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CERTARA

***Introduction to QSP in Drug
Discovery and Development: a
Historical, Current and Future
Perspective***

Piet van der Graaf

ASCPT, Washington DC
13th March 2019

The **birth** of QSP?

Quantitative and Systems Pharmacology in the Post-genomic Era: New Approaches to Discovering Drugs and Understanding Therapeutic Mechanisms

An NIH White Paper by the QSP Workshop Group – October, 2011

Peter K. Sorger (co-chair), Sandra R.B. Allerheiligen (co-chair)

Darrell R. Abernethy, Russ B. Altman, Kim L. R. Brouwer, Andrea Califano, David Z. D'Argenio, Ravi Iyengar, William J. Jusko, Richard Lalonde, Douglas A. Lauffenburger, Brian Shoichet, James L. Stevens, Shankar Subramaniam, Piet Van der Graaf and Paolo Vicini

Rebecca Ward (editor)

Definitions: QSP is defined as an approach to translational medicine that combines computational and experimental methods to elucidate, validate and apply new pharmacological concepts to the development and use of small molecule and biologic drugs. QSP will provide an integrated “systems-level” approach to determining mechanisms of action of new and existing drugs in preclinical and animal models and in patients. QSP will create the knowledge needed to change complex cellular networks in a specified way with mono or combination therapy, alter the pathophysiology of disease so as to maximize therapeutic benefit and minimize toxicity and implement a “precision medicine” approach to improving the health of individual patients.

NIH Workshop **follow up**



Quantitative Systems Pharmacology and Drug Discovery: Filling the Gaps in Current Models of the R&D

July 26, 2017 to July 27, 2017
Bethesda North Marriott Hotel & Conference Center (Conference Room: Salon C)
North Bethesda, Maryland
United States

Deliverables of Workshop

- White Paper: Merged Systems Biology and Pharmacology as an approach to CNS Drug Discovery: Filling the Gaps in Current Models of the R&D Process for Neurotherapeutics: NINDS/NIH Workshop, Summer, 2017
- Recommendations regarding the utility of this approach to CNS Drug Discovery and Development
- Posted WS summary on [ADDS website](#) to foster collaborations between data science and systems biology PI's (post → advertisement of WS to [ADDS blog](#) and its participants)

Summary and Recommendations from the 2017 NIH Workshop, "Quantitative Systems Pharmacology (QSP) and Drug Discovery: Filling the Gaps in Current Models of the R&D Process for Neurotherapeutics"

Hugo Geerts, In Silico Biosciences, Berwyn, PA

John Wiksw, Vanderbilt University, Nashville, TN, USA

Piet van der Graaf, Certara, Canterbury, UK

Jane Bai, Center for Drug Evaluation and Research, FDA, Silver Spring, MD, USA

Chris Gaiteri, Rush Alzheimer' Disease Center, Rush University, Chicago IL, USA

David Bennett, Rush Alzheimer' Disease Center, Rush University, Chicago, IL, USA

Susan Swalley, Novartis Biomedical Research Institute, Cambridge, MA, USA

Nancy Klimas, Miami VA Healthcare, Miami, FL, USA

Suzana Petanceska, Division of Neuroscience, NIA, Bethesda, MD, USA

MaryAnn Pellemounter, Division of Translational Research, NINDS, Bethesda, MD, USA



Pre-Conference Programming

ADVANCING QSP TOWARD PREDICTIVE DRUG DEVELOPMENT: FROM TARGETS TO TREATMENTS

WEDNESDAY, MARCH 13, 2019 | 8:00 AM – 5:00 PM

Co-Sponsors: ASCPT, US Food and Drug Administration (FDA), IQ Consortium, and the International Society of Pharmacometrics (ISoP)

Chairs: Cynthia J. Musante, PhD, Jane Bai, PhD, and Suzana Petanceska, PhD

Earlier versions of Systems Pharmacology



Peter C. Preusch, PhD
Pharmacological Sciences
Training Grant Program Director
NIGMS/NIH
Department of Health and
Human Services
E-mail preuschp@nigms.nih.gov.



LACDR



Universiteit
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Integrative and Organ Systems Pharmacology:

A New Initiative from the National Institute of General Medical Sciences

molecular
interventions

April 2004
Volume 4, Issue 2



BRITISH
PHARMACOLOGICAL
SOCIETY

About the Integrated Systems
Pharmacology Affinity Group

The Integrated Systems Pharmacology Affinity Group serves members who study complex systems or take a whole animal approach to understand drug action or toxicity at molecular, cellular or organ system levels. Those working across several systems, or working in systems without a defined Affinity Group, might be expected to align with Integrated Systems Pharmacology.

INNOVATION

Rescuing drug discovery: *in vivo* systems pathology and systems pharmacology

Jan van der Greef and Robert N. McBurney

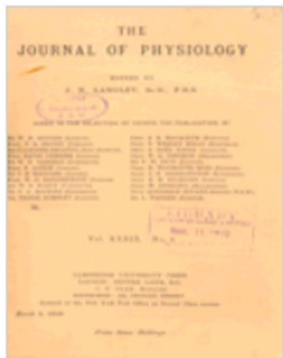
NATURE REVIEWS | DRUG DISCOVERY

VOLUME 4 | DECEMBER 2005 | 061

Even earlier versions of Systems Pharmacology

Volume 39, Issue 6

March 8, 1910

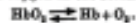


THE NATURE OF OXYHÆMOGLOBIN, WITH A NOTE ON ITS MOLECULAR WEIGHT. BY J. BARCROFT, M.A., B.Sc., *Fellow of King's College, Cambridge,* AND A. V. HILL, B.A., *Scholar of Trinity College, Cambridge.*

(From the Physiological Laboratory, Cambridge.)

¹ The responsibility for the mathematical portion of the work rests with Hill and for the oxygen estimations with Barcroft.

If $[O_2]$ be the concentration of oxygen in the solution we have the equation



Now at the surface of any bubbles of gas we have a diffusion of gas into the bubble from the solution and its rate of diffusion is proportional to $[O_2]$.

Hence if the rate of bubbling is always the same the rate of loss of oxygen from the solution is proportional at any moment to $[O_2]$.

Now let $100x\%$ be the percentage saturation, x the number of molecules of HbO_2 present (assuming the Hb to be present as molecules), a the total number of molecules present whether as Hb or HbO_2 .

The amount of O_2 present = x + the amount in simple solution. The latter term is small compared with the first and may be neglected.

$$\Delta - \frac{dx}{dt} = \text{the rate of loss of oxygen from the solution} = \lambda [O_2]$$

Now the laws of mass action give us

$$-\frac{dx}{dt} \propto k [HbO_2] - k' [Hb][O_2]$$

$$\text{i.e. } -\frac{dx}{dt} = kx [HbO_2] + \frac{\mu k'}{\lambda} [Hb] \frac{dx}{dt}$$

$$\text{i.e. } -\frac{dx}{dt} (\lambda + \mu k' [Hb]) = \mu k k x$$

$$\text{i.e. } -\frac{dx}{dt} = \frac{\mu k k x}{\lambda + \mu k' a - \mu k' x}$$

$$\text{i.e. } + \frac{dx (\mu k' x - \lambda - \mu k' a)}{\mu k k x} = dt$$

$$\text{i.e. } dx \left(\frac{k'}{kx} - \frac{\lambda + \mu k' a}{\mu k k x^2} \right) = dt$$

and integrating, $\left[x \frac{k'}{kx} - \frac{\lambda + \mu k' a}{\mu k k} \log_e x \right] = t$,

where the square brackets signify "between the limits of integration,"

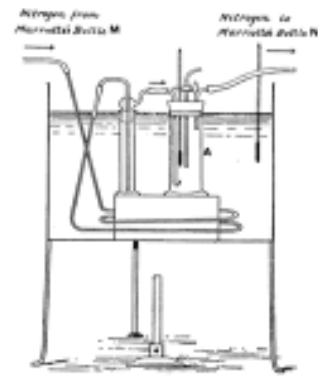


Fig. 1.

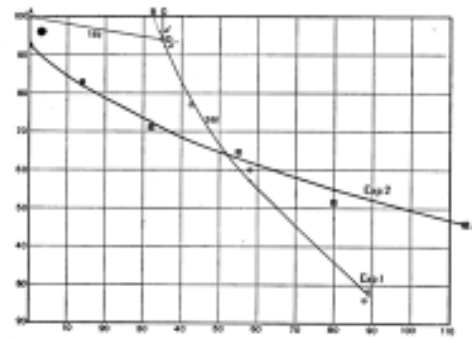
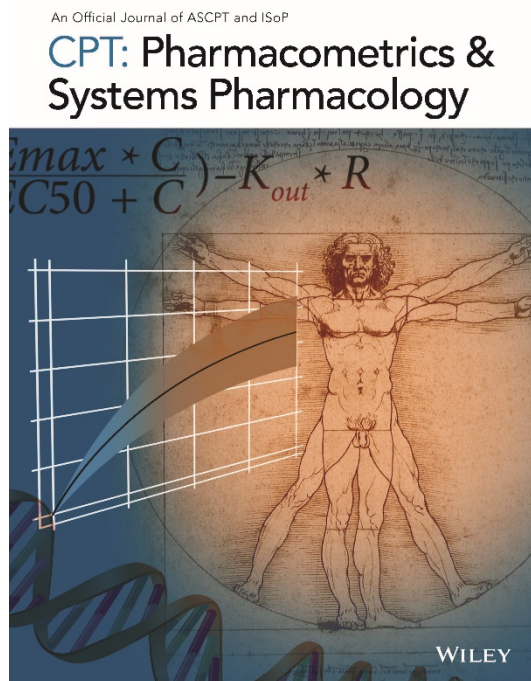


Fig. 2. Curves representing the calculated degree of dissociation at any time in Exps. 1 and 2. Percentage saturation plotted vertically, time in minutes horizontally. The points represent the actual observations.

