

Evaluating a systemic safety toolbox for use in Next Generation Risk Assessment

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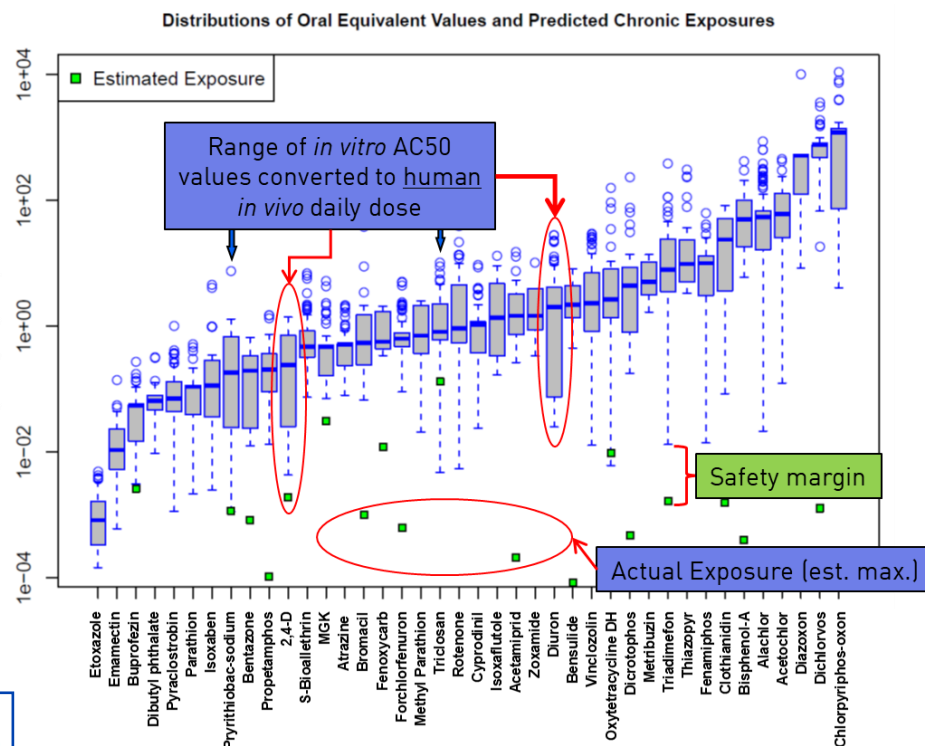


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Safety without animal testing - Next Generation Risk Assessment (NGRA)

NGRA is defined as *an exposure-led, hypothesis-driven* risk assessment approach that *integrates New Approach Methodologies (NAMs)* to assure *safety without the use of animal testing*

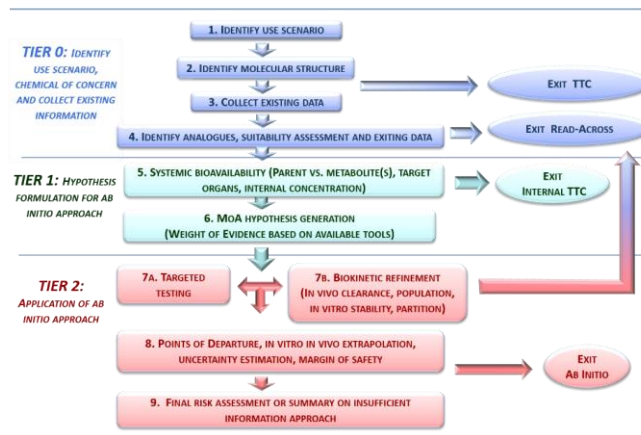


The hypothesis underpinning this type of NGRA is that **if there is no bioactivity observed at consumer-relevant concentrations, there can be no adverse health effects.**

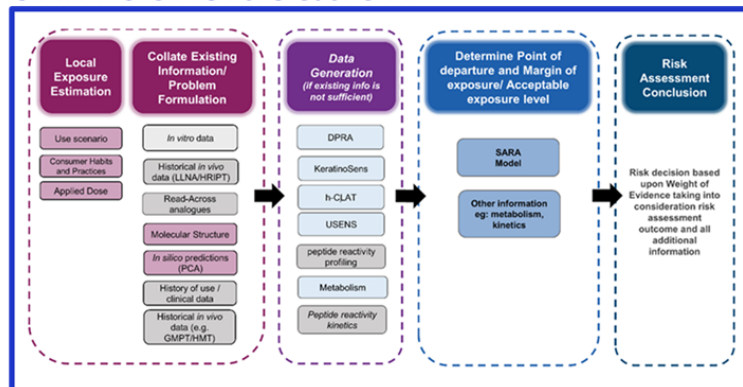
Slide from Dr Rusty Thomas,
EPA, with thanks

