



EPA's recently generated Absorption, Distribution, Metabolism, and Excretion (ADME) data and relevance to the Internal Threshold of Toxicological Concern (iTTC)

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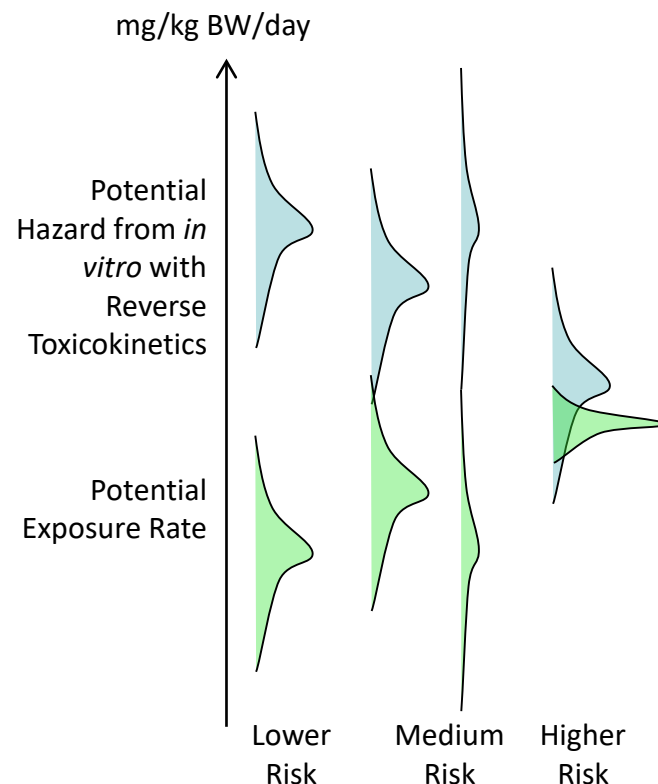
The views expressed in this presentation are those of the author and do not necessarily reflect the views or policies of the U.S. EPA

<https://orcid.org/0000-0002-4024-534X>

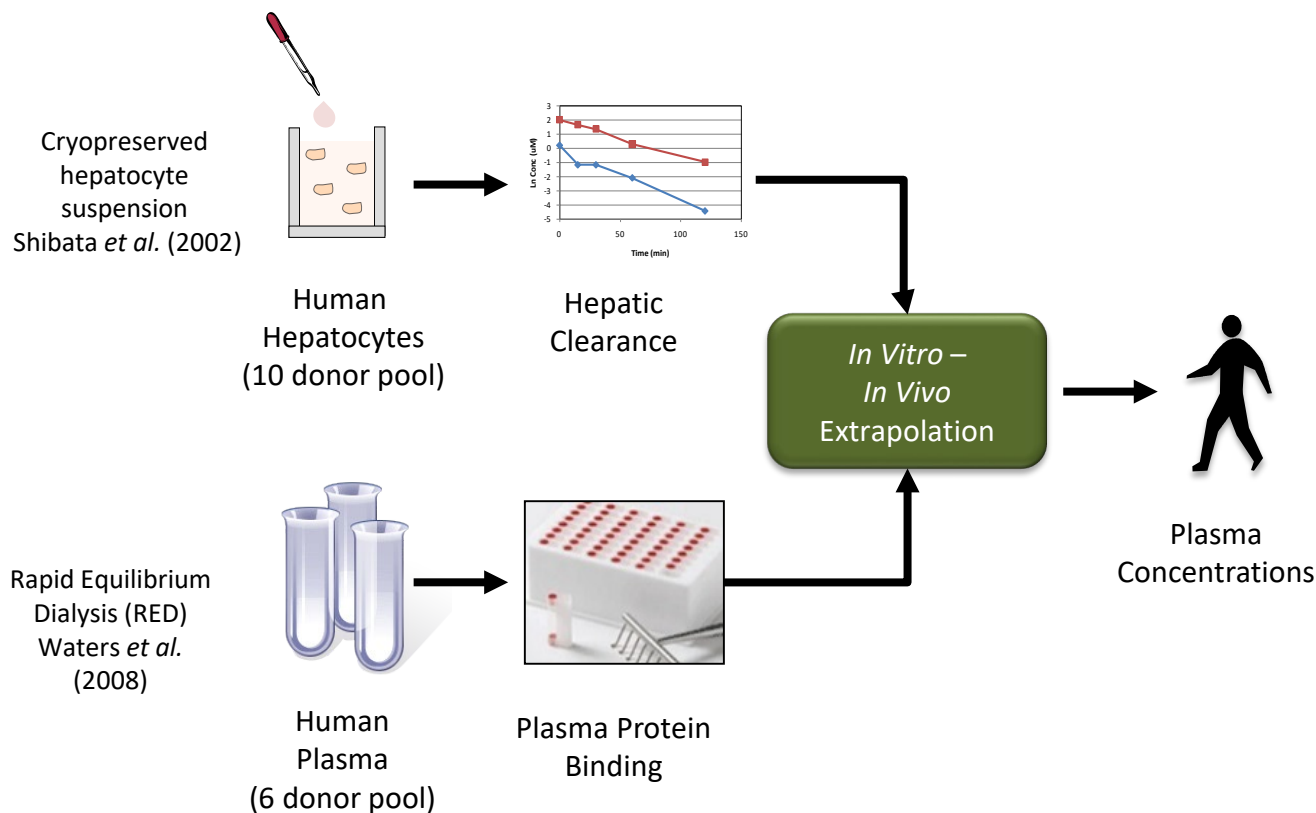
Internal TTC workshop II
26-27 September, 2018

High Throughput Toxicokinetics (HTTK)

- **Most chemicals do not have TK data** – Wetmore et al. (2012...) use *in vitro* methods adapted from pharma to fill gaps
- In order to address greater numbers of chemicals we collect *in vitro*, high throughput toxicokinetic (HTTK) data (Rotroff et al., 2010, Wetmore et al., 2012, 2015)
- HTTK methods have been used by the pharmaceutical industry to determine range of efficacious doses and to prospectively evaluate success of planned clinical trials (Jamei, *et al.*, 2009; Wang, 2010)
- The **primary goal** of HTTK is to provide a human dose context for bioactive *in vitro* concentrations from HTS (*i.e.*, *in vitro-in vivo* extrapolation, or **IVIVE**) (e.g., Wetmore et al., 2015)
- **Secondary goal** is to provide **open source data and models** for evaluation and use by the broader scientific community (Pearce et al, 2017a)

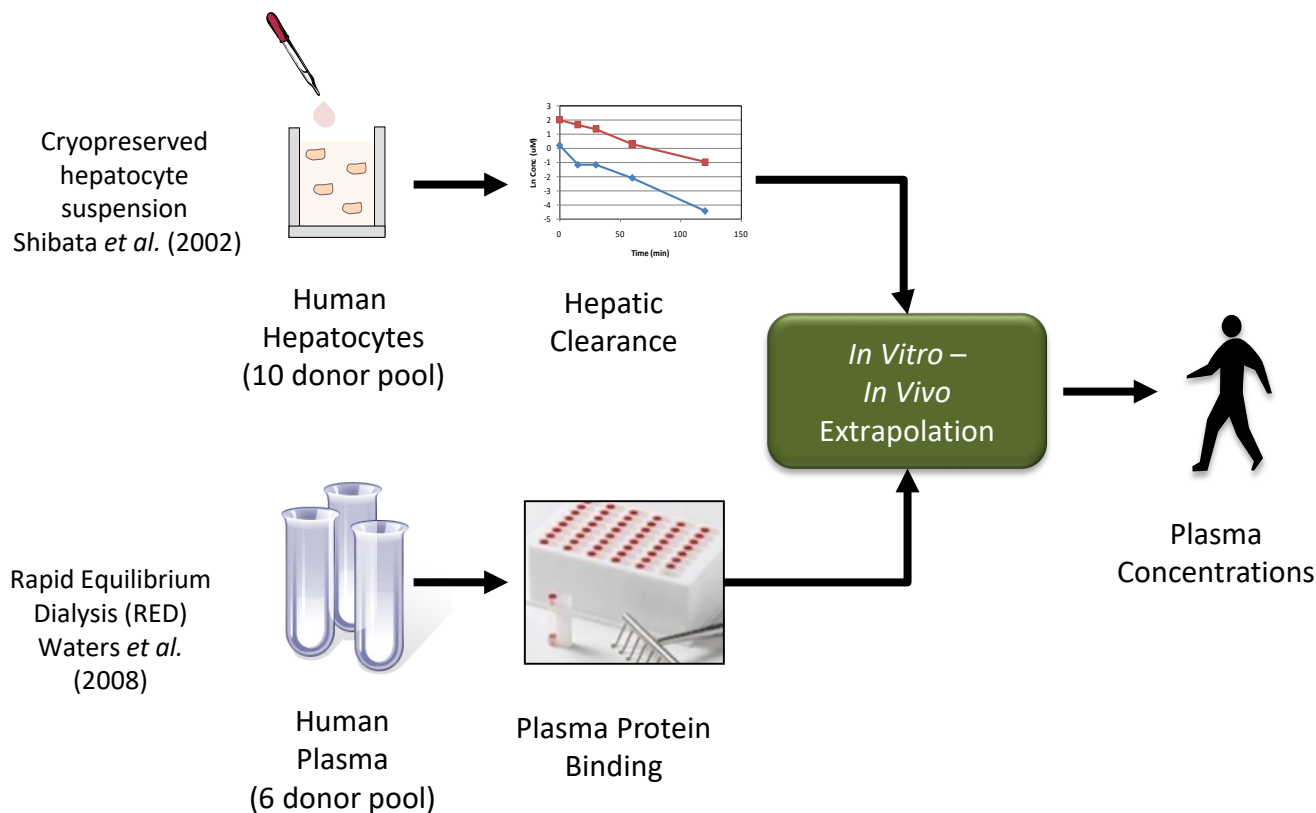


High-Throughput Toxicokinetics (HTTK) for *In Vitro-In Vivo* Extrapolation (IVIVE)



