

Evaluating New Approach Methodologies for use in Next Generation Risk Assessment

Dr Alistair Middleton

Science leader in Computational Toxicology

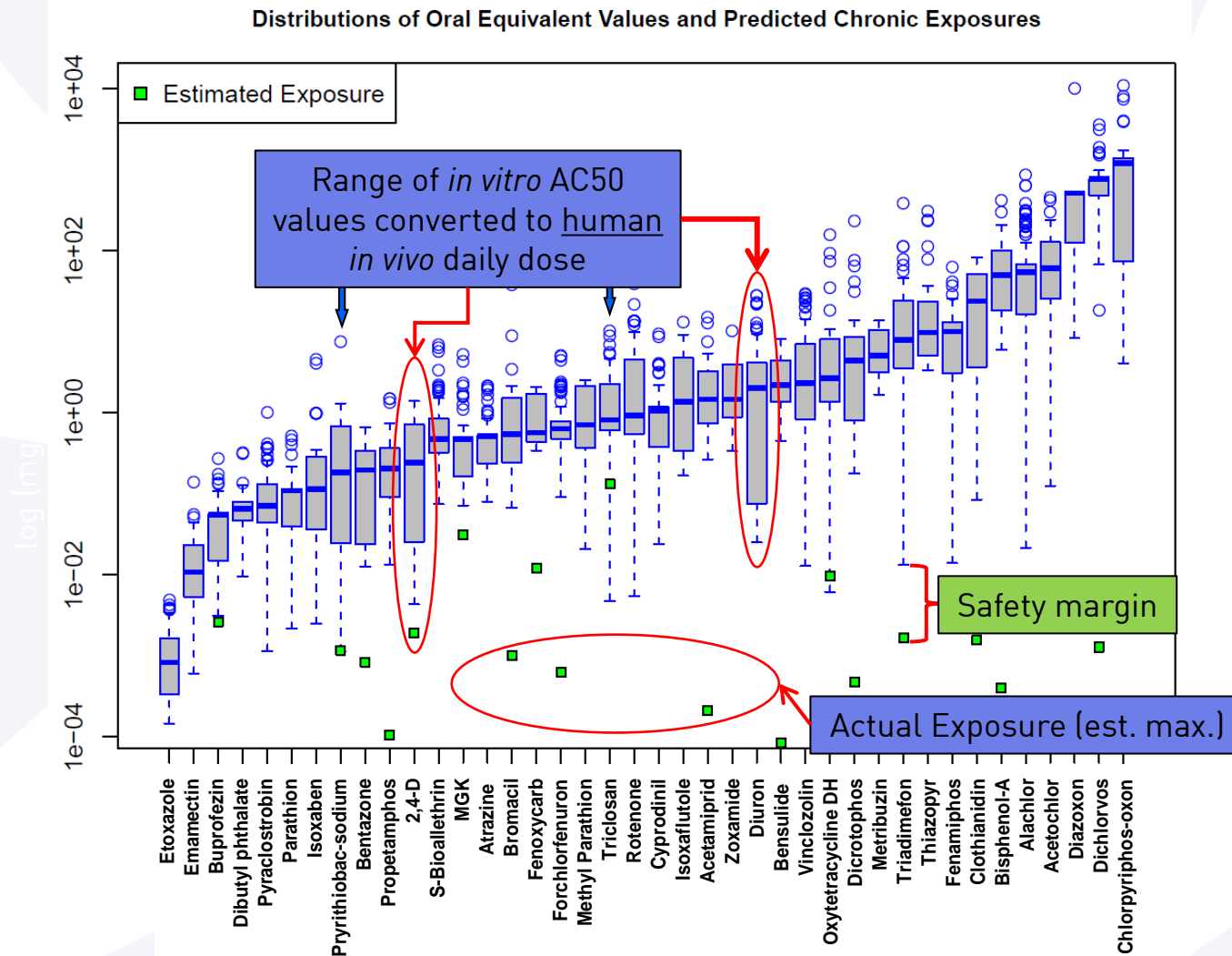
Unilever Safety & Environmental Assurance Centre (SEAC)



Unilever



Paradigm shift for systemic safety - Protection not Prediction



The hypothesis underpinning this type of NGRA is that **if there is no bioactivity observed at consumer-relevant concentrations, there can be no adverse health effects.**



Slide from Dr Rusty Thomas, EPA, with thanks

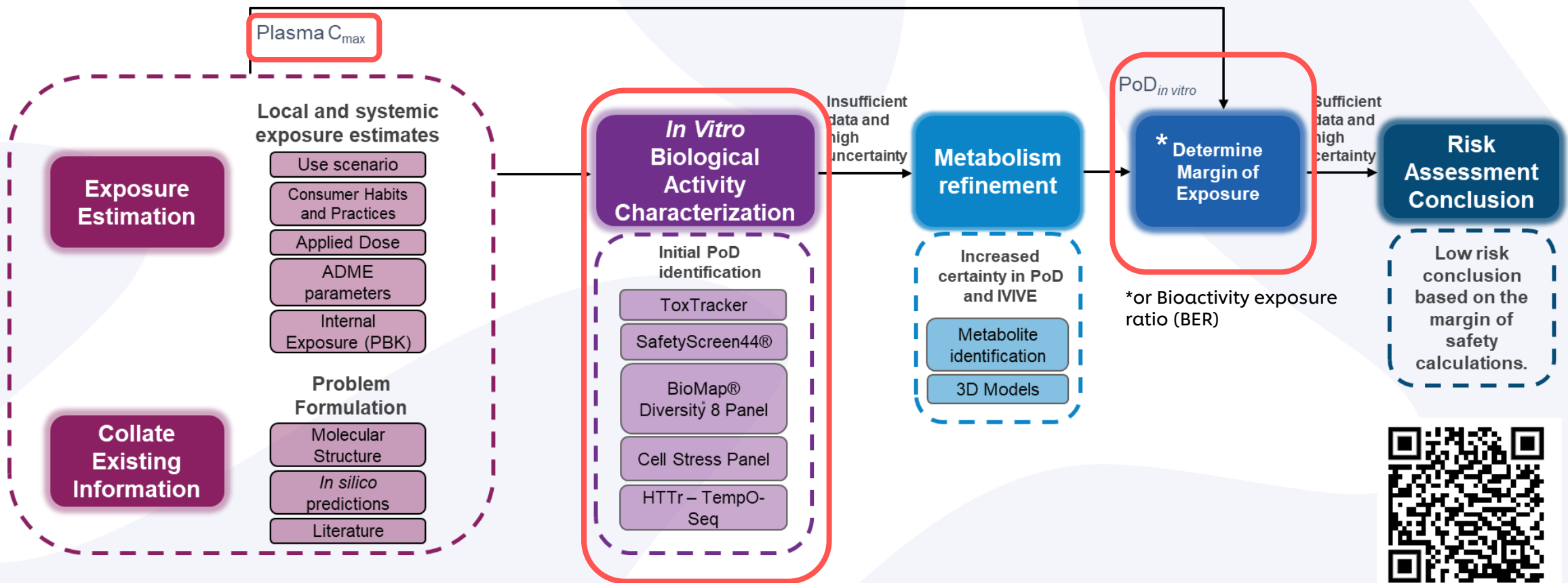
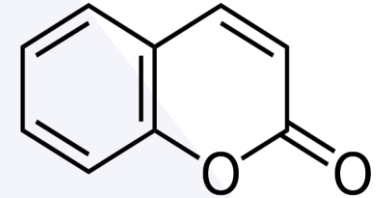
Rotroff, *et al.* Tox.Sci 2010

Thomas RS et al., 2019. Tox Sci. 1;169(2):317-332.