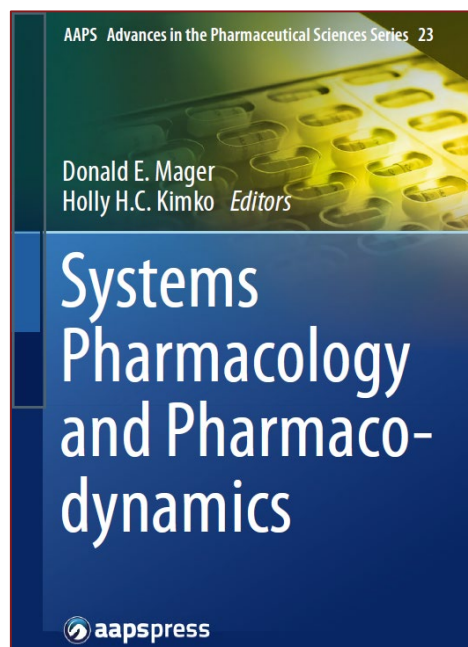
QSP: An established discipline

Journal



2012

Book



2016

Networks and Communities



Member Services > Networks and Communities > Quantitative Pharmacology > S

Systems Pharmacology (SP) Community



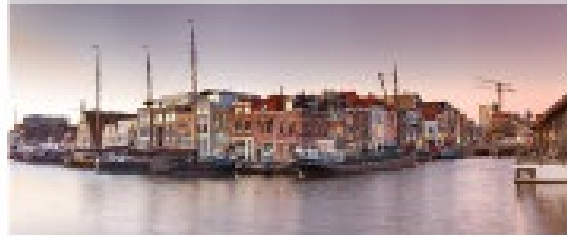
SYSTEMS PHARMACOLOGY COMMUNITY



UK | **Quantitative
Systems
Pharmacology**
NETWORK



...and a **meeting!**



QSPC2020

Quantitative Systems Pharmacology Conference 2020

April 22-24, 2020

Leiden, The Netherlands

www.qspc.eu



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Leiden
The Netherlands



Stadsgehoorzaal

LACDR

Application of QSP in Regulatory Review: FDA

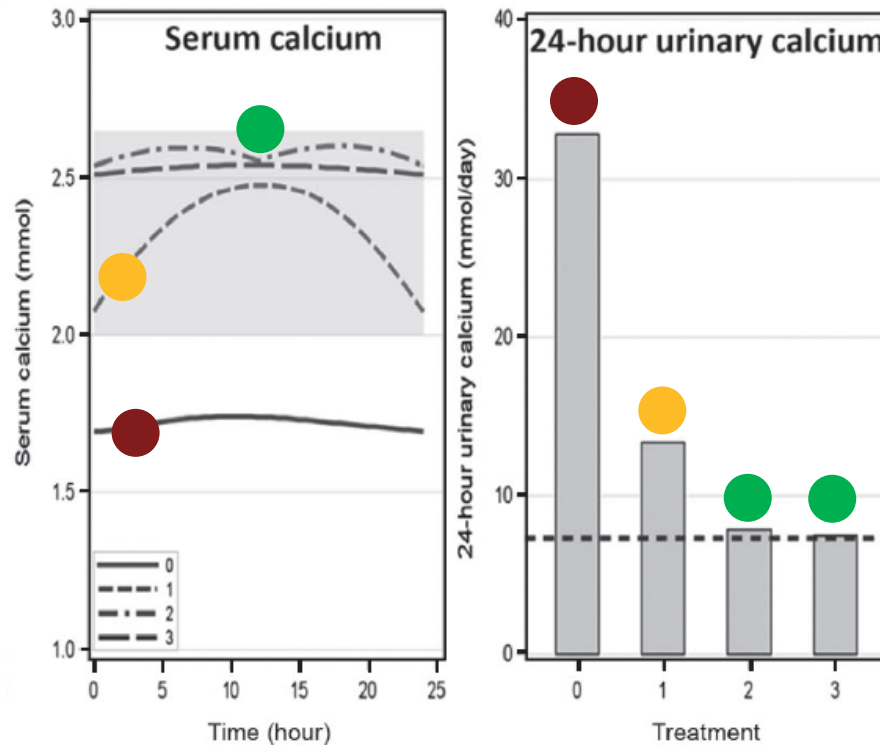
CLINICAL PHARMACOLOGY & THERAPEUTICS

doi:10.1002/cpt.1200

Use of a Systems Pharmacology Model Based Approach Toward Dose Optimization of Parathyroid Hormone Therapy in Hypoparathyroidism

Manoj Khurana¹, Immo Zadezensky^{2*}, Naomi Lowy³, Dragos Roman³, Jean-Marc Guettier^{4,*}, Liang Li¹, Jeffrey Florian³, Chandradas G. Sahajwalla¹, Vikram Sinha^{5,*} and Nitin Mehrotra^{5,*}

- Placebo
- Proposed regimen
- QSP suggested regimen



Application of QSP in Regulatory Review: **EMA**

CLINICAL PHARMACOLOGY & THERAPEUTICS | VOLUME 104 NUMBER 5 | NOVEMBER 2018

The Role of Quantitative Systems Pharmacology in the Design of First-in-Human Trials

Piet H. van der Graaf^{1,2} and Neil Benson²

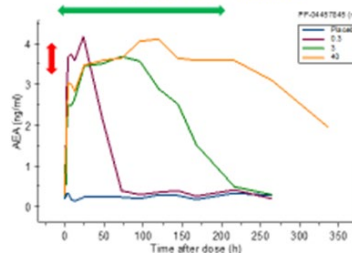
ORIGINAL ARTICLE

A Systems Pharmacology Perspective on the Clinical Development of Fatty Acid Amide Hydrolase Inhibitors for Pain

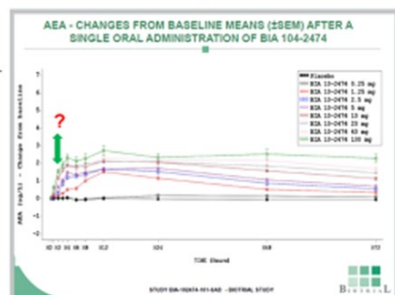
N Benson¹, E Metelkin², O Demin², GL Li², D Nichols⁴ and PH van der Graaf⁵

Elevation of AEA observed in healthy volunteers following a single oral dose

Data consistent with QSP model prediction for selective FAAH inhibition



Data could suggest additional clearance inhibition at higher doses?



Response to: “The Role of Quantitative Systems Pharmacology in the Design of First-in-Human Trials”

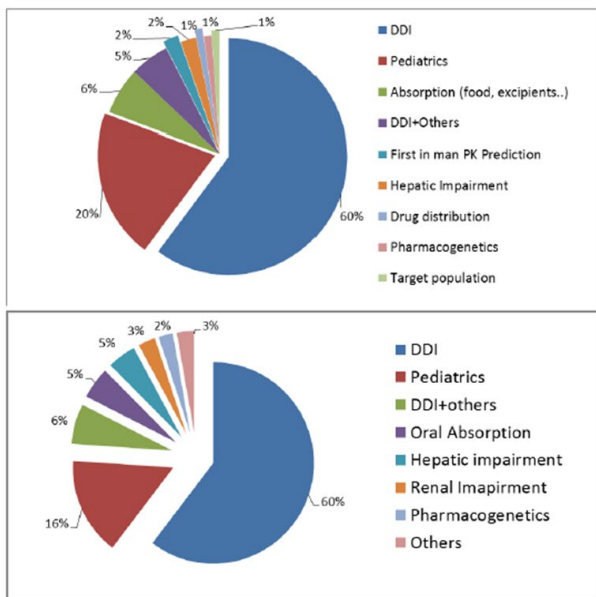
Kevin Blake¹, Milton Bonelli¹, Stefano Ponzano¹, Harald Enzmann², on behalf of the European Medicines Agency Committee for Human Medicinal Products “First-in-Human Guideline Drafting Group”[†]

EMA encourages the development and use of complementary mechanistic models in translational drug research, such as the present example. Guidance on novel technologies can be obtained from EMA under the form of a qualification advice or a qualification opinion (Qualification of novel methodologies for medicine development: http://www.ema.europa.eu/ema/index.jsp?curl=pages/regulation/document_listing/document_listing_000319.jsp&mid=WC0b01ac0580022bb0. Accessed 8 June 2018). Their inclusion in future guidelines will be further considered in line with increasing scientific knowledge.

Still some way to go: comparison with **PBPK**

Currently ~10 FDA submissions containing QSP per year

FDA submissions using PBPK modelling



Cumulative as of June 18, 2014 (n=96)

Sinha, MHRA PBPK Workshop 2014, London, UK

Cumulative as of Aug 1, 2016 (n=217)

41 Labels with *in-silico* substitutes for clinical data informed by Simcyp

Pfizer Revatio (Sildenafil) Pulmonary Arterial Hypertension	Johnson & Johnson Xarelto (Edoxaban) Deep Vein Thrombosis and Pulmonary Embolism	Tibotec Edurant (Rilpivirine) HIV Infection	Ariad Iclusig (Ponatinib) Chronic Myeloid Leukemia	GW Pharma Epidiolex (Cannabidiol) Epilepsy	Lilly Olaniant (Baricitinib) Rheumatoid Arthritis
Novartis Odomzo (Sonidegib) Basal Cell Carcinoma	Janssen Olysio (Simeprevir) Hepatitis C	Actelion Opsumit (Macitentan) Pulmonary Arterial Hypertension	Pharmacyclics Imbruvica (Brintnib) Mantle Cell Lymphoma and Chronic Lymphocytic Leukemia	AstraZeneca Movantik (Naloxegol) Opioid Induced Constipation	Genentech Cotelic (Cobimetinib) Metastatic Melanoma
Genzyme Cardega (Egflustat) Gaucher Disease	Sanofi Jevtana (Cabazitaxel) Prostate Cancer	Novartis Zykadia (Ceritinib) Metastatic Non-small Cell Lung Cancer	Pfizer Bosulif (Bosutinib) Chronic Myelogenous Leukemia	Alkermes Aristada (Aripiprazole lauroxil) Schizophrenia	AstraZeneca Lynparza (Olaparib) Advanced Ovarian Cancer
Novartis Farydak (Panobinostat) Multiple myeloma	Eisai Lenvima (Lenvatinib) Thyroid cancer	Genentech Alecensa (Alectinib) Non-small Cell Lung Cancer	AstraZeneca Tagrisso (Osimertinib) Metastatic NSCLC	Amgen Blincyto (Binatumomab) Acute Lymphoblastic Leukemia	AstraZeneca Calquence (Acalabrutinib) Mantle Cell Lymphoma
Eli Lilly Verzenio (Abemaciclib) Metastatic Breast Cancer	Intercept Ocaliva (Obecholic acid) Primary Biliary Cholangitis	Actelion Uptravi (Sellemping) Pulmonary Arterial Hypertension	Janssen Invokana (Canagliflozin) Type 2 Diabetes	Merck Prexvy (Letemovir) Cytomegalovirus	Merck Steglan (Ertugliflozin) Type 2 Diabetes
Novartis Kisqali (Ribociclib succinate) Metastatic Breast Cancer	PTC Therapeutics Emflaza (Deflazacort) Duchenne Muscular Dystrophy	Shionogi Symproic (Naldemedine) Opioid Induced Constipation	Spectrum Beloadaq (Belinostat) Peripheral T-cell Lymphoma	UCB Briviact (Brivaracetam) Epilepsy	Vertex Symdeko (Tezacaftor/cysteamine) Cystic Fibrosis
Novartis Rydapt (Midostaurin) Acute Myeloid Leukemia	Ariad Alunbrig (Brigatinib) Metastatic Non-small Cell Lung Cancer	Janssen Erlada (Apatinib) Non-metastatic Prostate Cancer	Helsinn Alyxzo (Ivosidenib) Acute and Delayed Neutropenia	AkaRx Doptelet (Avatrombopag mesilate) Thrombocytopenia	

