



Evaluating New Approach Methodologies for Toxicokinetics

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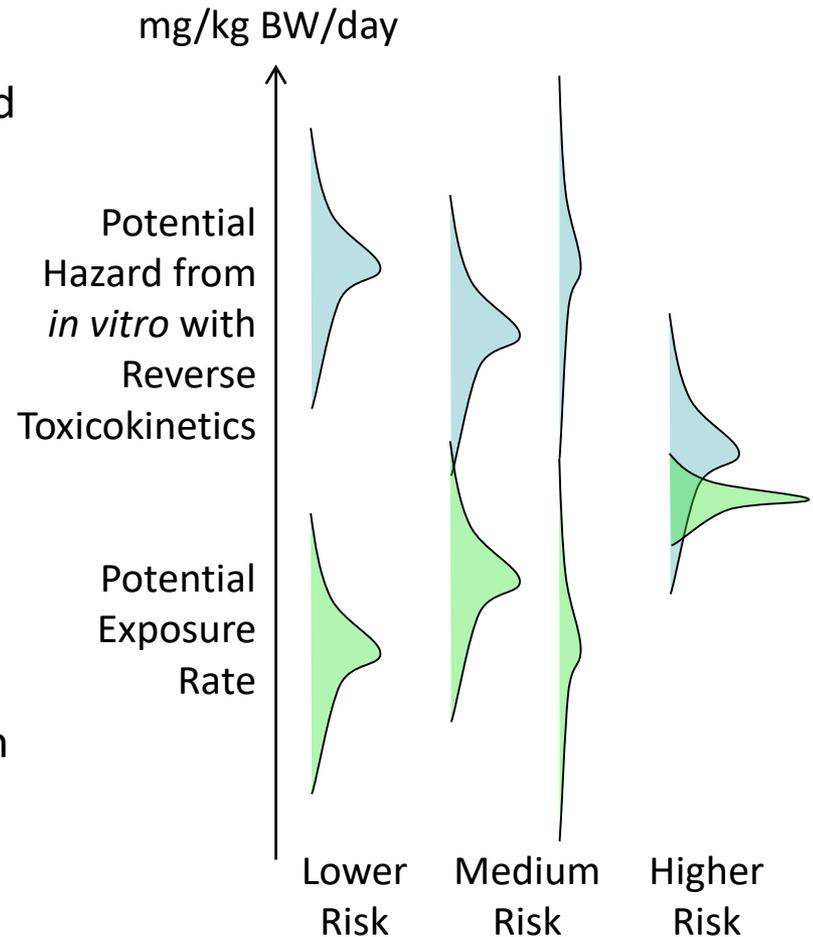
**Evaluating High-Throughput New
Approach Methods (NAM) for Exposure
ISES-ISEE Joint Annual Meeting
Ottawa, Canada
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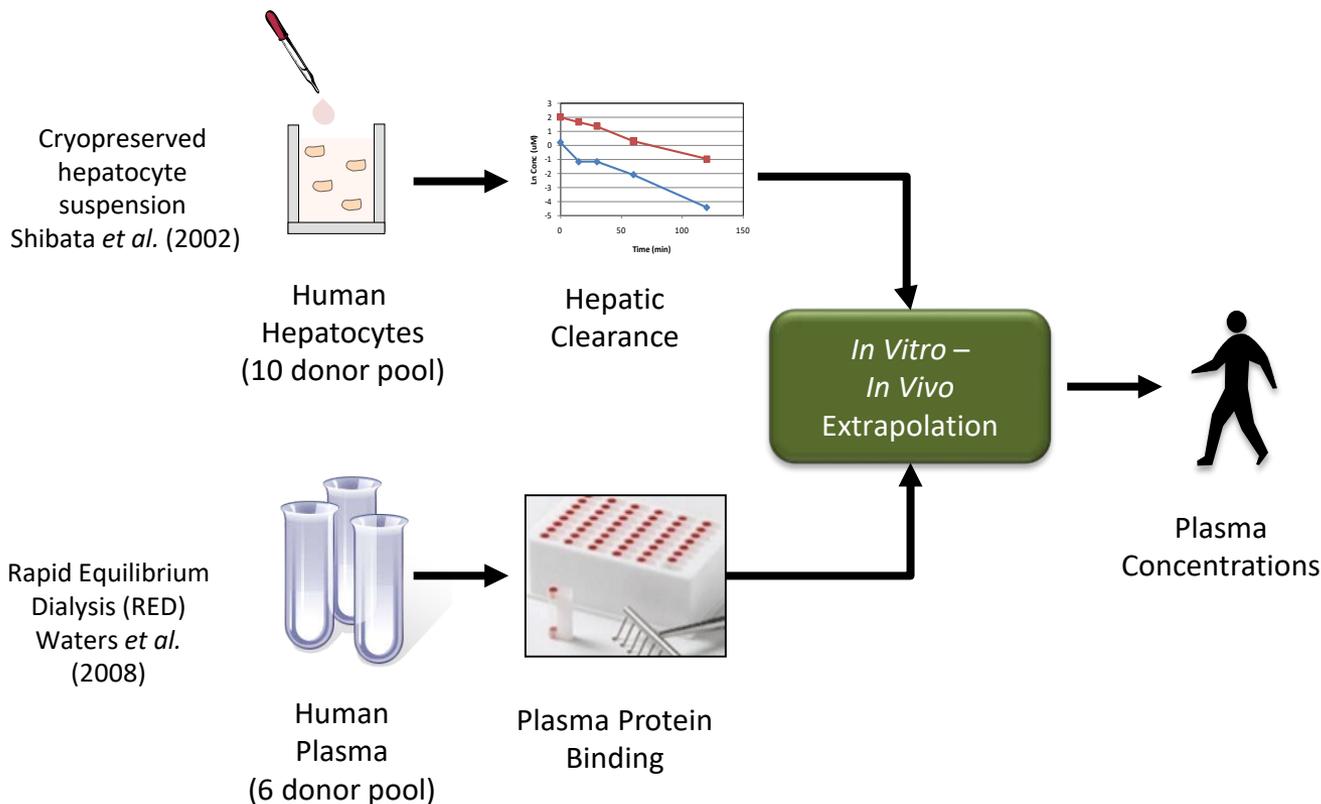
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Chemical Risk = Hazard x Exposure

- National Research Council (1983) identified chemical risk as a function of both inherent hazard and exposure
- To address thousands of chemicals, we need new approach methodologies that can prioritize those chemicals most worthy of additional study
- High throughput risk prioritization needs:
 1. high throughput hazard characterization (Dix et al., 2007, Collins et al., 2008)
 2. high throughput exposure forecasts (Wambaugh et al., 2013, 2014)
 3. high throughput toxicokinetics (i.e., dose-response relationship) linking hazard and exposure