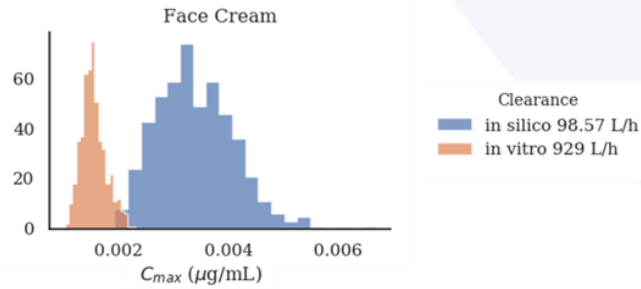
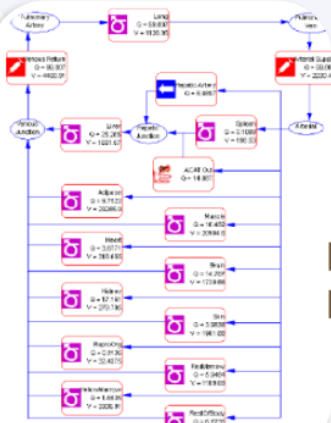
Example how to integrate NAMs for a NGRA: coumarin case study

0.1% COUMARIN IN FACE CREAM AND BODY LOTION (NEW FRAGRANCE)



The key NAMs in our NGRA approach

PBK Modelling



Toxicology in Vitro (2020), 63, 104746

In vitro pharmacological profiling

PERSPECTIVES

A GUIDE TO DRUG DISCOVERY — OPINION

Reducing safety-related drug attrition: the use of *in vitro* pharmacological profiling

Joanne Breen, Andrew J. Breen, Jacques Homan, Wolfgang Juratnik, Arun Sridhar, Gareth Waldron and Steven Whitbread

Abstract | *In vitro* pharmacological profiling is increasingly being used earlier in the drug discovery process to identify undesirable off-target activity profiles that could hinder or halt the development of candidate drugs or even lead to market withdrawal if discovered after a drug is approved. Here, for the first time, the rationale, strategies and methodologies for *in vitro* pharmacological profiling at four major pharmaceutical companies (AstraZeneca, GlaxoSmithKline, Novartis and Pfizer) are presented and illustrated with examples of their impact on the drug discovery process. We hope that this will enable other companies and academic institutions to benefit from this knowledge and consider joining us in our collaborative knowledge sharing.

Decreasing the high attrition rate in the drug discovery and development process is a primary goal of the pharmaceutical industry. One of the main challenges in achieving this goal is striking an appropriate balance between drug efficacy and potential adverse effects as early as possible in order to reduce safety-related attrition, particularly in the more expensive late stages of clinical development. Gaining a better understanding of the safety profile of drug candidates early in the process is also crucial for reducing the likelihood of safety issues limiting the use of approved drugs, or even leading to their market withdrawal, having to incur the massive financial and regulatory costs.

target (or targets), whose secondary effects are due to interactions with targets other than the primary target (or targets) that is off-target interactions. Off-target interactions are often the cause of ADRs in animal models or clinical studies, and careful characterization and identification of secondary pharmacology profiles of drug candidates early in the drug discovery process might help to reduce the incidence of type A ADRs.

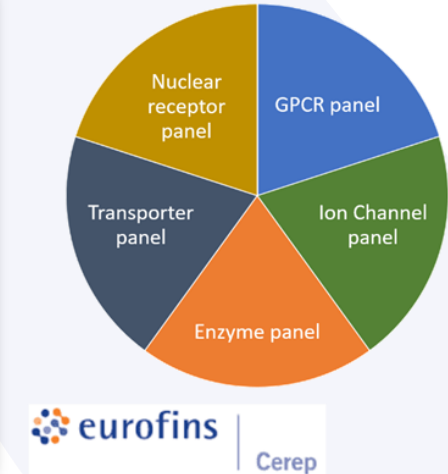
In vitro pharmacological profiling involves the screening of compounds against a broad range of targets (receptors, enzymes, ion channels, transporters, etc.) that are chosen from the scientific

safety testing of drug candidates and are designed to prevent serious ADRs from occurring in clinical studies.

The *in vitro* pharmacology assay that is absolutely required by regulatory authorities is one that measures the effects of new chemical entities on the ionotropic receptors (IC₅₀) or heterotrimeric G-protein-coupled receptor (GPCR) pathway (classically 11 member 2 (H2N12), also known as H2ERG). The mechanism by which blockade of H2ERG can affect potentially fatal cardiac arrhythmias (torsades de pointes) following a prolongation of the QT interval is well characterized^{1,2}, and the assessment of this ADR is one reason why this assay is a mandatory regulatory requirement. Receptor binding studies are also recommended as the first tier approach for the assessment of the dependence potential of novel chemical entities³.

However, current regulatory guidance does not describe which targets should constitute an *in vitro* pharmacological profiling panel and does not indicate the stage of the discovery process at which *in vitro* pharmacological profiling should occur. Nevertheless, the general view for most pharmaceutical companies is to perform this testing early in drug discovery to reduce attrition and to facilitate better production of ADRs in the later stages of drug discovery and development.

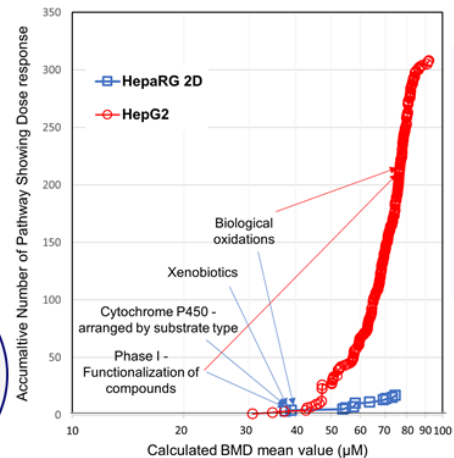
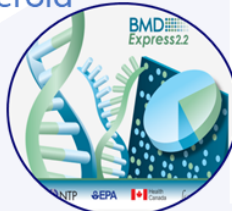
Here, for the first time, four major pharmaceutical companies (AstraZeneca, GlaxoSmithKline, Novartis and Pfizer) share their knowledge and experience of their innovative application of existing screening technologies to detect off-target interactions of compounds. The objective of this article is to describe the rationale and main strategies for the use of *in vitro* pharmacological profiling to reduce drug attrition and to



Transcriptomics

- Use of full human gene panel ~ 21k
- 24 hrs exposure
- 7 concentrations
- 3 cell lines HepG2/ HepaRG/ MCF7
- 3D HepaRG spheroid