Rotroff, et al. *Tox.Sci* 2010

Decision frameworks in NGRA

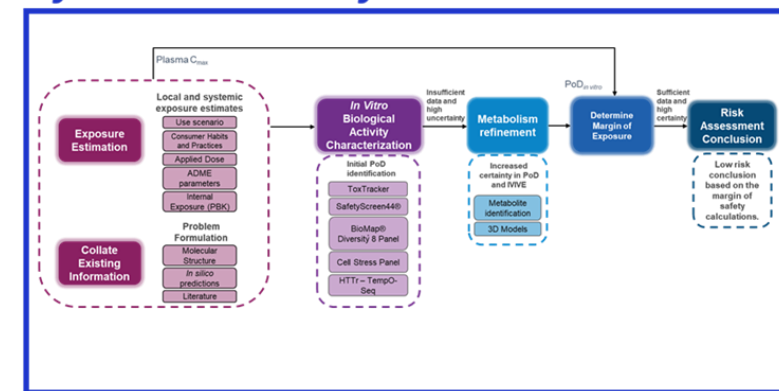


Skin Sensitisation

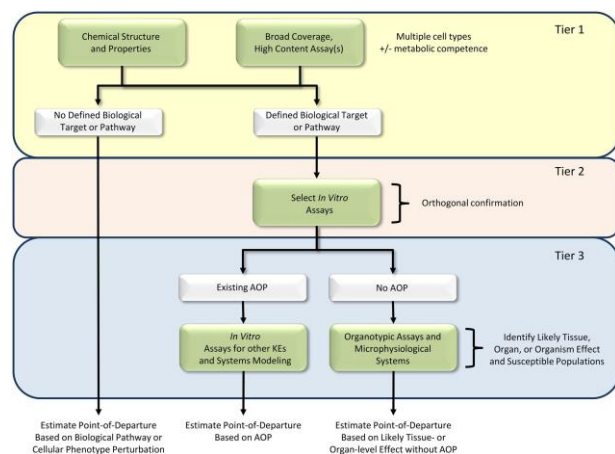


Reynolds et al (2021) Reg Tox Pharmacol, 127, 105075

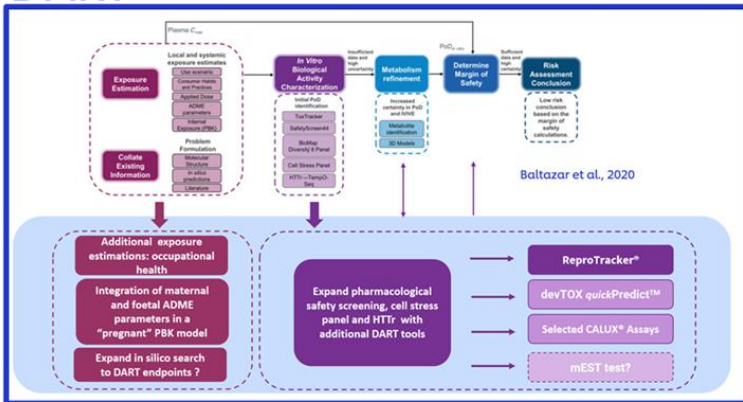
Systemic safety



Baltazar et al., (2020) Tox Sci, Volume 176, Issue 1, Pages 236–252

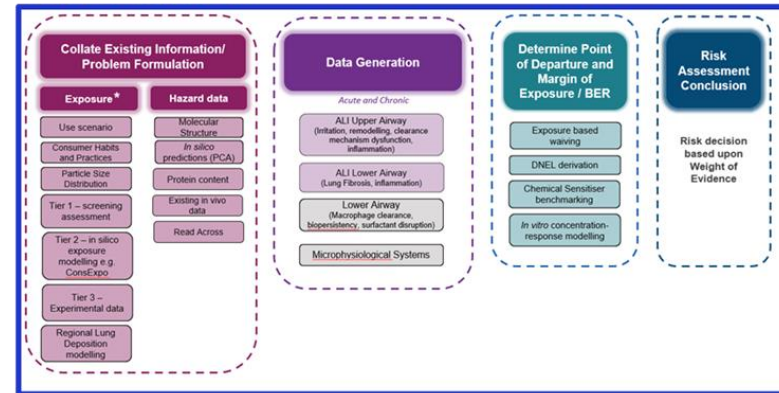


DART



Rajagopal et al (2022). Front. Toxicol., 07 March 2022

Inhalation



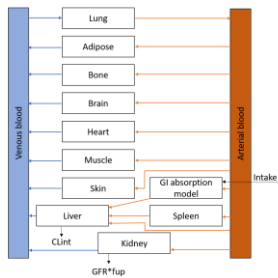
Decision frameworks in NGRA

Problem formulation – Tier 0

Initial BER estimate – Tier 1

Use case

Exposure route

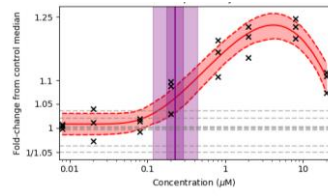


Exposure

Assay 1

Assay 2

Assay 3



Hazard

- Internal exposure levels estimated using PBK models
- Set of assays covering broad range of biological effects & PODs.
- Key output: Bioactivity Exposure Ratio ($BER = \text{POD} / \text{Exposure}$)

Uncertain risk: exposure may trigger bioactivity

Small BER (i.e., Exposure close to POD)

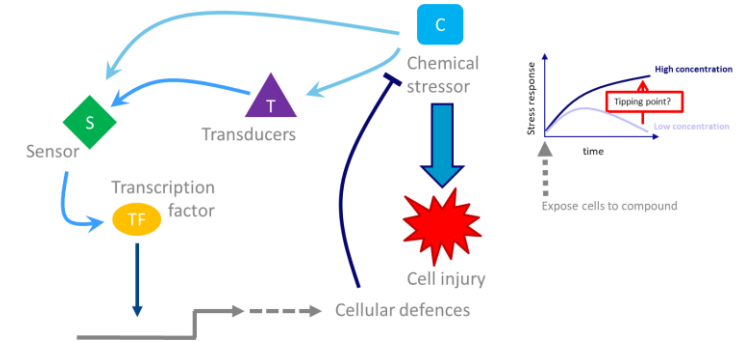
Low risk of exposure causing any bioactivity

Large BER (i.e., Exposure \ll POD)

Hypothesis: If a chemical ingredient does not trigger any bioactivity at human relevant exposure levels, then there can be no adverse health effects.

Safety Decision

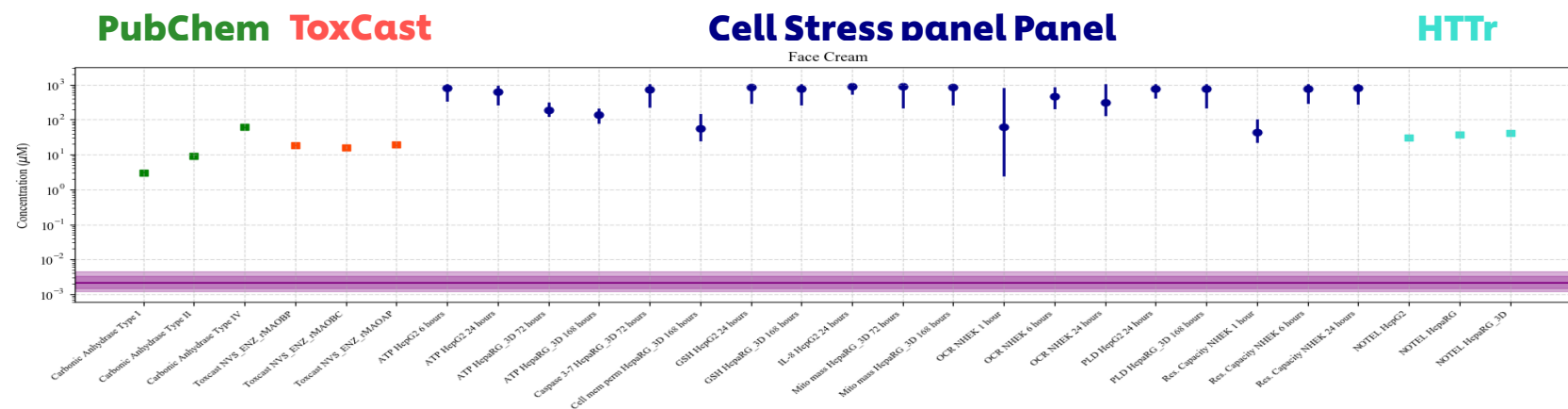
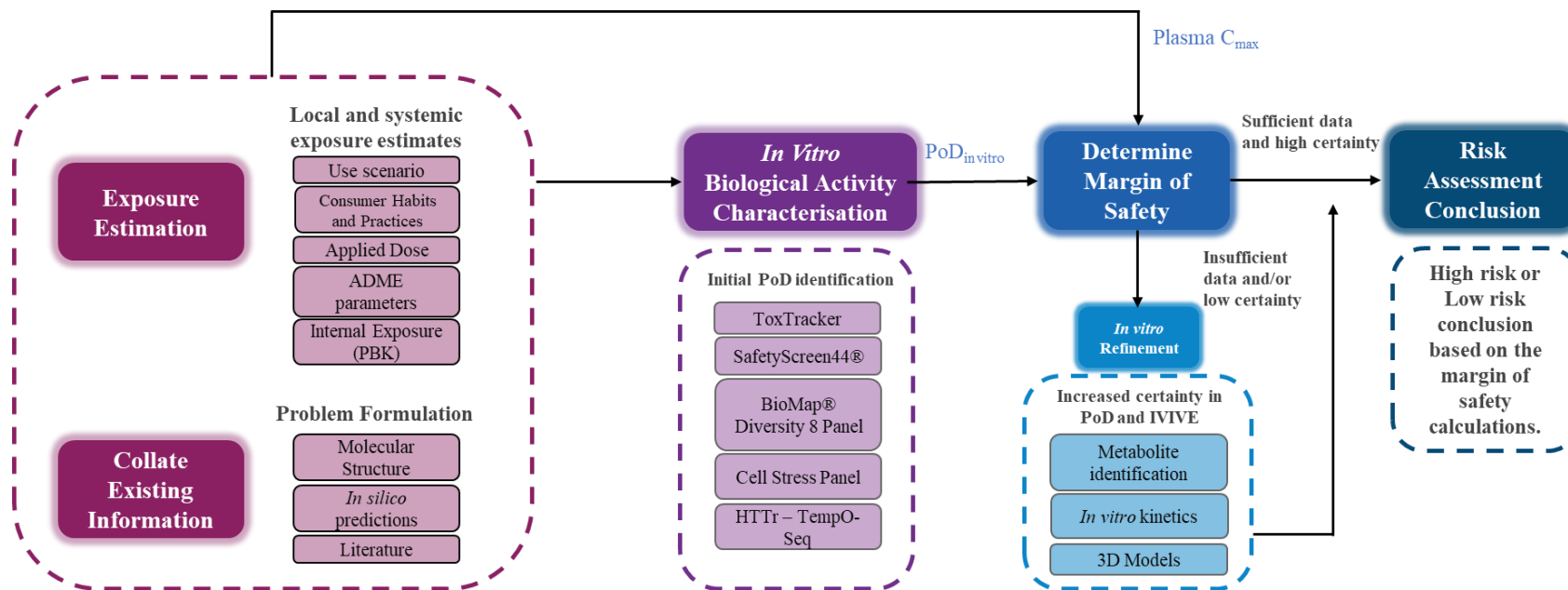
BER refinement – Tier 2