Majority related to drug-drug interactions (DDIs, ~ 60%); pediatrics ranks the second

Ping Zhao



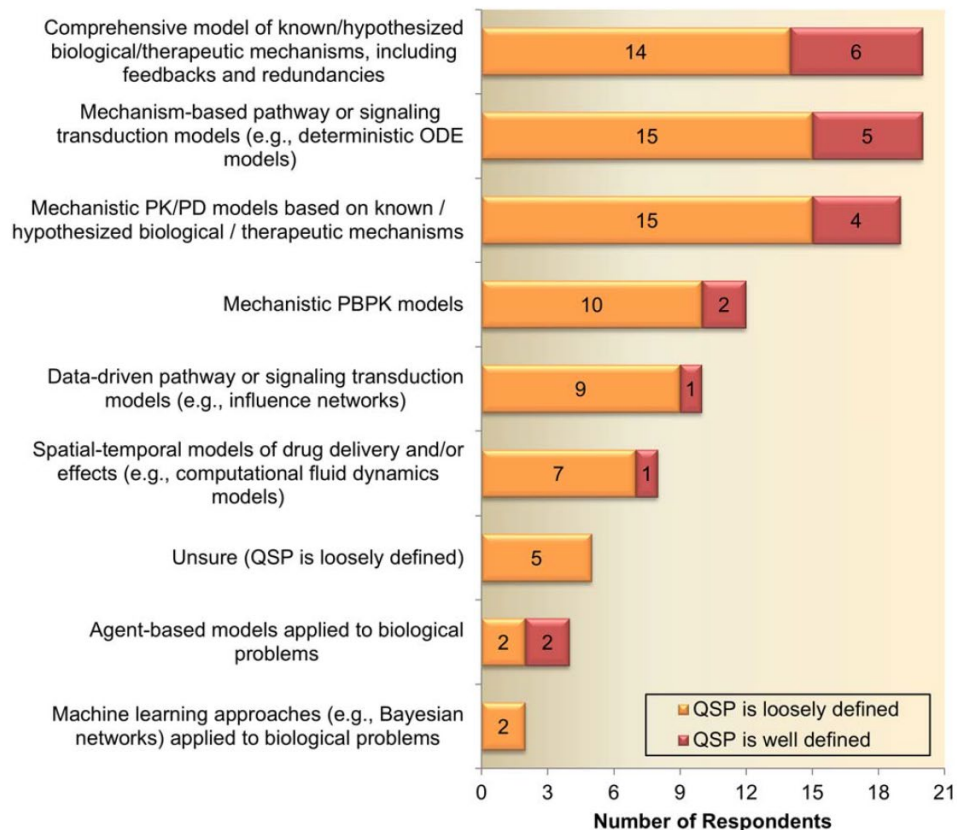
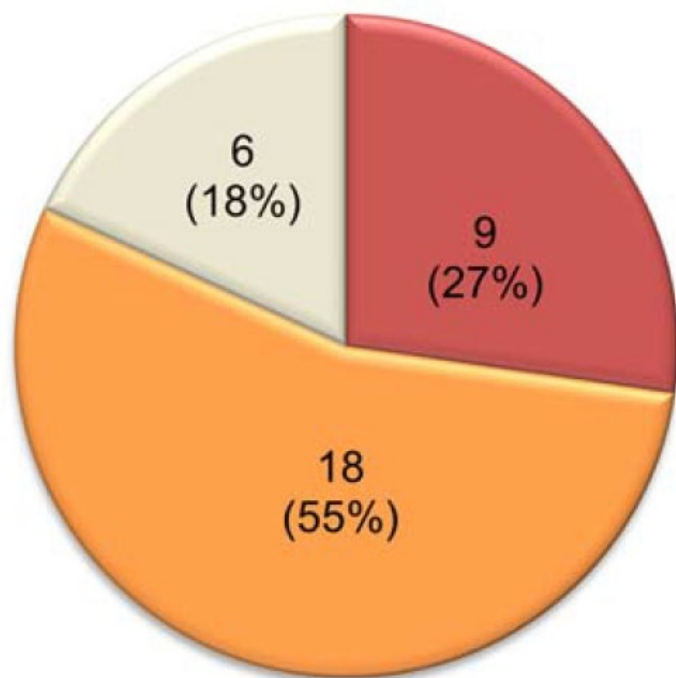
QSP: Dose (regimen) prediction and Target validation?

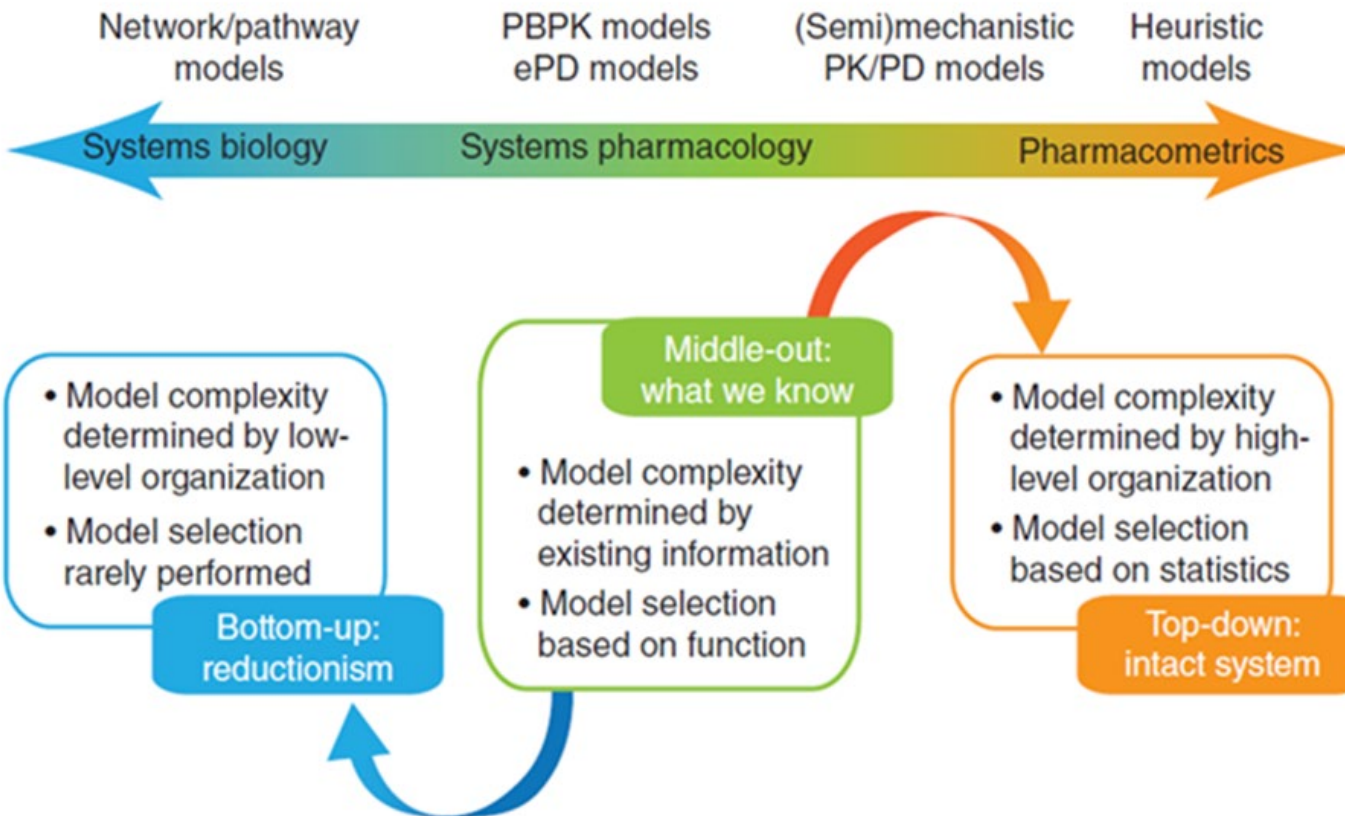
QSP definition: confusion?

REVIEW

Preclinical QSP Modeling in the Pharmaceutical Industry: An IQ Consortium Survey Examining the Current Landscape

Use & Clarity of QSP





nature publishing group

PERSPECTIVES

COMMENTARIES

Systems Pharmacology for Drug Discovery and Development: Paradigm Shift or Flash in the Pan?