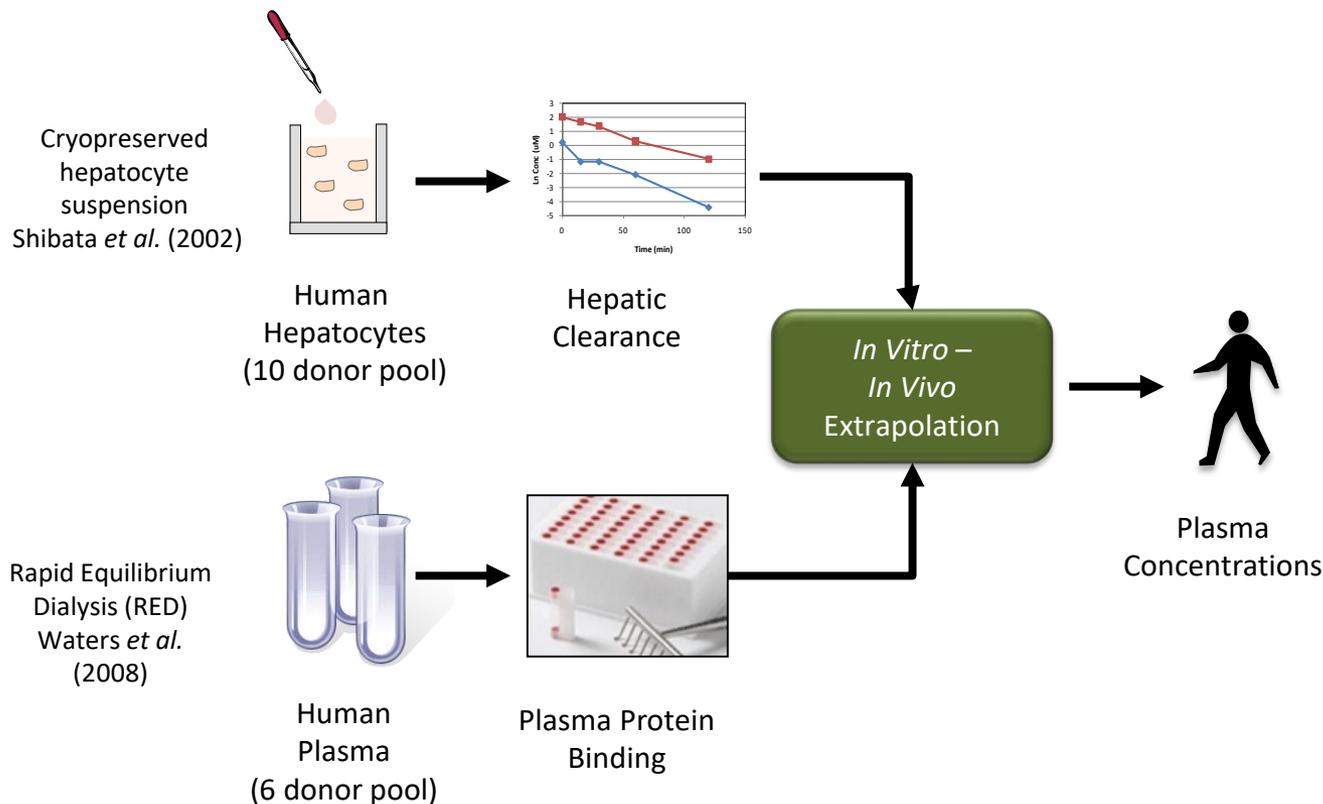


High-Throughput Toxicokinetics (HTTK) for *In Vitro-In Vivo* Extrapolation (IVIVE)



- **Most chemicals do not have TK data** – we use *in vitro* HTTK methods adapted from pharma to fill gaps
- 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)

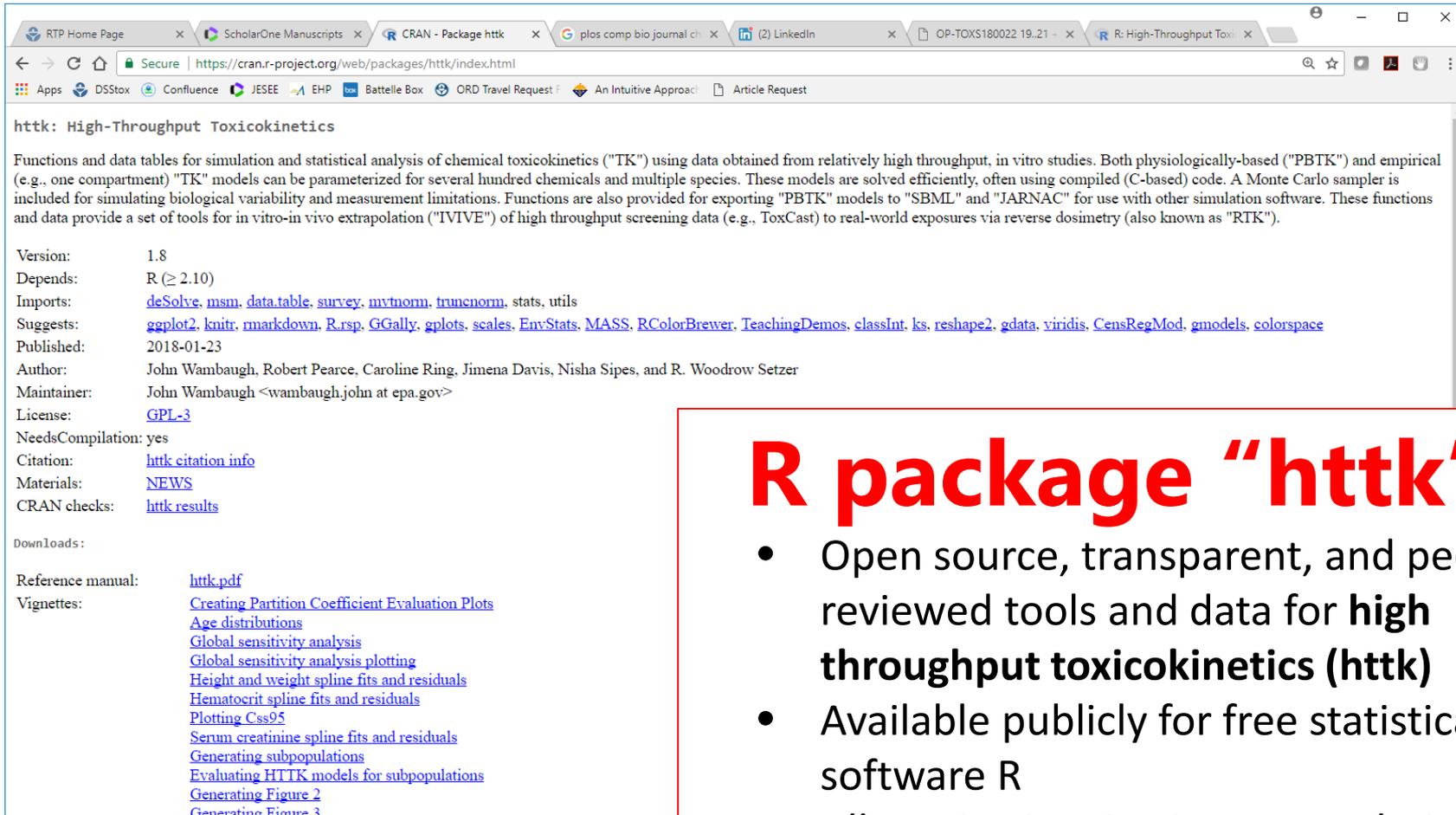
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Environmental chemicals:

Rotroff *et al.* (2010) 35 chemicals
 Wetmore *et al.* (2012) +204 chemicals
 Wetmore *et al.* (2015) +163 chemicals
 Wambaugh *et al.* (in prep.) + ~300 chemicals

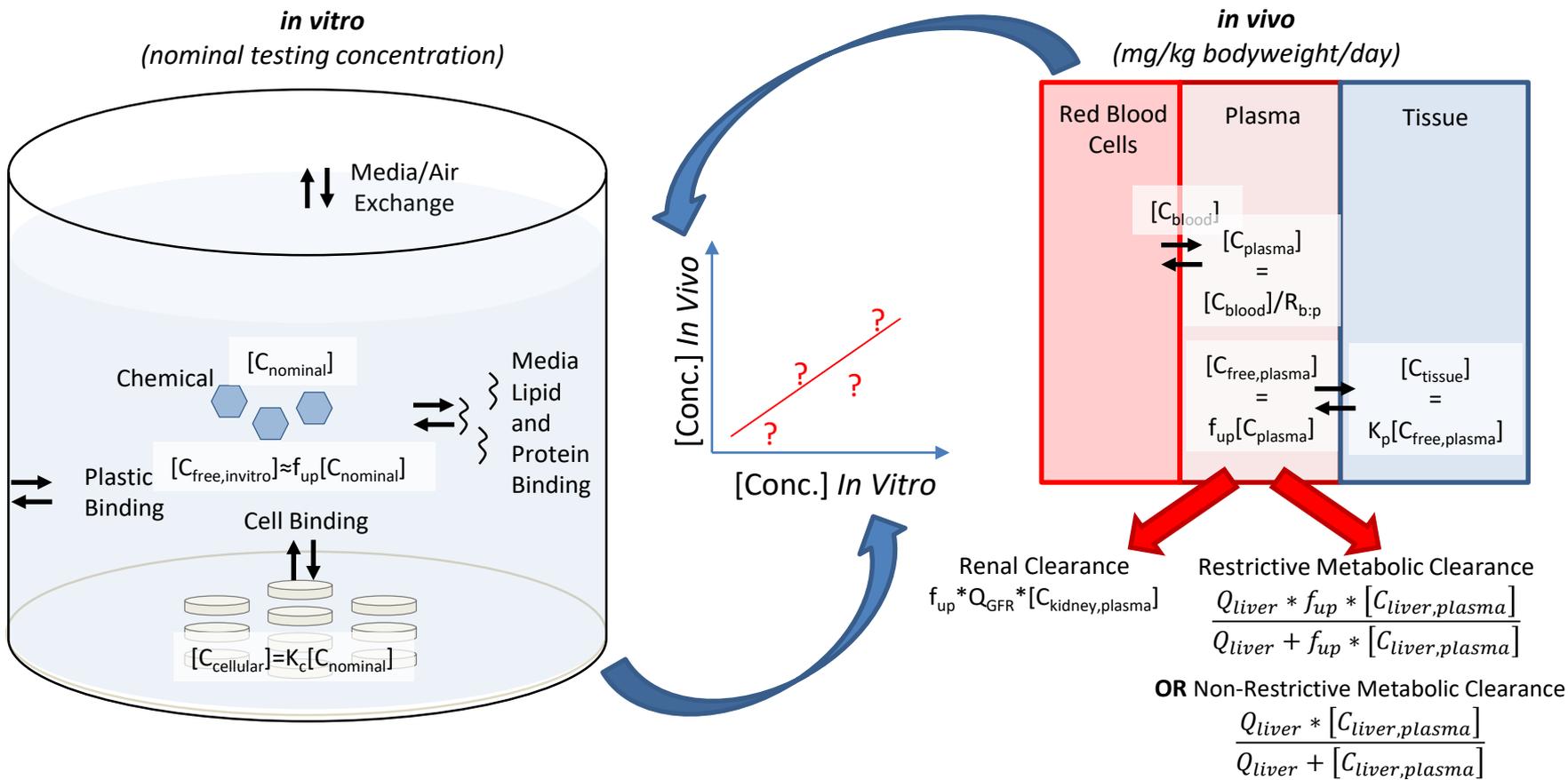


The screenshot shows the CRAN website for the 'httk' R package. The page title is 'httk: High-Throughput Toxicokinetics'. The main text describes the package's functions and data tables for simulation and statistical analysis of chemical toxicokinetics. Below the description, there is a list of package details including version (1.8), dependencies (R ≥ 2.10), imports (deSolve, msm, data.table, survey, mytnorm, truncnorm, stats, utils), suggests (ggplot2, knitr, markdown, R.rsp, GGally, gplots, scales, EnvStats, MASS, RColorBrewer, TeachingDemos, classInt, ks, reshape2, gdata, viridis, CensRegMod, gmodels, colorspace), published date (2018-01-23), author (John Wambaugh, Robert Pearce, Caroline Ring, Jimena Davis, Nisha Sipes, and R. Woodrow Setzer), maintainer (John Wambaugh <wambaugh.john at epa.gov>), license (GPL-3), needs compilation (yes), citation (httk citation info), materials (NEWS), CRAN checks (httk results), and a list of vignettes including 'httk.pdf', '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_{ss95}', 'Serum creatinine spline fits and residuals', 'Generating subpopulations', 'Evaluating HTTK models for subpopulations', 'Generating Figure 2', and '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-base toxicokinetics (PBTK)