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- In drug development, HTTK methods allow IVIVE to estimate therapeutic doses for clinical studies – predicted concentrations are typically on the order of values measured in clinical trials (Wang, 2010)

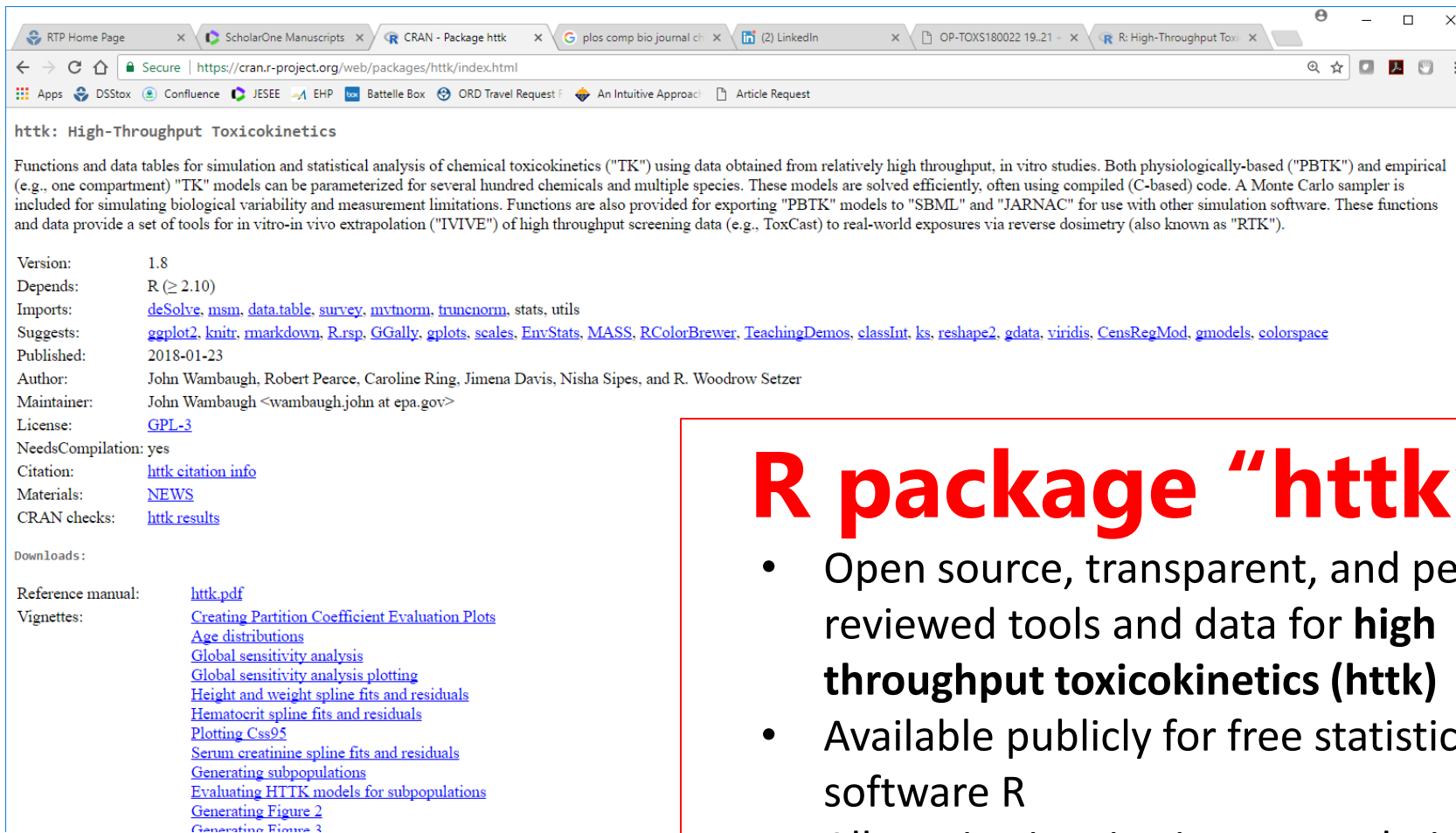
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Environmental chemicals:

Rotroff *et al.* (2010) 35 chemicals
Wetmore *et al.* (2012) +204 chemicals
Wetmore *et al.* (2015) +163 chemicals
Wambaugh *et al.* (in prep.) + ~400 chemicals



The screenshot shows the CRAN package page for 'httk'. The browser tabs include 'RTP Home Page', 'ScholarOne Manuscripts', 'CRAN - Package httk', 'plos comp bio journal ch', '(2) LinkedIn', 'OP-TOXS180022 19.21', and 'R: High-Throughput Tox'. The address bar shows 'https://cran.r-project.org/web/packages/httk/index.html'. The page title is 'httk: High-Throughput Toxicokinetics'. The description states: 'Functions and data tables for simulation and statistical analysis of chemical toxicokinetics ("TK") using data obtained from relatively high throughput, in vitro studies. Both physiologically-based ("PBTK") and empirical (e.g., one compartment) "TK" models can be parameterized for several hundred chemicals and multiple species. These models are solved efficiently, often using compiled (C-based) code. A Monte Carlo sampler is included for simulating biological variability and measurement limitations. Functions are also provided for exporting "PBTK" models to "SBML" and "JARNAC" for use with other simulation software. These functions and data provide a set of tools for in vitro-in vivo extrapolation ("IVIVE") of high throughput screening data (e.g., ToxCast) to real-world exposures via reverse dosimetry (also known as "RTK").'

Version: 1.8
Depends: R (≥ 2.10)
Imports: [deSolve](#), [msm](#), [data.table](#), [survey](#), [mytnorm](#), [truncnorm](#), stats, utils
Suggests: [ggplot2](#), [knitr](#), [markdown](#), [R.spc](#), [GGally](#), [gplots](#), [scales](#), [EnvStats](#), [MASS](#), [RColorBrewer](#), [TeachingDemos](#), [classInt](#), [ks](#), [reshape2](#), [gdata](#), [viridis](#), [CensRegMod](#), [gmodels](#), [colorspace](#)
Published: 2018-01-23
Author: John Wambaugh, Robert Pearce, Caroline Ring, Jimena Davis, Nisha Sipes, and R. Woodrow Setzer
Maintainer: John Wambaugh <wambaugh.john@epa.gov>
License: [GPL-3](#)
NeedsCompilation: yes
Citation: [httk citation info](#)
Materials: [NEWS](#)
CRAN checks: [httk results](#)

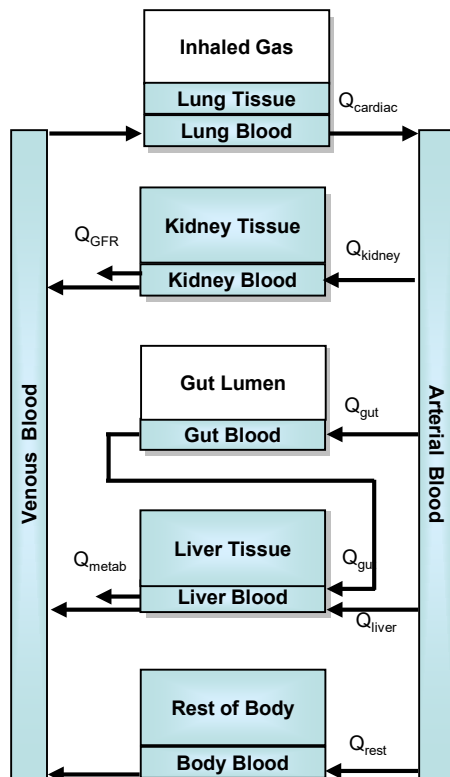
Downloads:

Reference manual: [httk.pdf](#)
Vignettes: [Creating Partition Coefficient Evaluation Plots](#), [Age distributions](#), [Global sensitivity analysis](#), [Global sensitivity analysis plotting](#), [Height and weight spline fits and residuals](#), [Hematocrit spline fits and residuals](#), [Plotting C_{ss}95](#), [Serum creatinine spline fits and residuals](#), [Generating subpopulations](#), [Evaluating HTHK models for subpopulations](#), [Generating Figure 2](#), [Generating Figure 3](#)

R package "httk"