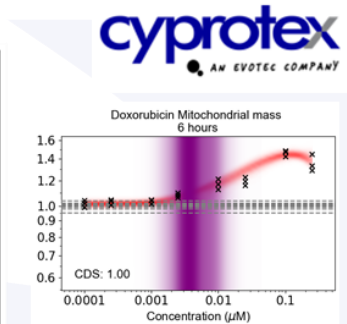
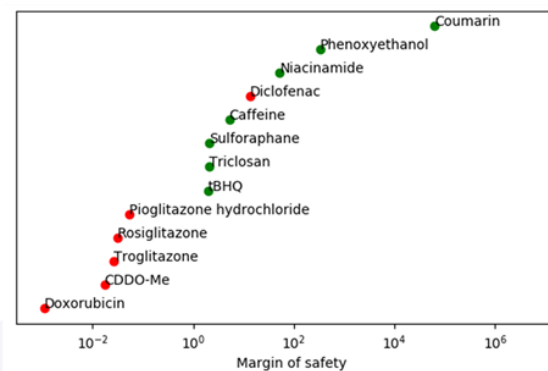
BMDexpress 2



Cellular Stress Pathways

13 chemicals, 36 Biomarkers; 3 Timepoints; 8 Concentrations; ~10 Stress Pathways

- Exposure scenario adopted for chemical is 'low risk'** (from consumer goods perspective)
- Nicotinamide [food, cosmetics]
 - Caffeine [beverages, cosmetics]
 - Phenoxyethanol [cosmetics]
 - Sulfuraphane [food]
 - tBHQ [antioxidant]
 - Triclosan [antimicrobial]
- Exposure scenario adopted for chemical is 'high risk'** (from consumer goods perspective)
- CDDO-Me [drug]
 - DEM [industrial chemical]
 - Doxorubicin [drug]
 - Diclofenac [drug]
 - Troglitazone [drug]
 - Pioglitazone [drug]
 - Rosiglitazone [drug]



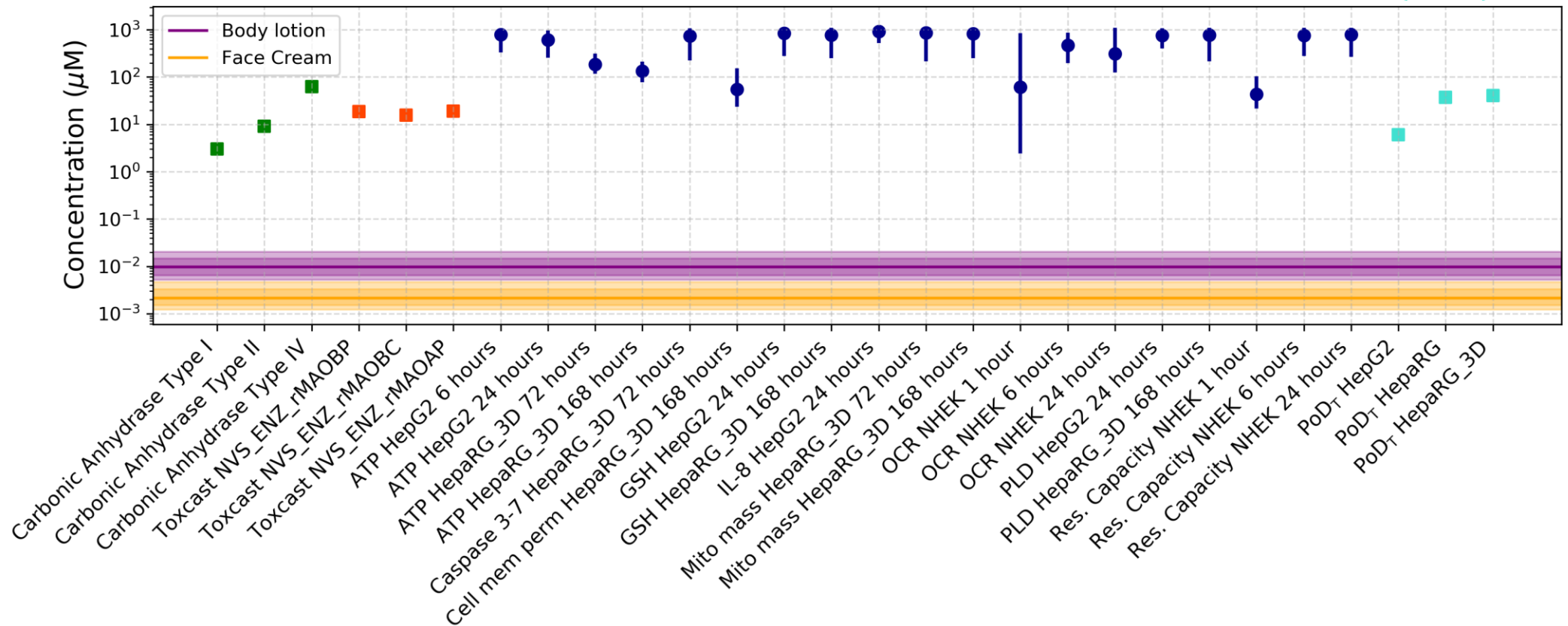
Toxicol Sci (2020), 176, 11-33

Exposure and PoD are plotted and used to derive a Bioactivity-Exposure Ratio (BER)

In-vitro pharmacological profiling (IPP)

Cell Stress Panel

High throughput transcriptomics (HTTR)



The 5th percentile of the BER distribution ranged between 158 and 96738

In this case study: Weight of evidence suggested that the inclusion of 0.1% coumarin in face cream or a body lotion is safe for the consumer

How do we build scientific confidence in a systemic safety toolbox?

1. Determine whether the toolbox is fit for purpose (leads to safety decisions that are protective of human health).
2. Take into account human safety in assessing the approach (where possible)
3. Identify what an appropriate safety decision might be (e.g., BER threshold).

Accelerating the Pace of Chemical Risk Assessment (APCRA)

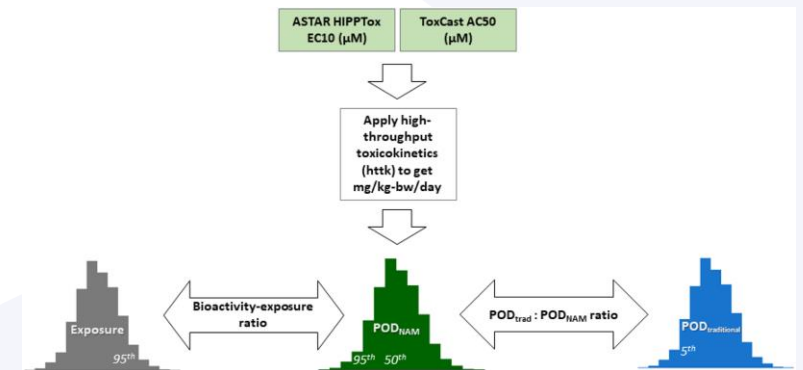


TOXICOLOGICAL SCIENCES, 173(1), 2020, 202–225

doi: 10.1093/toxsci/kfz201
Advance Access Publication Date: September 18, 2019
Research Article

Utility of *In Vitro* Bioactivity as a Lower Bound Estimate of *In Vivo* Adverse Effect Levels and in Risk-Based Prioritization

