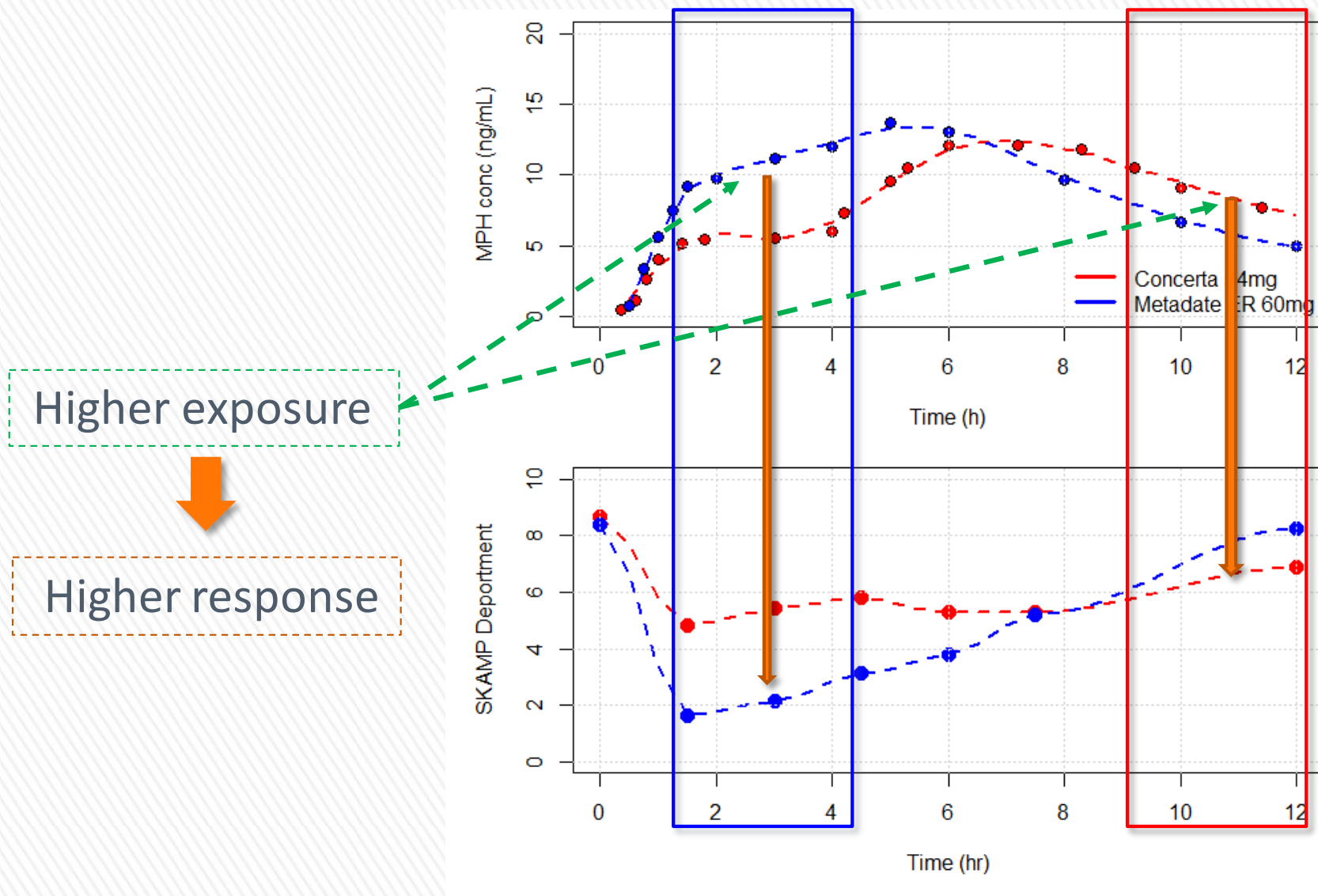
Daytrana (methylphenidate transdermal system) film label - FDA https://www.accessdata.fda.gov/drugsatfda_docs/label/2010/021514s011lbl.pdf



Wigal S.B. Efficacy and Safety Limitations of Attention-Deficit Hyperactivity Disorder Pharmacotherapy in Children and Adults. *CNS Drugs* 2009; 23 Suppl.1: 21-31

Legend: **CON**=CONCERTA®, **MCD**=METADATE ER®, **MTS**=DAYTRANA®

ER Oral Formulations: Treatment Effect



Higher exposure



Higher response

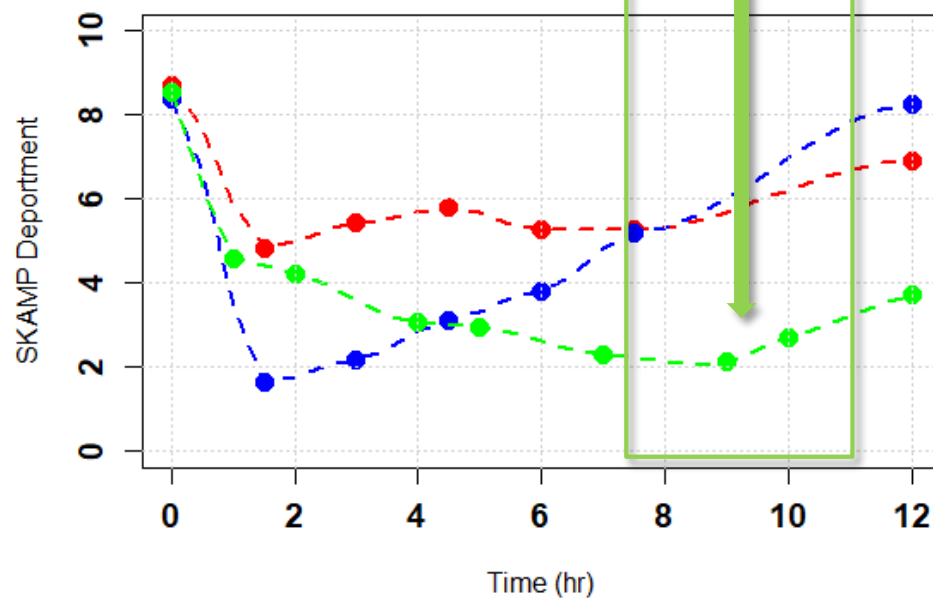
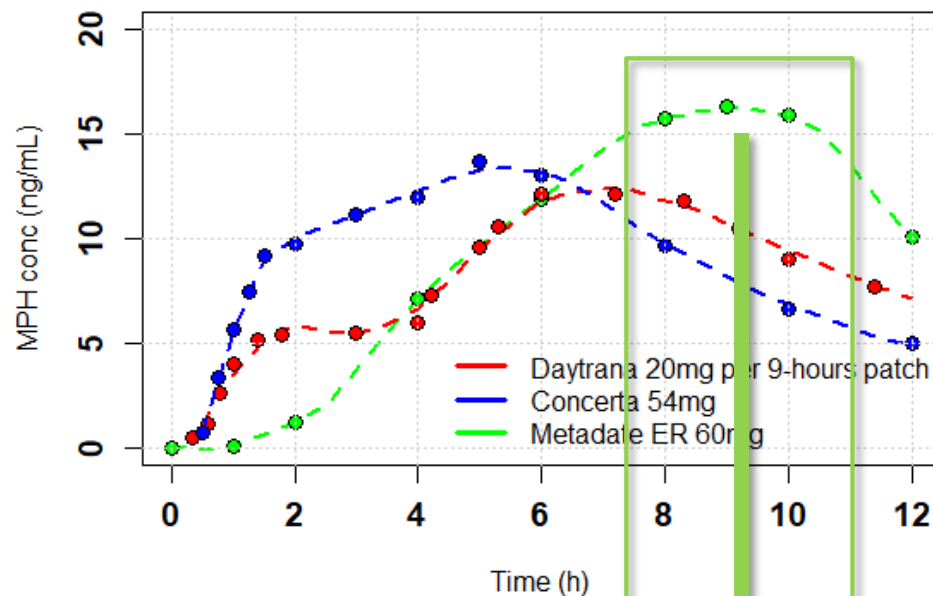
Use a meta-analytic approach for aggregating PK and PD data generated in different studies

Treatment Effect

Higher exposure



Higher response



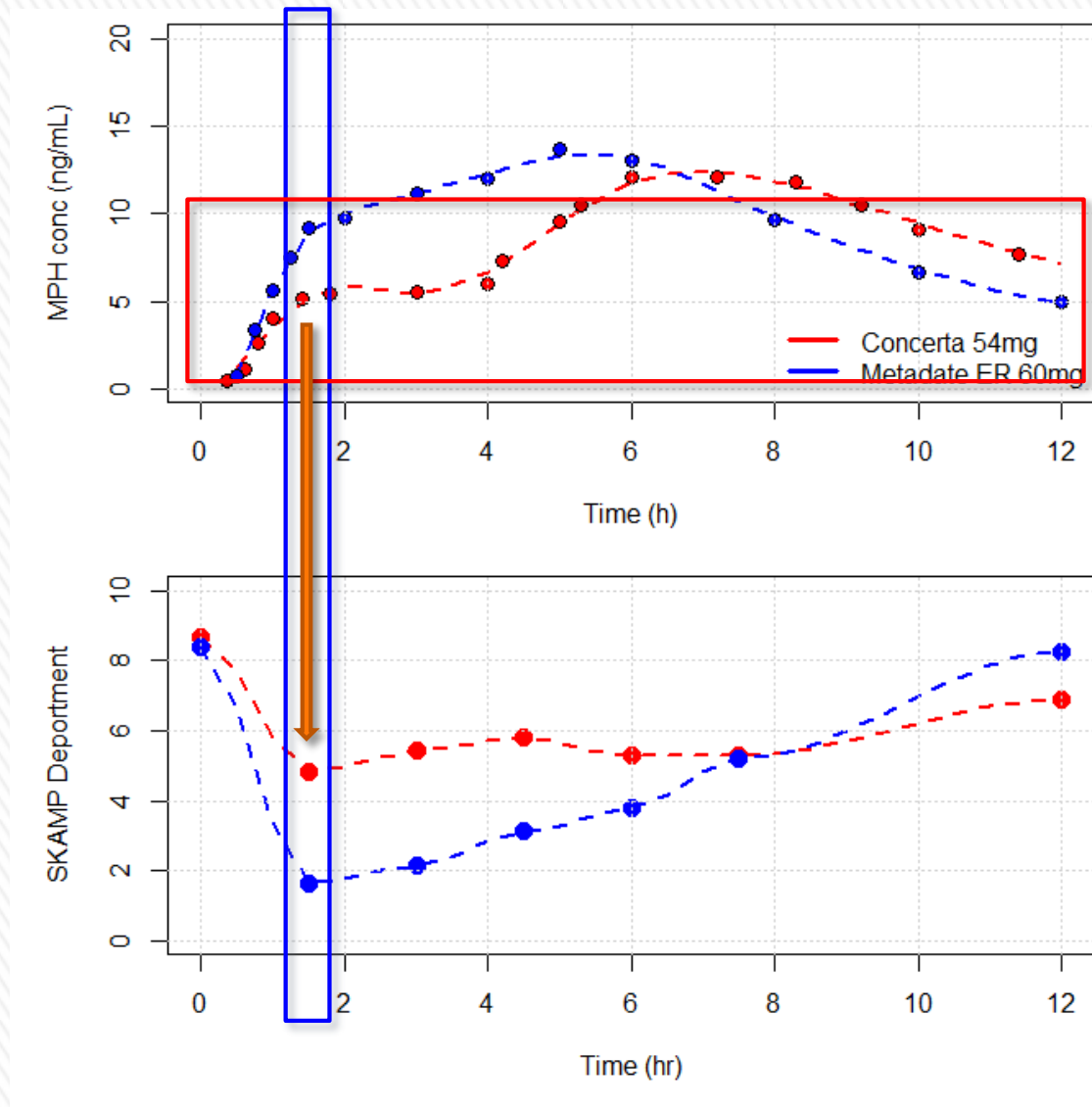
Use a meta-analytic approach for aggregating PK and PD data generated in different studies

ER Oral Formulations: Treatment Effect

For a given treatment, the larger response is not associated with the larger exposure