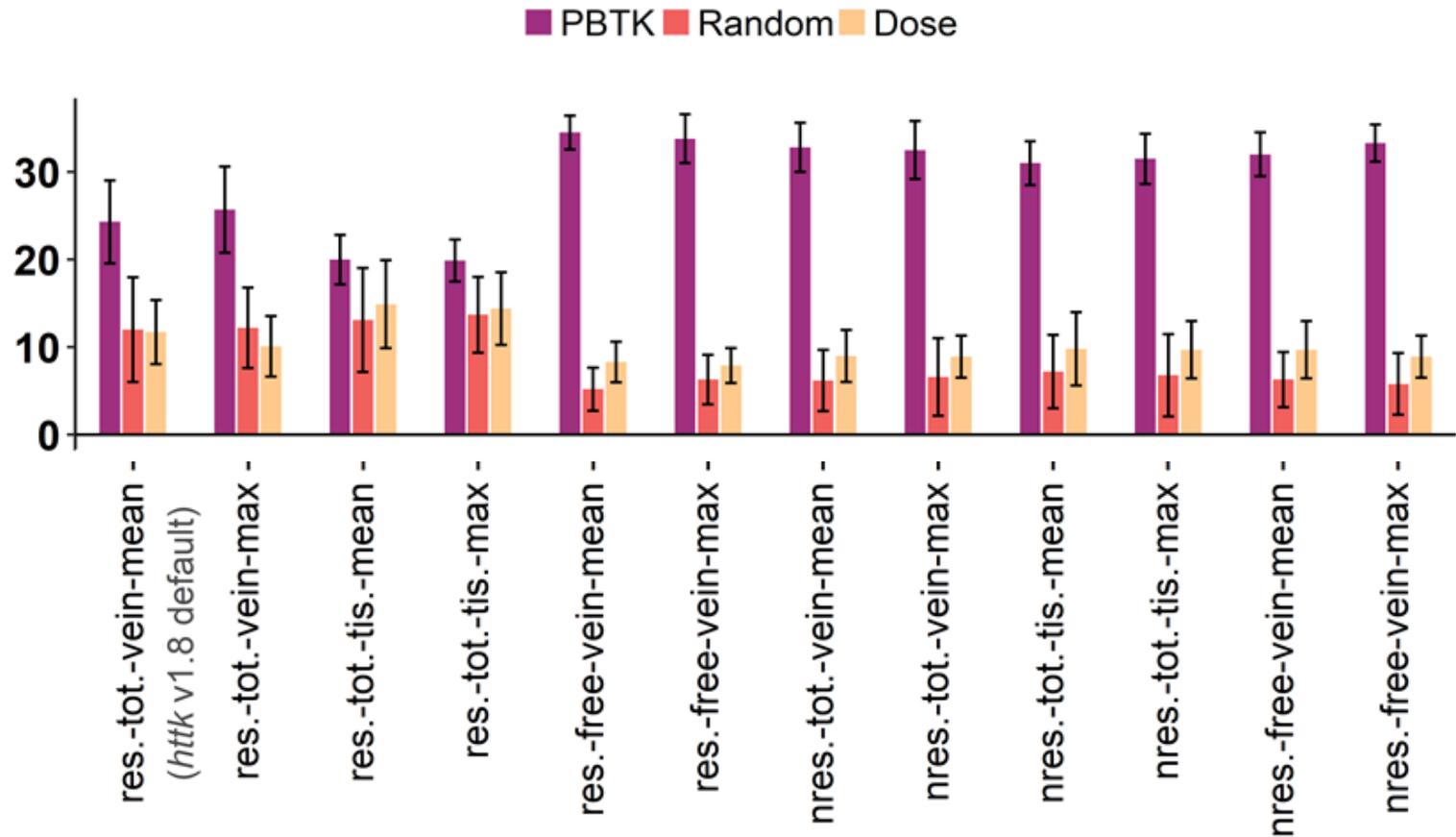
High-Throughput Toxicokinetics (HTTK) for *In Vitro-In Vivo* Extrapolation (IVIVE)



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

Optimizing HHTK-based IVIVE

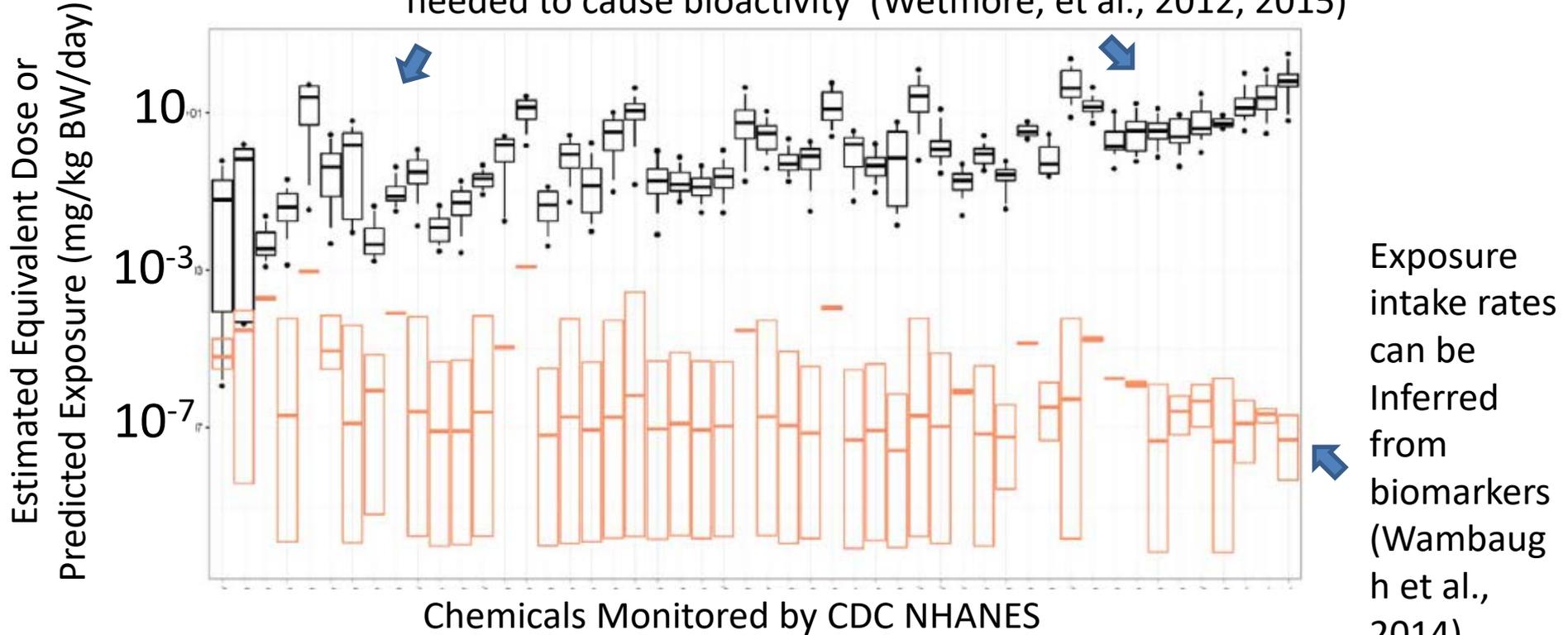
Times model selected as best for predicting *in vivo* endpoints