- Refinement of internal exposure estimates:
 - Using clinical data
- Refined understanding of biological effects:
 - Repeat dose dynamics
 - Distinguishing bioactivity and adversity
 - Microphysiological systems

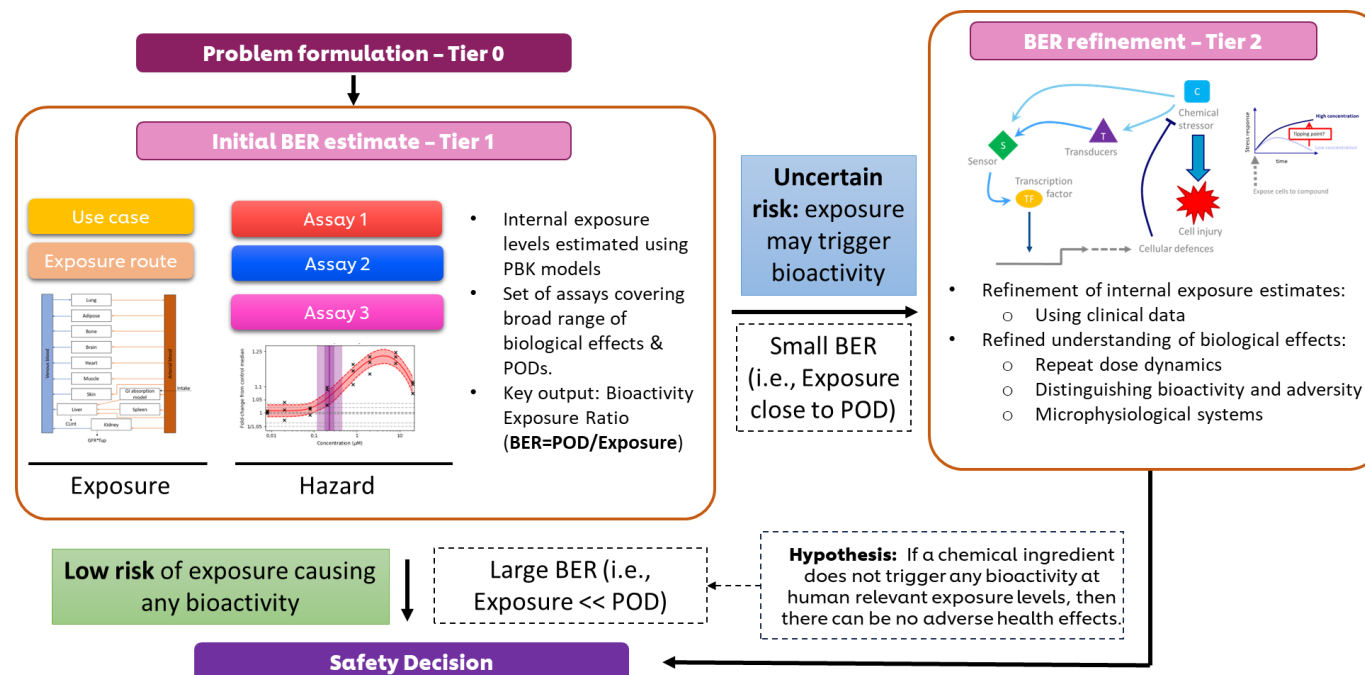
Gaining confidence in NAMs: first case study with coumarin

For coumarin, a safety assessment based on NAMs was at least as protective as the risk assessment based on traditional approaches

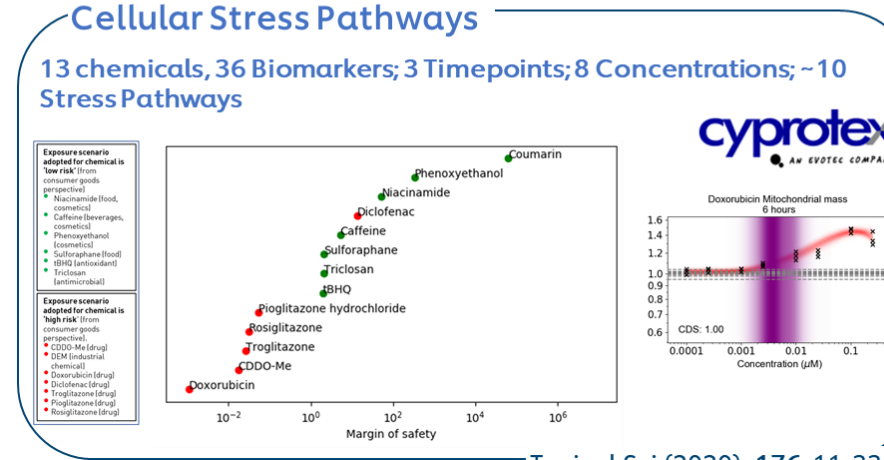
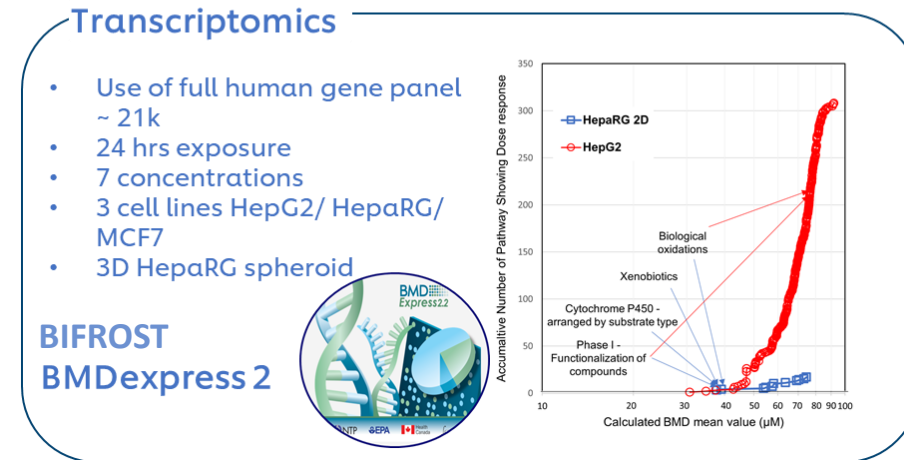
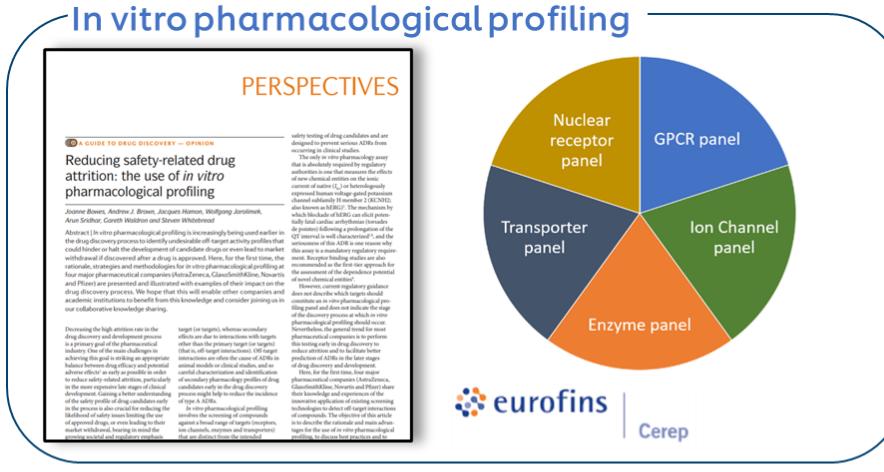
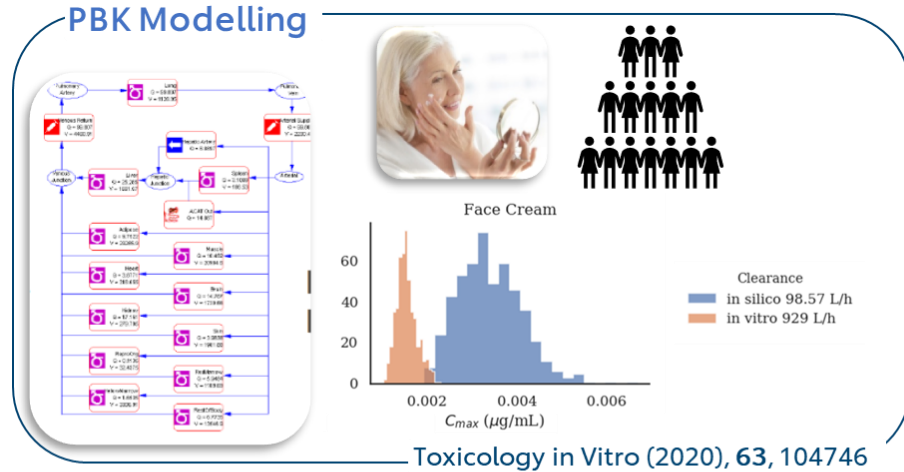