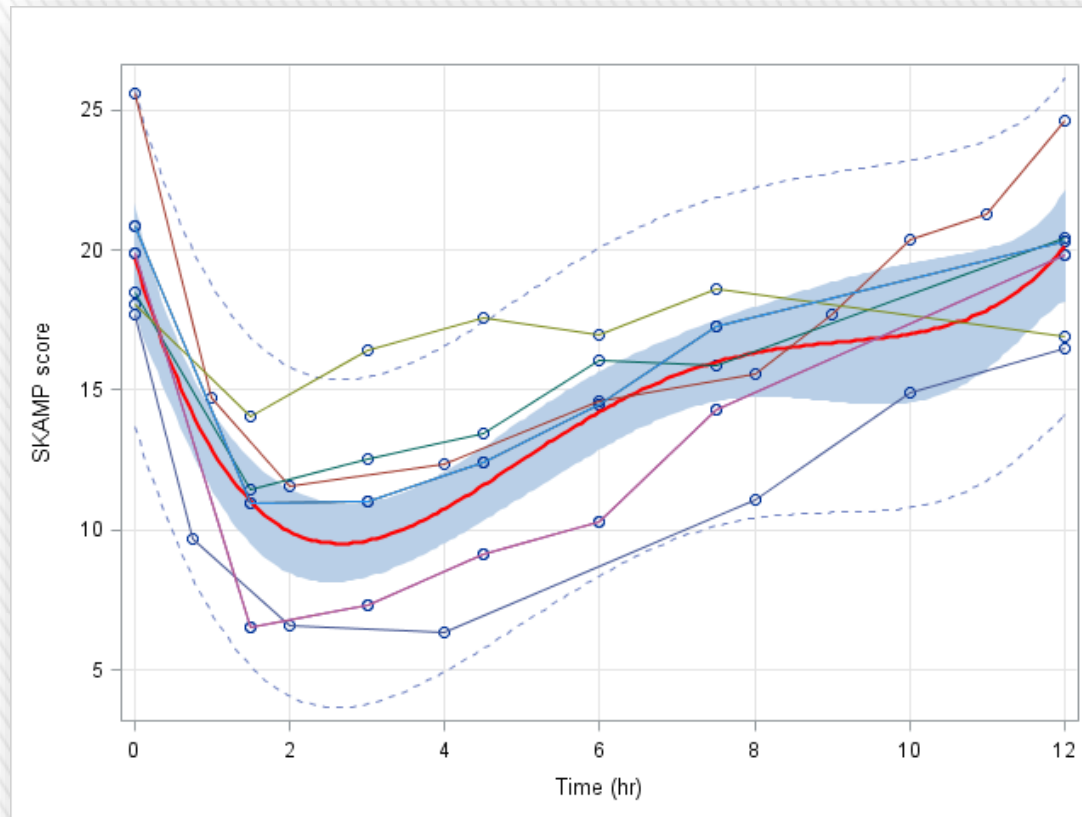
- Time-course of the longitudinal placebo response ?
- Tachyphylaxis?



Use a meta-analytic approach for aggregating PK and PD data generated in different studies

MPH response – Meta-analytic approach

Smoothing function describing the typical SKAMP trajectories from 4 clinical trials after Concerta[®] (16mg, 36mg, and 54mg), Metadate CD[®] (20 mg, 40 mg, and 60 mg), Focalin XR[®] (20 mg), and Quillivant XR[®] (60 mg).

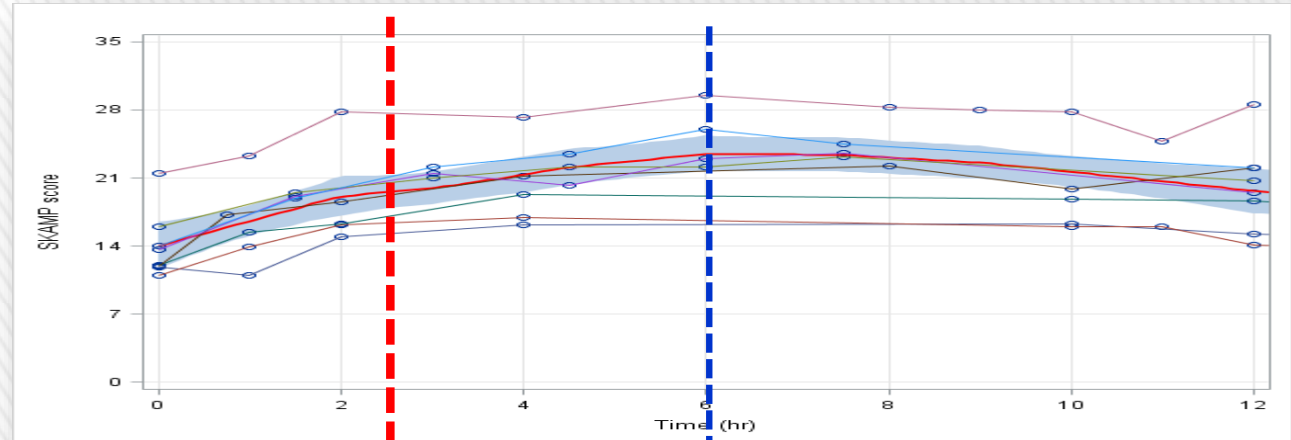


- Common longitudinal trajectory
- Dose related drop from baseline

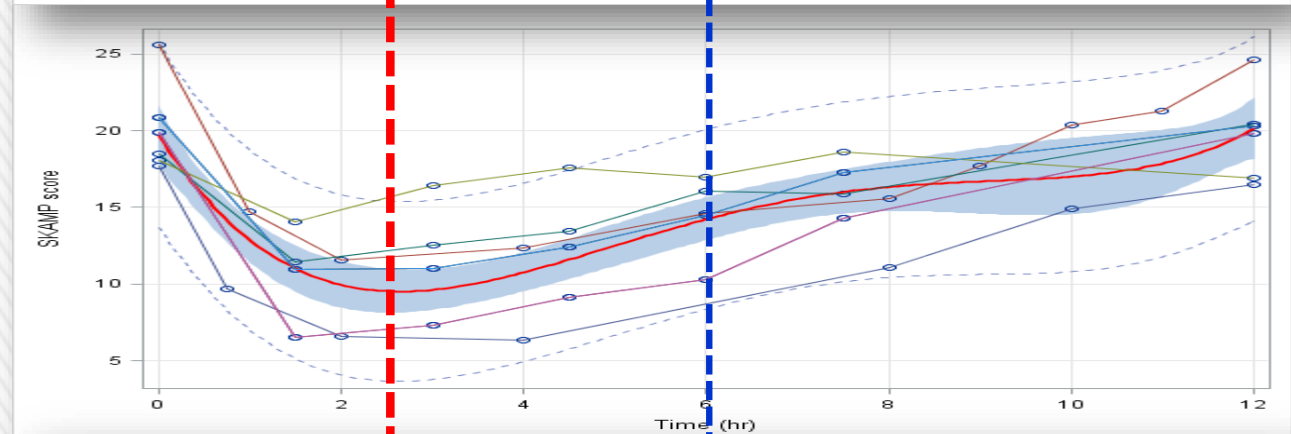
- 1) Kimko H. et al. *J Pharmacokinet Pharmacodyn.* 2012;39(2):161-76.
- 2) Sharon B. Wigal et al., *Journal Of Child And Adolescent Psychopharmacology* 2013. 23(1),
- 3) Sonuga-Barke E.J. Et al *BMC Psychiatry.* 2004. 4:28
- 4) Raul R. Silva et al. *Journal Of Child And Adolescent Psychopharmacology* . 2006. 16(3)

Placebo and MPH related response

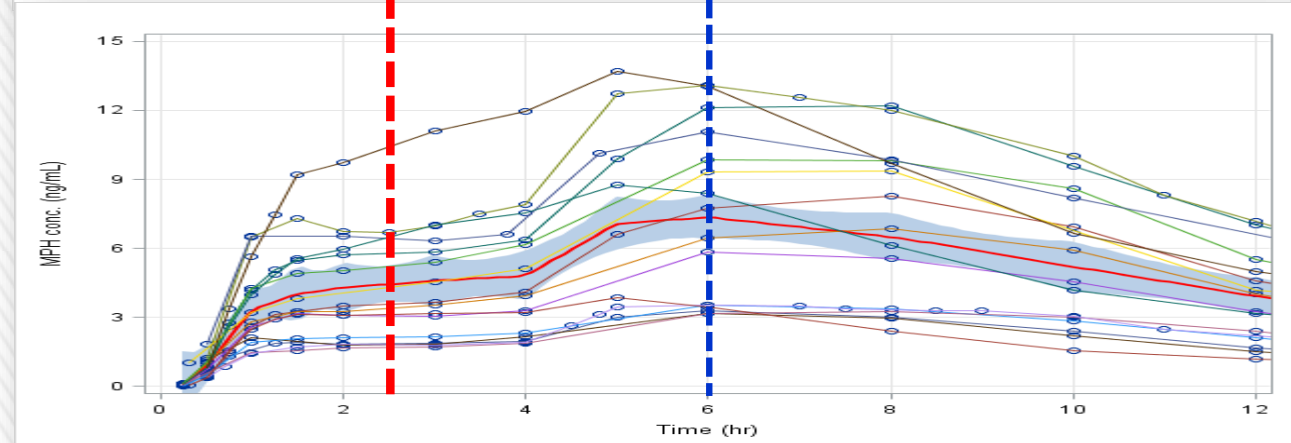
Placebo
SKAMP
score



MPH
SKAMP
score



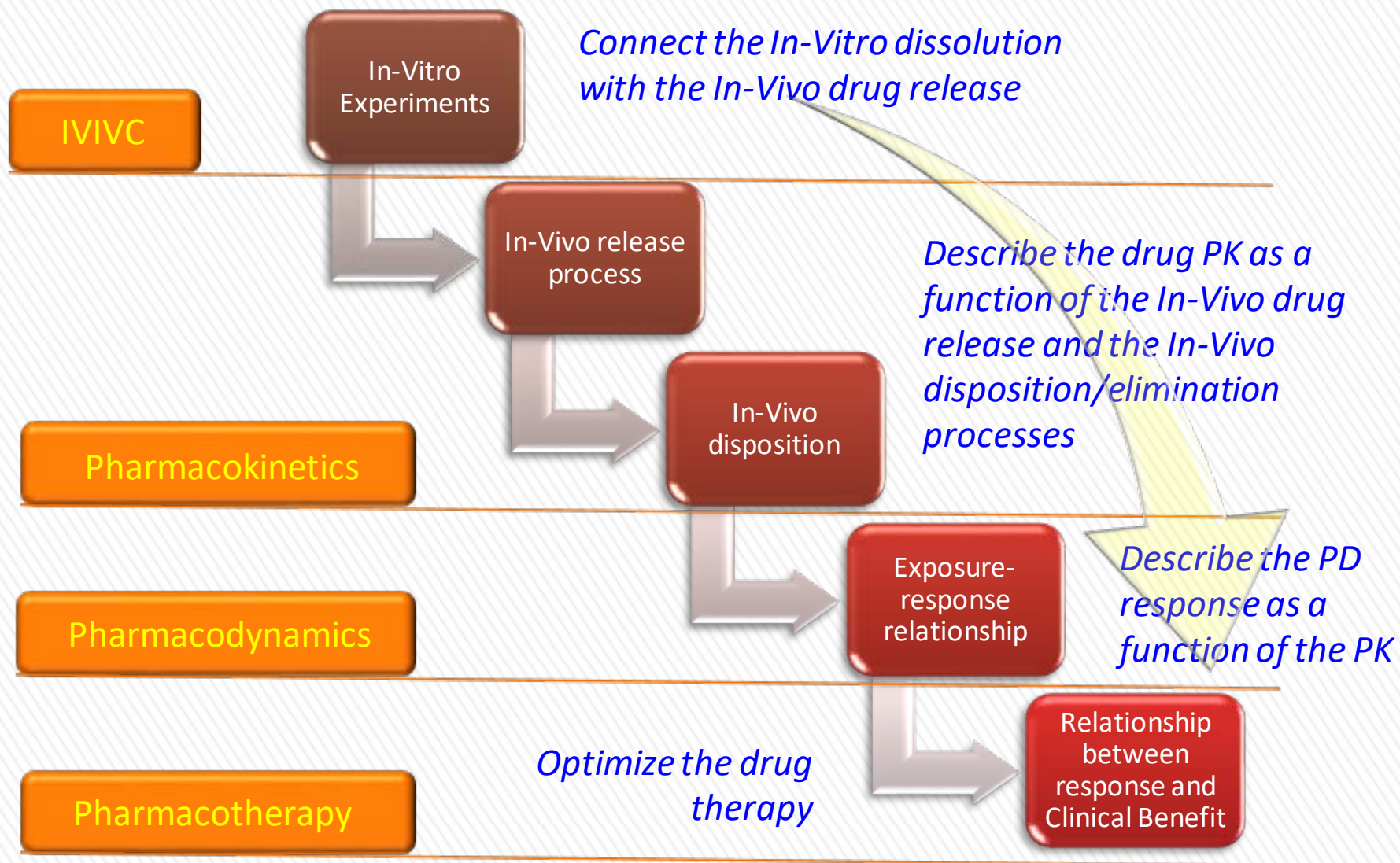
MPH
PK
Concerta®
Metadate®



Modeling strategy

- I. Characterize the complex PK of the MPH extended release products
- II. Characterize the time course of the Placebo response
- III. Characterize the MPH related effect on the SKAMP scores accounting for tachyphylaxis
- IV. Estimate the optimal MPH dose and release rate for better controlling early morning clinical response