P Vicini¹ and PH van der Graaf^{1,2}

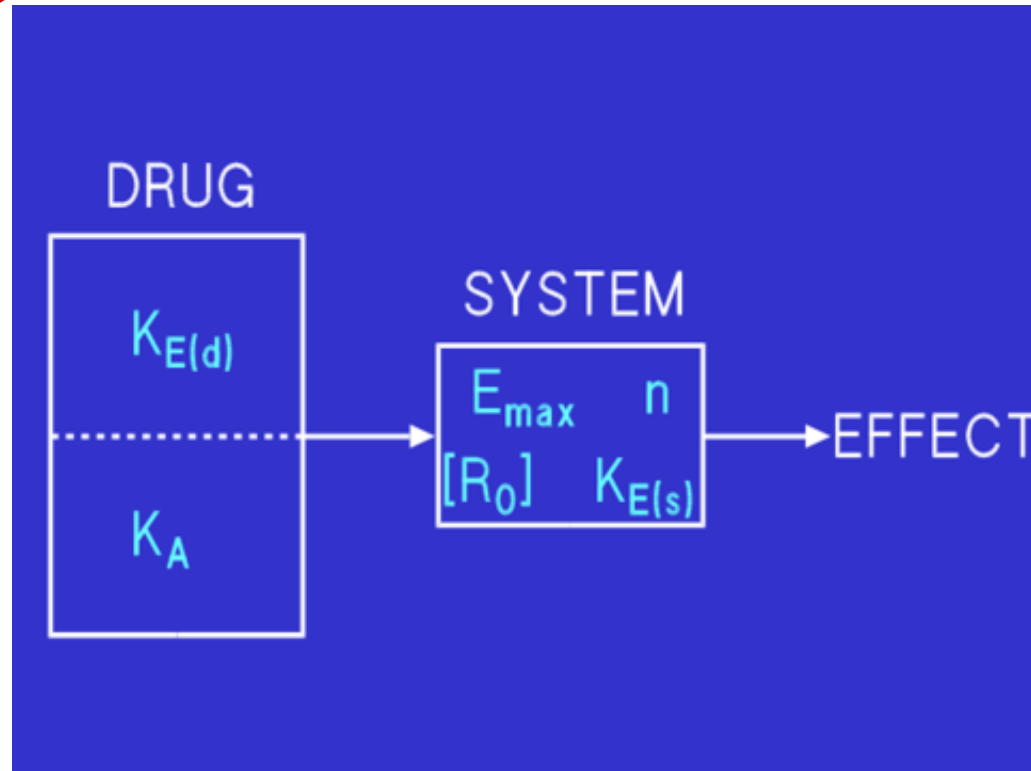
Systems pharmacology is an emerging approach that sets out to use quantitative concepts rooted in the synergy between modeling and simulation on one hand and large-scale data collection and analysis on the other to investigate biological systems and design therapeutic interventions. Interest in this field has recently increased substantially, but so have perceived challenges and misunderstandings, especially related to implementation. This Commentary explores the opportunities and challenges in the rapidly evolving area of systems pharmacology from a drug discovery and development perspective.

Statistical regression techniques teach us that the more components a model has, the greater its explanatory power will be; however, the individual contributions of those components will become more and more difficult to discern. Systems-pharmacology models attempt to inject biological realism early in the process—not necessarily with the intent to build a “bedfellow out of quarks” but, rather, to bring molecular or cellular detail closer to high-level, functional behavior. Systems pharmacology was defined in a recent white paper by the National Institutes of Health (NIH) Quantitative Systems Pharmacology (QSP) workshop group¹ as “an approach to translational medicine that combines computational and experimental methods to elucidate, validate and apply new pharmacological concepts to the development and use of small molecule and biologic drugs [with the aim of] determining mechanisms of action of new and existing drugs in preclinical and animal models and in patients.”

QSP



Extrapolation



PMx



Interpolation

EXAMPLES

- Efficacy/safety of novel MOA
- Preclinical \rightarrow clinical (dose; schedule)
- Novel combination therapy
- Biomarker identification
- Novel patient populations
- Novel indications

QSP case study: Immunogenicity (IG)



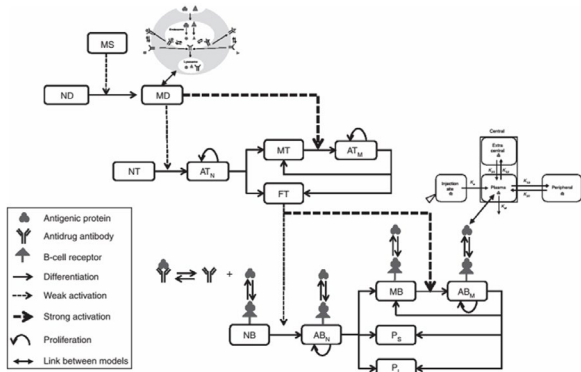
Citation: CPT Pharmacometrics Syst. Pharmacol. (2014) 3, e134; doi:10.1038/psp.2014.31
 © 2014 ASCPT All rights reserved 2163-8306/14

www.nature.com/psp

ORIGINAL ARTICLE

A Mechanistic, Multiscale Mathematical Model of Immunogenicity for Therapeutic Proteins: Part 2—Model Applications

X Chen¹, TP Hickling² and P Vicini³



IG Model



IG Simulator

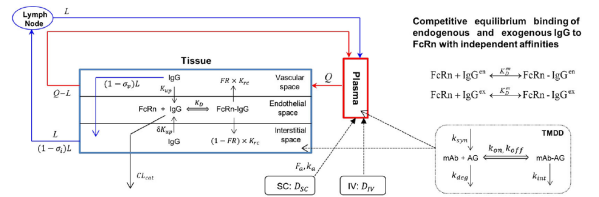


The AAPS Journal, Vol. 16, No. 5, September 2014 (© 2014)
 DOI: 10.1208/s12248-014-9640-5

Research Article

Simulation of Monoclonal Antibody Pharmacokinetics in Humans Using a Minimal Physiologically Based Model

Linzhong Li,^{1,3} Iain Gardner,¹ Miroslav Dostalek,² and Masoud Jamei¹



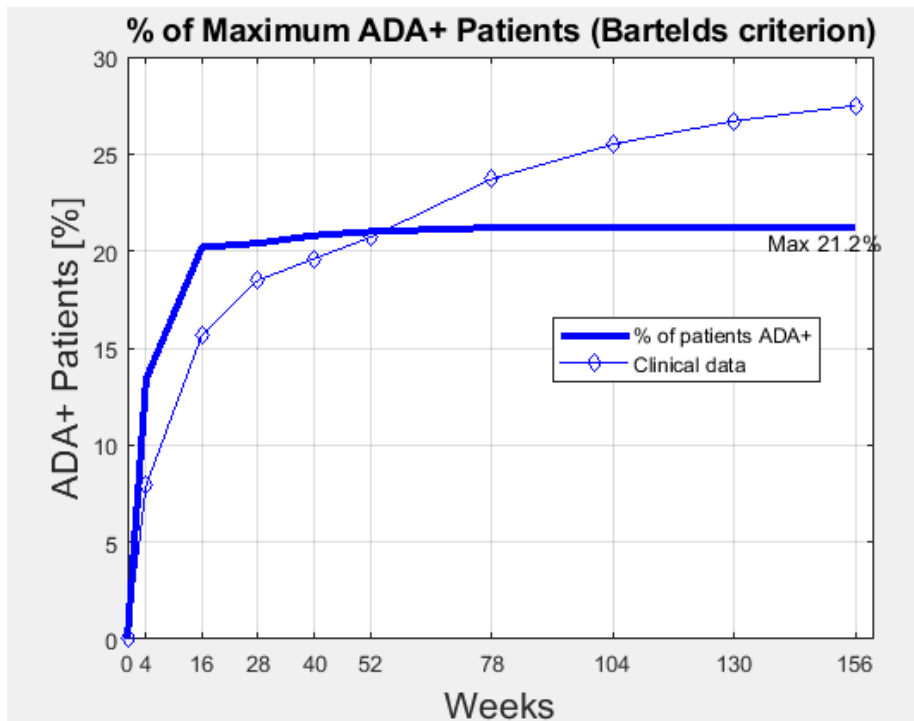
Bioinformatics predictions

Number of epitopes = 2

MHC Class II allele	Allele frequency (North American population)	Epitope 1 binding affinity (include BC loop region)	Epitope 2 binding affinity (include FG loop loop region)
DPA1*01:03/DPB1*02:01	0.5200	203	4000
DPA1*01:01/DPB1*04:01	0.2364	426	4000
DRB1*03:01	0.0689	472	4000
DRB1*11:01	0.0472	278	4000
DRB1*15:01	0.0696	447	218
DRB1*04:05	0.0231	4000	118
DRB1*07:01	0.0872	4000	161
DRB1*08:02	0.0742	4000	854
DRB1*13:02	0.0394	4000	168

Example: Adalimumab

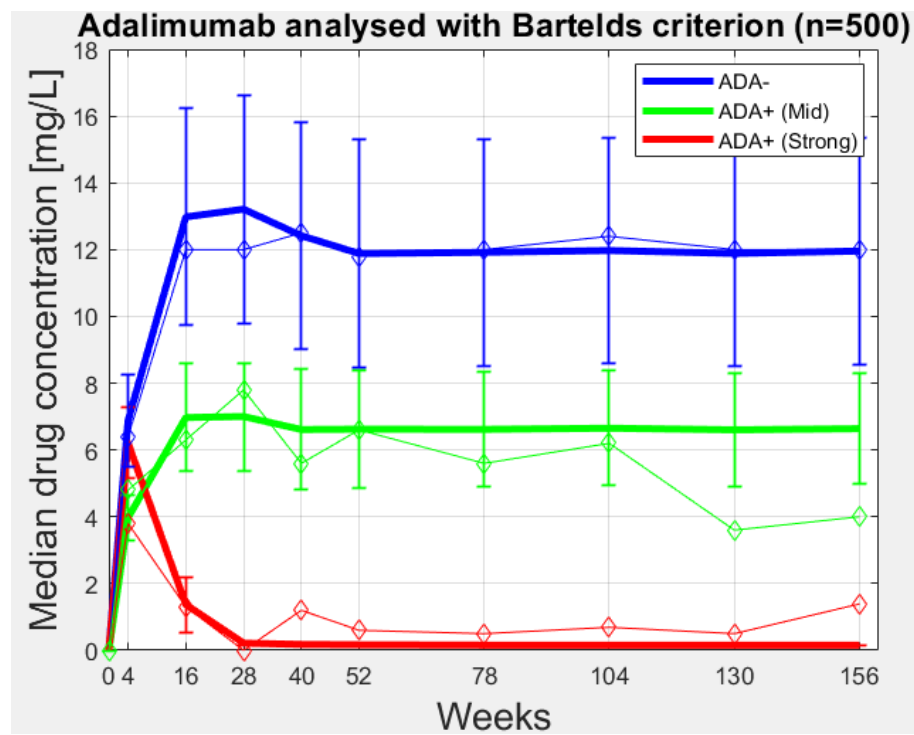
ADA response



(error bars = median absolute deviation)