Building and evaluating a systemic safety toolbox

1. Focus the tools and workflows used to make decisions at Tier 1 of an NGRA framework. Includes both:
 - *In vitro* cell assays
 - Exposure models
2. Decisions that can be made with the toolbox are either that a given exposure level is **low risk**, or that the exposure scenario is of **uncertain risk**.
3. In principle, the toolbox will be used as part of a wider tiered assessment framework, which uses e.g., other data through **Tier 0**.



The key NAMs in our NGRA approach



Key aims: 1) select *in vitro* assays that can cover both specific and non-specific mechanisms of toxicity, and 2) can be used to detect early perturbations that may lead to relevant toxicity effects, before the onset of adversity.



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

1. Determine whether the toolbox is fit for purpose
 - Can the toolbox be used to make safety decisions that are protective of human health?
 - Do the various assays and cell types used in the toolbox provide sufficient biological coverage?
 - Are the PBK exposure estimates sufficiently accurate?
2. When evaluating the toolbox, take into account all relevant safety data in assessing the approach:
 - Where available/possible, take into account human safety data.
 - Consider both chronic and acute exposure scenarios
 - Ensure we are protective for a broad range of systemic toxicities.
3. Identify an appropriate safety decision model
 - For example, setting a threshold value on the bioactivity exposure ratio.

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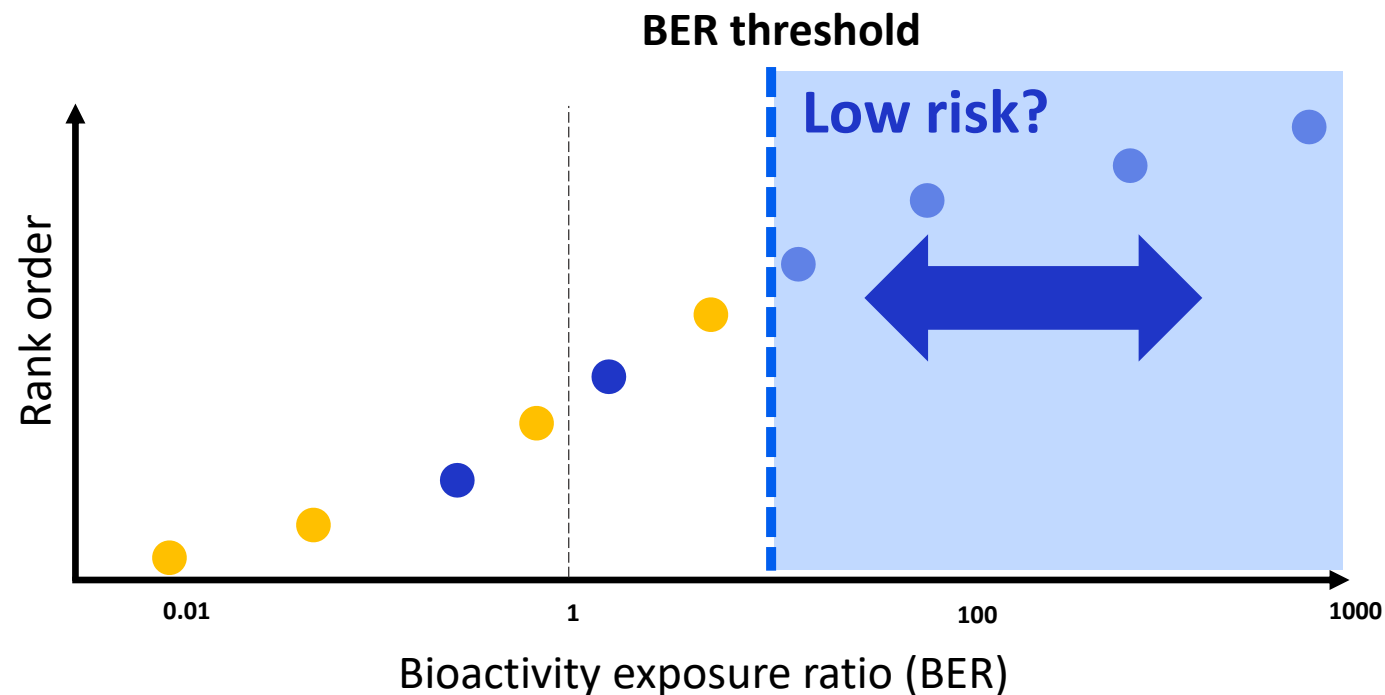
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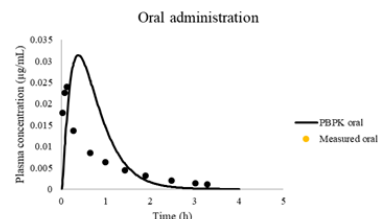
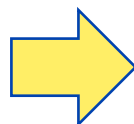
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Chemical exposures scenarios

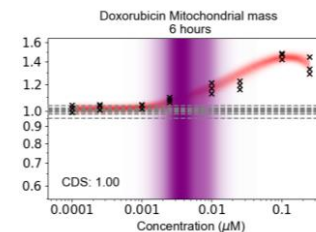
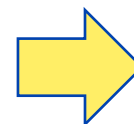
- 'Low' risk (from consumer goods perspective) – e.g. foods, cosmetics
- 'High' risk (from consumer goods perspective) – e.g. drugs



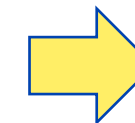
Define typical use-case scenarios benchmark chemical-exposures; Mixture of High and low risk



PBK models of systemic exposure



In-vitro cell assays, estimate PoDs



Calculate the bioactivity exposure ratio

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 (extended 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.