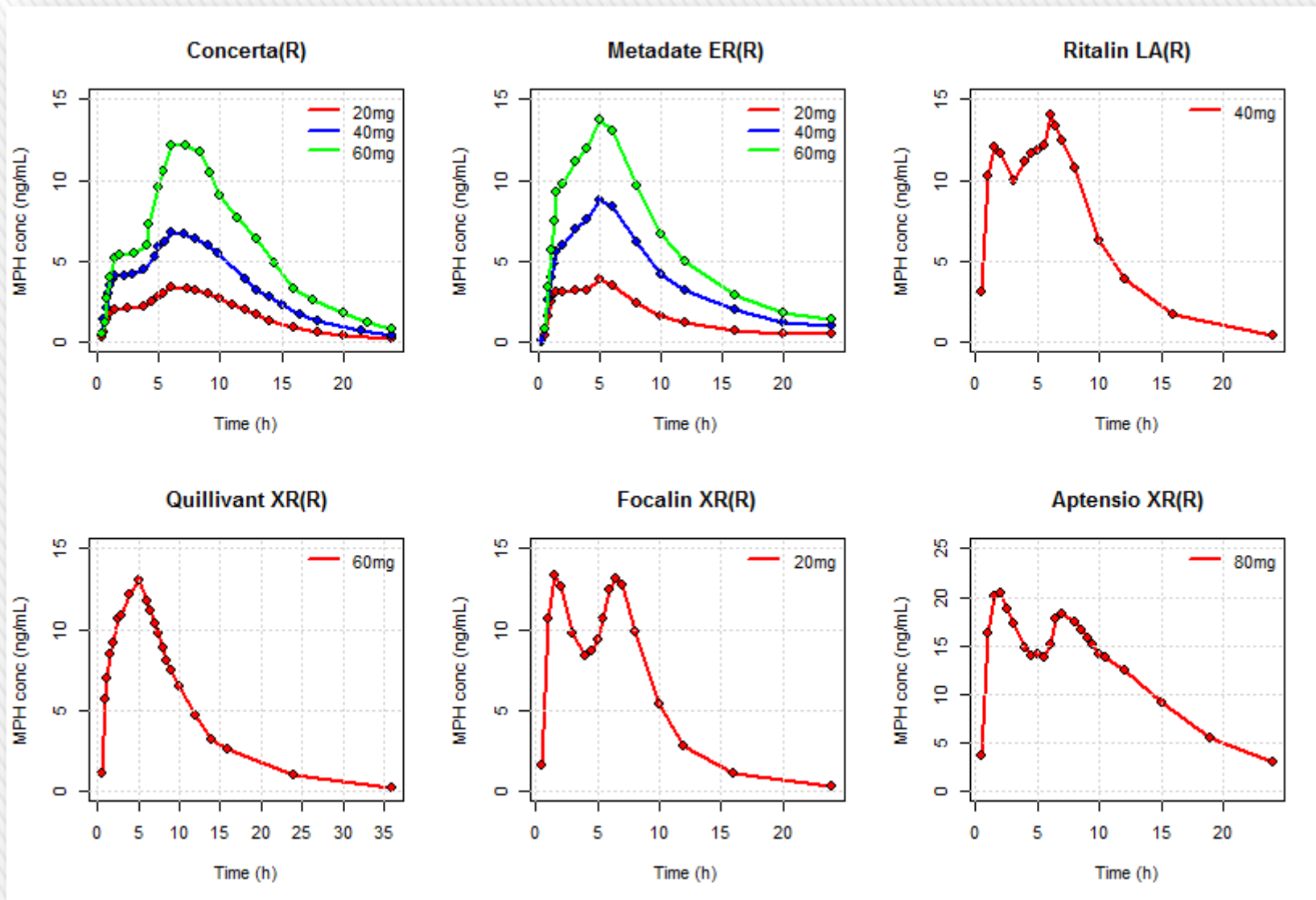
Integrated modelling framework



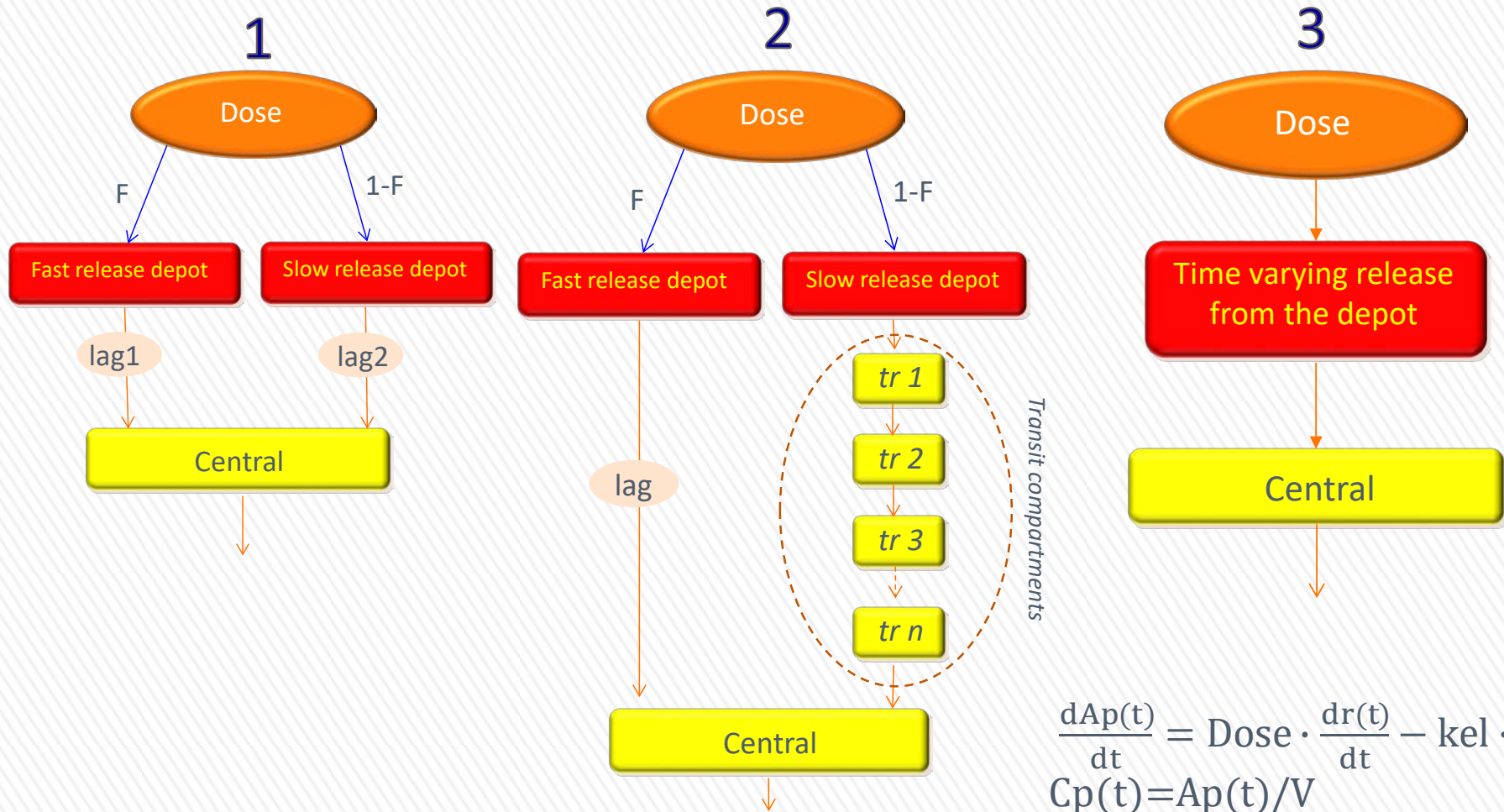
Part I

Characterize the complex PK of the MPH extended release products

PK profiles of different extender release formulations of MPH



Alternative models

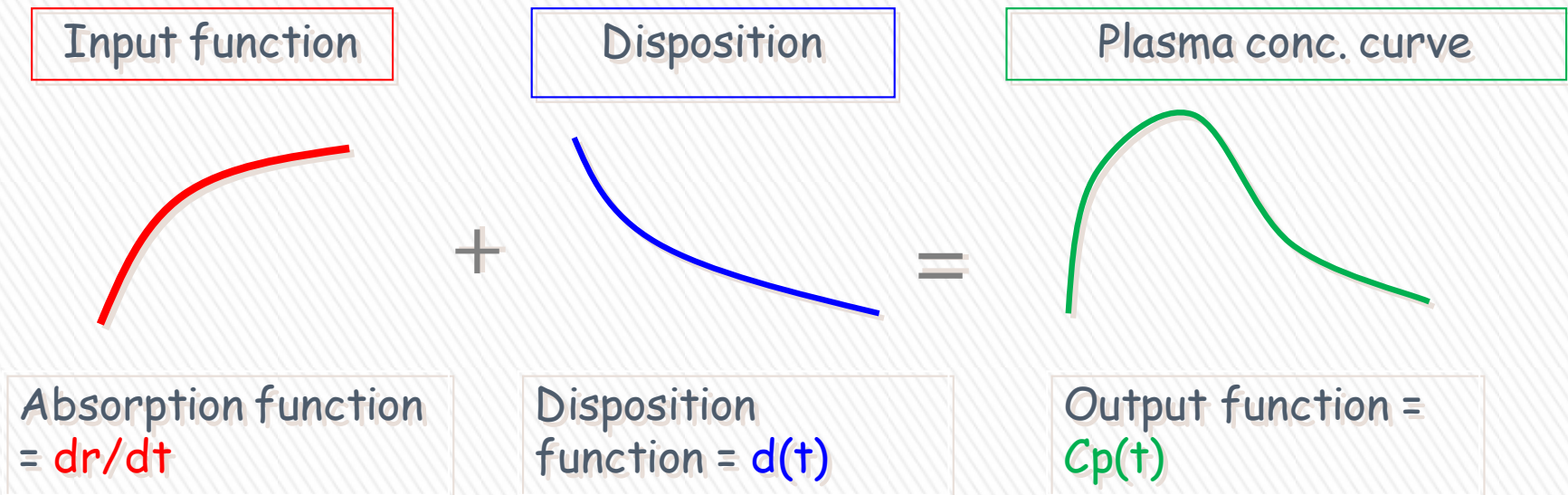


$$r(t) = 1 - \left(FF \cdot e^{-\left(\left(\frac{t}{TD}\right)^{ss}\right)} + (1 - FF) \cdot e^{-\left(\left(\frac{t}{TD1}\right)^{ss1}\right)} \right)$$

Convolution integral theory

$$C_p(t) = \int_0^t \frac{dr(\tau)}{dt} \cdot d(t - \tau) \cdot d\tau$$

The plasma drug-concentration-vs.-time curve can be viewed as the resultant of the combined processes relating drug absorption, distribution and elimination



The output function $C_p(t)$ can be estimated as the convolution of a input function dr/dt , with a disposition function $d(t)$ (drug disposition after IV dose)

Convolution-based PK model

The integrated PK model linking in-vivo drug release with the disposition and elimination processes can be developed using a convolution-based approach. The drug concentration (C_p), resulting from an arbitrary dose, can be described by convolution as:

$$C_p(t) = \int_0^t \frac{dr(\tau)}{dt} \cdot d(t - \tau) \cdot d\tau$$

where $f(t)$ is the rate of in-vivo drug delivery, $d(t)$ is the unit impulse response and $*$ is the symbol defining the convolution.