Various Combinations of IVIVE Assumptions

High Throughput Risk Prioritization

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

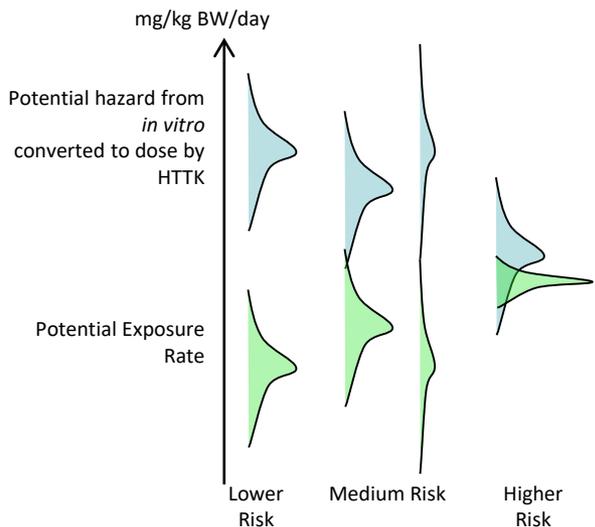
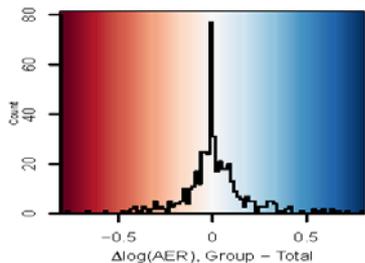


National Health and Nutrition Examination Survey (NHANES) is an ongoing survey that covers ~10,000 people every two years

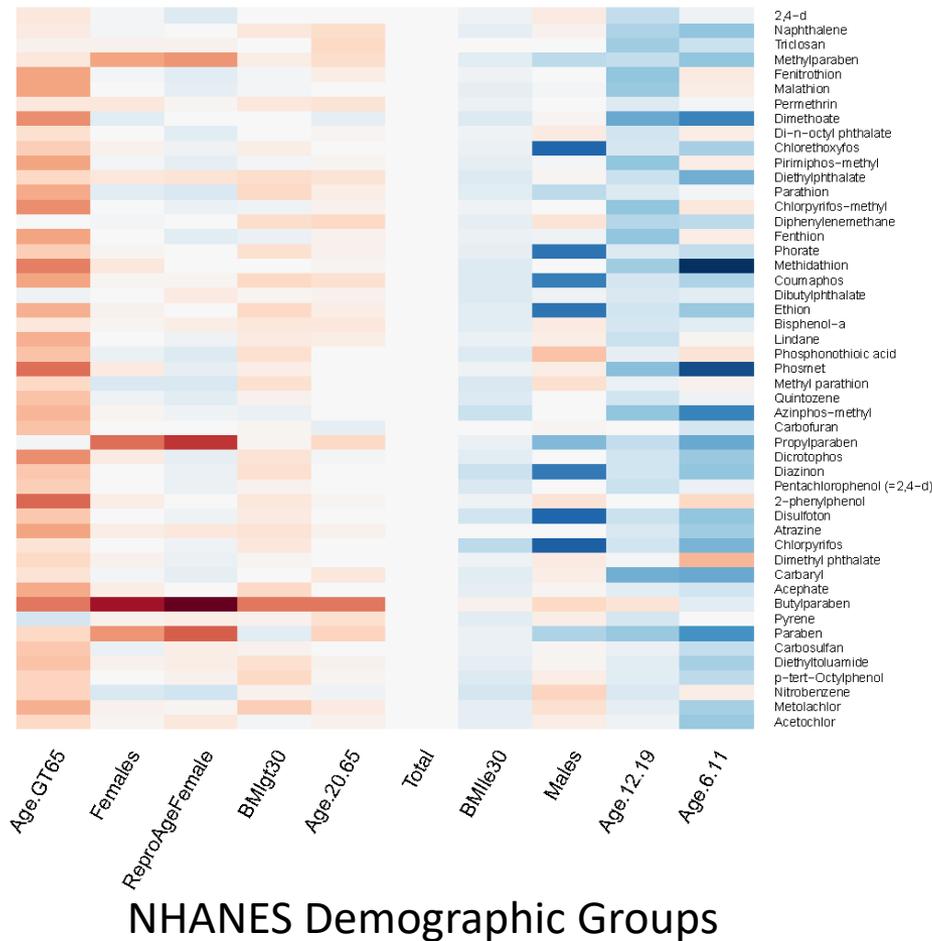
Most NHANES chemicals do not have traditional PK models (Strope et al., 2018)

Life-stage and Demographic Specific Predictions

- We use HHTK to calculate margin between bioactivity and exposure for specific populations