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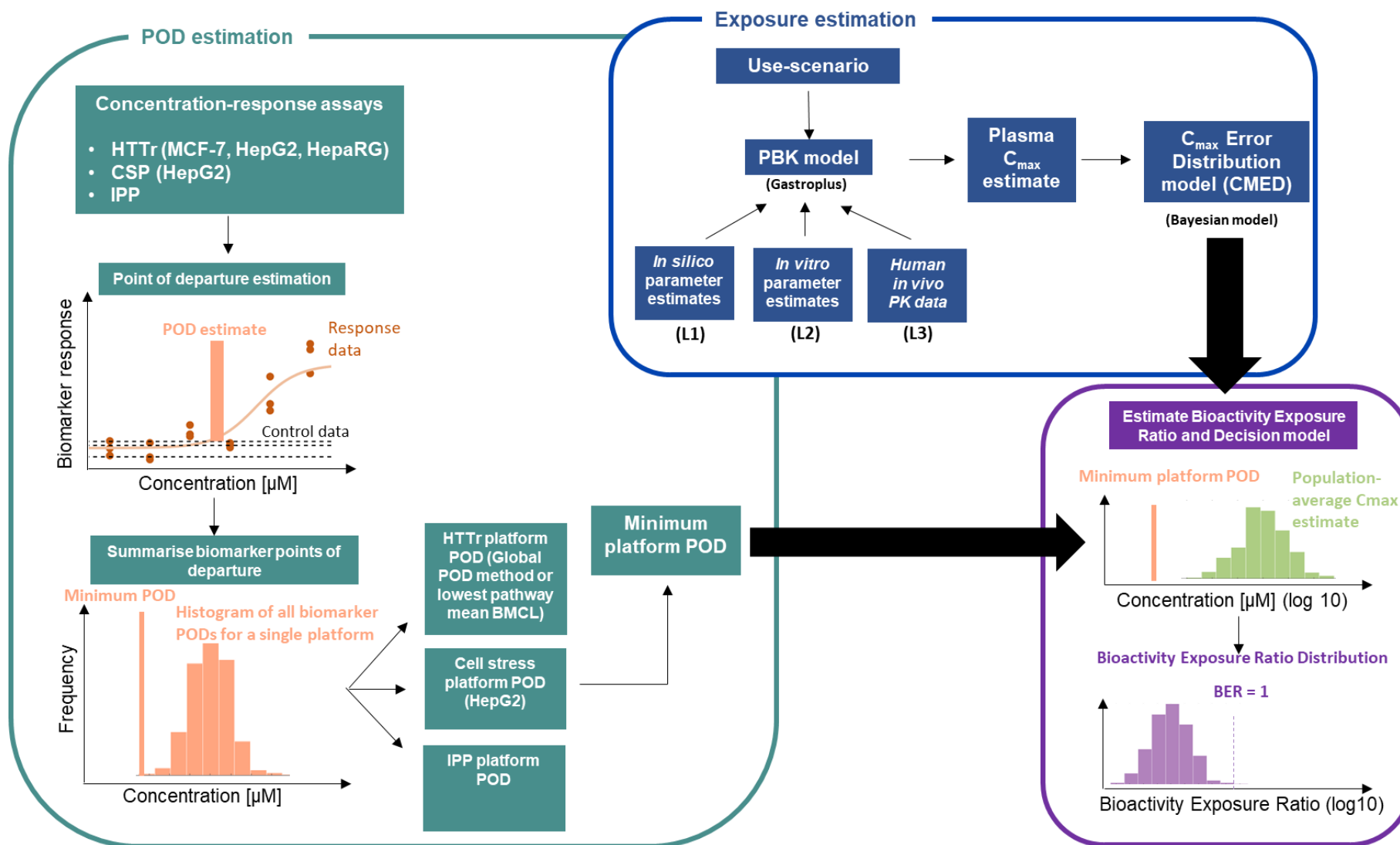
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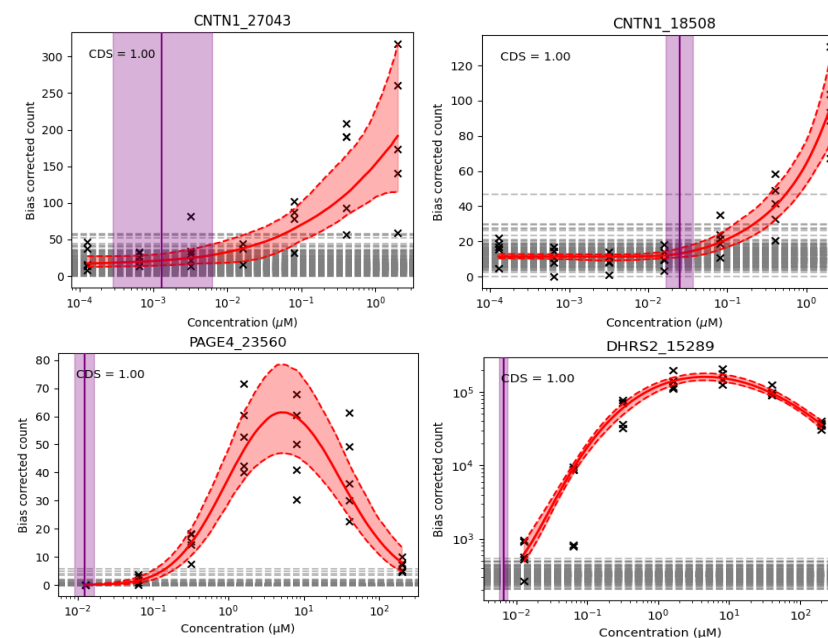
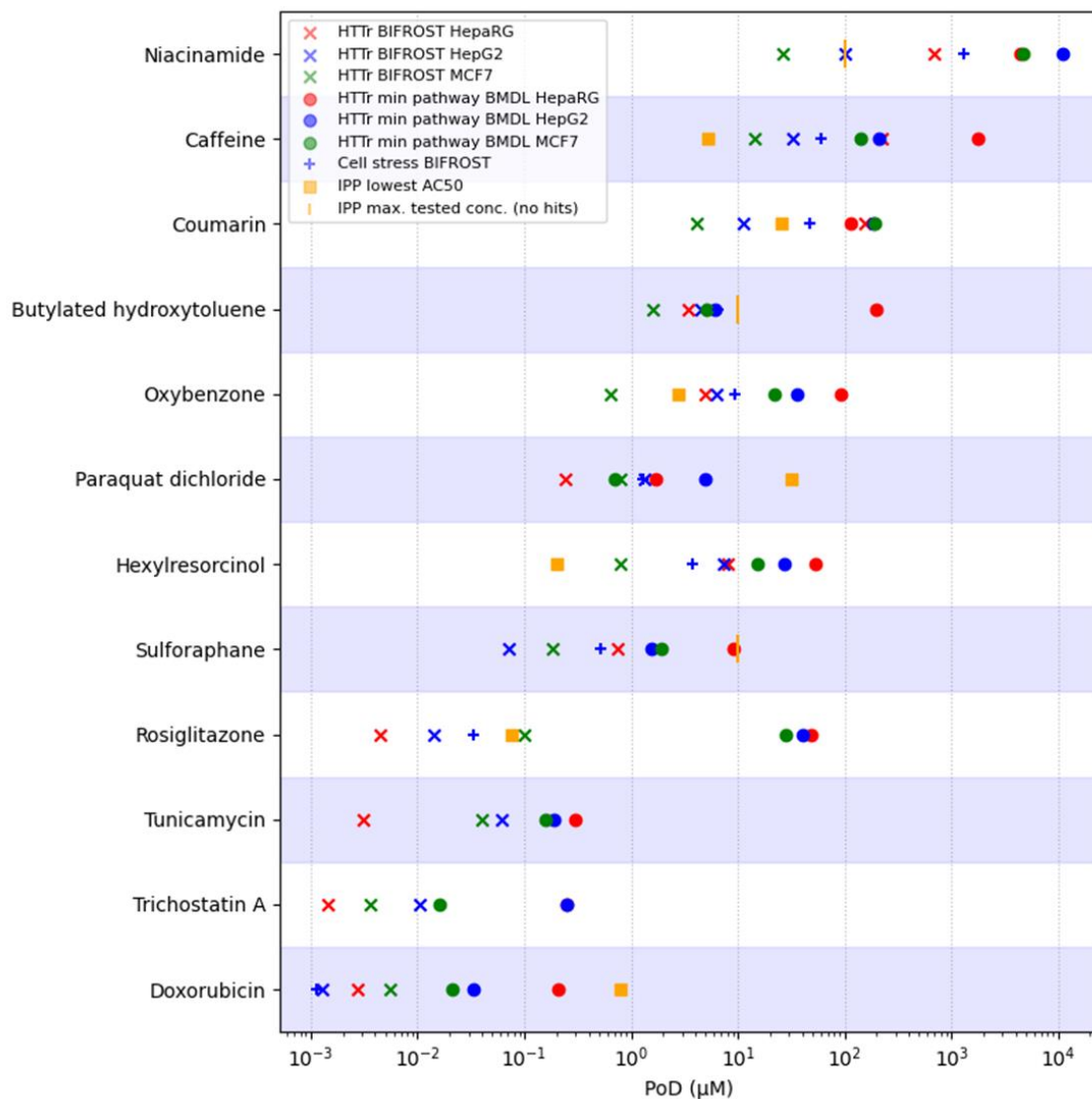
Stage 1: defining the benchmark chemical exposure scenarios