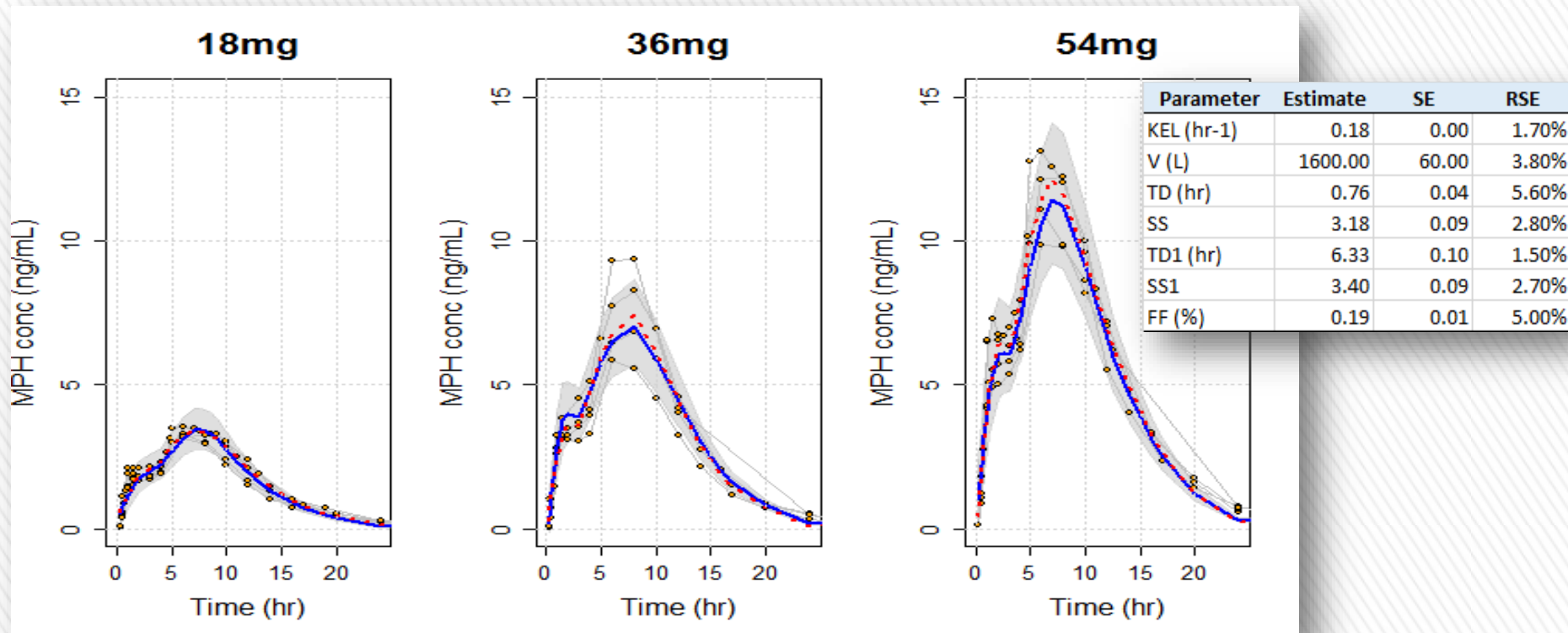
In case of a simple disposition process (say one compartment), the model equation describing $C_p(t)$ can be written as

$$\frac{dC_p}{dt} = Dose \cdot \frac{dr}{dt} - Kel \cdot C_p$$

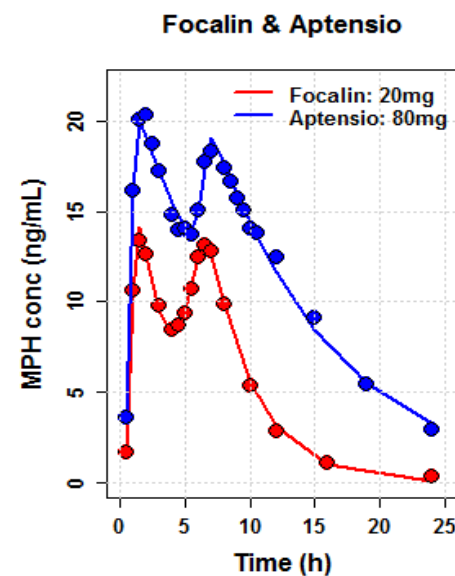
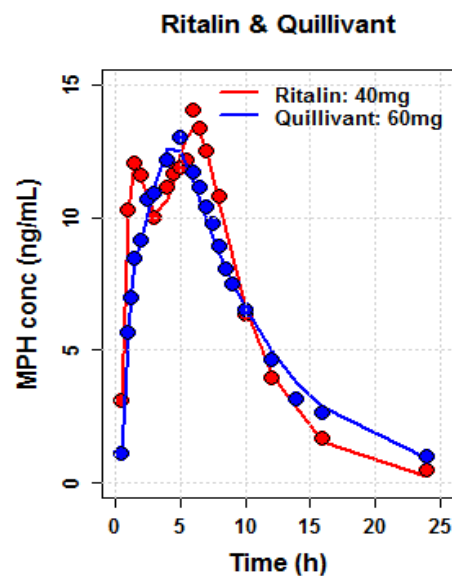
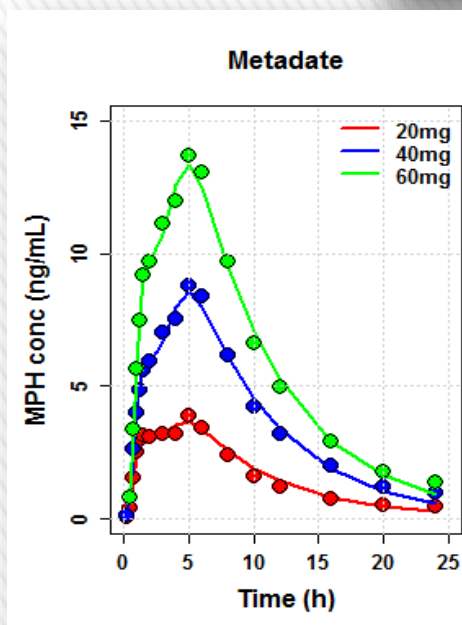
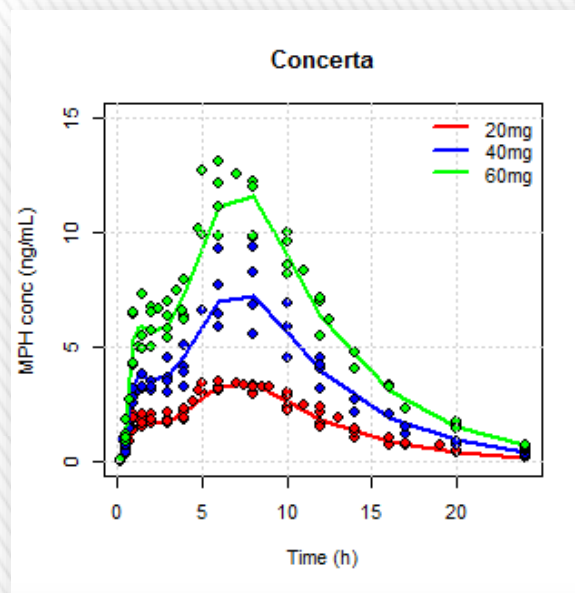
Assuming that the time-varying fraction of the dose released can be described by the function $r(t)$ (input function). This can be computed analytically or can be approximated using the finite difference approach (see an example of implementation in NONMEM)

Convolution-based model used to fit Concerta[®] data

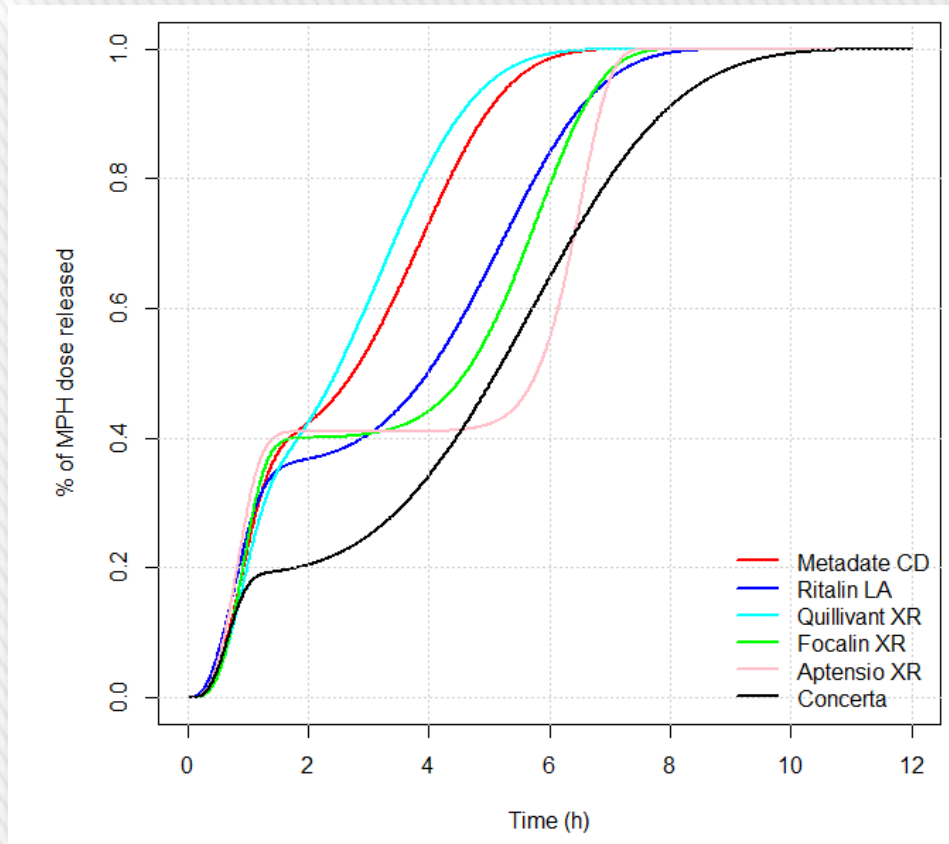
- Data extracted from 9 publications describing the MPH PK following administration of Concerta[®] (16mg, 36mg, and 54mg) were used for model building
- The convolution-based approach performed better than the other models



Fit 6 MPH formulations PK data with the same model



Comparing in-vivo release of 6 MPH products



f = fraction of the dose released in the 1st process
 td = time to absorb 63.2% of the dose released in the 1st process
 $td1$ = time to absorb 63.2% of the dose released in the 2nd process
 ss = sigmoidicity factor for the 1st process
 $ss1$ = sigmoidicity factor for the 2nd process

	td (hours)	ss	ss1	td1 (hours)	ff (%)	td_99% (hours)	td1_99% (hours)
Metadate CD [®]	1.04	2.87	3.68	4.21	39%	1.64	6.19
Ritalin LA [®]	0.92	2.6	4.18	5.55	36%	1.51	7.8
Quillivant XR [®]	1.03	2.99	2.98	3.64	32%	1.57	5.91
Focalin XR [®]	1.02	3.41	6.56	5.96	40%	1.5	7.39
Aptensio XR [®]	0.93	3.15	6.56	14.3	59%	1.46	7.2
Concerta [®]	0.76	3.18	3.4	6.33	19%	1.07	9.79