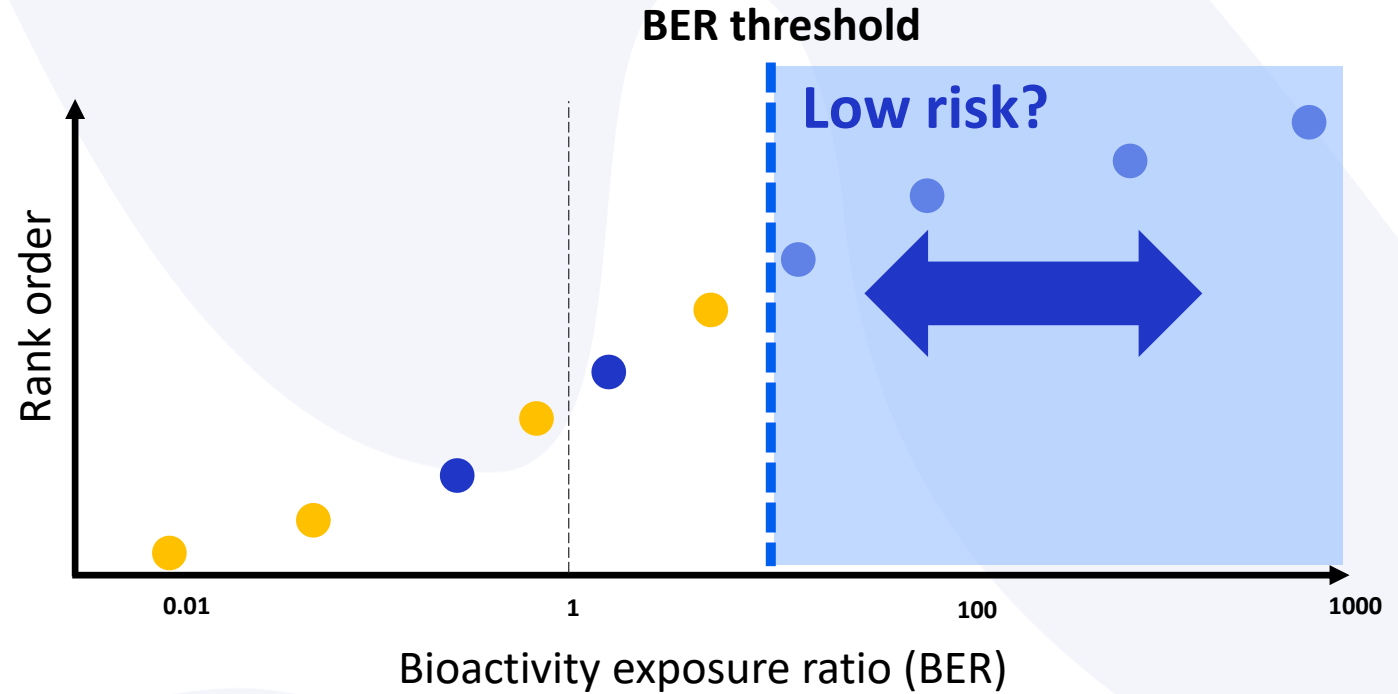
Katie Paul Friedman ,^{*,1} Matthew Gagne,[†] Lit-Hsin Loo,[‡] Panagiotis Karamertzanis,[§] Tatiana Netzeva,[§] Tomasz Sobanski,[§] Jill A. Franzosa,[¶] Ann M. Richard,^{*} Ryan R. Lougee,^{*,||} Andrea Gissi,[§] Jia-Ying Joey Lee,[‡] Michelle Angrish,^{|||} Jean Lou Dorne,^{|||} Stiven Foster,[#] Kathleen Raffaele,[#] Tina Bahadori,^{||} Maureen R. Gwinn,^{*} Jason Lambert,^{*} Maurice Whelan,^{**} Mike Rasenberg,[§] Tara Barton-Maclaren,[†] and Russell S. Thomas *



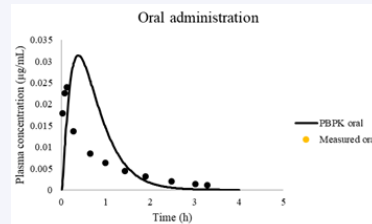
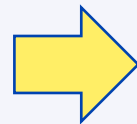
How do we build scientific confidence in a systemic safety toolbox?

Chemical exposures scenarios

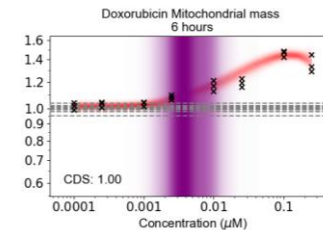
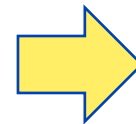
- 'Low' risk (from consumer goods perspective) – e.g. foods, cosmetics
- 'High' risk (from consumer goods perspective) – e.g. drugs



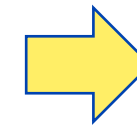
Define typical use-case scenarios benchmark chemical-exposures; Mixture of High and low risk