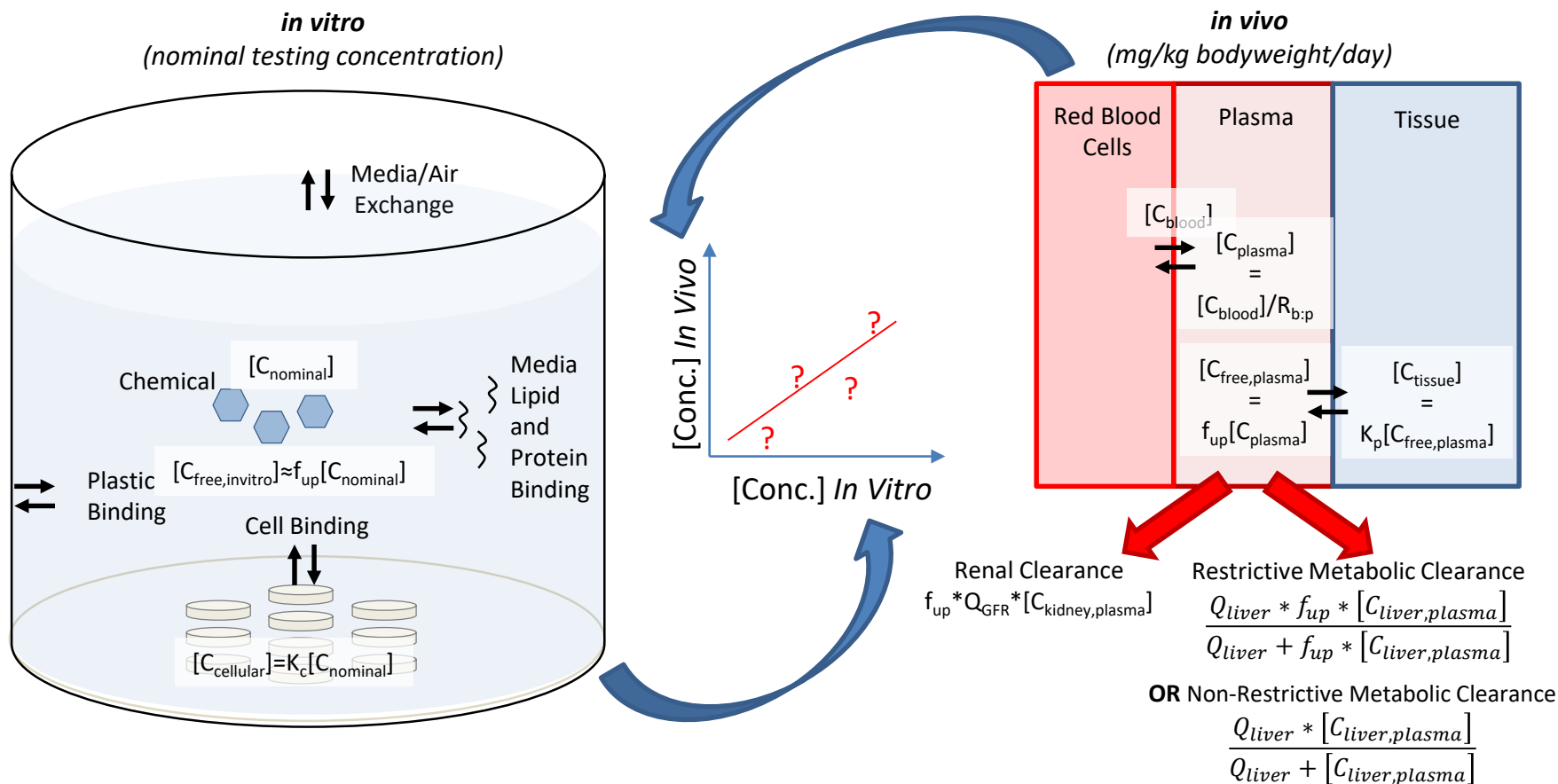
- Open source, transparent, and peer-reviewed tools and data for **high throughput toxicokinetics (httk)**
- Available publicly for free statistical software R
- Allows *in vitro-in vivo* extrapolation (IVIVE) and physiologically-based toxicokinetics (PBTK)

A General Physiologically-based Toxicokinetic (PBTK) Model



- “httk” includes a generic PBTK model
- Some tissues (e.g. arterial blood) are simple compartments, while others (e.g. kidney) are compound compartments consisting of separate blood and tissue sections with constant partitioning (i.e., tissue specific partition coefficients)
- Exposures are absorbed from reservoirs (gut lumen)
- Some specific tissues (lung, kidney, gut, and liver) are modeled explicitly, others (e.g. fat, brain, bones) are lumped into the “Rest of Body” compartment.
- The only ways chemicals “leave” the body are through metabolism (change into a metabolite) in the liver or excretion by glomerular filtration into the proximal tubules of the kidney (which filter into the lumen of the kidney).

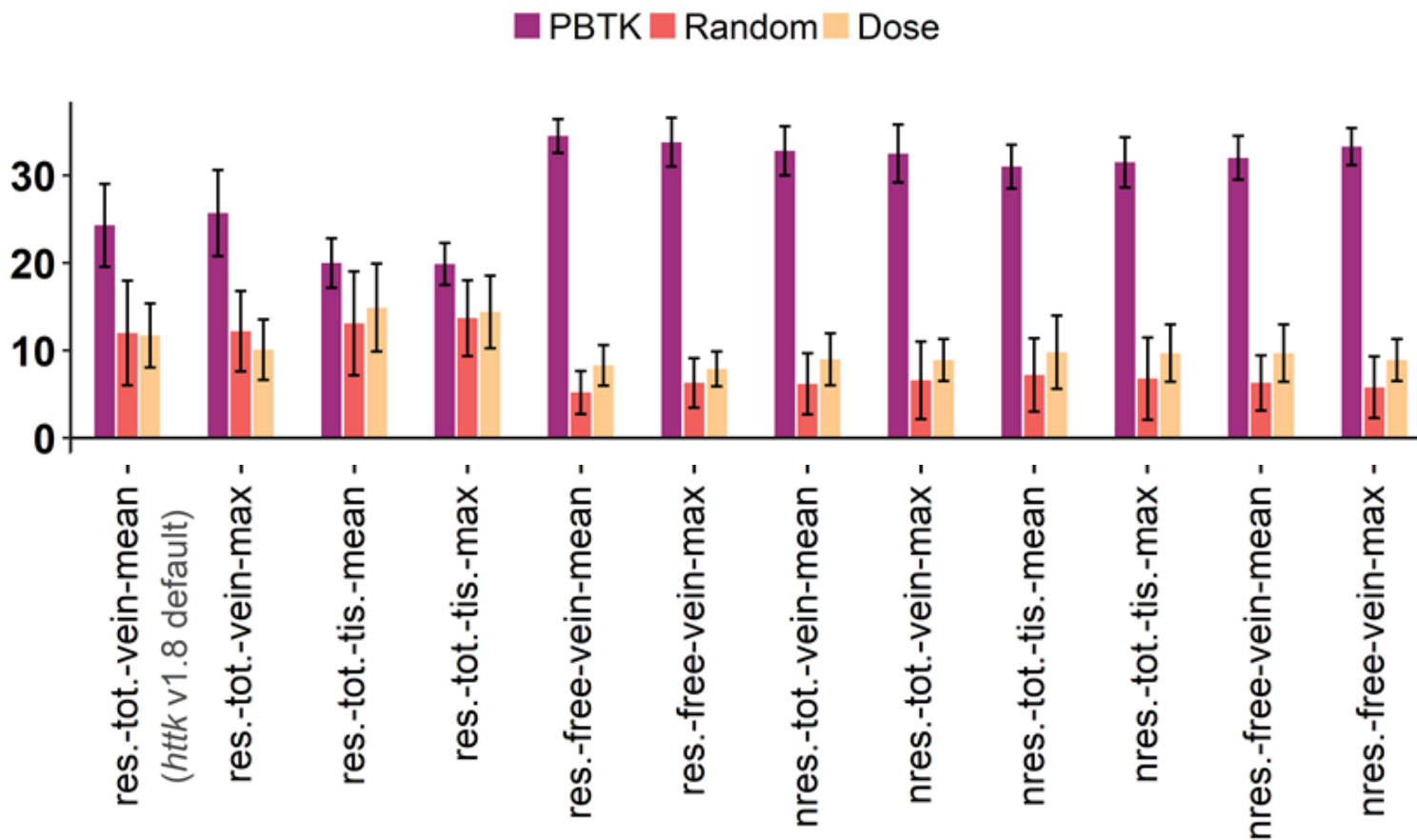
High-Throughput Toxicokinetics (HTTK) for *In Vitro-In Vivo* Extrapolation (IVIVE)



Selecting the appropriate *in vitro* and *in vivo* concentrations for extrapolation