Linking In-vitro and in-vivo release: the convolution-based model

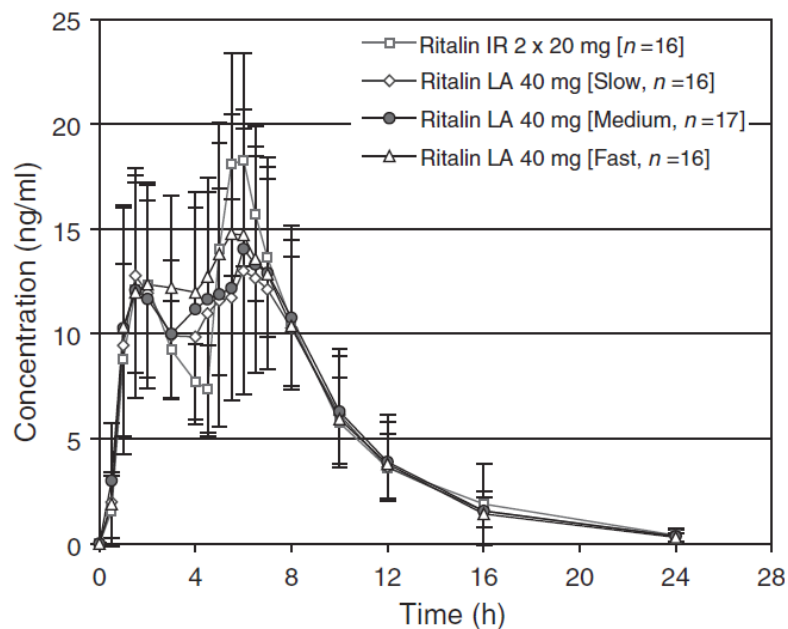


Figure 1. Plasma concentration (mean \pm SD)-time profile of MPH after a single dose of Ritalin LA slow-, medium- or fast-release formulation and Ritalin tablets (immediate release formulation) given 4 h apart

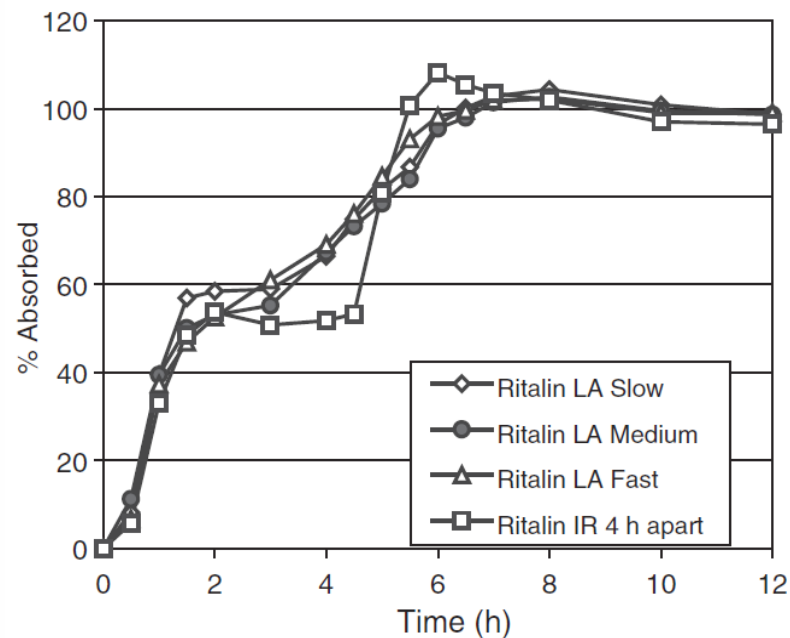
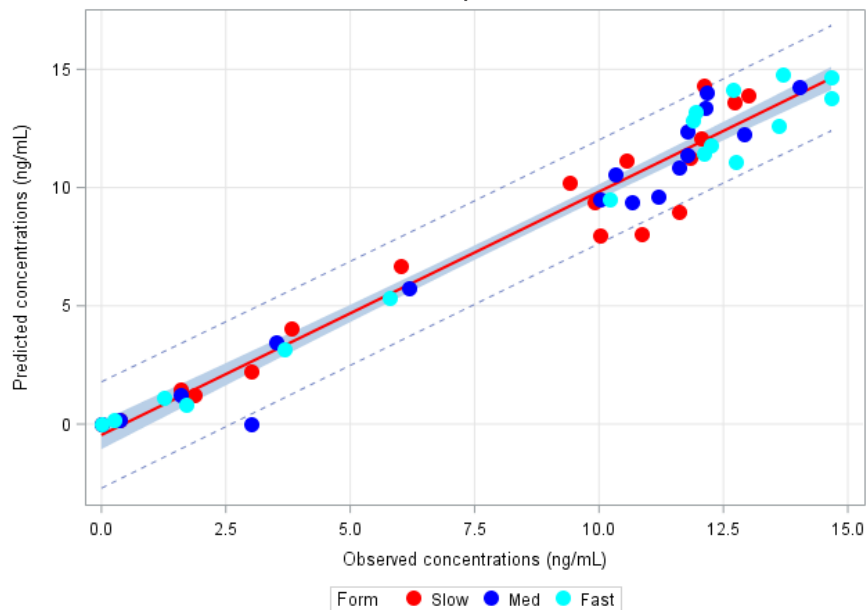


Figure 3. Cumulative fraction of MPH absorbed from Ritalin LA slow-, medium- and fast-release formulations and from Ritalin-IR tablet given 4 h apart

Evaluating the IVIVC

In-vivo prediction



Parameter Estimates						
Variable	DF	Parameter Estimate	Standard Error	t Value	Pr > t	95% Confidence Limits
Intercept	1	-0.45953	0.29739	-1.55	0.1286	-1.05685 0.13779
Observed PK	1	1.02828	0.03058	33.63	<.0001	0.96686 1.08969

$$PE = \frac{1}{n} \sum_{i=1}^n \frac{|\text{Obs. value} - \text{Pred. value}|}{\text{Obs. value}} \cdot 100$$

0 - 24 hr						
Formulation	cmax_o	auc_o	cmax_p	auc_p	pe_cmax	pe_auc
Slow	13.01	127.46	14.35	127.36	10.28	0.08
Medium	14.02	130.86	14.25	122.92	1.64	6.07
Fast	14.67	133.44	14.79	126.75	0.87	5.01
Average					4.26	3.72

Predictability criteria:

- PE <15% for each formulation,
- PE <10% for mean values

0 - 3 hr			
Formulation	auc_o	auc_p	pe_auc
Slow	14.70	15.09	2.62
Medium	16.99	16.30	4.05
Fast	15.18	14.79	2.62
Average			3.10

3 - 7 hr			
Formulation	auc_o	auc_p	pe_auc
Slow	44.70	41.71	6.68
Medium	46.70	46.15	1.18
Fast	45.96	47.16	2.62
Average			3.49

7 - 12 hr			
Formulation	auc_o	auc_p	pe_auc
Slow	26.25	28.44	8.35
Medium	25.96	23.80	8.36
Fast	36.63	32.59	11.03
Average			9.24

Part II

Characterize the time course of the
Placebo response

Disease Progression Model

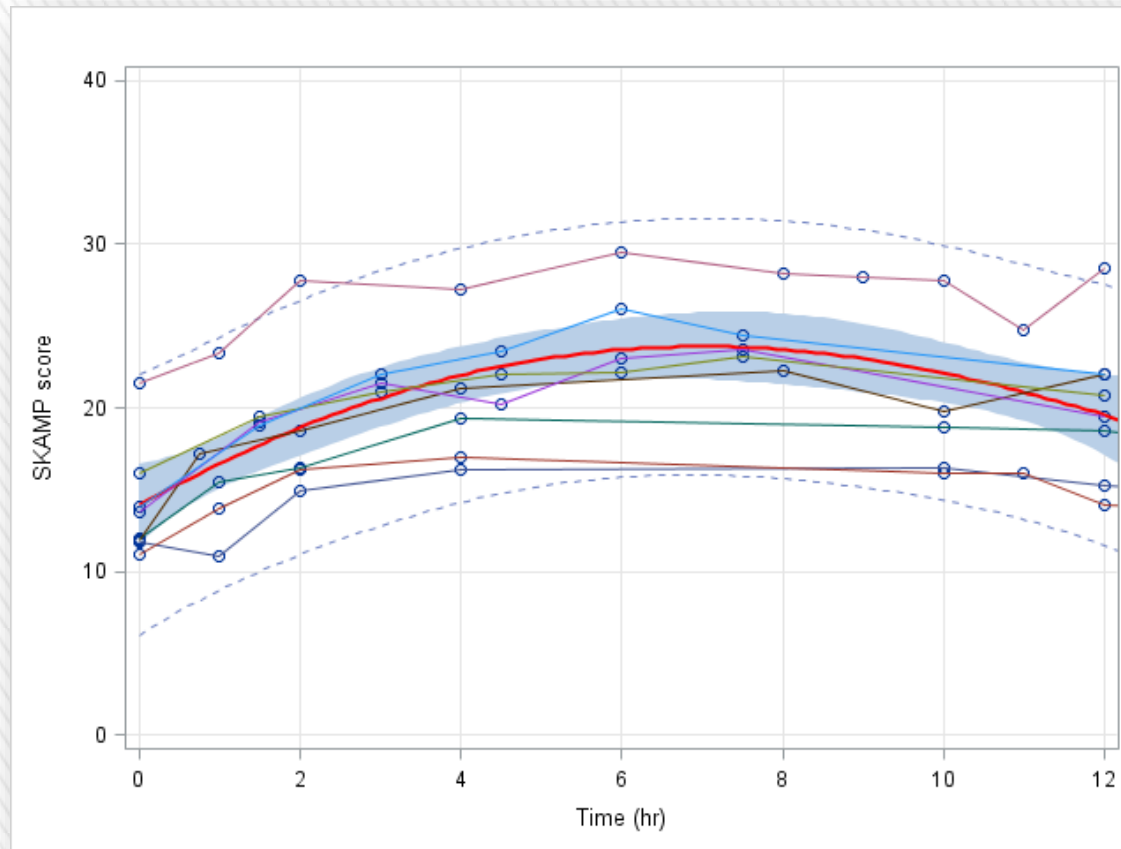
Disease Progression Model: A quantitative model that accounts for the time course of disease status, $S(t)$:

- *Baseline Disease State*
- *Natural History*
- *Placebo Response*
- *Active Treatment Response*

$$S(t) = \text{Natural History} + \text{Placebo} + \text{Active}$$

Placebo data – Meta-analytical approach

Smoothing function describing the typical mean SKAMP placebo trajectories from 4 clinical trials



- 1) Kimko H. et al. *J Pharmacokinet Pharmacodyn.* 2012;39(2):161-76.
- 2) Sharon B. Wigal et al., *Journal Of Child And Adolescent Psychopharmacology* 2013. 23(1),
- 3) Sonuga-Barke E.J. Et al *BMC Psychiatry.* 2004. 4:28
- 4) Raul R. Silva et al. *Journal Of Child And Adolescent Psychopharmacology* . 2006. 16(3)

Semi-mechanistic Placebo response model

The rate of change of the response (R = SKAMP score) was described by:

$$\frac{dR}{dt} = K_{in} \cdot (1 + f(t)) - K_{out} \cdot R$$

where k_{in} represents the zero-order rate constant for onset of response, R , and k_{out} is the first-order rate constant for the loss of response variable.