ADA+ Cohort
ADA+ 12-100AU/ml
Strong ADA+ >100AU/ml

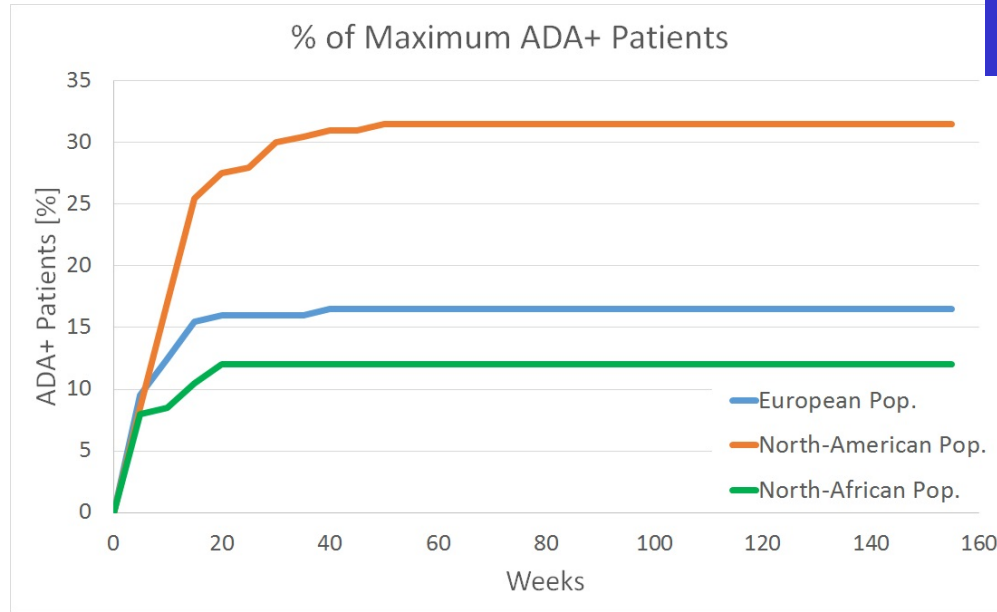
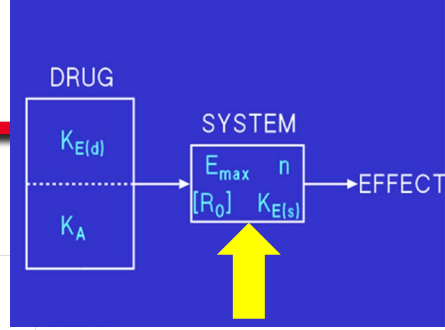
Pharmacokinetics



Model	Data
60%	59%
40%	41%

Change System

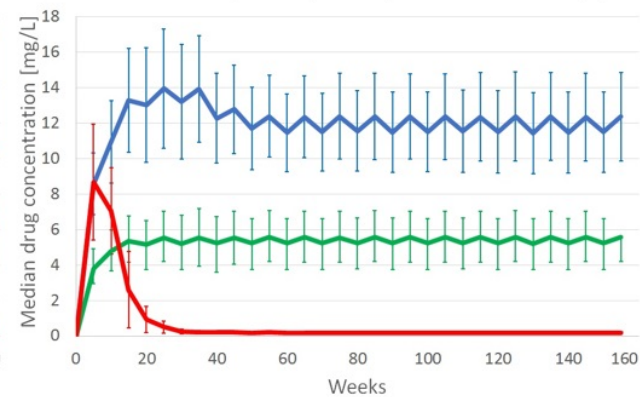
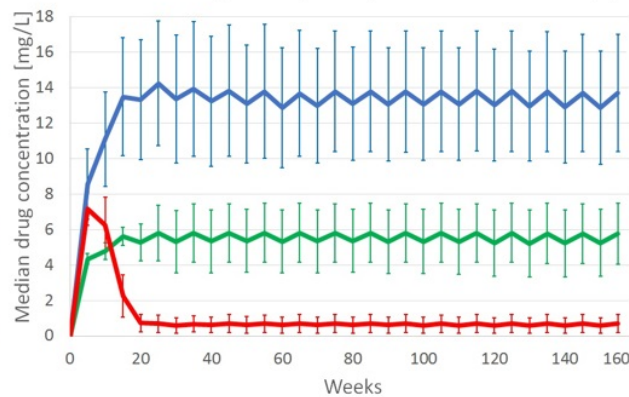
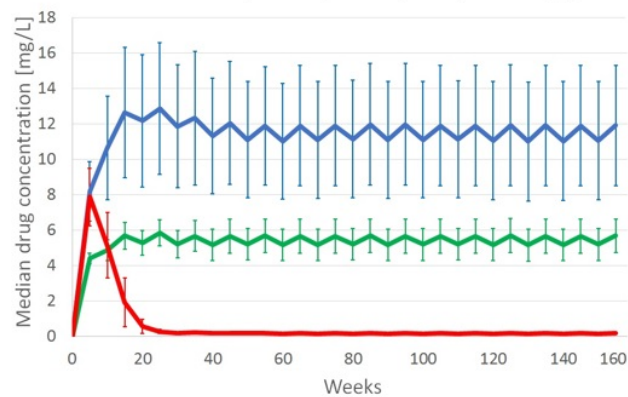
Compound X (adalimumab-like)



Generic compound (n=200, European Pop.)

Generic compound (n=200, North-American Pop.)

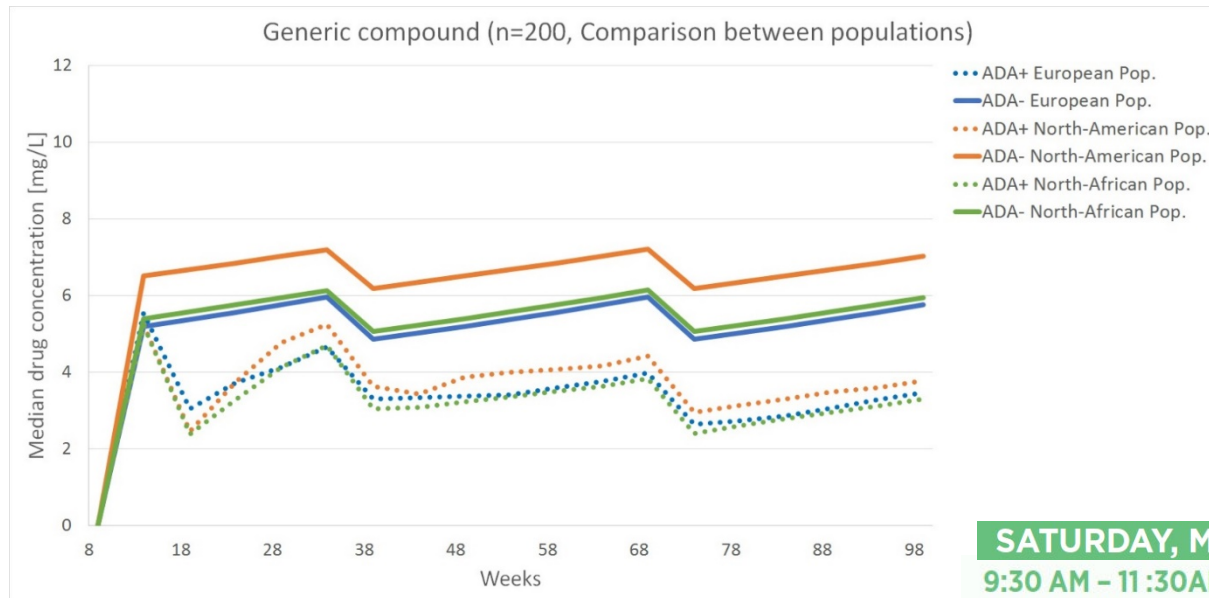
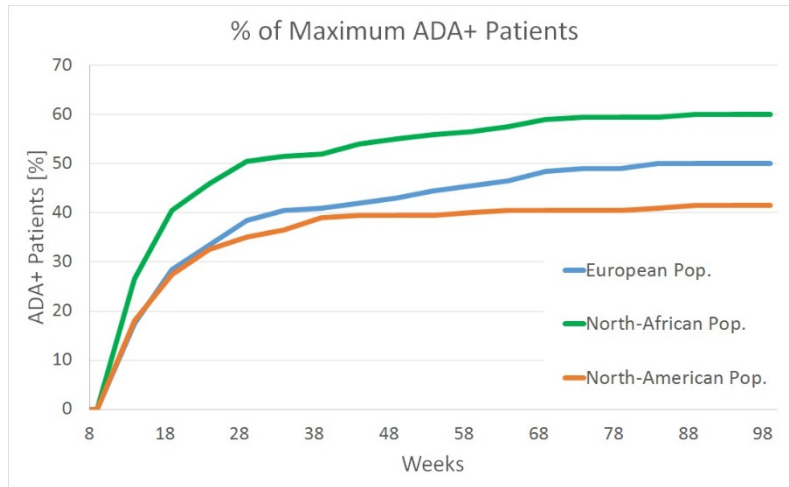
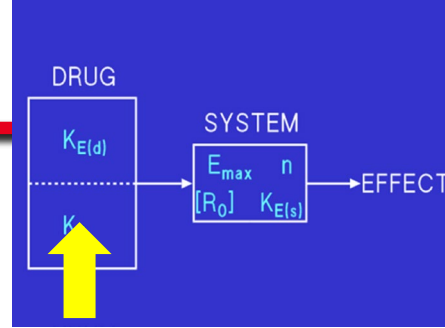
Generic compound (n=200, North-African Pop.)



— ADA- — ADA+ (Mid) — ADA+ (Strong)

Change Compound

Compound Y



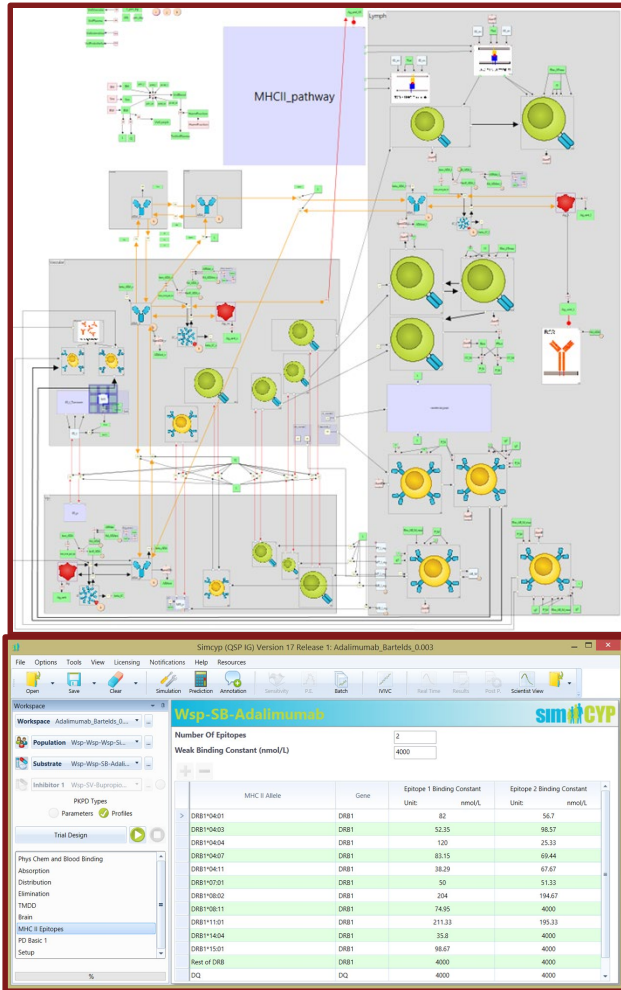
SATURDAY, MARCH 16, 2019

9:30 AM - 11:30 AM

SYMPOSIUM

Immunogenicity in Clinical Practice and Drug Development: When is it Significant?