Optimizing HTK-based IVIVE

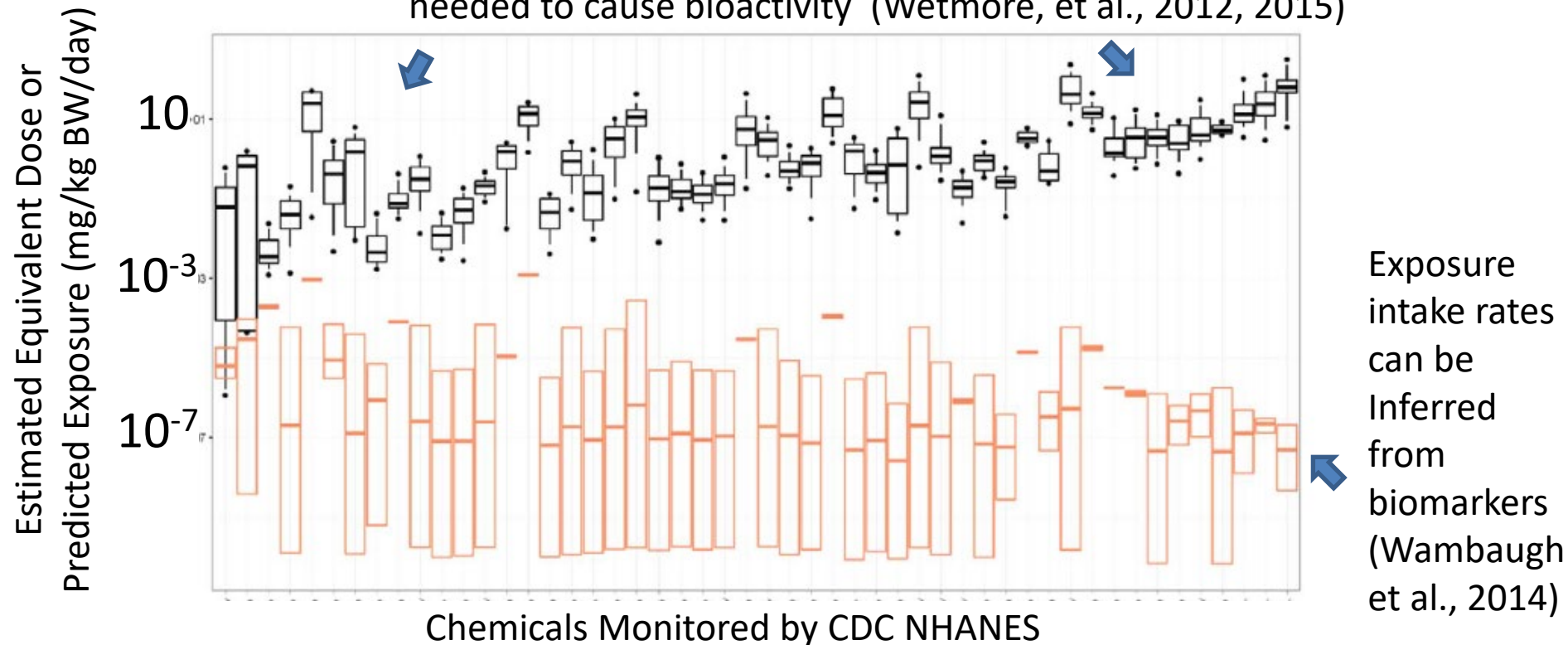
Times model selected as best for
predicting *in vivo* endpoints



Various Combinations of IVIVE Assumptions

Selecting Candidates for Prioritization

High Throughput Screening + HTTK can estimate doses needed to cause bioactivity (Wetmore, et al., 2012, 2015)



National Health and Nutrition Examination Survey (NHANES) is an ongoing survey that covers ~10,000 people every two years

Most NHANES chemicals do not have traditional PK models (Strope et al., 2018)

Why Build Another Generic PBTK Tool?

	SimCYP	ADMET Predictor / GastroPlus	MEGen	IndusChemFate	httk
Maker	SimCYP Consortium / Certara	Simulations Plus	UK Health and Safety Laboratory	Cefic LRI	US EPA
Availability	License, but inexpensive for research	License, but inexpensive for research	Free: http://xnet.hsl.gov.uk/megen	Free: http://cefic-lri.org/lri_toolbox/induschemfate/	Free: https://CRAN.R-project.org/package=httk
Open Source	No	No	Yes	No	Yes
Default PBPK Structure	Yes	Yes	No	Yes	Yes
Expandable PBPK Structure	No	No	Yes	No	No
Population Variability	Yes	No	No	No	Yes
Batch Mode	Yes	Yes	No	No	Yes
Graphical User Interface	Yes	Yes	Yes	Excel	No
Physiological Data	Yes	Yes	Yes	Yes	Yes
Chemical-Specific Data Library	Many Clinical Drugs	No	No	15 Environmental Compounds	543 Pharmaceutical and ToxCast Compounds
Ionizable Compounds	Yes	Yes	Potentially	No	Yes
Export Function	No	No	Matlab and AcslX	No	SBML and Jarnac
R Integration	No	No	No	No	Yes
Easy Reverse Dosimetry	Yes	Yes	No	No	Yes
Future Proof XML	No	No	Yes	No	No

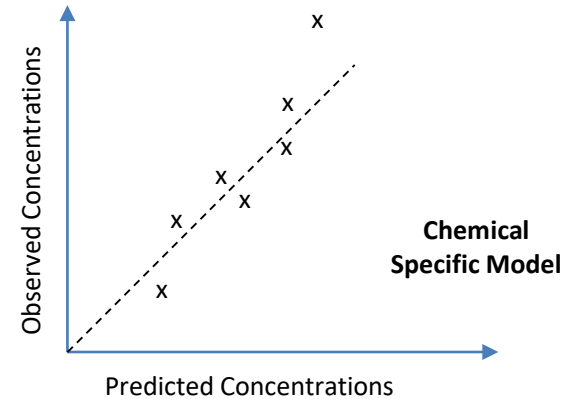
We want to do a statistical analysis (using R) for as many chemicals as possible

Doing Statistical Analysis with HTK

- If we are to use HTK, we need confidence in predictive ability
- In drug development, HTK methods estimate therapeutic doses for clinical studies – predicted concentrations are typically on the order of values measured in clinical trials (Wang, 2010)
 - For most compounds in the environment there will be no clinical trials
- Uncertainty must be well characterized
 - We compare to *in vivo* data to get **empirical estimates of HTK uncertainty**
 - ORD has both compiled existing (literature) TK data (Wambaugh *et al.*, 2015) and conducted new experiments in rats on chemicals with HTK *in vitro* data (Wambaugh *et al.*, submitted)
 - Any approximations, omissions, or mistakes should work to increase the estimated uncertainty when evaluated systematically across chemicals

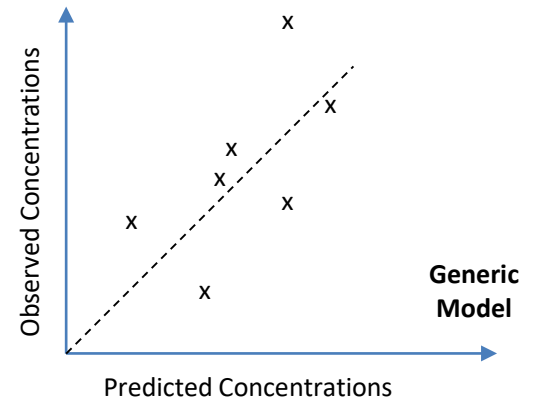
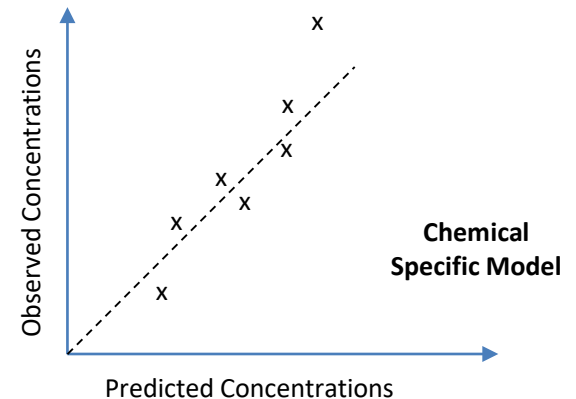
Building Confidence in TK Models

- In order to evaluate a **chemical-specific TK model** for “chemical x” you can compare the predictions to *in vivo* measured data
 - Can estimate bias
 - Can estimate uncertainty
 - Can consider using model to extrapolate to other situations (dose, route, physiology) where you don’t have data
- However, we do not typically have TK data



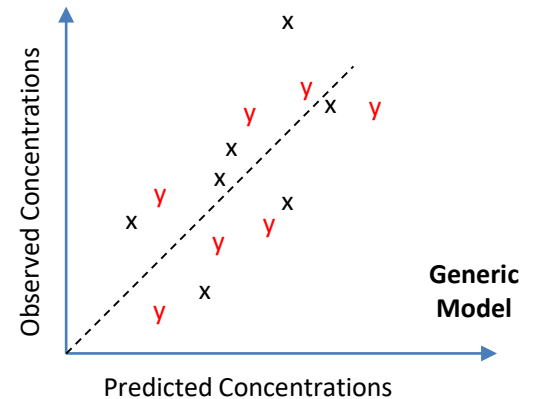
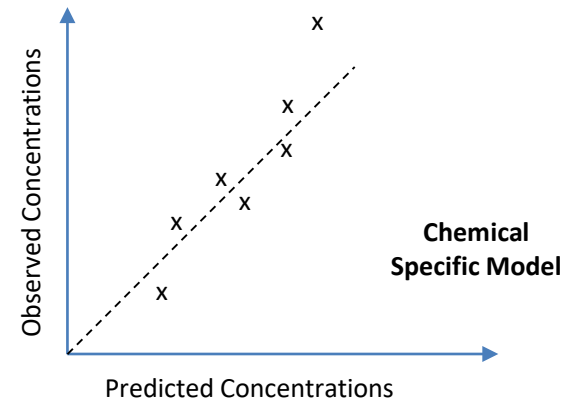
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- However, we do not typically have TK data
- We can parameterize a **generic TK model**, and evaluate that model for as many chemicals as we do have data
 - We do expect larger uncertainty, but also greater confidence in model implementation
 - Estimate bias and uncertainty, and try to correlate with chemical-specific properties
 - Can again consider using model to extrapolate to other situations (chemicals without *in vivo* data)