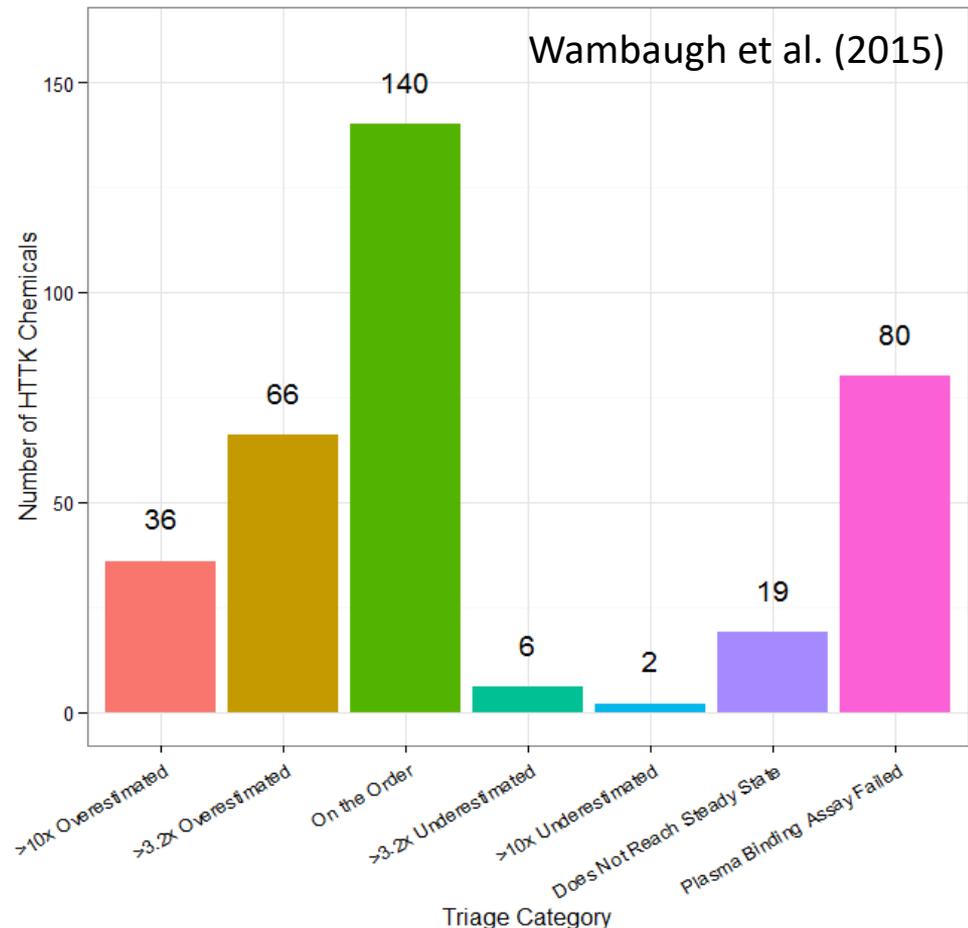


Change in Activity : Exposure Ratio



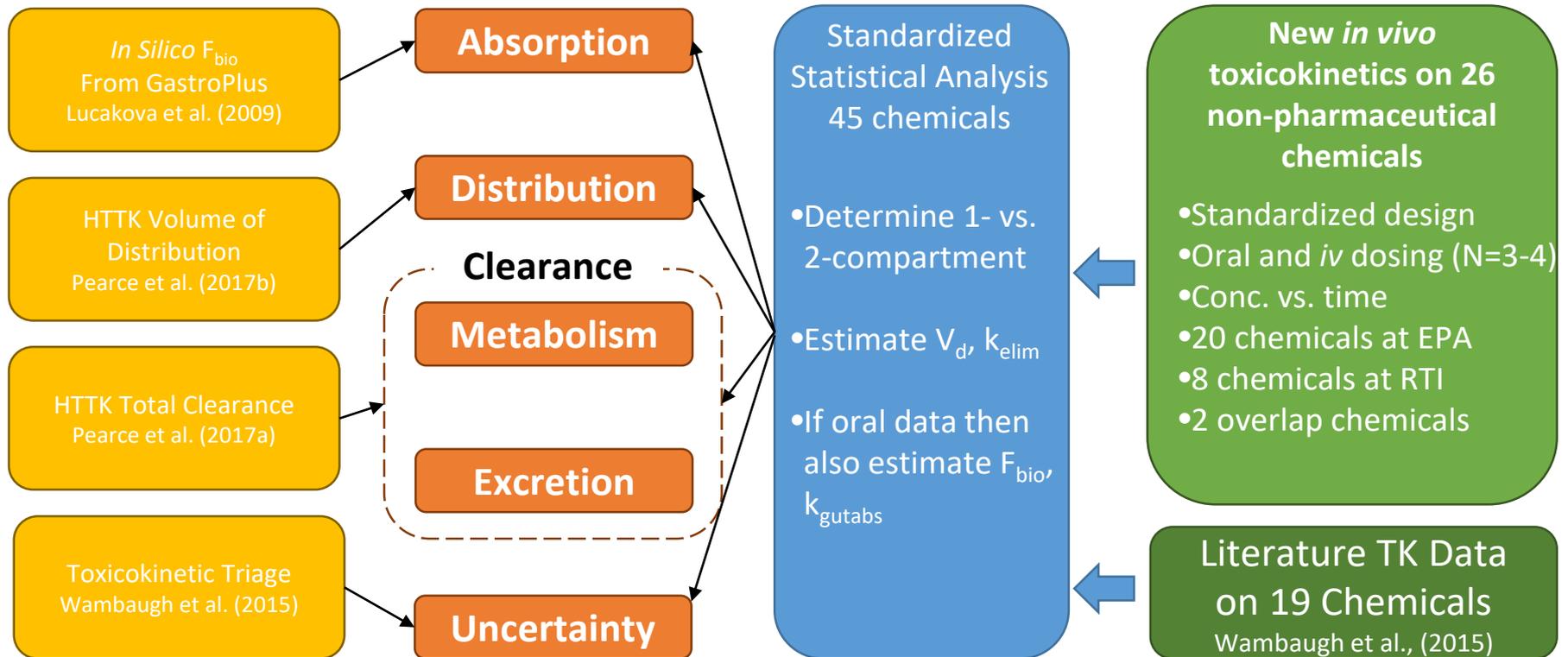
Toxicokinetic Triage

- Through comparison to existing *in vivo* data, a cross-validated (random forest) predictor of success or failure of HTTK was constructed
- We added categories for chemicals that do not reach steady-state or for which plasma binding assay fails
- All chemicals can be placed into one of seven confidence categories
- Plurality of chemicals end up in the “on the order” bin (within a factor of 3.2x) which is consistent with Wang (2010)