Applications: Extrapolation



Extrapolation to **population** with different HLA allele frequencies.

Personalised & Precision medicine: Prediction of PK and IG for **genotyped individual**.

Extrapolation to larger populations. (**Phase III, IV**)

IG Management: Extrapolation to different **dosing regimes**.

Extrapolation to **paediatric population** or individual children.

Extrapolation to **disease population**.

Extrapolation to **age group**.

Prediction of the effect of **co-therapy**

Some additional **Points for Consideration**

- Open Source?
- Experimentation?
- Pharmacometrics \leftrightarrow QSP?
-

Open Source?

Reproducibility of Quantitative Systems Pharmacology (QSP) Models: Current Challenges and Future Opportunities

Daniel C. Kirouac^{1,3,*}, Brian Cicali^{2,3} and Stephan Schmidt^{2,3}

ABSTRACT

Provision of model code is required for publication in *CPT:PSP*, enabling QSP model *availability*. A searchable repository of published QSP models would enhance model *accessibility*. We assess the feasibility of establishing such a resource, based on 18 QSP models published in this journal. However, due to the diversity of software platforms (9), file formats and functionality, such a resource is pre-mature. We evaluated 12 of the models (those coded in R, PK-Sim/MOBI and Matlab) for functionality. **Of the 12, only 4 were executable**, in that figures from the associated manuscript could be generated via a 'run' script. Many researchers are aware of the challenges involved in re-purposing published models. We offer some ideas to enable model sharing going forward, including annotation guidelines, standardized formats, and inclusion of 'run' scripts. If practitioners can agree to some minimum standards for provision of model code, this would accelerate model re-use and extension.

Open Science

- Sharing knowledge/resources/data
- Consensus on science and common tools/standards
- Engaging pharma, academics and regulators through publication & education
- Rigorous approach to IT and QA: interoperable, future proof and meeting regulatory requirements
- Ability to integrate confidential data/knowledge/models in a seamless and incremental manner



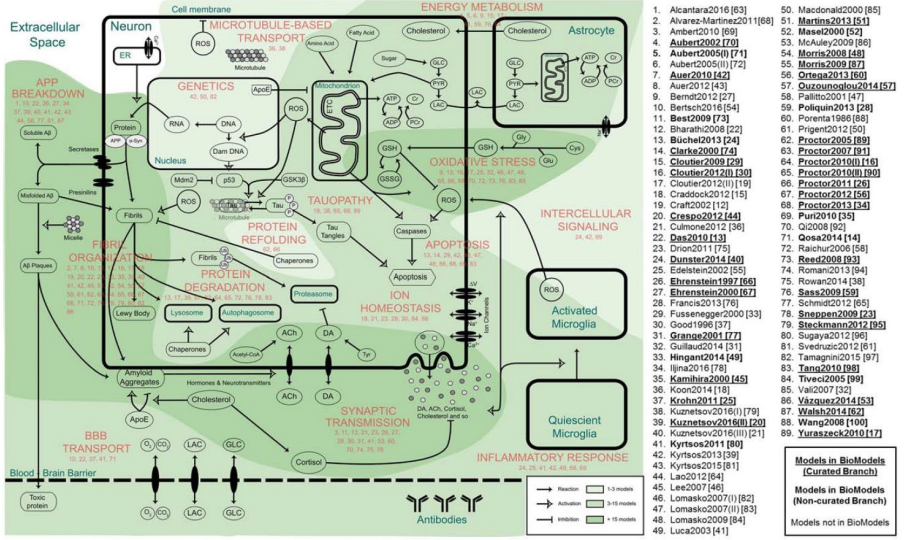
EUROPEAN MEDICINES AGENCY
SCIENCE MEDICINES HEALTH

13 December 2018
EMA/CHMP/458101/2016
Committee for Medicinal Products for Human Use (CHMP)

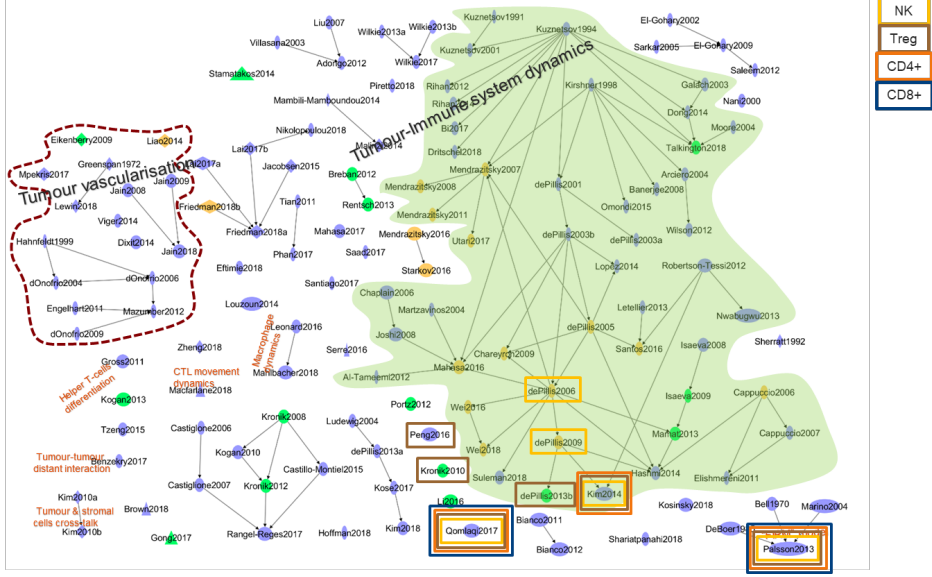
Guideline on the reporting of physiologically based pharmacokinetic (PBPK) modelling and simulation



Neurodegeneration



Immuno-Oncology



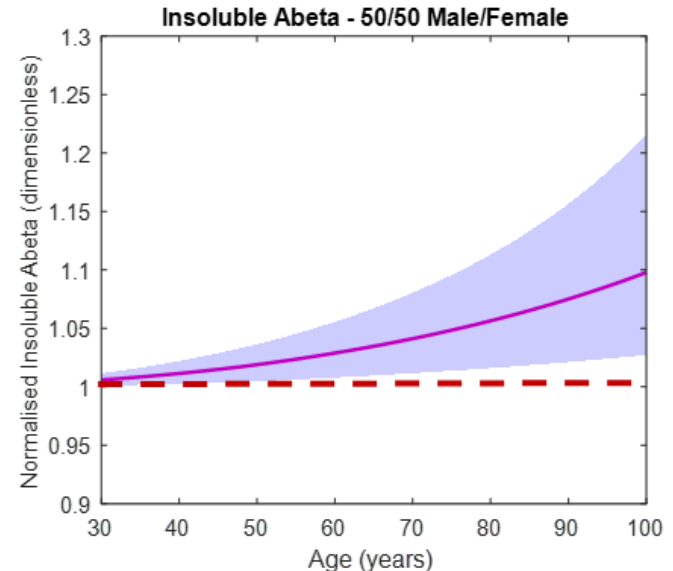
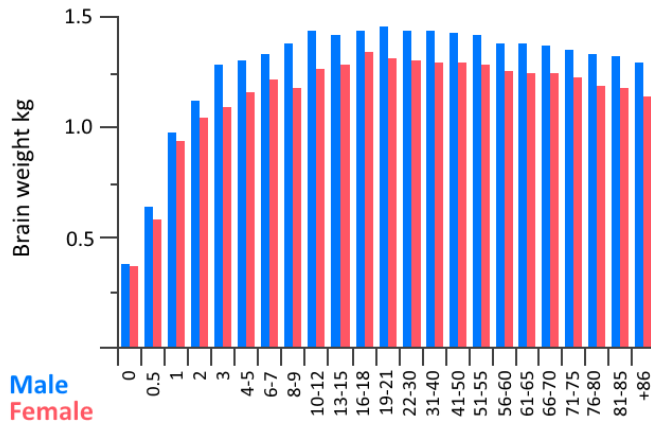
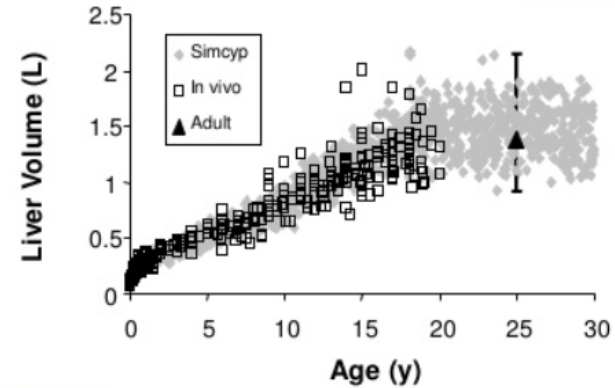
REVIEW
The Impact of Mathematical Modeling in Understanding the Mechanisms Underlying Neurodegeneration: Evolving Dimensions and Future Directions

A Lloret-Villas¹, TM Varusai¹, N Juty¹, C Laibe¹, N Le Novère², H Hermjakob¹ and V Chelliah¹



Open Science example: organ size

- Impact of variable organ size (age, gender etc.) well-established in PBPK
- Similar scholarship needs to be developed for QSP
- Example: Brain volume in models of AD



Open Science example: target expression

ORIGINAL ARTICLE

Application of a systems pharmacology model for translational prediction of hERG-mediated QTc prolongation

Verena Gotta^{1,2}, Zhiyi Yu³, Frank Cools⁴, Karel van Ammel⁴, David J. Gallacher⁴, Sandra A. G. Visser⁵, Frederick Sannajust⁶, Pierre Morissette⁶, Meindert Danhof¹ & Piet H. van der Graaf^{1,7}

1.2. External evaluation of translational predictions (*sotalol* & *moxifloxacin*)

Good clinical predictions in adults and children were obtained (<5-10 ms prediction discrepancy from clinical regression model until ΔQT_c of 35 ms). However, **QTc- effects in neonates were under-predicted** (>20 ms prediction discrepancy).