PBK models of systemic exposure



In-vitro cell assays, estimate PoDs



Calculate the bioactivity exposure ratio

Visualising how the toolbox performs against the pilot study data



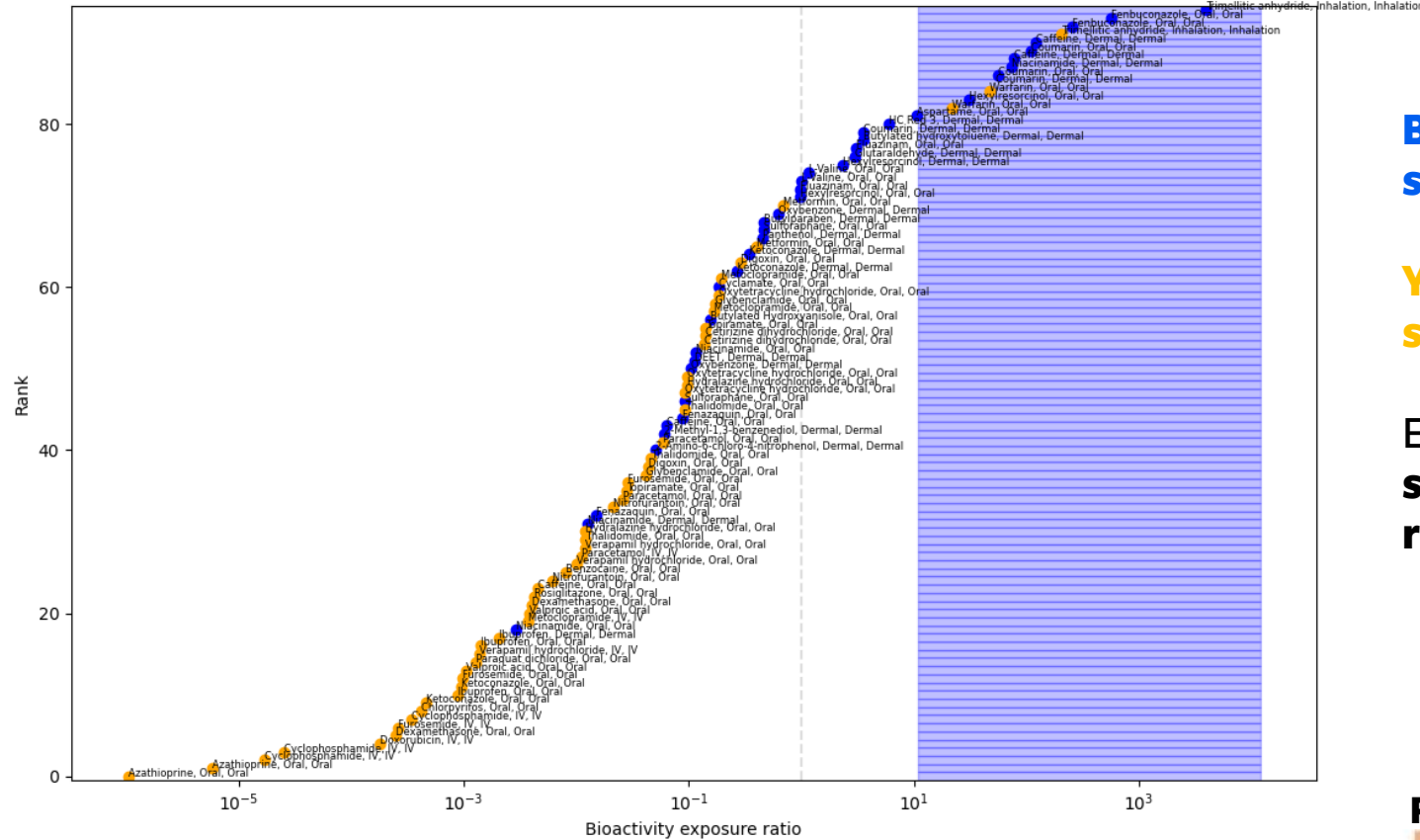
low risk chemical-exposure

low risk chemical-exposure

scenarios within the **blue**
 region are identified as **low**

Extending the evaluation to 38 chemicals and 70 exposure scenarios

PBK level: L2
 PoD types: IPP lowest IC50, CSP global PoD, HTTr global PoD, Minimum pathway BMDL
 Protectiveness: 52/55 (95%), Utility: 9/39 (23%)
 Correlation: -0.58



Blue: low risk chemical-exposure scenario

Yellow: high risk chemical-exposure scenario

Exposure scenarios within the **blue shaded region** are identified as **low risk**.

Protectiveness and utility metrics

$$\text{Protectiveness} = \frac{H_U}{H_U + H_L}$$

$$\text{Utility} = \frac{L_L}{L_L + L_U}$$

H_U - # of high risk exposures identified as uncertain risk
 H_L - # of high risk exposures identified as low risk

L_U - # of low risk exposures identified as uncertain risk
 L_L - # of low risk exposures identified as low risk

Discussion

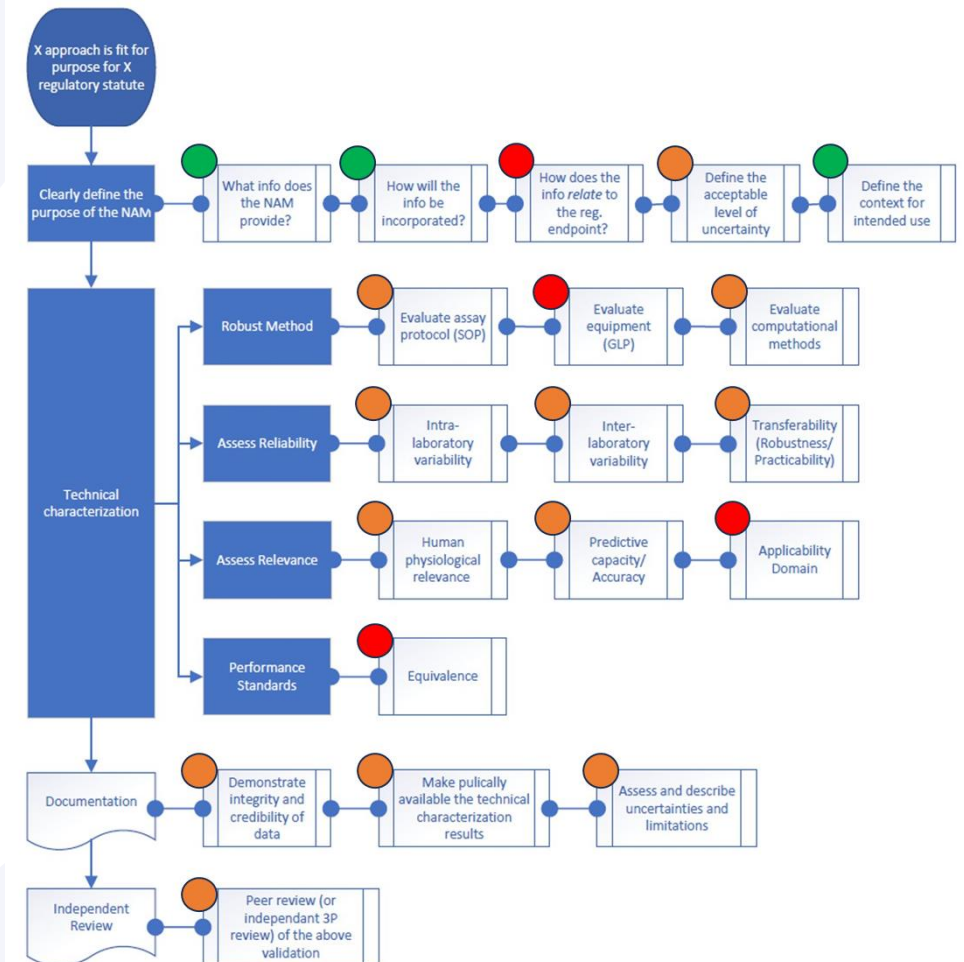
- Have now extended the evaluation to 38 chemicals with 70 associated **high risk** and **low risk** exposure scenarios.
- Adopt **iterative approach** to evaluating and then identifying potential improvements to the toolbox.
- **Unilever-EPA CRADA**: Generating data for 10 cell lines, using high-throughput transcriptomics and phenotypic profiling.
- The overall objective is to establish the **scientific confidence** that the toolbox is fit for purpose.
- In the process of mapping activities against existing NAM validation criteria (inc van der Zalm (2022) and OECD TG34)



A framework for establishing scientific confidence in new approach methodologies

Anna J. van der Zalm¹ · João Barroso² · Patience Browne³ · Warren Casey⁴ · John Gordon⁵ · Tala R. Henry⁶ · Nicole C. Kleinstreuer⁷ · Anna B. Lowit⁶ · Monique Perron⁸ · Amy J. Clippinger¹

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US-EPA: Richard Judson, Josh Harrill, Logan Everett, Imran Shah