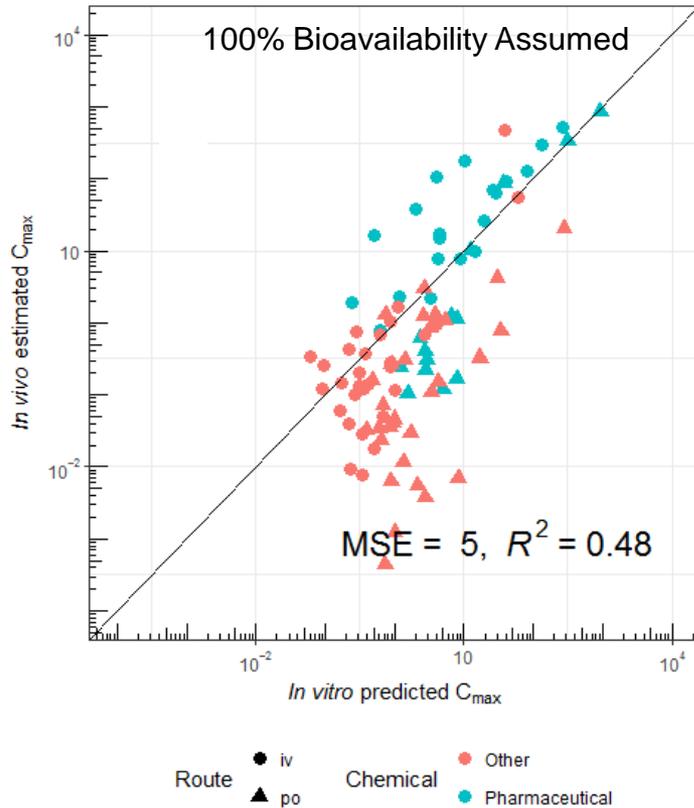
New Data for Evaluation

Available literature *in vivo* TK evaluation data was heavily biased toward pharmaceuticals



Wambaugh et al. (2018)

Evaluating HTTK

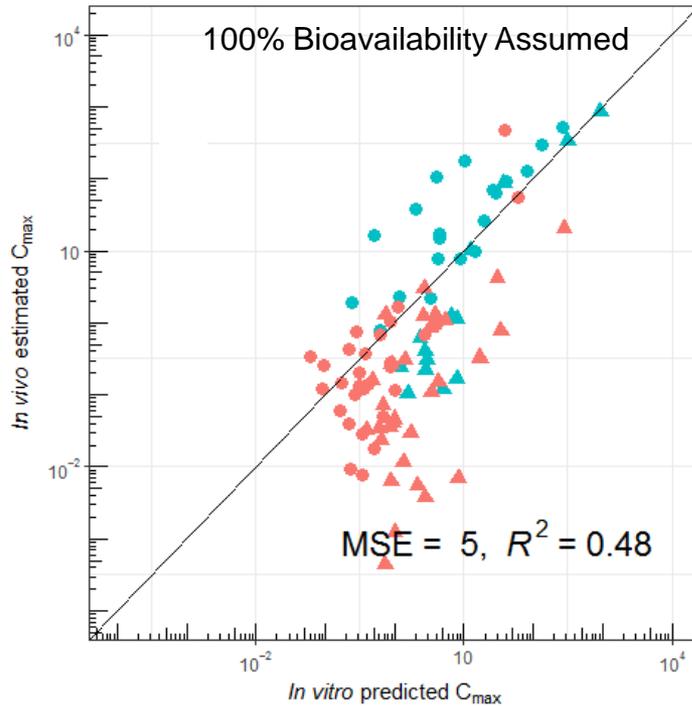


We evaluate HTTK by comparing predictions with observations for as many chemicals as possible

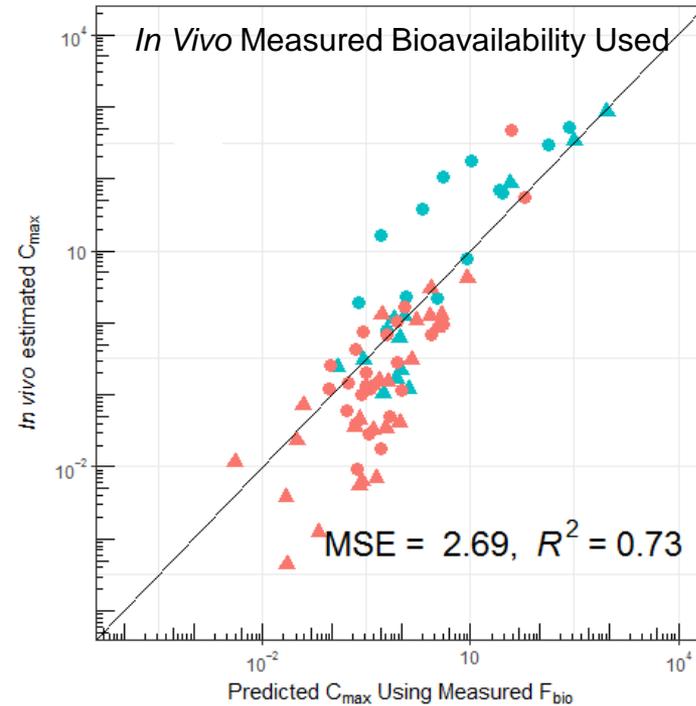
Wambaugh et al. (2018)