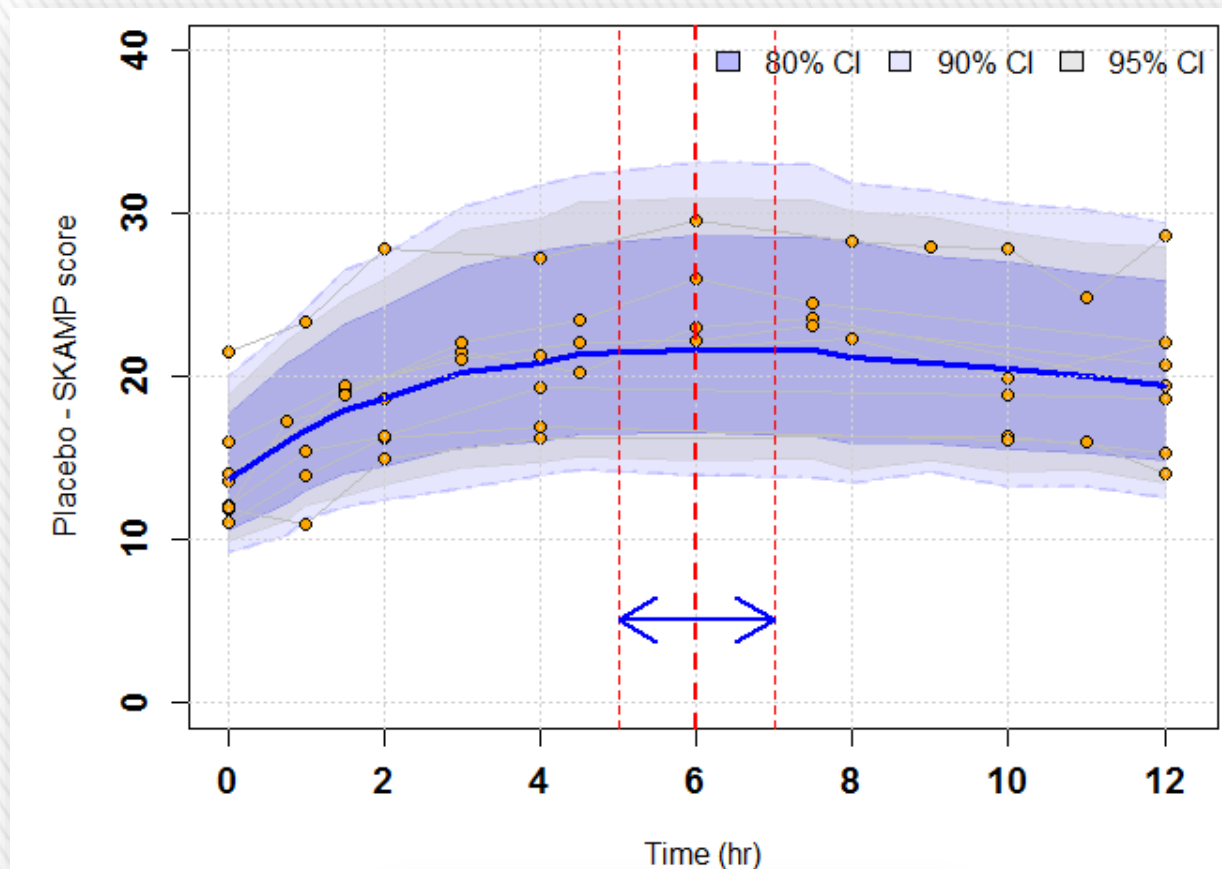
As the system is assumed to be stationary, the response (R) begins at a predetermined baseline value (Bas), changes with time, and eventually returns back to R_0 .

$$f(t) = AA \cdot e^{-time \cdot A1} \quad \text{Time varying placebo effect}$$

$$R(t = 0) = Bas = \frac{K_{in}}{K_{out}} \quad \text{Baseline SKAMP score}$$

$$K_{in} = K_{out} \cdot Bas$$

Placebo response – Modelling results



Parameter	Estimate	SE	RSE
BAS	12.70	1.040	8.20%
KOUT	0.15	0.008	5.00%
A1	0.16	0.008	5.10%
AA	1.85	0.162	8.80%

The dark blue area represent the 80% prediction intervals and the light blue area represent the 90% and the 95% prediction intervals

Part III

Characterize the MPH related effect
on the SKAMP scores

Tachyphylaxis

Methylphenidate exhibit acute tolerance :concentrations measured soon after an initial dose cause a greater pharmacodynamic effect then concentrations occurring at a later time

- The EC_{50} (the MPH concentration giving 50% of the maximal effect) increase with time
- This assumption translate the observation that the effect associated with a constant MPH exposure decline with time: as time passes, higher EC_{50} -> more drug is needed to achieve the same effect*

$$EC_{50}(t) = EC_{50b} \left(1 + \frac{time^{ga}}{t50^{ga} + time^{ga}} \right)$$

Where:

$EC_{50}(t)$ = effective time varying EC_{50} value

EC_{50b} = EC_{50} at time 0

$t50$ = time at which 50% of the maximal change in $EC_{50}(t)$ is reached

ga = rate of change in the $EC_{50}(t)$

*Kimko et al. Journal of pharmacokinetics and pharmacodynamics 2012;39:161-76

MPH clinical response model

$$SKAMP(effect) = R(t) + Delta - \frac{Emax \cdot C_p}{EC_{50}(t) + C_p}$$

Where:

$R(t)$ is the placebo response defined by the model

$$\frac{dR}{dt} = k_{in} \cdot (1 + f(t)) - k_{out}R$$

$EC_{50}(t)$ is the time varying EC_{50} defined by the model

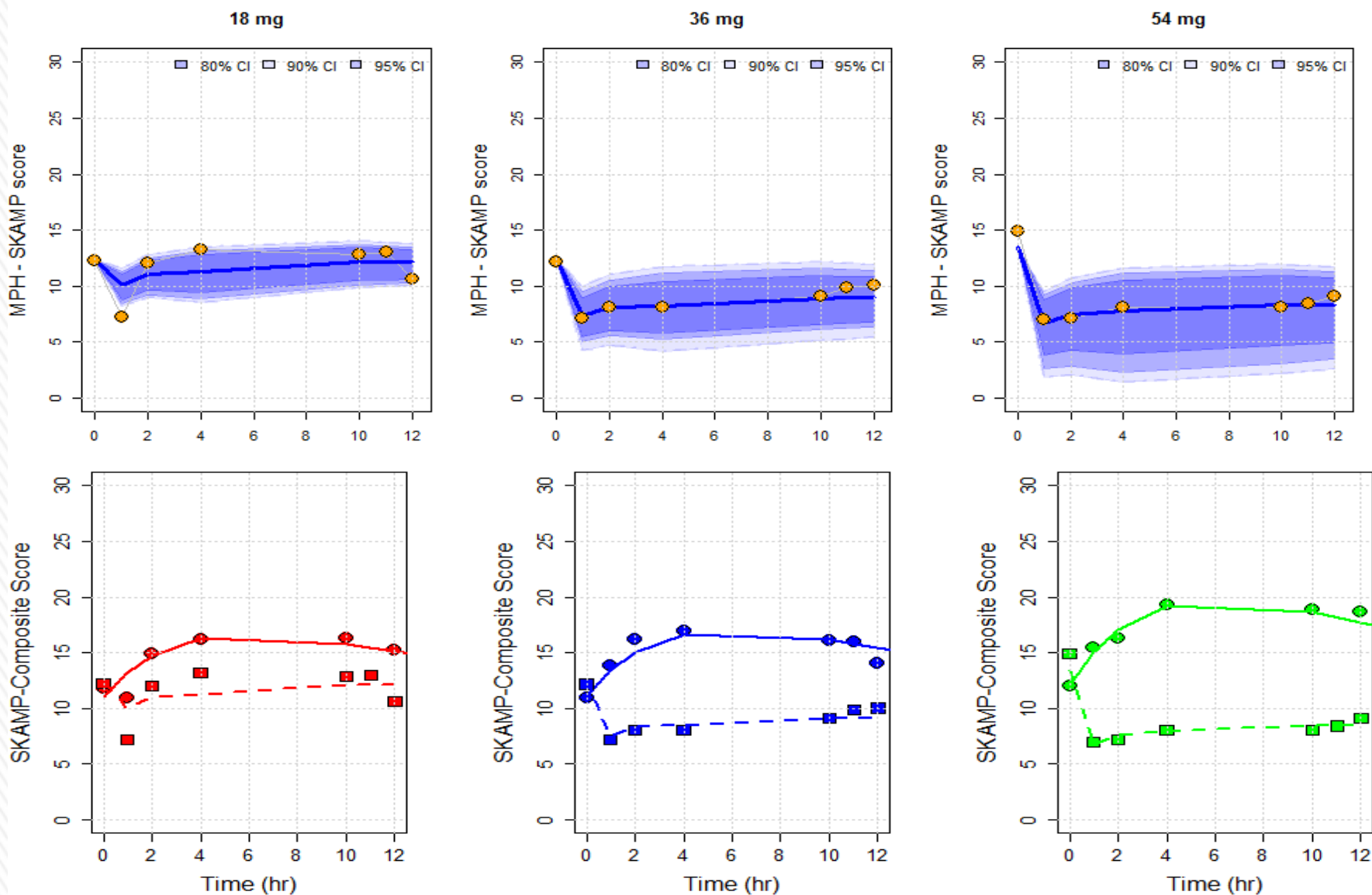
$$EC_{50}(t) = EC_{50b} \left(1 + \frac{time^{ga}}{t_{50}^{ga} + time^{ga}} \right)$$

$Delta$ is the score difference at baseline depending on the treatment between assessment days

$Emax$ is the maximal MPH related effect

C_p is the MPH drug concentration

The PK/PD model



Part IV

Estimate the optimal MPH dose and iv-vivo release rate for optimizing the clinical benefit of a treatment

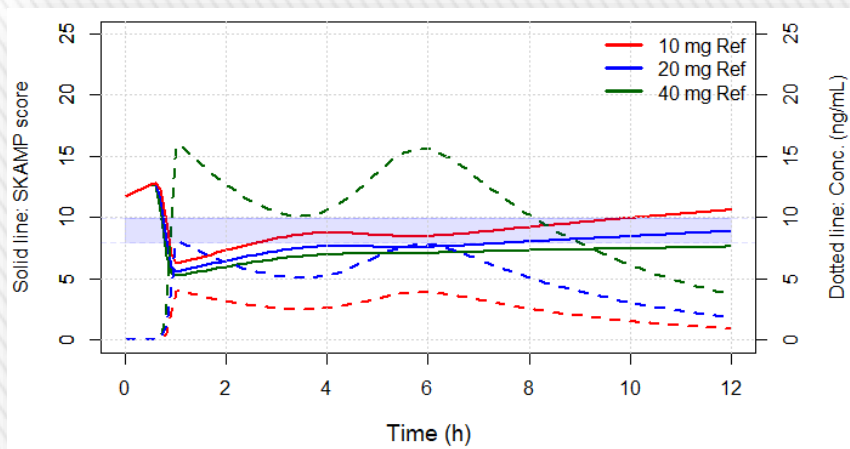
Clinical benefit

$$SKAMP(effect) = P(t) + Delta - \frac{Emax \cdot C_p}{EC_{50} + C_p}$$

$$r_{vivo}(t) = Dose \cdot \left[1 - \left(ff \cdot e^{-\left(\frac{time}{td} \right)^{ss}} + (1 - ff) \cdot e^{-\left(\frac{time}{td1} \right)^{ss1}} \right) \right]$$

$$f(t) = \frac{dr_{vivo}}{dt}$$

$$\frac{dA}{dt} = F_i * Dose * f(t) - k_{el} \cdot A \quad \rightarrow \quad Cp = \frac{A}{V}$$

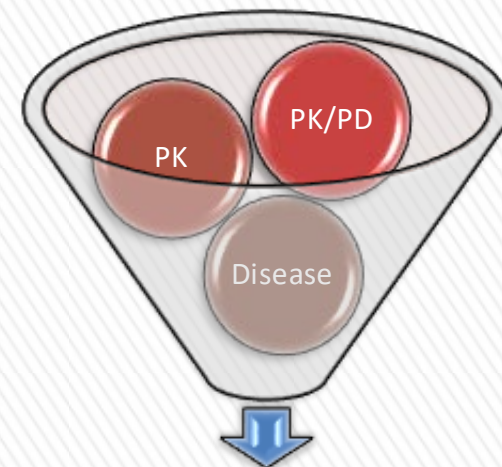


CB: maintenance of SKAMP scores from 8 to 10 during 12 hours was considered as the target clinical response