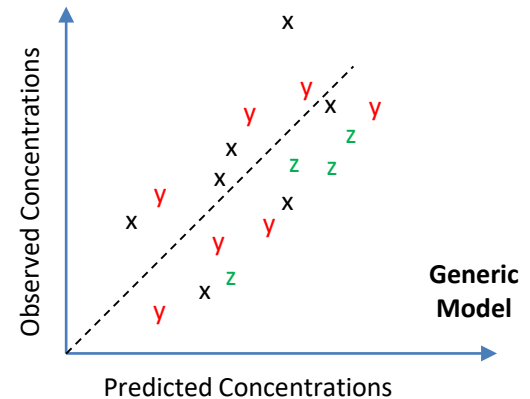
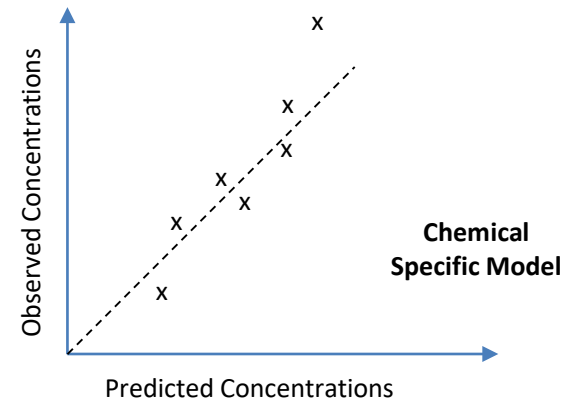
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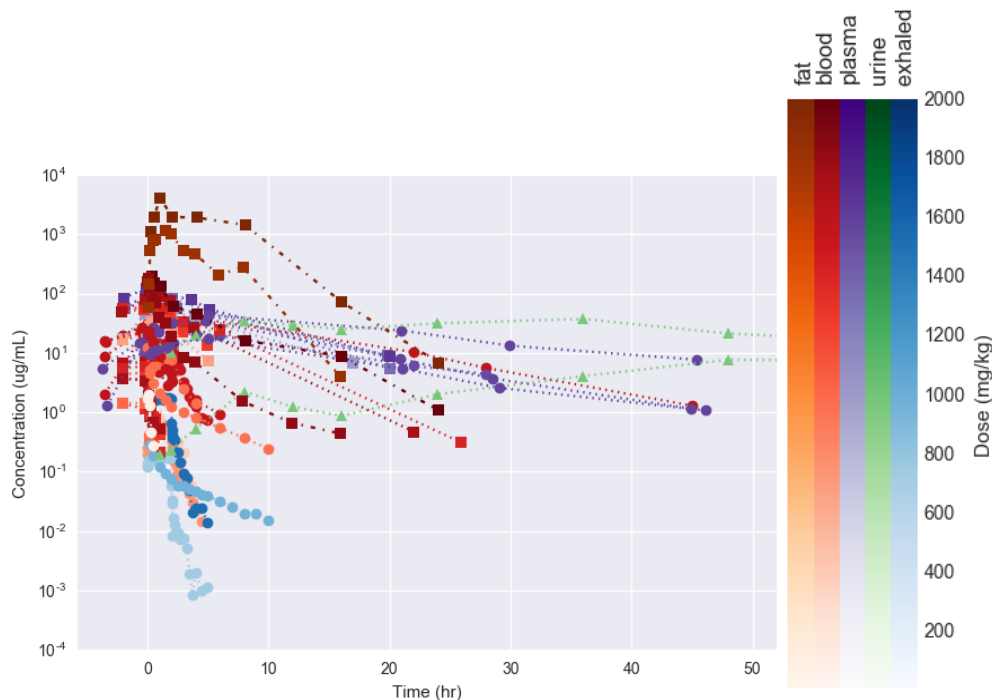
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In Vivo TK Database

- EPA is developing a public database of concentration vs. time data for building, calibrating, and evaluating TK models
- Curation and development ongoing, but to date includes:
 - 198 analytes (EPA, National Toxicology Program, literature)
 - Routes: Intravenous, dermal, oral, sub-cutaneous, and inhalation exposure
- Database will be made available through web interface and through the “httk” R package
- Standardized, open source curve fitting software *invivoPKfit* used to calibrate models to all data:

<https://github.com/USEPA/CompTox-ExpoCast-invivoPKfit>



Building Confidence in HTK

“...the steady-state, peak, and time-integrated plasma concentrations of non-pharmaceuticals were predicted with reasonable accuracy... HTK and IVIVE methods are adequately robust to be applied to high throughput *in vitro* toxicity screening data of environmentally-relevant chemicals for prioritizing based on human health risks.”

Toxicological Sciences



SOT | Society of
Toxicology
www.toxsci.oxfordjournals.org



TOXICOLOGICAL SCIENCES, 2018, 1–18

doi: 10.1093/toxsci/kfy020
Advance Access Publication Date: January 27, 2018
Research Article

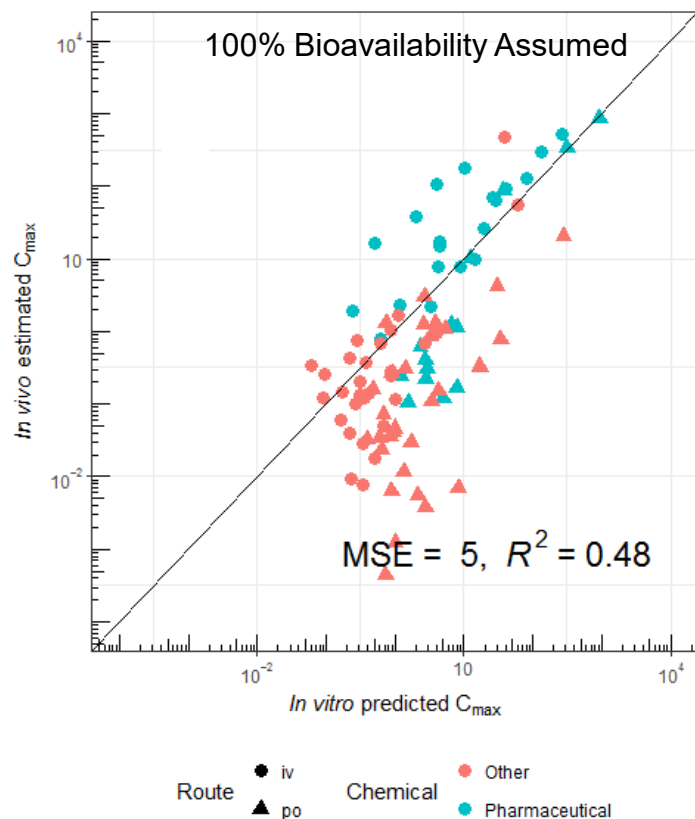
Evaluating *In Vitro*-*In Vivo* Extrapolation of Toxicokinetics

John F. Wambaugh,^{*,1} Michael F. Hughes,[†] Caroline L. Ring,^{*,‡,2} Denise K. MacMillan,[†] Jermaine Ford,[†] Timothy R. Fennell,[§] Sherry R. Black,[§] Rodney W. Snyder,[§] Nisha S. Sipes,[¶] Barbara A. Wetmore,^{||} Joost Westerhout,^{|||} R. Woodrow Setzer,^{*} Robert G. Pearce,^{*,‡} Jane Ellen Simmons,[†] and Russell S. Thomas^{*}

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Evaluating HTKK

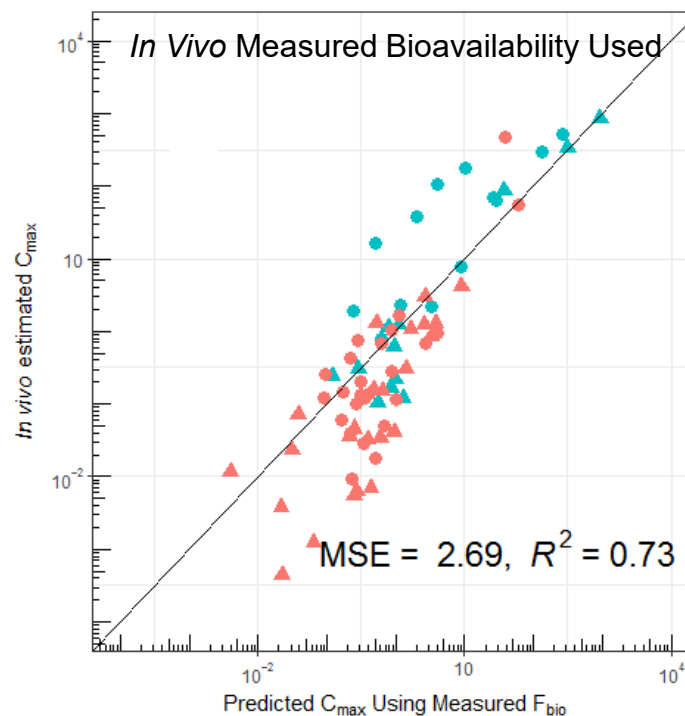
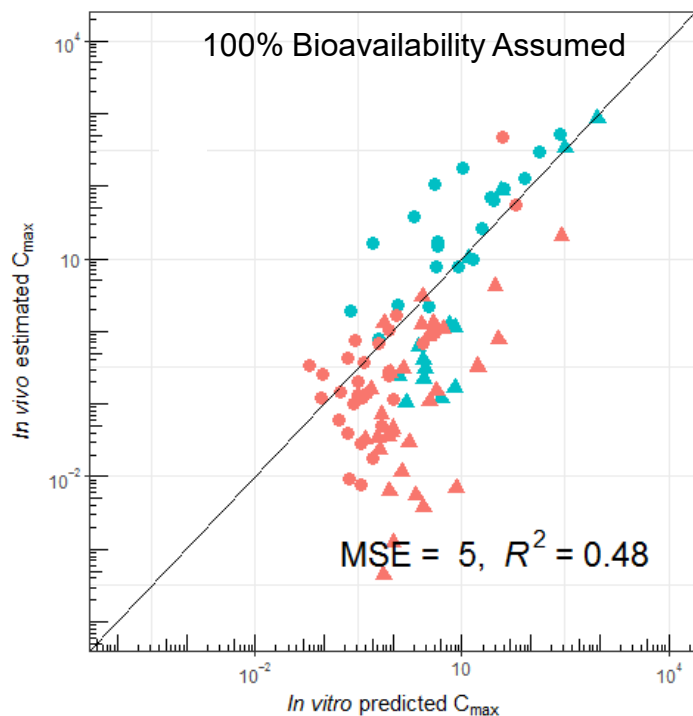