Thank You



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Overall evaluation strategy

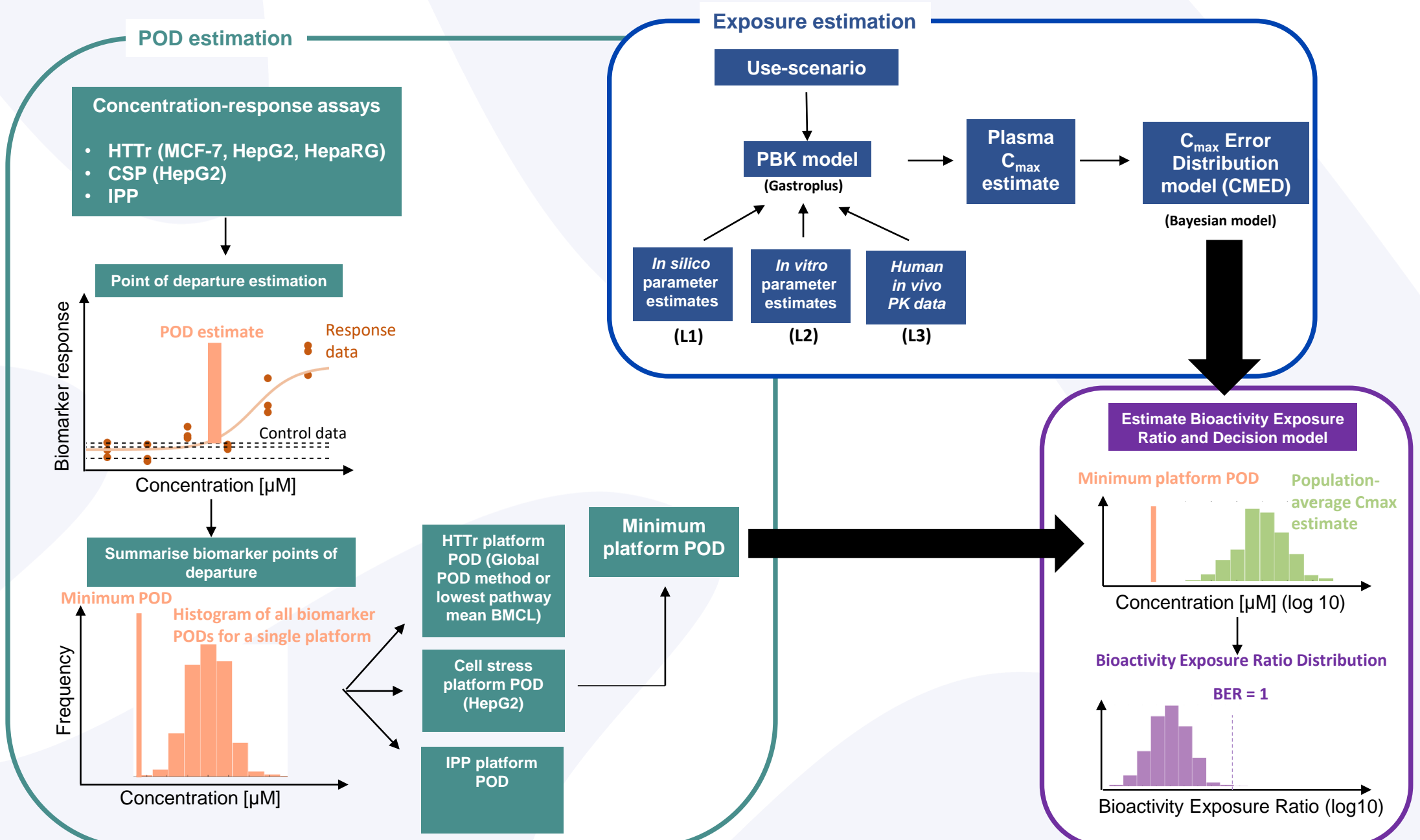
Step 1 (pilot study)*

- Define what the toolbox contains (which NAMs) and the workflow through which they should be used.
- Define process of how the toolbox will be evaluated, and the metrics that will be used to determine its 'performance'
- Explore using a small set of chemicals and exposure scenarios (11 chemicals, 25 exposure scenarios)
- Define **prototype decision model** for determining the BER threshold.

Step 2 (full evaluation)

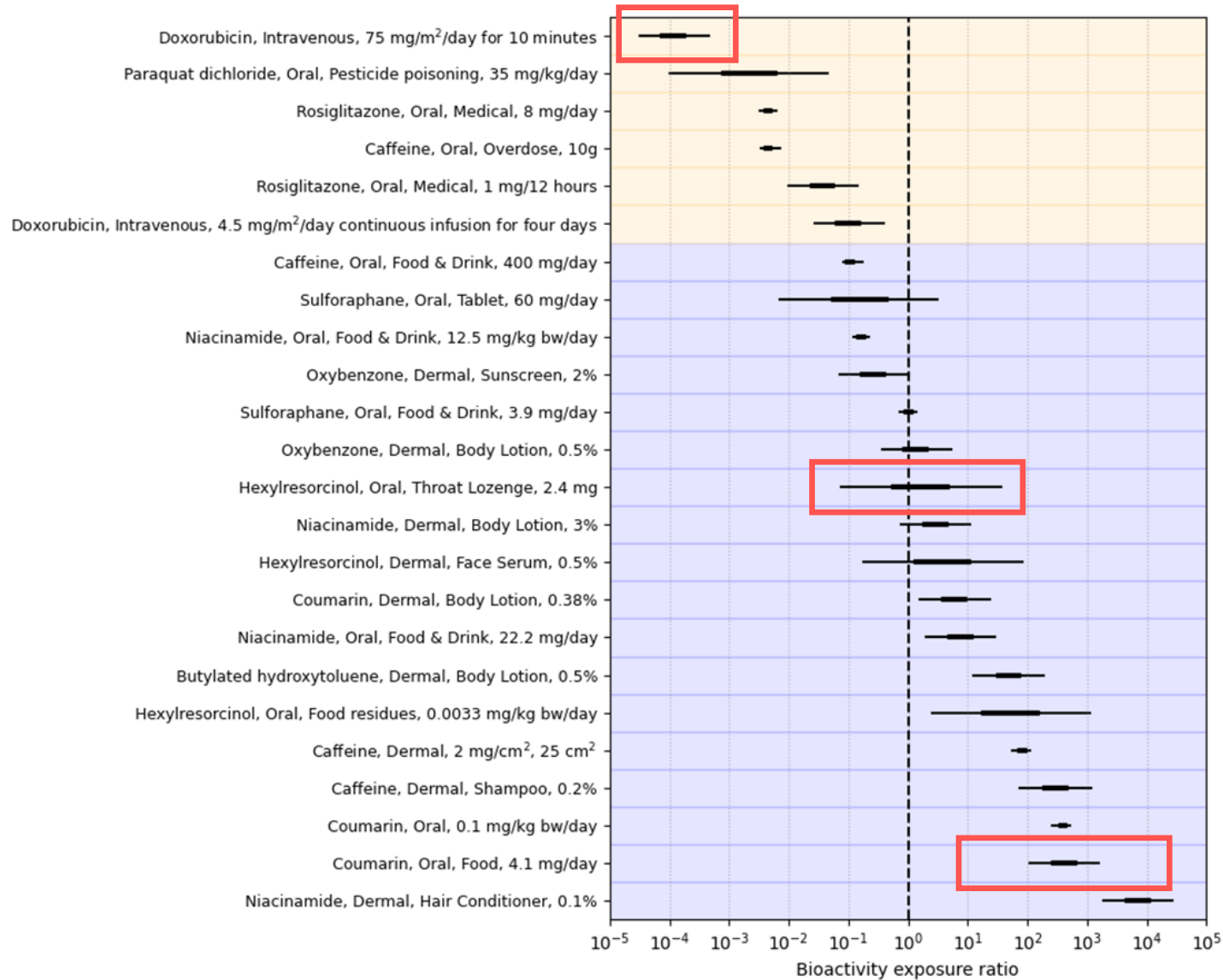
- Evaluate the toolbox using ~40 chemicals with ~100 exposure scenarios based on the toolbox established in the pilot study.
- Use learnings from the toolbox evaluation to refine the toolbox in terms of **NAM composition** and the **decision model**.