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

The systemic toolbox workflow to estimating a BER



POD estimation

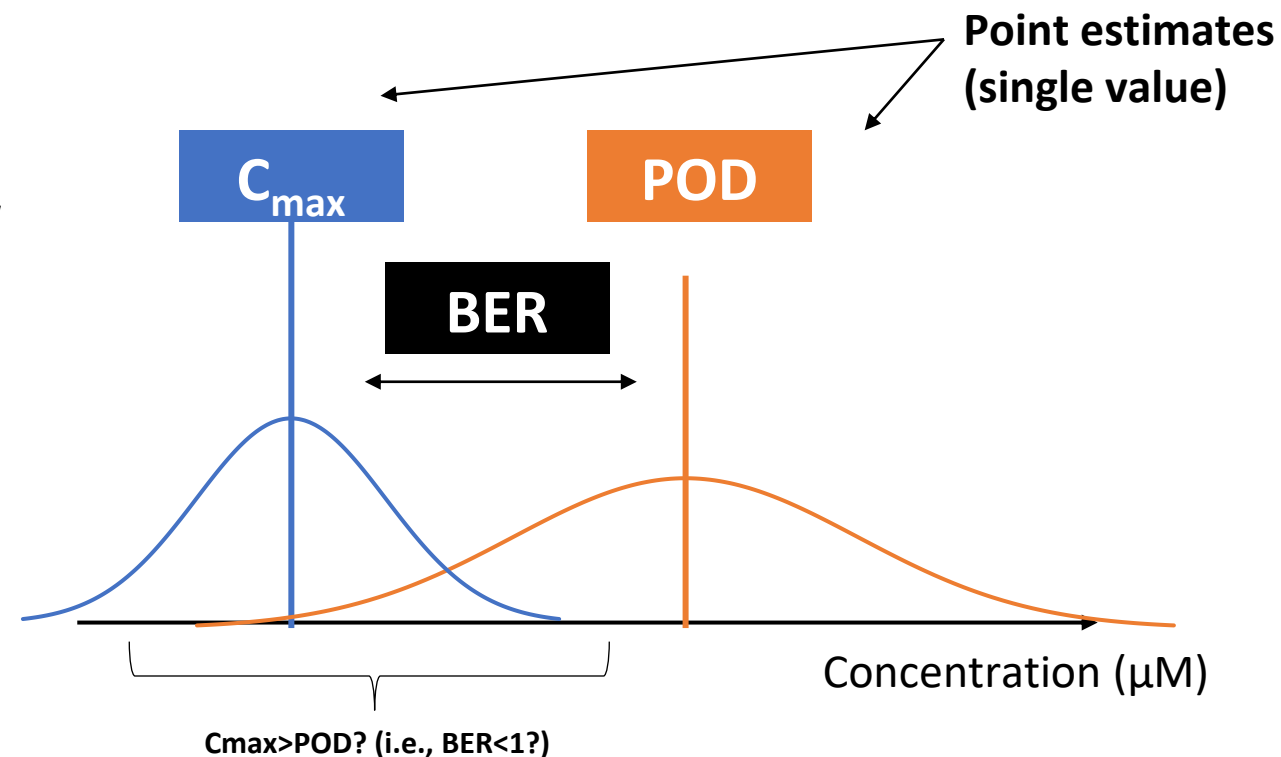


- For 8/10 of compounds tested in the pilot study, HTTr provided the most conservative (lowest POD) when basing the POD on individual genes.
- Pathway based PODs from the HTTr data were typically much higher.

Uncertainty quantification and decision making

Why do we care about quantifying uncertainty?

- Using the point estimates, C_{max} appears to be below the POD.
- The true values of both metrics are subject to uncertainty.
- These uncertainties can be captured in terms of distributions.
- The distributions show the range of plausible values for the C_{max} and POD.
- Quantifying uncertainty in quantities like C_{max} and the POD can be helpful to determine when a safety decision can be made with confidence, or when more refinement is needed.



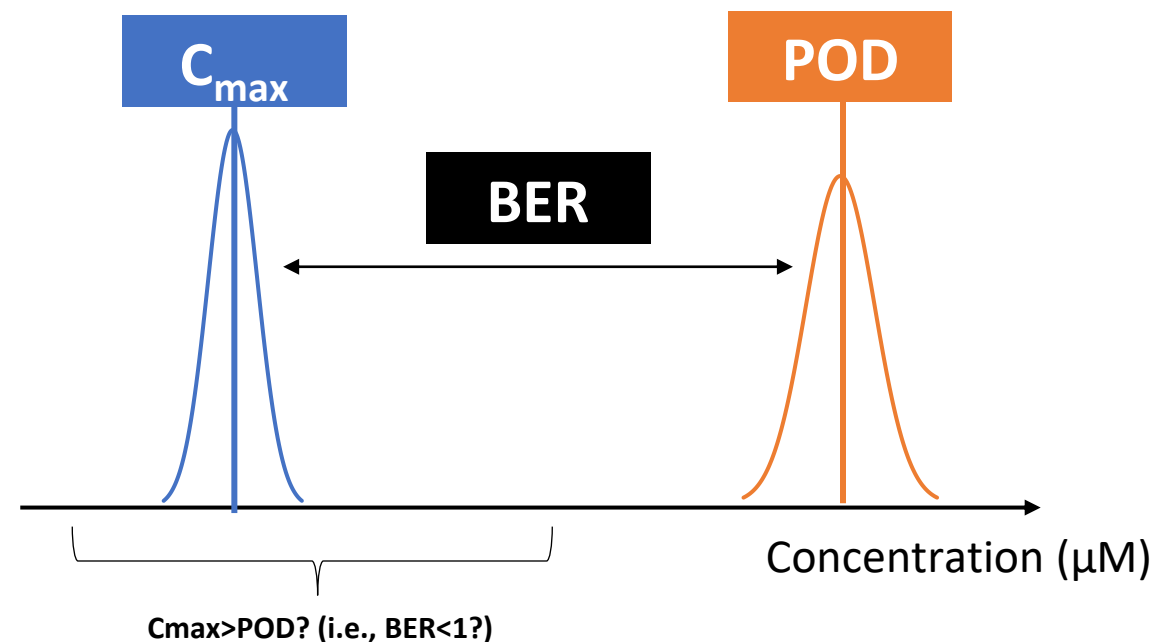
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$$\text{Prob}(\text{BER} > 1) = ?$$