We evaluate HTKK by comparing predictions with observations for as many chemicals as possible

Wambaugh et al. (2018)

Evaluating HTTK

Impact of Oral Bioavailability Data



Wambaugh et al. (2018)

Characterizing Bioavailability *In Vitro*

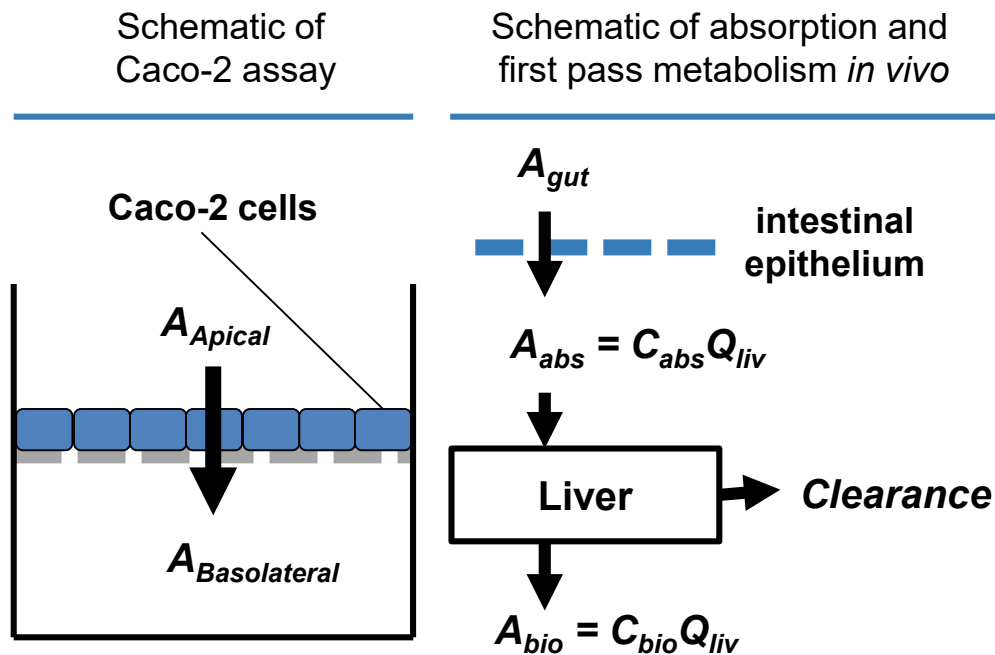
Greg Honda (NCCT) made a SOT2018 presentation on using Caco2 *in vitro* data to predict absorption for ~300 ToxCast chemicals

$$P_{AB} = \frac{1}{area * C_{Apical}} \frac{dA_{Basolateral}}{dt}$$

$$F_{abs} = A_{abs} / A_{gut} \approx \text{func.} (P_{AB})$$

$$F_{FP} \approx \frac{Q_{liv}}{Q_{liv} + f_{up} Cl / R_{b2p}}$$

$$F_{bio} = F_{abs} F_{FP}$$



Characterizing Bioavailability *In Vitro*

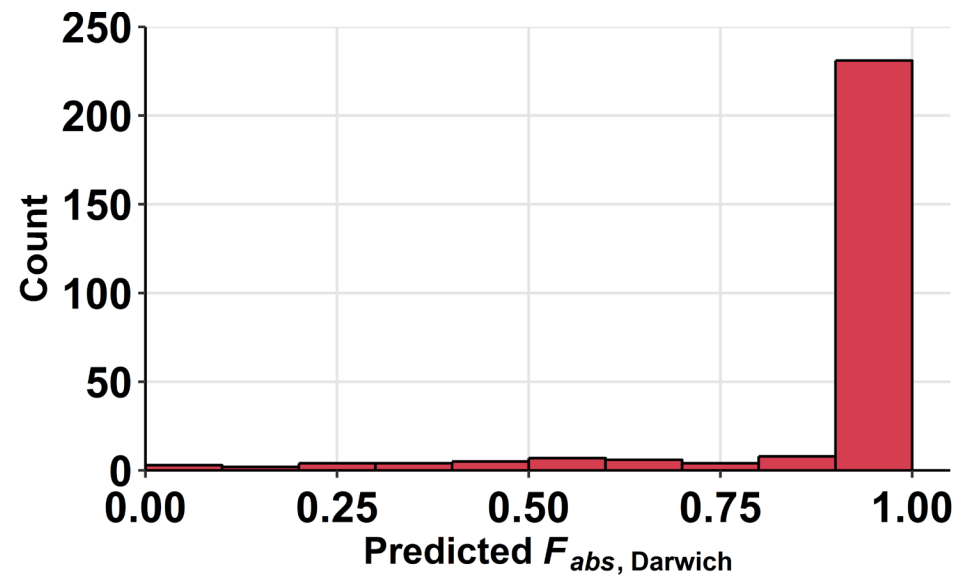
$$P_{AB} = \frac{1}{area * C_{Apical}} \frac{dA_{Basolateral}}{dt}$$

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Darwich et al. 2010



Characterizing Bioavailability *In Vitro*

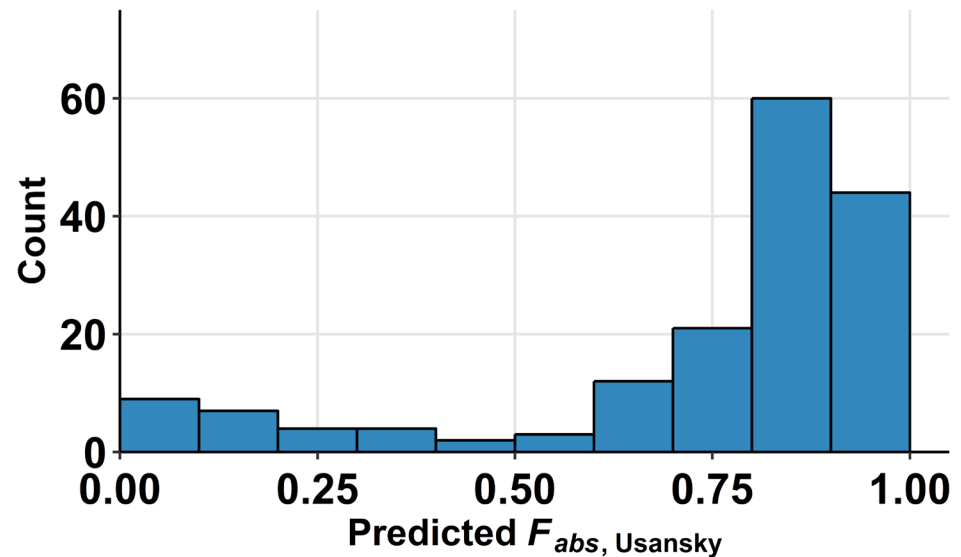
$$P_{AB} = \frac{1}{area * C_{Apical}} \frac{dA_{Basolateral}}{dt}$$