Stage 3: Estimating the BER from the toolbox



Stage 3: Estimating the BER from the toolbox

BER=1



Blue: low risk chemical-exposure scenario

Yellow: high risk chemical-exposure scenario

BER=1:
Cmax estimates coincide with the minimum POD

Progress in the application of NAMs in NGRA for systemic safety

NAMs applied in an *ab initio* hypothetical/NGRA case study (e.g. coumarin and phenoxyethanol)

 **SOT** | Society of Toxicology
academic.oup.com/toxsci


TOXICOLOGICAL SCIENCES, 176(1), 2020, 236–252
doi: 10.1093/toxsci/kfaa048
Advance Access Publication Date: April 10, 2020
Research article

A Next-Generation Risk Assessment Case Study for Coumarin in Cosmetic Products

Maria T. Baltazar,¹ Sophie Cable, Paul L. Carmichael, Richard Cubberley, Tom Cull, Mona Delagrange, Matthew P. Dent, Sarah Hatherell, Jade Houghton, Predrag Kukic, Hequn Li, Mi-Young Lee, Sophie Malcomber, Alistair M. Middleton, Thomas E. Moxon, Alexis V. Nathanail, Beate Nicol, Ruth Pendlington, Georgia Reynolds, Joe Reynolds, Andrew White, and Carl Westmoreland

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 **OECD**
Organisation for Economic Co-operation and Development

ENV/CBC/MONO(2021)35

Unclassified English - Or. English
27 October 2021

ENVIRONMENT DIRECTORATE
CHEMICALS AND BIOTECHNOLOGY COMMITTEE

Case Study on use of an Integrated Approach for Testing and Assessment (IATA) for Systemic Toxicity of Phenoxyethanol when included at 1% in a body lotion

Series on Testing and Assessment,
No. 349

NAMs applied in real-life chemical safety assessments

APPLIED IN VITRO TOXICOLOGY
Volume 7, Number 2, 2021
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Use of the MucilAir Airway Assay, a New Approach Methodology, for Evaluating the Safety and Inhalation Risk of Agrochemicals

Marie McGee Hargrove,^{1,1} Bob Parr-Dobrzanski,² Lei Li,³ Samuel Constant,⁴ Joanne Wallace,⁵ Paul Hinderliter,^{1,*} Douglas C. Wolf,¹ and Alex Charlton²

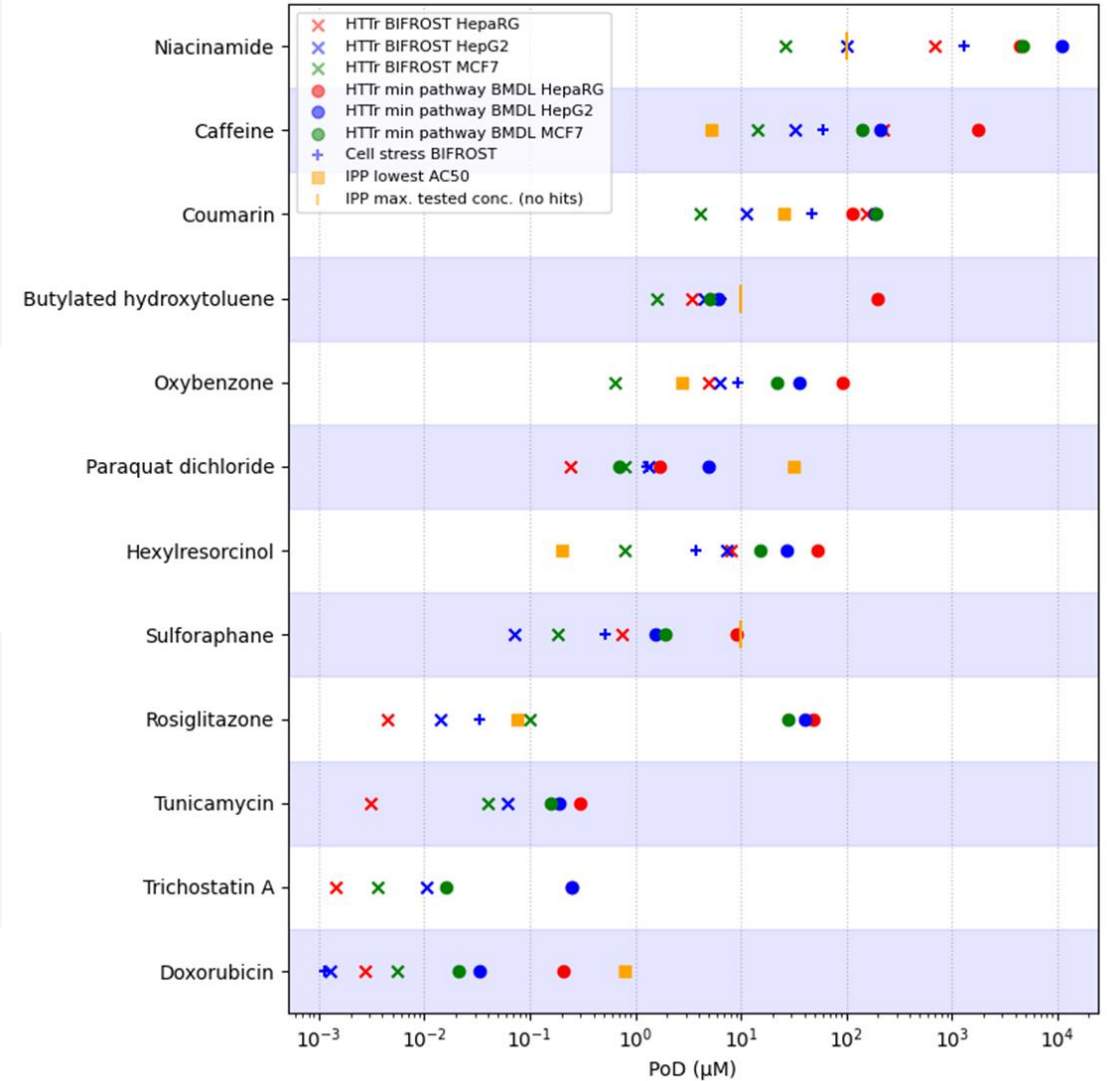
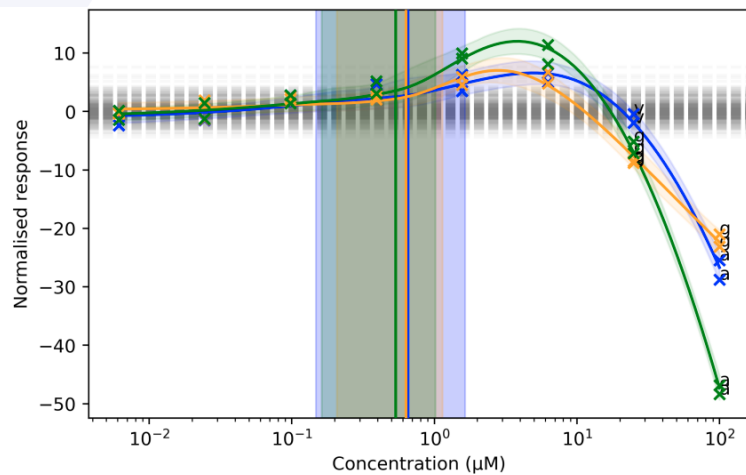
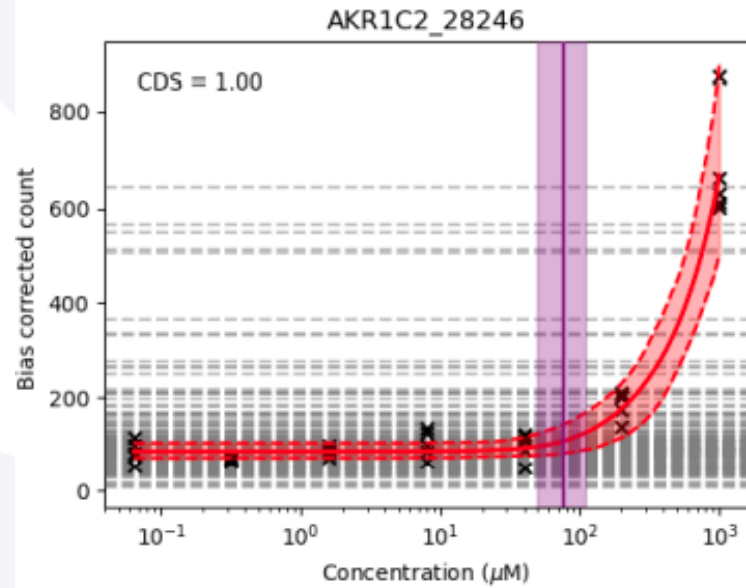