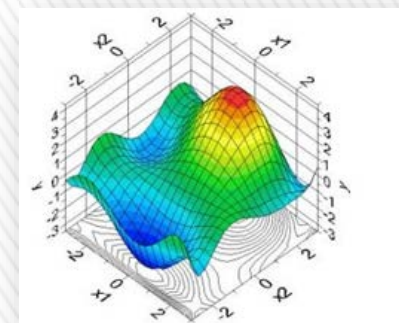
$$CB = f(SKAMP)$$

$$SKAMP = f(C_p)$$

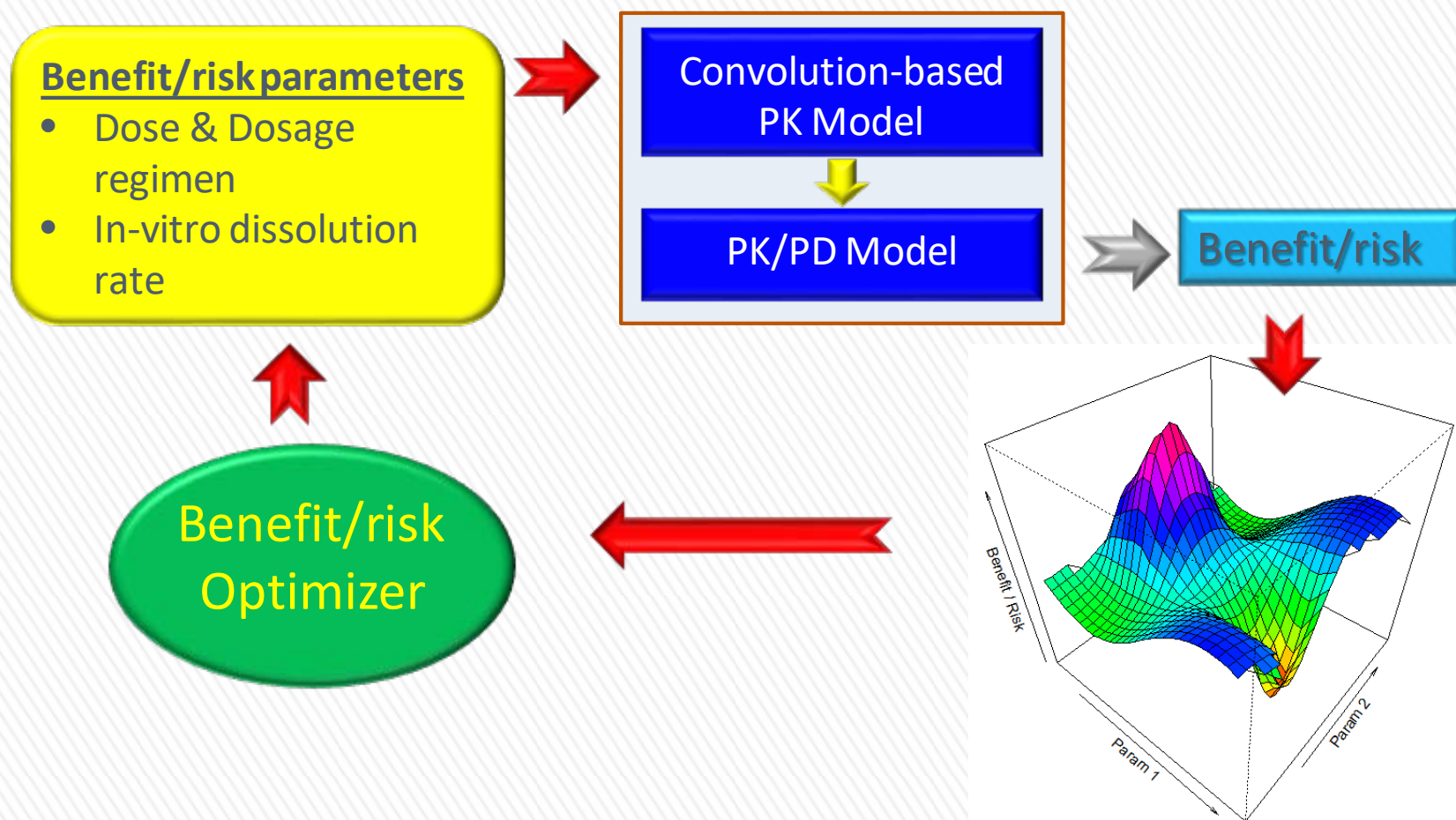
$$C_p(ff, ss, ss1, td, td1, Dose)$$



Response variable = Clinical benefit

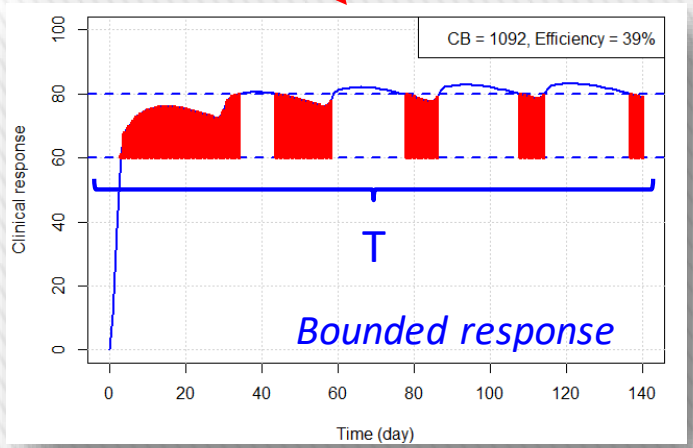
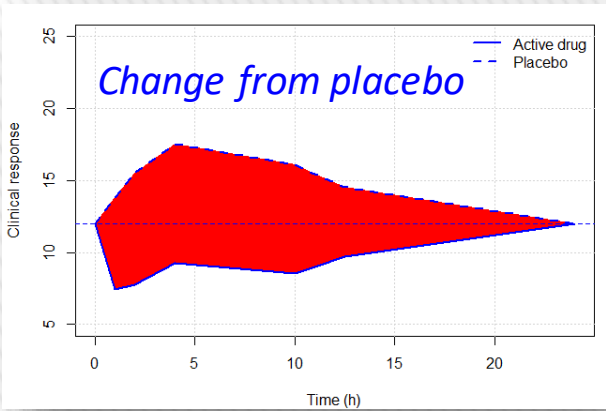
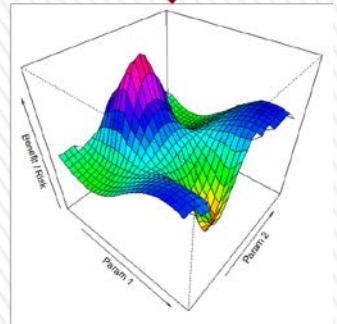
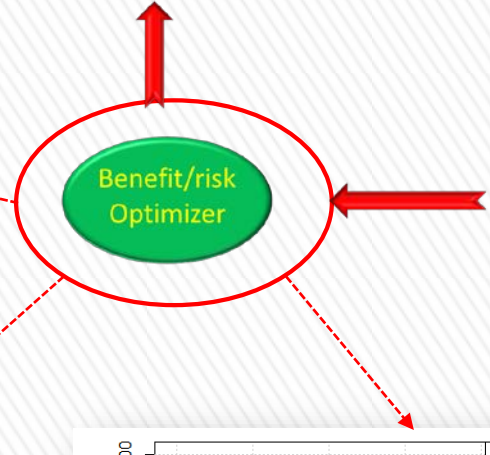
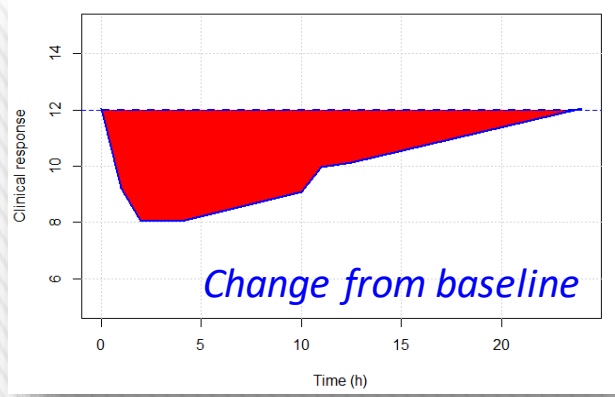
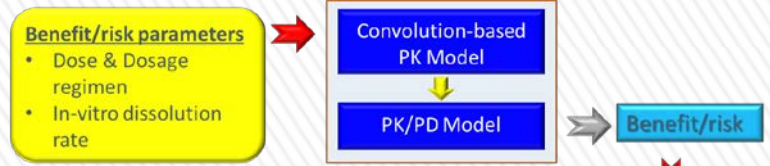


Response Surface Analysis and Nonlinear Optimization Algorithm for Maximization of Clinical Benefit



R Gomeni, FMM Bressolle-Gomeni, TJ Spencer, SV Faraone, L Fang, A Babiskin. Model-Based Approach for Optimizing Study Design and Clinical Drug Performances of Extended-Release Formulations of Methylphenidate for the Treatment of ADHD. Clin Pharmacol Ther. 2017 Mar 29. doi: 10.1002/cpt.684