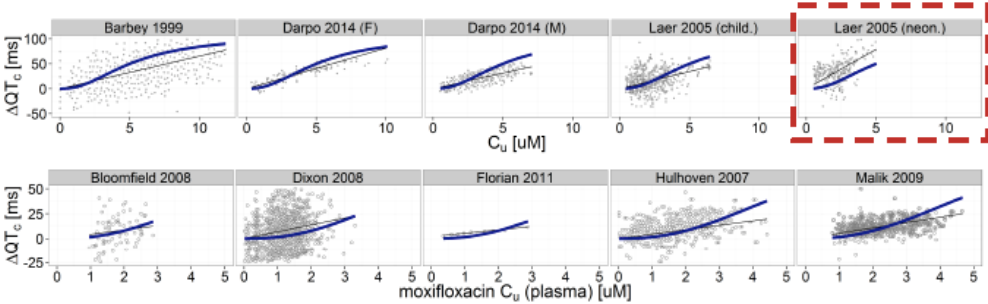


Fig.2: Translational predictions from preclinical data and system-specific scaling parameters only (blue lines) are contrasted with reported clinical ΔQT_c from indicated references (grey dots: digitized observations. black lines: predictions from respective clinical regression model).

Rothmond et al. BMC Neuroscience 2012, 13:18
<http://www.biomedcentral.com/1471-2202/13/18>



RESEARCH ARTICLE Open Access

Developmental changes in human dopamine neurotransmission: cortical receptors and terminators

Debora A Rothmond^{1,2,3*}, Cynthia S Weickert^{1,2,3} and Maree J Webster⁴

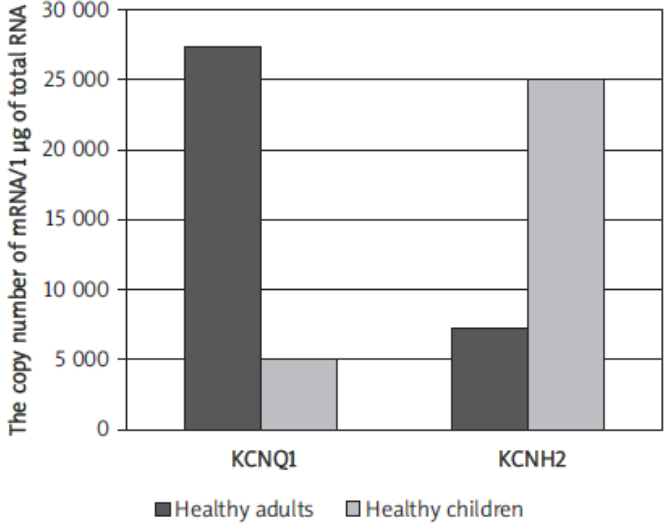
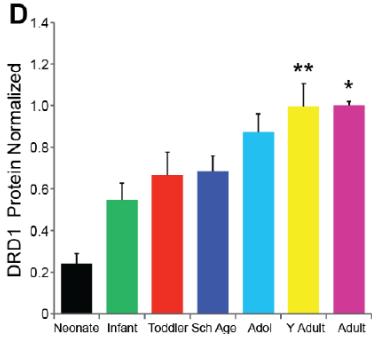


Figure 3. Comparison of KCNQ1 and KCNH2 mRNA levels between healthy adults and healthy children. Results are expressed as copy numbers per 1 μ g of total RNA

Arch Med Sci 2011; 7, 6: 941-947
 DOI: 10.5114/AOMS.2011.26604
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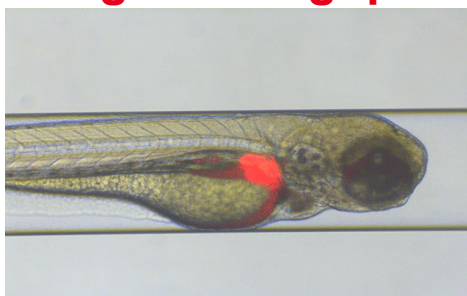
Experimentation?

Quantitative and Systems Pharmacology in the Post-genomic Era: New Approaches to Discovering Drugs and Understanding Therapeutic Mechanisms

An NIH White Paper by the QSP Workshop Group – October, 2011

Definitions: QSP is defined as an approach to translational medicine that combines computational and experimental methods to elucidate, validate and apply new pharmacological concepts to the development and use of small molecule and biologic drugs. QSP will provide an integrated “systems-level” approach to determining mechanisms of action of new and existing drugs in preclinical and animal models and in patients. QSP will create the knowledge needed to change complex cellular networks in a specified way with mono or combination therapy, alter the pathophysiology of disease so as to maximize therapeutic benefit and minimize toxicity and implement a “precision medicine” approach to improving the health of individual patients.

High Throughput



High Content

British Journal of Pharmacology (2018) ••••• 1

RESEARCH PAPER

Fingerprints of CNS drug effects: a plasma neuroendocrine reflection of D₂ receptor activation using multi-biomarker pharmacokinetic/pharmacodynamic modelling

Correspondence Elizabeth CM de Lange, Division of Systems Biomedicine and Pharmacology, Leiden Academic Center for Drug Research, Leiden University, Leiden, The Netherlands. E-mail: ecmde Lange@act.leidenuniv.nl

Received 2 January 2018; Revised 6 July 2018; Accepted 11 July 2018

Willem J van den Brink¹, Dick-Jan van den Berg¹, Floor E M Bonse¹, Robin Hartman¹, Yin-Cheong Wong¹, Piet H van der Graaf^{1,2} and Elizabeth CM de Lange¹



Cybernetics, Redux: An Outside-In Strategy for Unraveling Cellular Function

Mohan Malleshaiah¹ and Jeremy Gunawardena^{1,*}

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<http://dx.doi.org/10.1016/j.devcel.2015.12.025>

² Developmental Cell 36, January 11, 2016 ©2016 Elsevier Inc.

Preclinical QSP Modeling in the Pharmaceutical Industry: An IQ Consortium Survey Examining the Current Landscape

This lack of dedicated experimental support is currently a gap within most companies and may hinder the successful implementation of QSP models, especially when such mod-

British Journal of Pharmacology (2018) ••••• 1

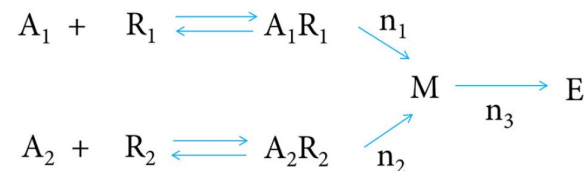
RESEARCH PAPER

Quantitative systems pharmacology analysis of drug combination and scaling to humans: the interaction between noradrenaline and vasopressin in vasoconstriction

Correspondence Piet H van der Graaf, Certara QSP, Canterbury Innovation Centre, Unit 43, University Road, Canterbury CT2 7FG, UK. E-mail: piet@certara.com

Received 18 May 2018; Accepted 27 May 2018

Anyue Yin^{1,2}, Akihiro Yamada^{1,3}, Wiro B Stam¹, Johan G C van Hasselt¹ and Piet H van der Graaf^{1,5}



Slope	
N _H (observed)	2.3-2.7
N ₁ & N ₂	~0.5-0.7
N ₃	4

Pharmacometrics ↔ QSP?

EDITORIAL

Pharmacometrics and/or Systems Pharmacology

Piet H. van der Graaf^{1,2,*}

COMMENTARY

QSP Versus the Rest: Let the Competition Commence!

Hitesh B. Mistry^{1,2}

PERSPECTIVE

Quantitative Systems Pharmacology and Empirical Models: Friends or Foes?

Neil Benson^{1,*}

PERSPECTIVE

Benchmarking QSP Models Against Simple Models: A Path to Improved Comprehension and Predictive Performance

Andrew M. Stein^{1*} and Michael Looby²

My model is better than yours

Use of a Systems Pharmacology Model Based Approach Toward Dose Optimization of Parathyroid Hormone Therapy in Hypoparathyroidism

Manoj Khurana¹, Immo Zadezensky^{2,*}, Naomi Lowy³, Dragos Roman³, Jean-Marc Guettier^{4,*}, Liang Li¹, Jeffrey Florian³, Chandrahas G. Sahajwalla¹, Vikram Sinha^{5,*} and Nitin Mehrotra^{5,*}

pharmacodynamics (PDs) of PTH dose and dosing regimen. **Although other modeling approaches may be feasible, in this specific case,** QSP model-based simulations fulfilled the information gap to support recommendations of this postmarketing trial.

Pharmacometrics ↔ QSP?