<https://www.regulations.gov/document/EPA-HQ-OPP-2011-0840-0080>

Stage 1: defining the benchmark chemical exposure scenarios

Chemical	Exposure scenario	Risk classification
Oxybenzone	2 scenarios: 0.5%; 2% sunscreen	Low risk
Caffeine	2 scenarios: 0.2% shampoo & coffee oral consumption 50 mg	Low risk
Caffeine	10g – fatal case reports	High risk
Coumarin	3 scenarios: 4 mg/d oral consumption; 1.6% body lotion (dermal); TDI 0.1 mg/kg oral	Low risk
Hexylresorcinol	3 scenarios: Food residues (3.3 ug/kg); 0.4% face cream; throat lozenge 2.4 mg	Low risk
BHT	Body lotion 0.5%	Low risk
Sulforaphane	2 scenarios: Tablet 60 mg/day; food 4.1-9.2 mg/day	Low risk
Niacinamide	4 scenarios: oral 12.5-22 mg/kg; dermal 3% body lotion and 0.1 % hair condition	Low risk
Doxorubicin	75 mg/m ² IV bolus 10 min; 21 days cycles; 8 cycles	High risk
Rosiglitazone	8 mg oral tablet	High risk
Valproic Acid (VPA)	2 scenarios: oral tablet 1000 mg & > 60 mg/kg	High risk
Paraquat	Accidental ingestion 35 mg/kg	High risk

Stage 2: Estimating PODs from the different bioactivity assays

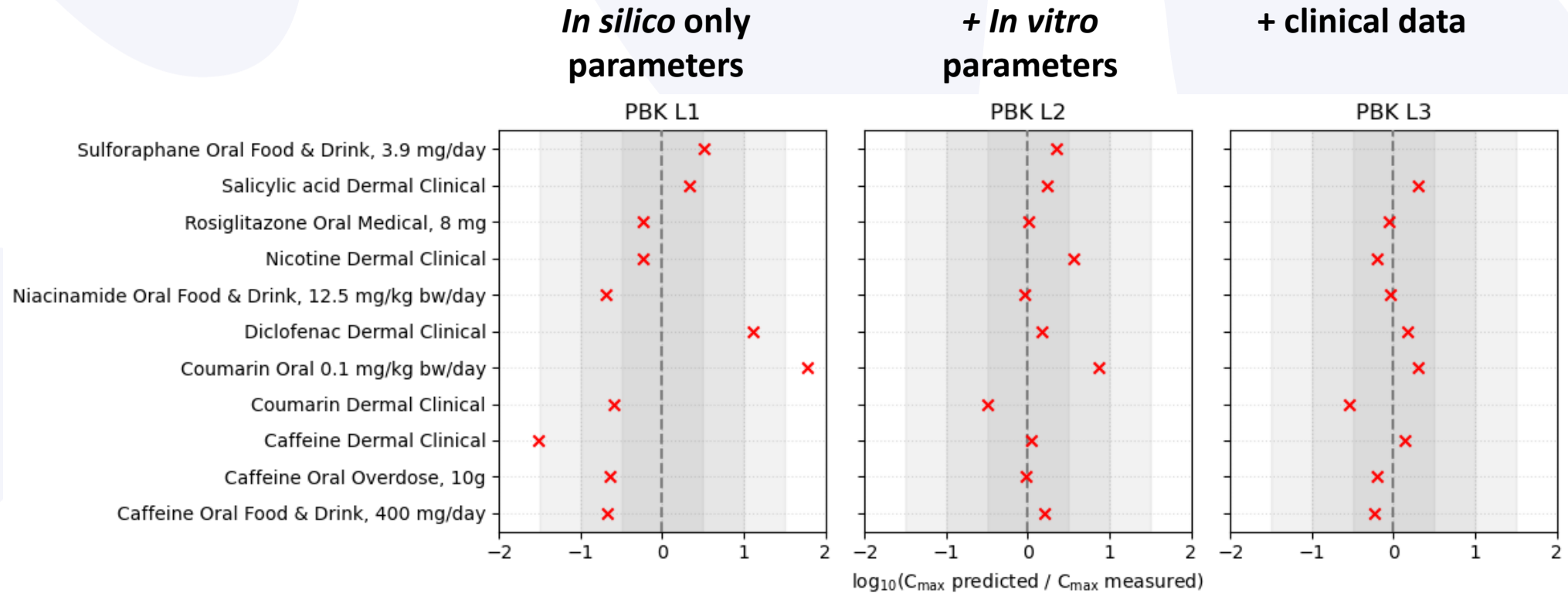


HTTr: High-throughput transcriptomics

CSP: Cell Stress Panel

IPP: In vitro pharmacological profiling

Considering the error in PBK models based on parameterisation level



- The PBK prediction error decreases as we go 'up' parameterisation levels
- Developed a Bayesian statistical model to quantify the error for a novel chemical