Uncertainty quantification and decision making

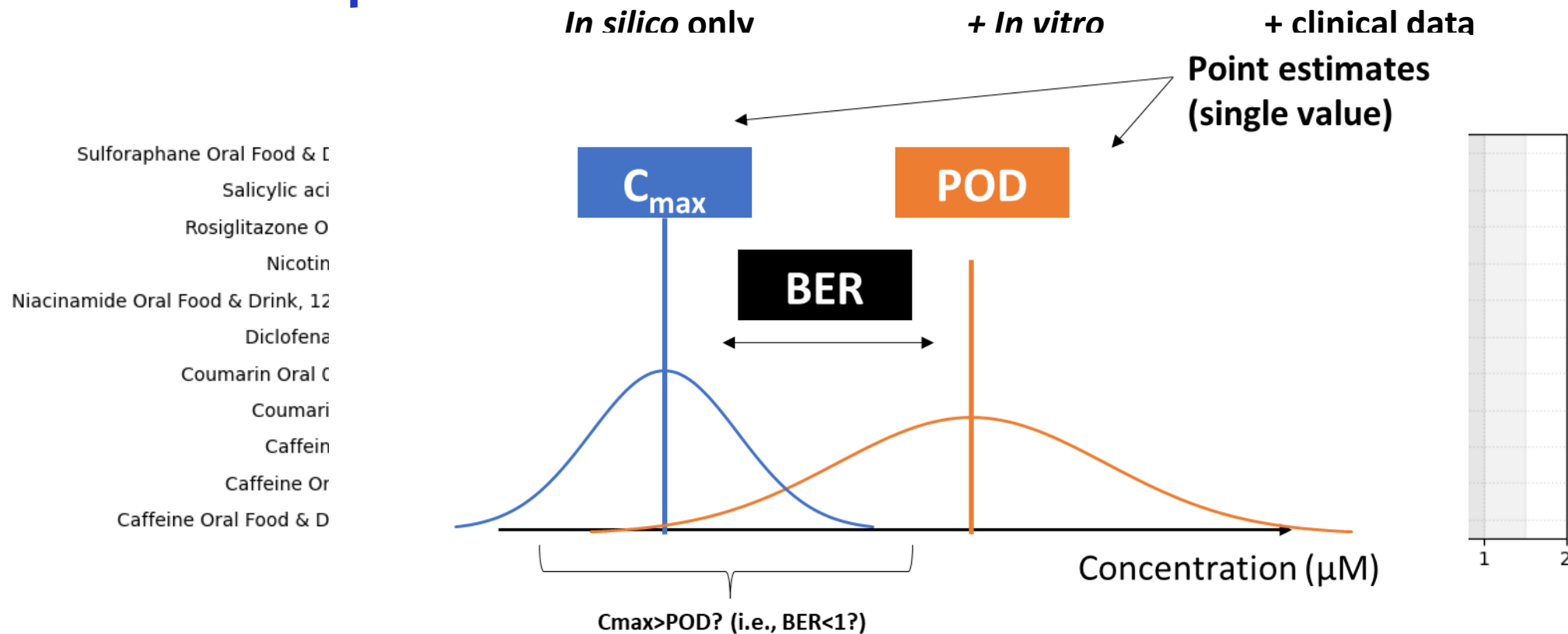
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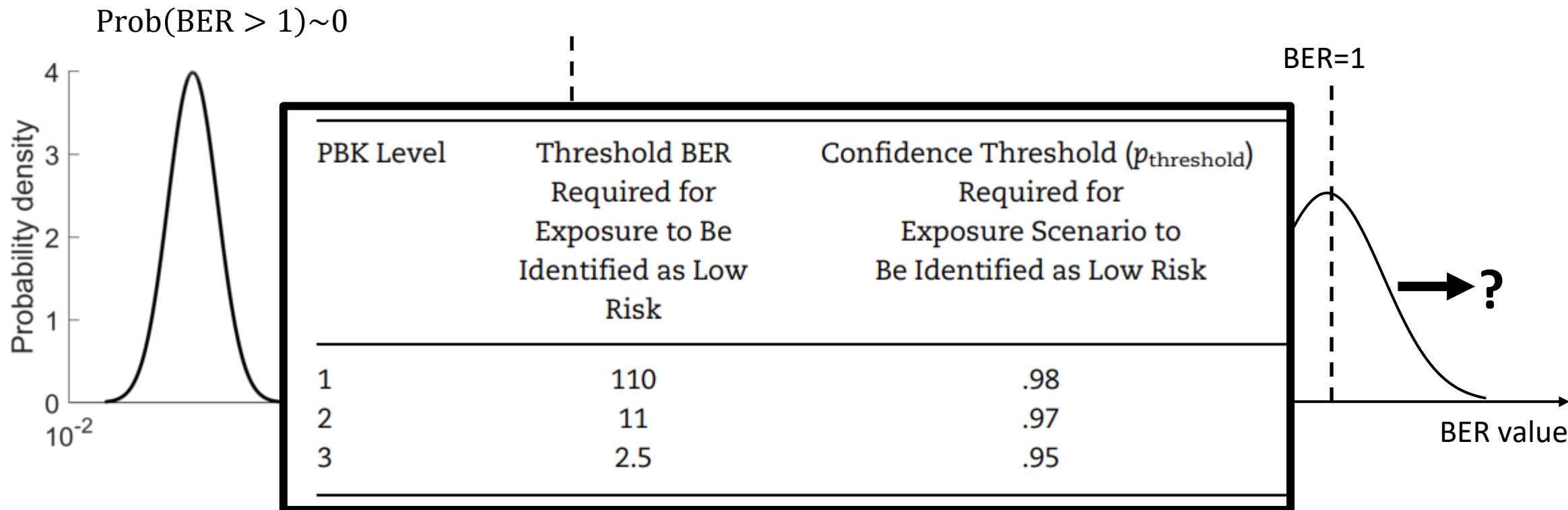
$$\text{Prob}(\text{BER} > 1) = ?$$

Quantifying PBK model accuracy and uncertainty for different chemical exposure scenarios



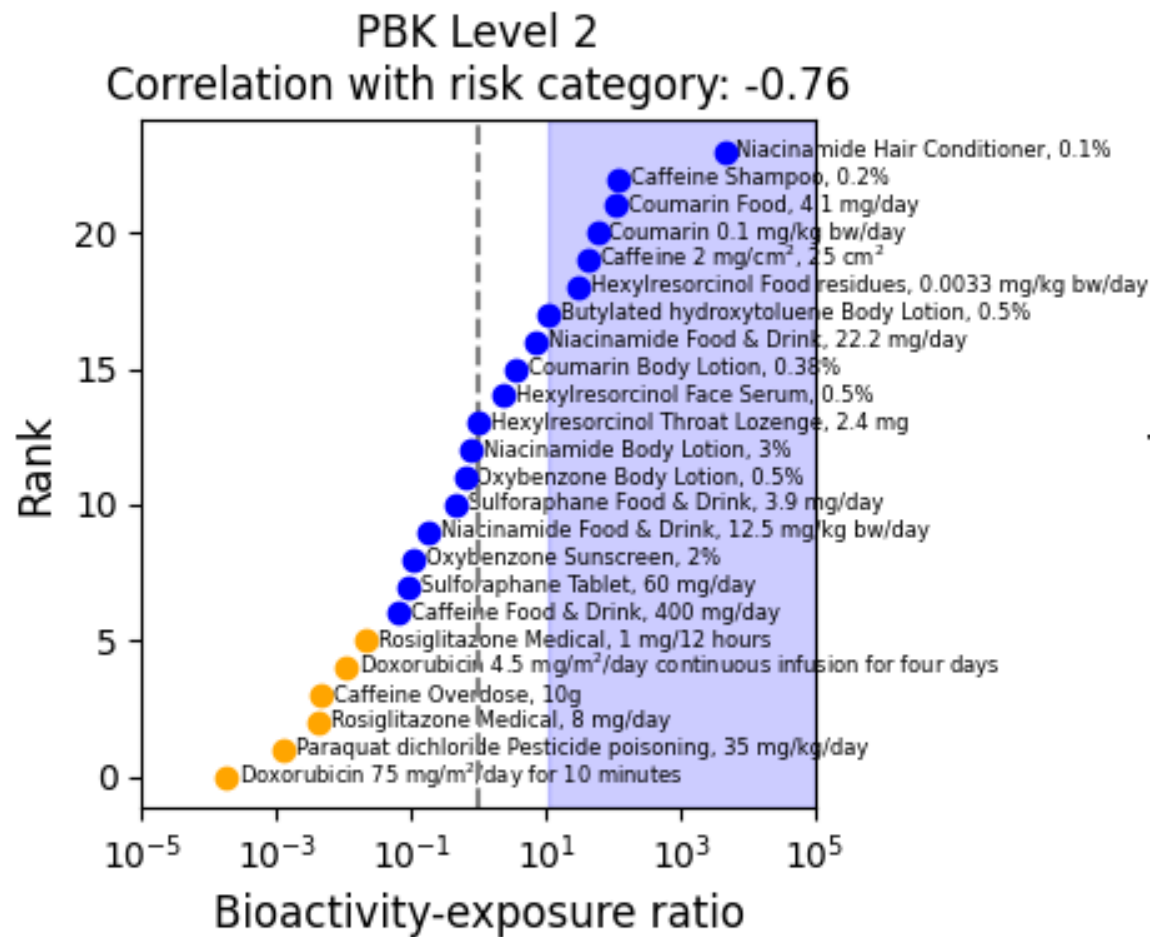
- The accuracy of PBK model C_{max} estimates can be quantified by comparing the predicted C_{max} value to measured values for different clinical datasets.
- The C_{max} Error Distribution (CMED) model was developed using these data to quantify the uncertainty in a PBK C_{max} prediction novel substance or exposure scenario, depending on how the PBK model had been parameterised.