Integrated Pharmacometrics and Systems Pharmacology (iPSP)

PERSPECTIVE

Perspective on the State of Pharmacometrics and Systems Pharmacology Integration

Mirjam N. Trame¹, Matthew Riggs², Konstantinos Biliouris¹, Dhananjay Marathe³, Jerome Mettetal⁴, Teun M. Post^{5,6}, Matthew L. Rizk⁷, Sandra A. G. Visser⁸ and Cynthia J. Musante^{9*}

228 PSP papers 2012-2017:
19% iPSP

Linking QSP to Clinical Endpoints

Citation: CPT: Pharmacometrics & Systems Pharmacology (2012) 1, e11; doi:10.1038/psp.2012.10
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www.nature.com/psp

Integrated Pharmacometrics and Systems Pharmacology Model-Based Analyses to Guide GnRH Receptor Modulator Development for Management of Endometriosis

MM Riggs¹, M Bennetts², PH van der Graaf² and SW Martin³



Model Reduction

Snowden et al. BMC Systems Biology (2017) 11:17
DOI 10.1186/s12918-017-0397-1

BMC Systems Biology

METHODOLOGY ARTICLE

Open Access

A combined model reduction algorithm for controlled biochemical systems

Thomas J. Snowden^{1,2}, Piet H. van der Graaf^{3,2} and Marcus J. Tindall^{1,4*}

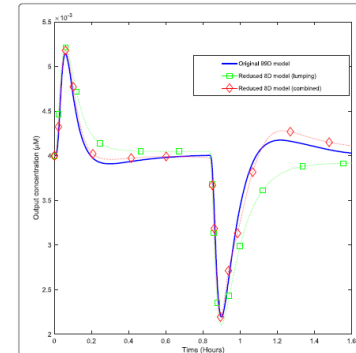
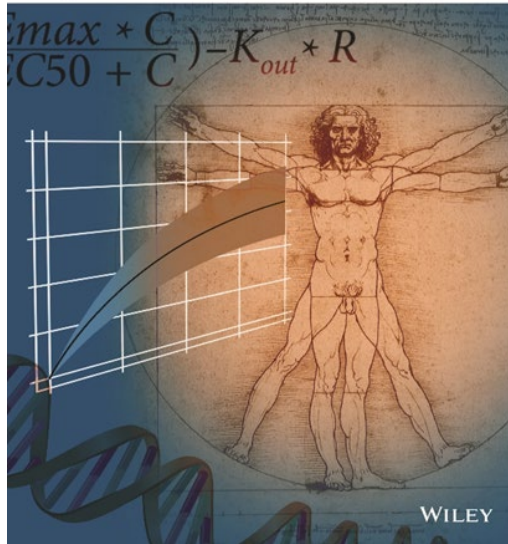


Fig. 6 Timecourses of the output from the original 99-dimensional and the reduced 8-dimensional systems. This plot emphasises the fact that the reduced model is designed to remain valid for any reasonable change in input. The system starts by being affected by an agonist that increases the rate of EGF binding by 25% for 50 minutes, at this point the input flips to an antagonist decreasing the rate of EGF binding by 50% and runs for the same time period. At any given time point the error between the original and reduced model exceeds no more than 5%.

CPT: PSP 2.0

An Official Journal of ASCPT and ISO-P

CPT: Pharmacometrics &
Systems Pharmacology



EDITORIAL

CPT: Pharmacometrics & Systems Pharmacology 2.0

France Mentre^{1,*}

Backups

PBPK guidelines a template for QSP?



EUROPEAN MEDICINES AGENCY
SCIENCE MEDICINES HEALTH

13 December 2018
EMA/CHMP/458101/2016
Committee for Medicinal Products for Human Use (CHMP)

Guideline on the reporting of physiologically based
pharmacokinetic (PBPK) modelling and simulation



Adobe Acrobat
Document

PBPK guidelines a template for QSP?

- 4. Reporting of PBPK modelling and simulation.....**
- 4.1. Objective and regulatory purpose
- 4.2. Background information
- 4.3. Qualification
- 4.4. Model parameters.....
- 4.4.1. Assumptions
- 4.4.2. System-dependent parameters
- 4.4.3. Drug parameters and the drug model.....
- 4.5. Model development
- 4.6. Simulation of the intended scenario
- 4.7. Platform and drug model evaluation
- 4.7.1. Sensitivity analyses
- 4.7.2. Evaluation of the predictive performance of the drug model
- 4.8. Results.....
- 4.9. Discussion of the regulatory application

Model reduction

Snowden et al. *BMC Systems Biology* (2017) 11:17
DOI 10.1186/s12918-017-0397-1

BMC Systems Biology

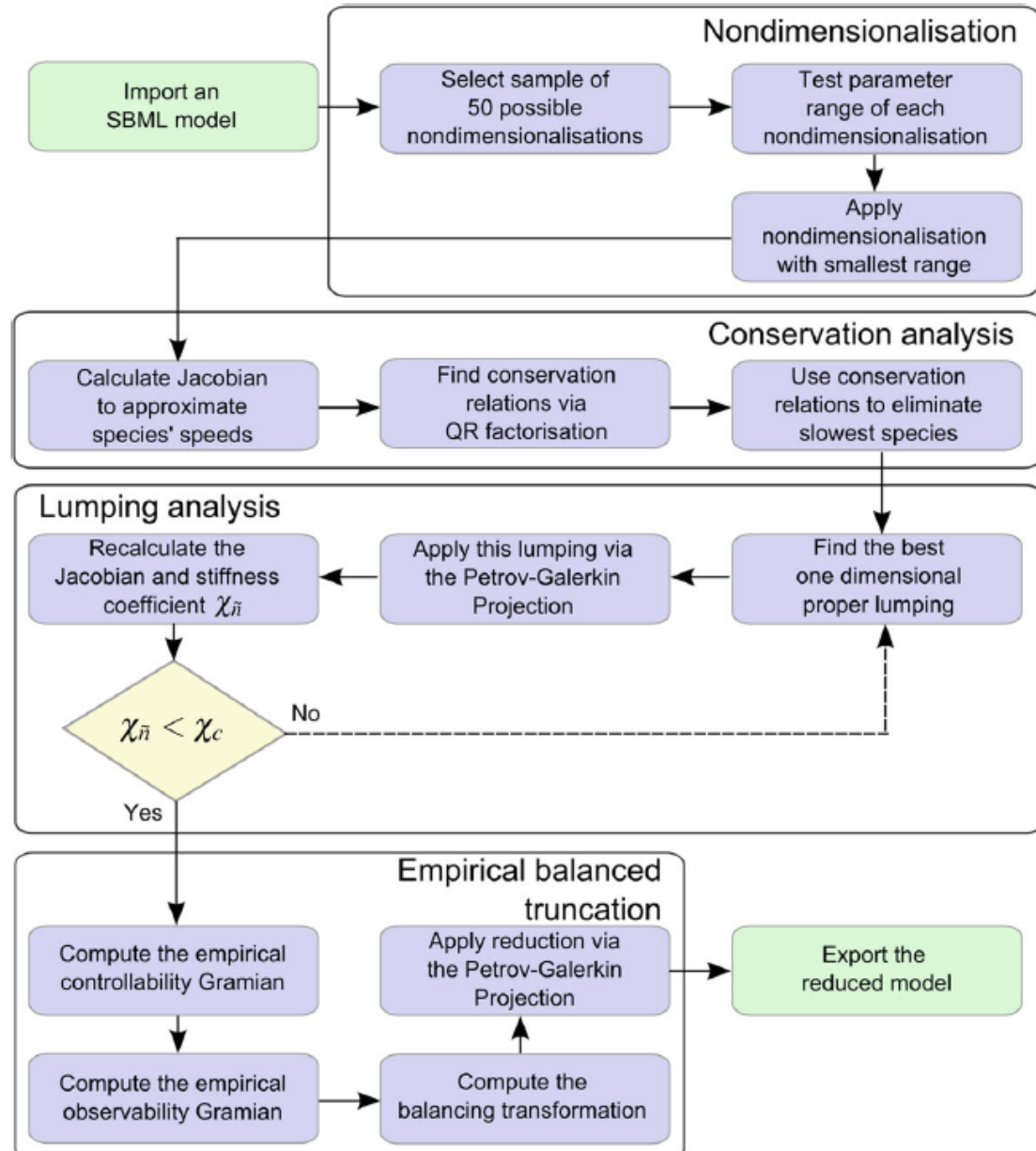
METHODOLOGY ARTICLE

Open Access

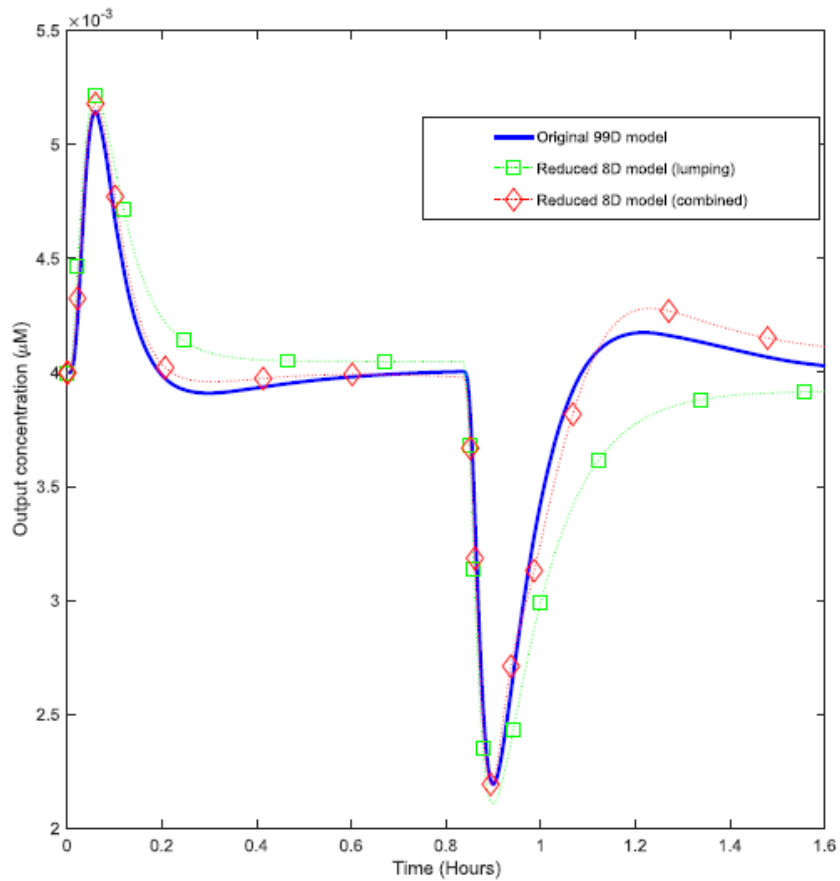
A combined model reduction algorithm for controlled biochemical systems



Thomas J. Snowden^{1,2}, Piet H. van der Graaf^{3,2} and Marcus J. Tindall^{1,4*}



NGF model reduction



Dimension	EBT error	Lumping error	Stiffness	Combined error
75	0.76%	≈ 0%*	42658	—
50	#	0.01%	42633	—
25	#	0.52%	10664	—
15	#	1.26%	7934	—
14	#	2.21%	7934	—
13	#	2.29%	7934	—
12	#	1.21%	1591	—
11	#	3.07%	236	—
10	#	6.02%	264	2.84%
9	#	10.96%	211	4.02%
8	#	13.12%	43	4.32%
7	#	14.18%	42	4.77%
6	#	29.53%	44	13.08%
5	#	39.03%	45	20.81%
4	#	46.47%	212	31.09%
3	#	54.67%	42	34.58%
2	#	53.52%	18	41.10%
1	#	55.73%	1	50.46%