$$F_{abs} = A_{abs} / A_{gut} \approx \text{UsaSin} (P_{AB})$$

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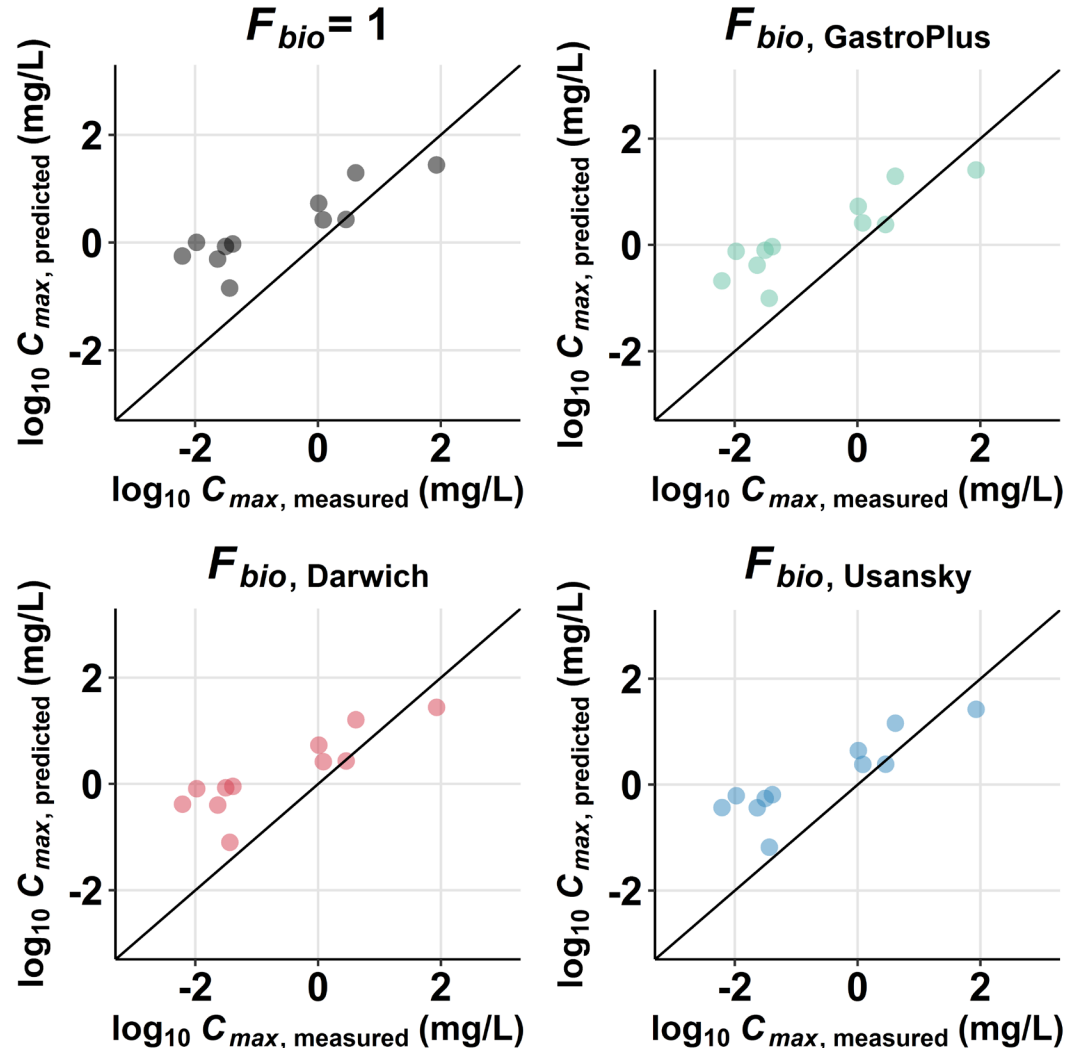
$$F_{bio} = F_{abs} F_{FP}$$

Usansky and Sinko 2005

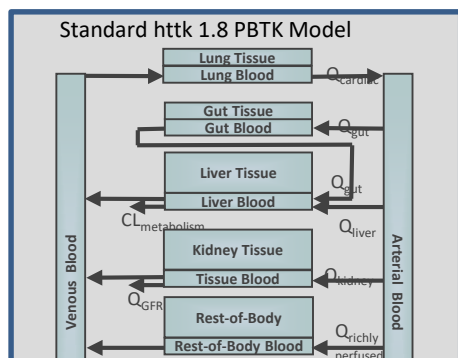


Predicting F_{bio} for Toxicokinetics

- C_{max} predicted using a 1 compartment model (Wambaugh *et al.* 2018)
- Minimal difference when using estimated F_{bio} in prediction of toxicokinetics observed for this limited set of chemicals

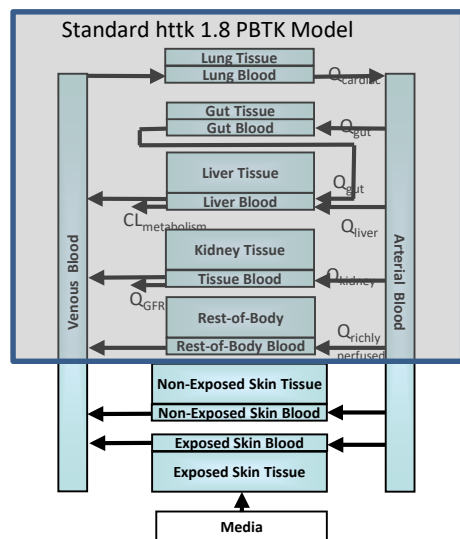


New HT-PBTK Models



- We are working to augment the basic HT-PBTK model with new PBTK models
- Each model will be released publicly upon peer-reviewed publication
- Pre-publication models can be shared under a MTA
- We assume there will be coding errors and over-simplifications, so each publication involves curation of evaluation data from the scientific literature and through statistical analysis
- Cvt database (Sayre et al.) is critical to these efforts

New HT-PBTK Models

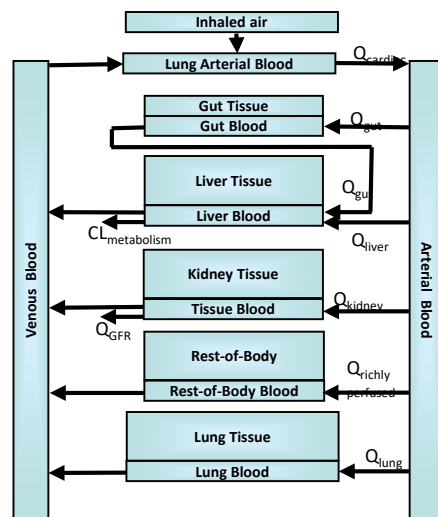
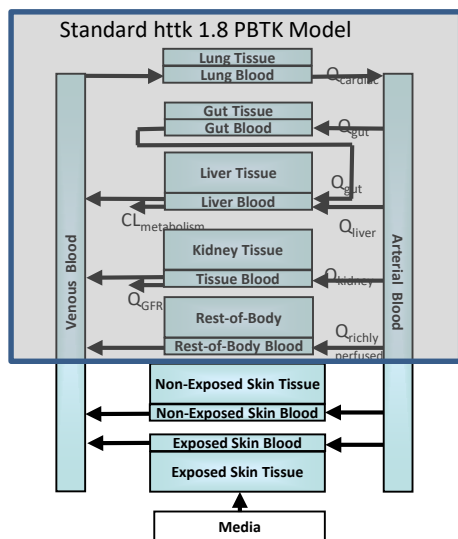


Dermal Exposure Route

EPA, Unilever, INERIS

New HT-PBTK Models

Gas Inhalation Exposure Route EPA, USAFSAM

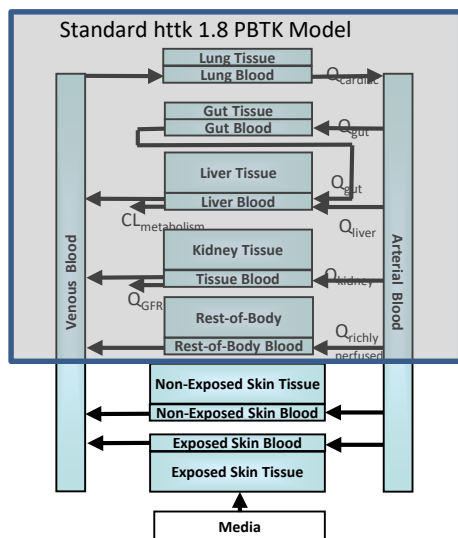


Dermal Exposure Route

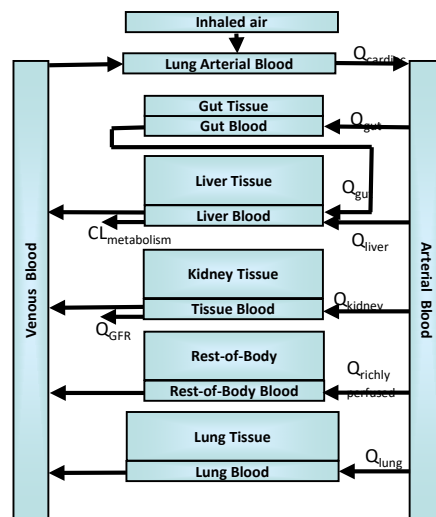
EPA, Unilever, INERIS

New HT-PBTK Models

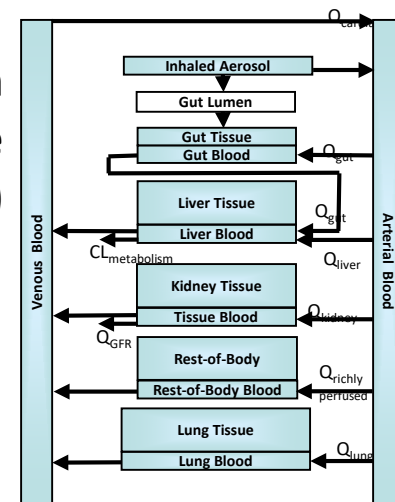
Gas Inhalation Exposure Route EPA, USAFSAM



Dermal Exposure Route EPA, Unilever, INERIS

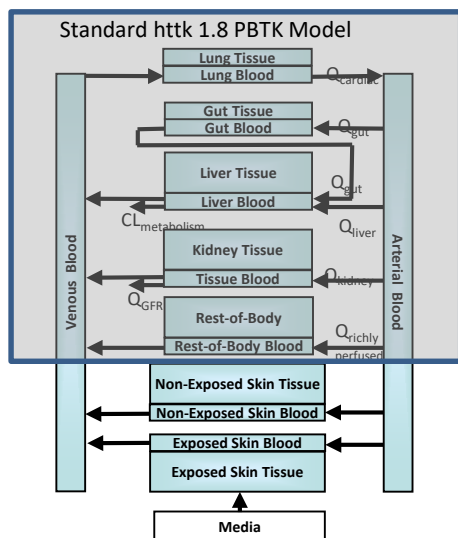


Aerosol Inhalation Exposure Route (with APEX model) EPA, USAFSAM

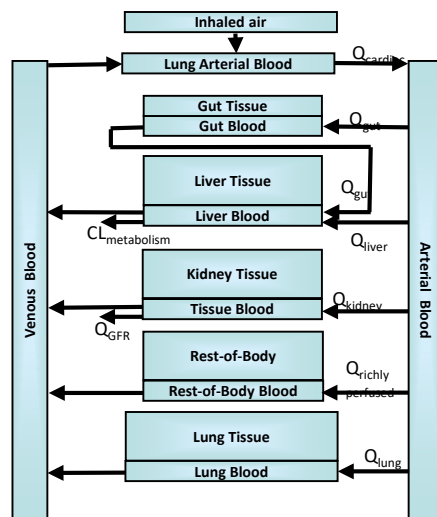


New HT-PBTK Models

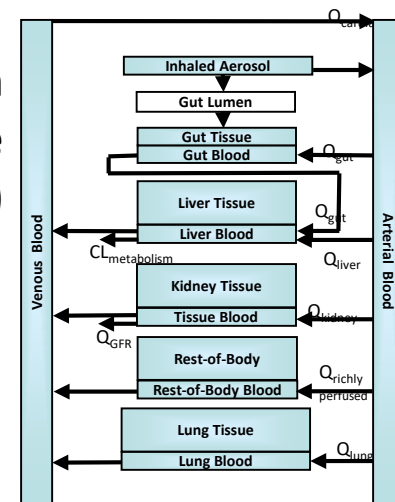
Gas Inhalation Exposure Route EPA, USAFSAM