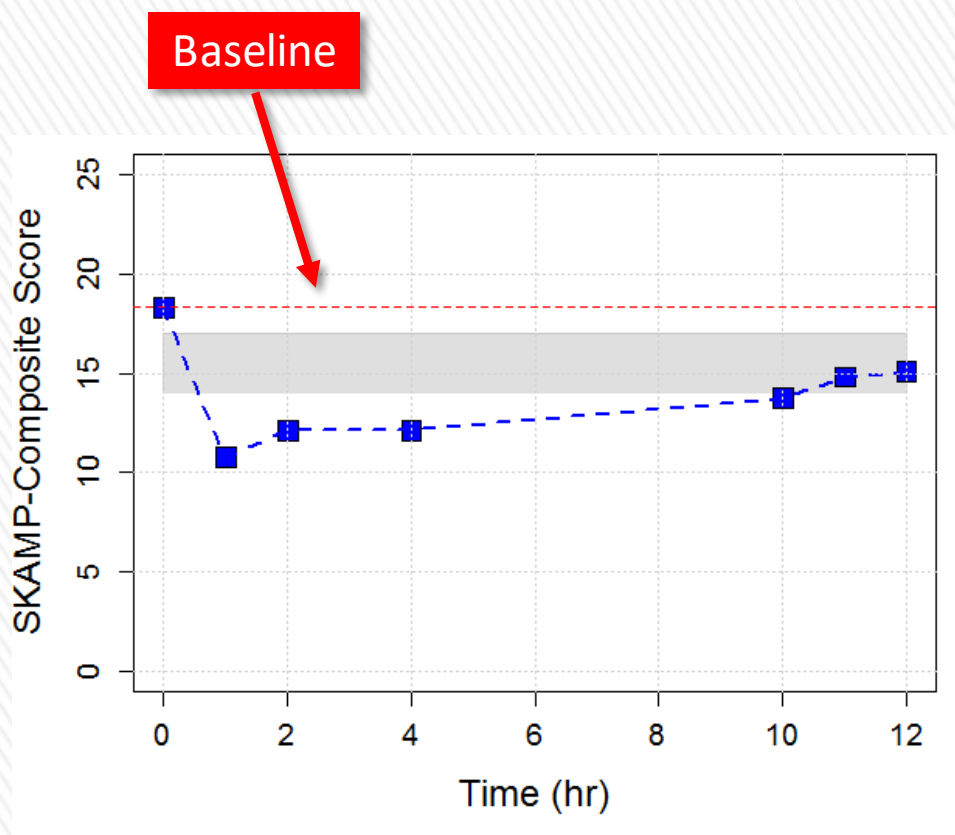
Clinical Benefit



} Δ

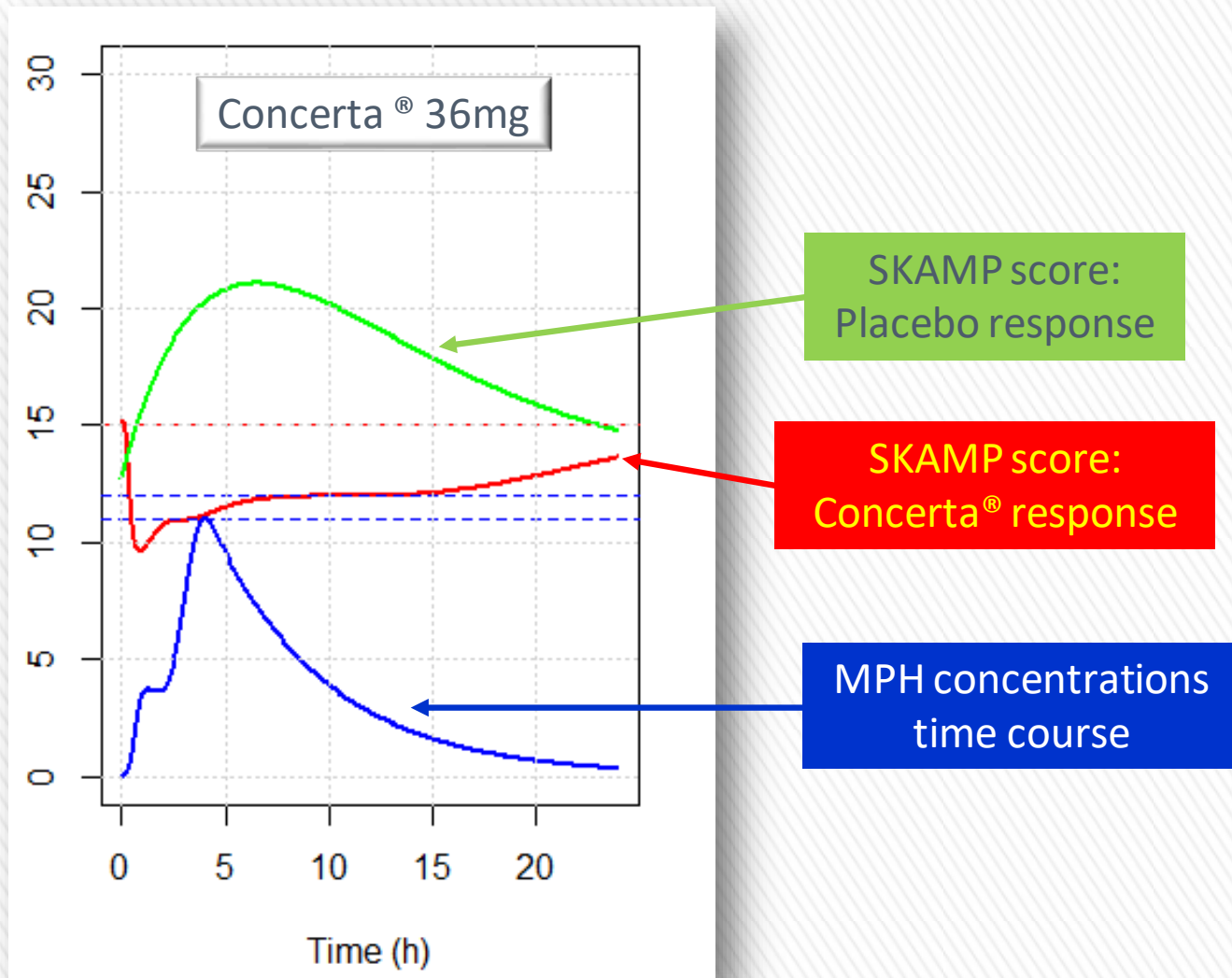
Clinical benefit in the treatment of ADHD

- Question : what is the ideal MPH ER dose and in-vivo release for
- an initial improvement in the SKAMP score (say for example a drop of 15% of the baseline value after one hour from drug intake)
 - a maintenance of this improvement during 8 consecutive hours?



Target region for the clinical response

Reference treatment

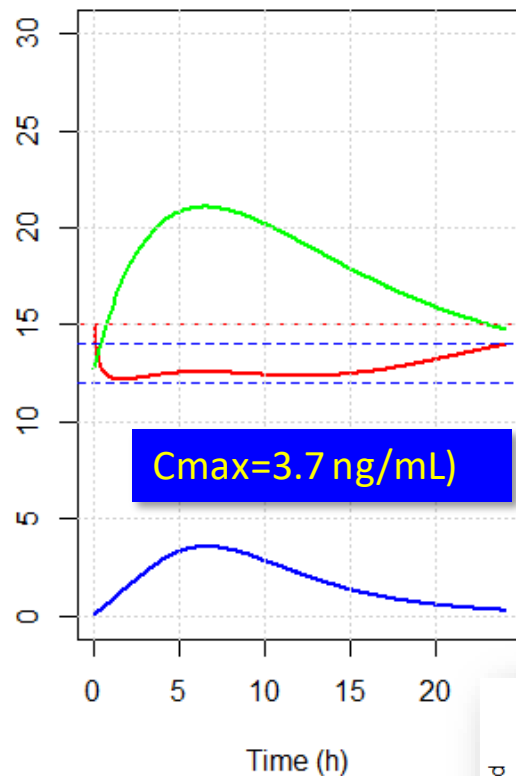
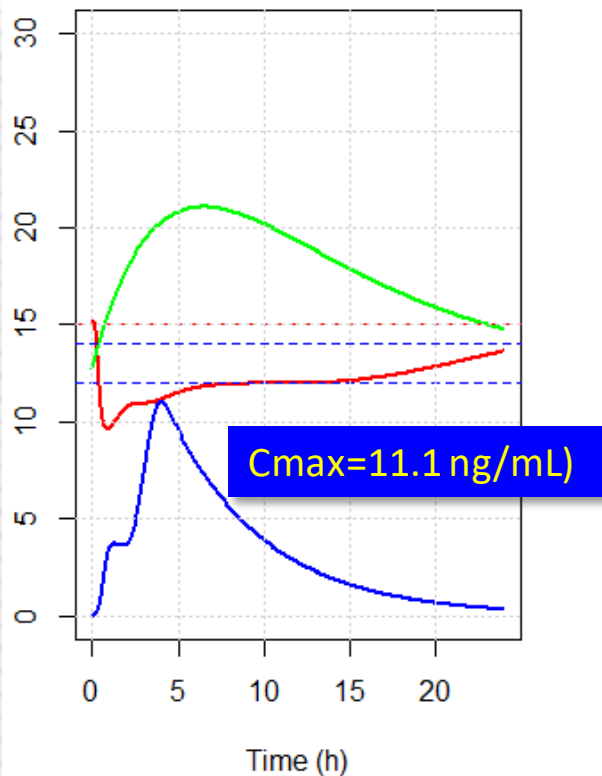


SKAMP score:
Placebo response

SKAMP score:
Concerta[®] response

MPH concentrations
time course

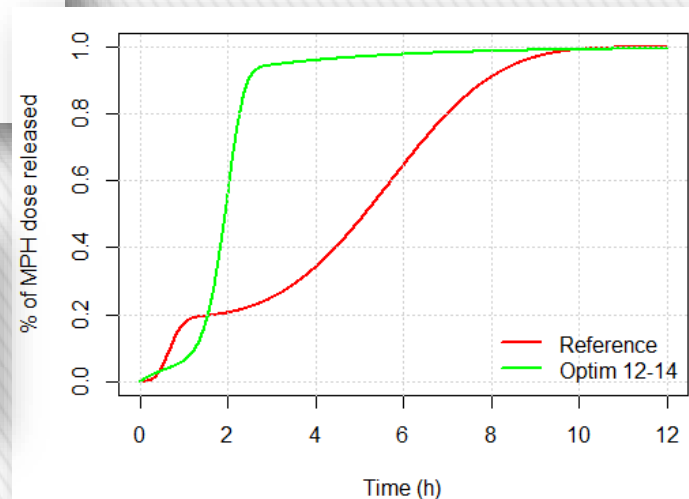
Model prediction



Optimize MPH dose and MPH in-vivo release rate

- *SKAMP score between 12 and 14*
- *Variable dose*

	SKAMP		In-Vivo release					Dose (mg)
	Min	Max	td (hr)	ss	td1 (nr)	ss1	ff (%)	
Reference			0.76	3.18	3.40	6.33	0.19	36
Optimal	12	14	2.72	0.87	6.36	2.06	0.16	20




Exposure-response of blood pressure and heart rate for methylphenidate in healthy adults

J Pharmacokinet Pharmacodyn (2017) 44:245–262
DOI 10.1007/s10928-017-9513-5



ORIGINAL PAPER

Exposure–response analyses of blood pressure and heart rate changes for methylphenidate in healthy adults


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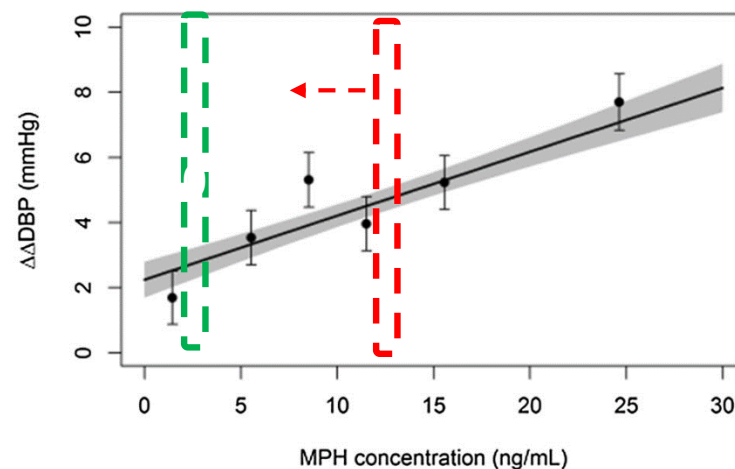
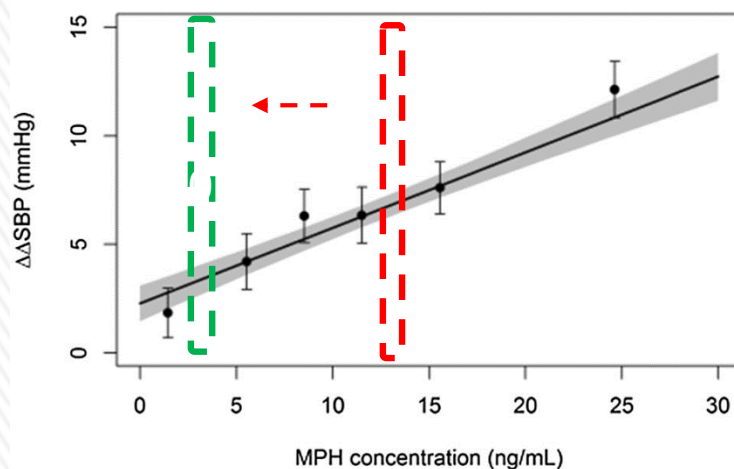
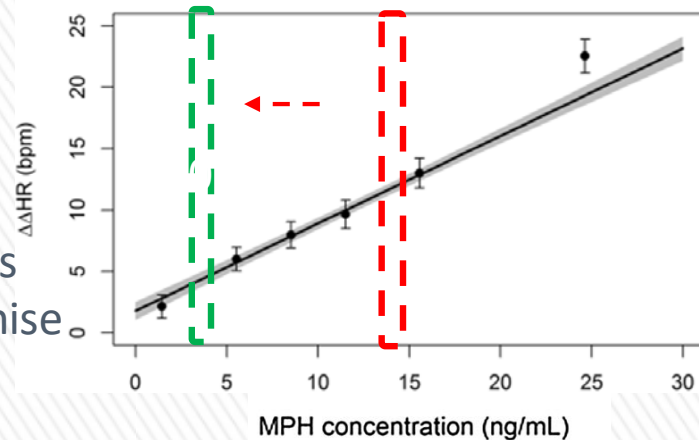
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 **Conclusion:** The developed models adequately characterized the circadian rhythm and the MPH induced effects on BP and HR. The changes in BP and HR were highly correlated with MPH blood levels with no apparent delay.

Exposure–response analyses of blood pressure and heart rate changes for MPH in healthy adults

The exposure-response of blood pressure (BP) and heart rate (HR) for MPH in healthy adults indicated that the BP and HR changes were directly related and highly dependent on the MPH plasma concentration. These safety issues associated with MPH treatment may compromise the treatment course of ADHD in children and also raise parents' concerns over them.



Li L, Wang Y, Upoor RS, Mehta MU, Farchione T, Mathis MV, Zhu H. Exposure-response analyses of blood pressure and heart rate changes for methylphenidate in healthy adults. *J Pharmacokinet Pharmacodyn.* 2017 Jun;44(3):245-262. doi: 10.1007/s10928-017-9513-5.

Conclusion

- A model-informed approach can be used for identifying the best performing *in-vivo* delivery rate appropriate for maximizing the benefit-risk ratio and for facilitating the development of a formulation with the required characteristics using *in-vitro/in-vivo* correlation.
- The surface-response analysis can be prospectively applied for optimizing the drug development process by identifying the drug properties associated with an optimized benefit-risk.
- The proposed model-informed approach provides the pharmaceutical companies with a methodological framework for developing drugs with drug delivery and a dose selection suitable to produce a clinical benefit prospectively defined by the clinicians and not just a clinical response better than the placebo response.

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Thank you

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