Evaluating HTTK

Impact of Oral Bioavailability Data



Route ● iv Chemical ● Other
 ▲ po ● Pharmaceutical

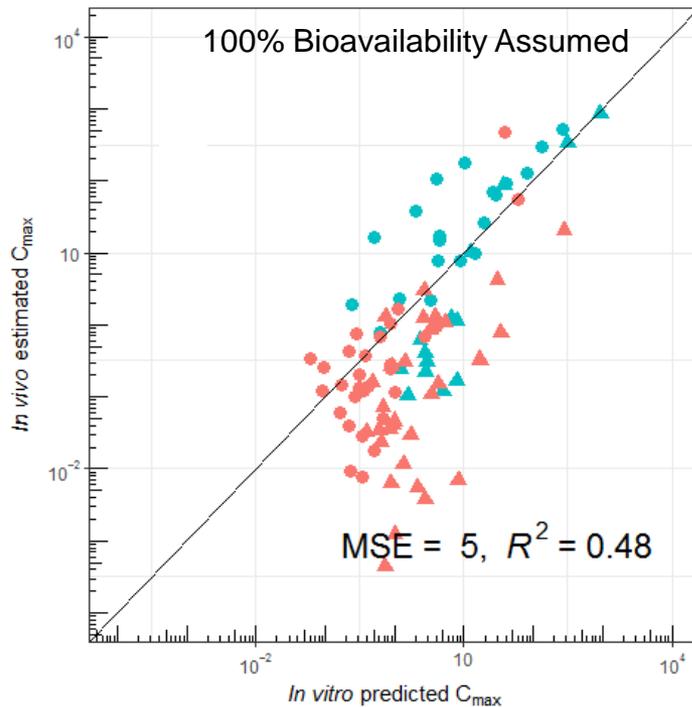


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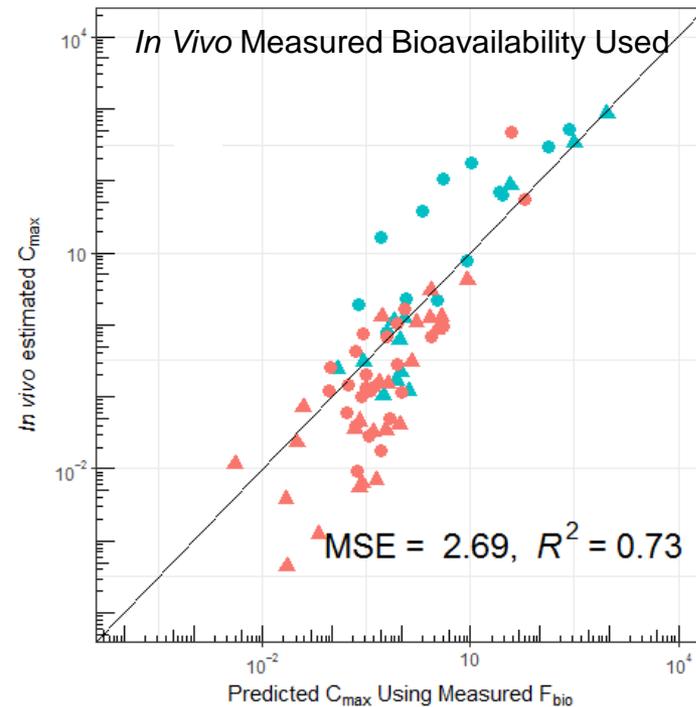
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Evaluating HTTK

Impact of Oral Bioavailability Data



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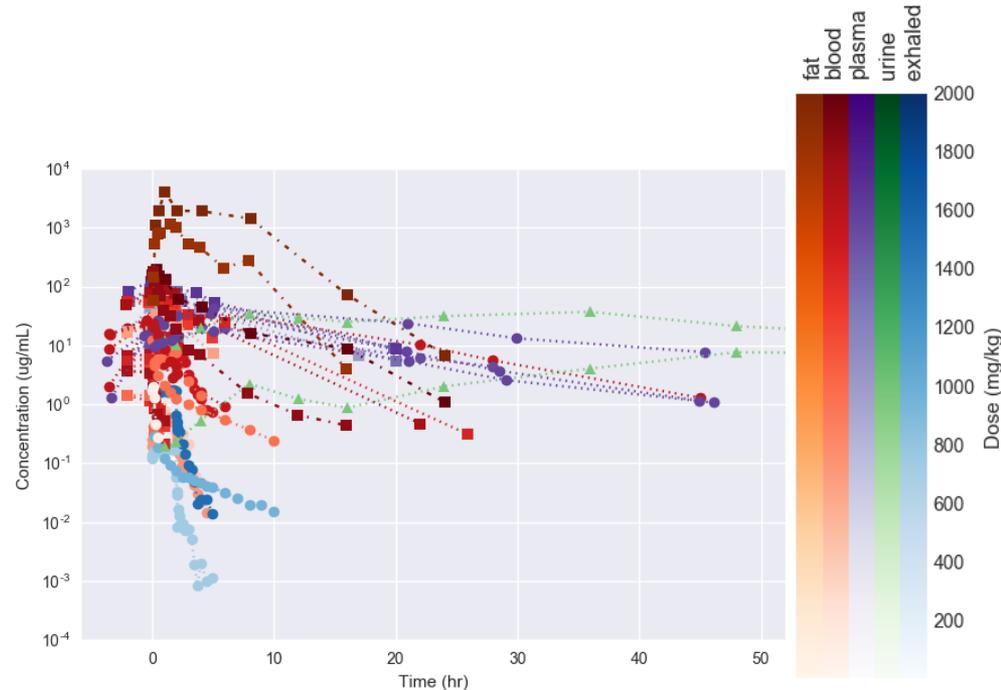
Route ● iv Chemical ● Other
▲ po ● Pharmaceutical

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

In Vivo TK Database

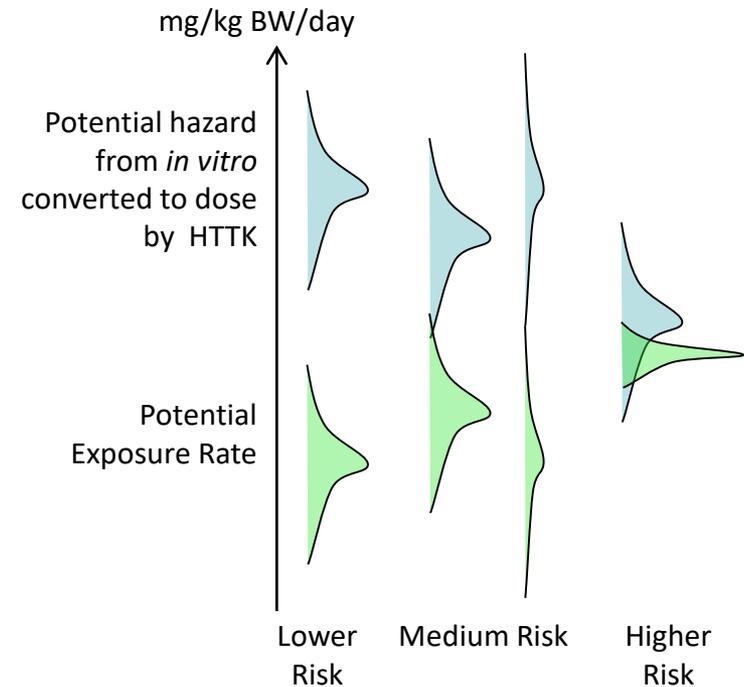
- EPA is developing a public database of concentration vs. time data for building, calibrating, and evaluating TK models
- Curation and development ongoing, but to date includes:
 - 175 analytes (EPA, National Toxicology Program, literature)
 - Routes: Intravenous, dermal, oral, sub-cutaneous, and inhalation 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>



Conclusions

- We would like to know more about the risk posed by thousands of chemicals in the environment – which ones should we start with?
- HTK 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 (i.e., “points of departure”).
- Comparison between HTK predicted time course concentrations in plasma and *in vivo* data indicate that some properties (e.g. average and maximum concentration) can be predicted with confidence.
- Comparison between *in vitro* bioactivity data and HTK-adjusted internal dose predictions for *in vivo* points of departure has refined assumptions of the HTK NAMs.
- NAMs for TK allow risk-based prioritization of large numbers of chemicals.



Rapid Exposure and Dosimetry (RED) Project

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***Trainees**

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