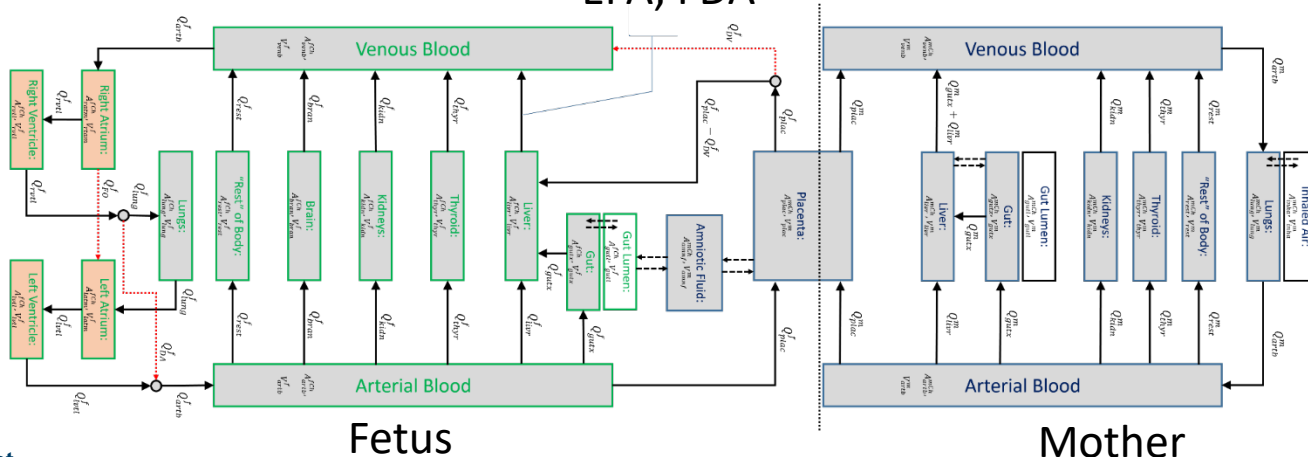
Dermal Exposure Route EPA, Unilever, INERIS



Aerosol Inhalation Exposure Route (with APEX model) EPA, USAFSAM

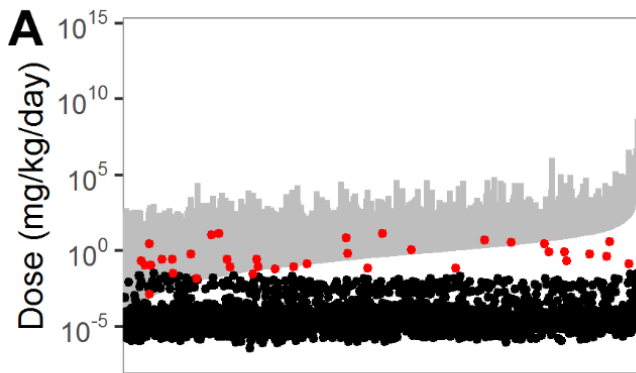


Human Gestational Model EPA, FDA

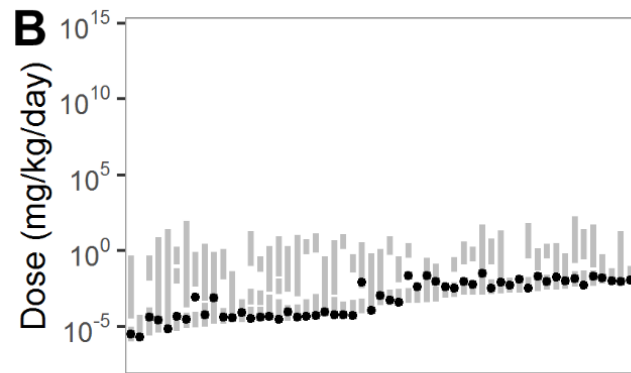


In Silico HTKK Predictions

- Tox21 has screened >8000 chemicals – Sipes *et al.* (2017) wanted to compare *in vitro* active concentrations with HTKK predicted maximum plasma concentrations with high throughput exposure predictions from Wambaugh *et al.* (2014)
 - “httk” package only has ~500 chemicals
- Used Simulations Plus ADMet Predictor to predict for entire library (supplemental table) and used add_chemtable() function to add into “httk” package
- Predictions available in httk v1.8



Dose range for all 3925
Tox21 compounds
eliciting a ‘possible’-to-
‘likely’ human *in vivo*
interaction alongside
estimated daily exposure

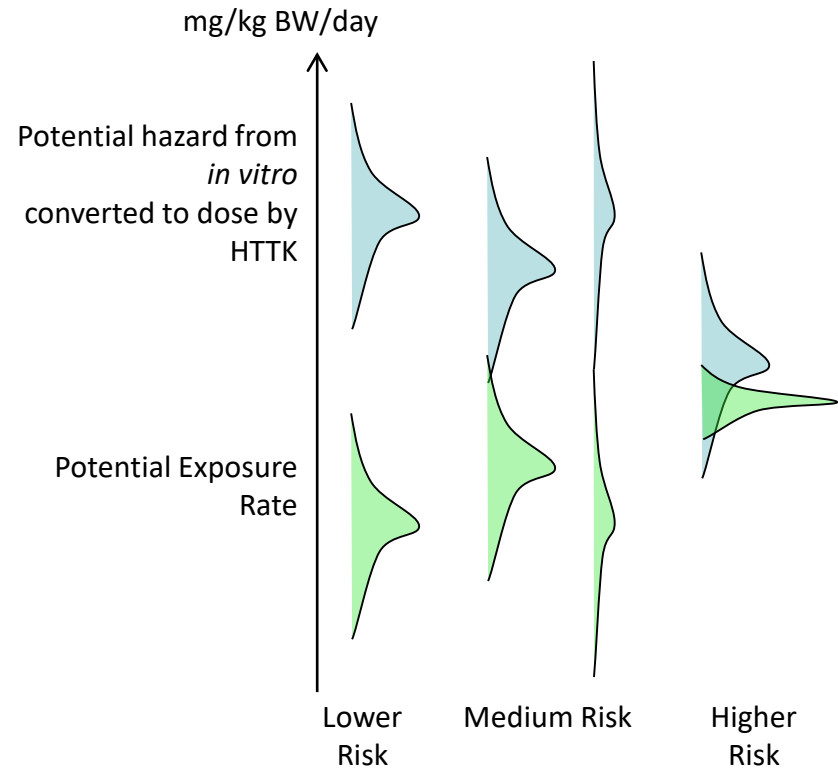


56 compounds with
potential *in vivo*
biological interaction
at or above estimated
environmental
exposures

Figure from Sipes *et al.*, (2017)

Conclusions

- We would like to know more about the risk posed by thousands of chemicals in the environment – which ones should we start with?
- In addition to toxicity, we need information on **Toxicokinetics**:
 - HTTK New approach methodologies (NAMs) are being evaluated through 1) uncertainty analysis and 2) comparison between *in vitro* predictions and *in vivo* measurements of both plasma concentrations and doses associated with the onset of effects
 - Modeling various exposure routes (e.g., inhalation of gasses and aerosols) allows extrapolation to important scenarios



Chemical Safety for Sustainability (CSS) Research Program

Rapid Exposure and Dosimetry (RED) Project

NCCT

Chris Grulke
Greg Honda*
Richard Judson
Matthew Linakis*
Andrew McEachran*
Ann Richard
Risa Sayre*
Woody Setzer
Rusty Thomas
John Wambaugh
Antony Williams

NRMRL

Xiaoyu Liu

NHEERL

Linda Adams
Christopher Ecklund
Marina Evans
Mike Hughes
Jane Ellen Simmons

NERL

Cody Addington*
Craig Barber
Namdi Brandon*
Peter Egeghy
Hongtai Huang*
Kristin Isaacs
Ashley Jackson*
Charles Lowe*
Dawn Mills*
Seth Newton

Katherine Phillips

Paul Price
Jeanette Reyes*
Randolph Singh*
Jon Sobus
John Streicher*
Mark Strynar
Mike Tornero-Velez
Elin Ulrich
Dan Vallero
Barbara Wetmore

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John Cowden (NCCT)

***Trainees**

Collaborators

Arnot Research and Consulting
Jon Arnot
Johnny Westgate
Institut National de l'Environnement et des Risques (INERIS)
Frederic Bois
Integrated Laboratory Systems
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