Estimating the Bioactivity Exposure Ratio distribution



- The distribution representing uncertainty C_{max} estimate can be combined with the minimum PODs to form a single BER distribution. (Currently this distribution does not take into account POD uncertainty).
- The minimum POD was selected in order to ensure safety decisions are sufficiently conservative.

Systemic safety toolbox pilot study results: 100% protective for all PBK levels



Blue: low risk chemical-exposure scenario

Yellow: high risk chemical-exposure scenario

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

Across the various PBK parameterisation levels:

- **100% protective** (i.e., 100% of all high-risk exposure scenarios were correctly identified as not low risk)
- **Up to 69% utility** (i.e., 69% of all low-risk exposures were correctly identified as low risk).

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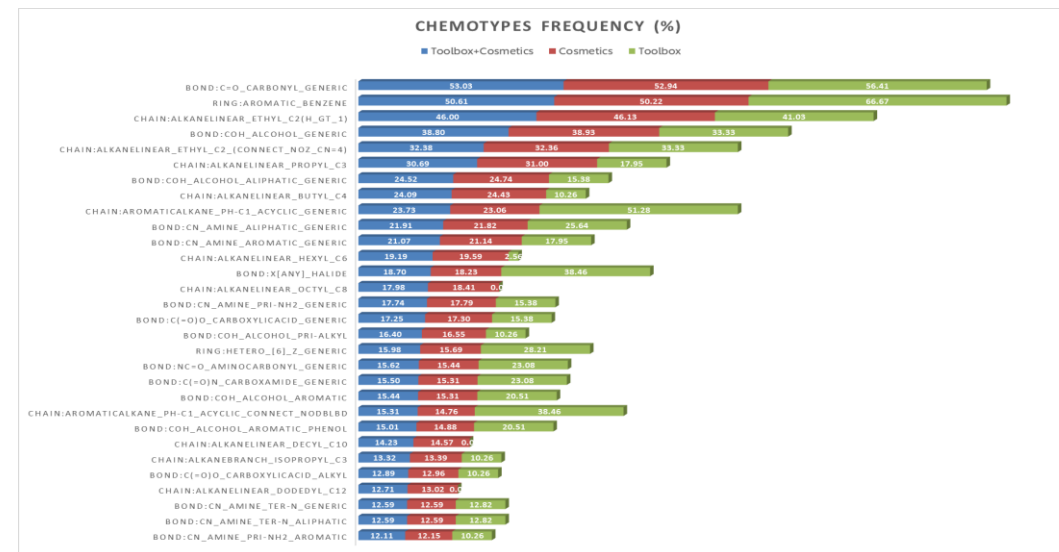
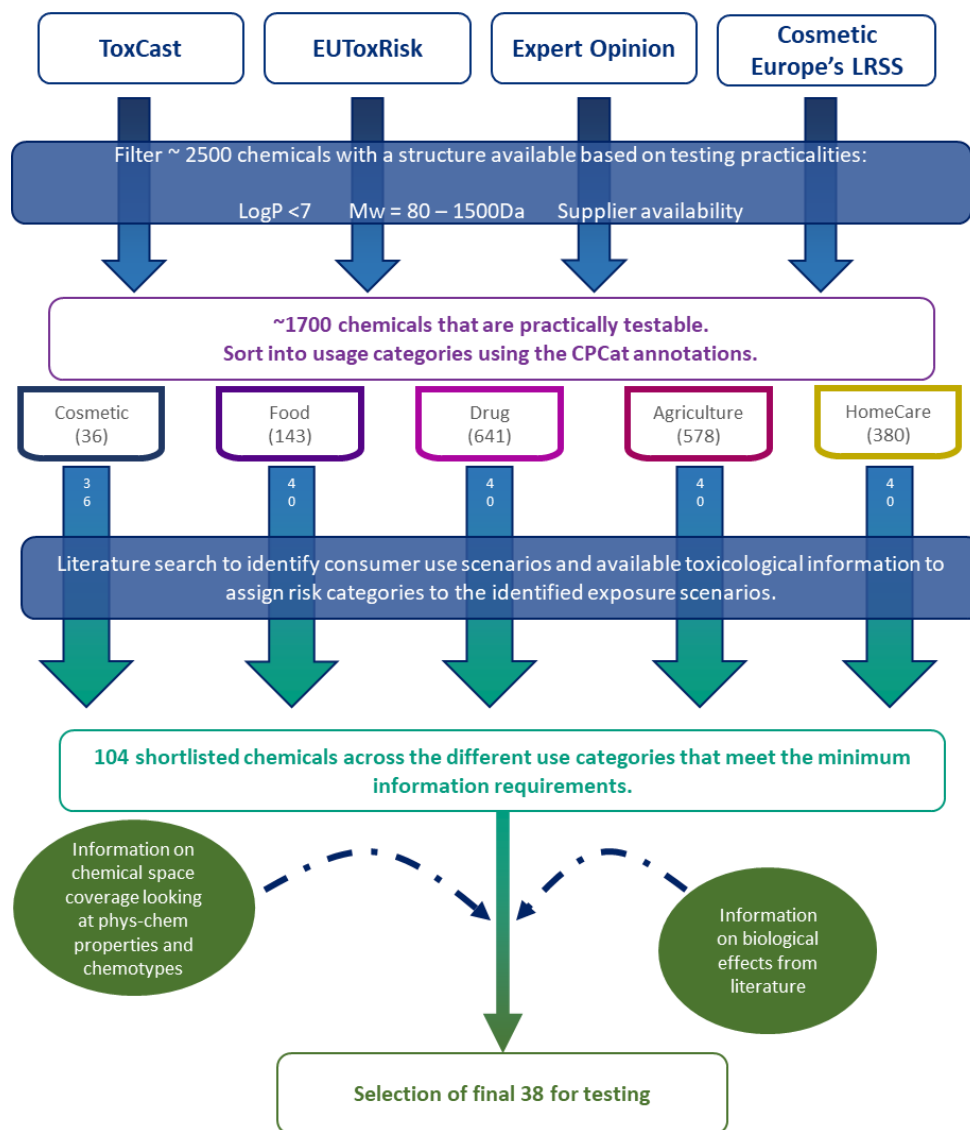
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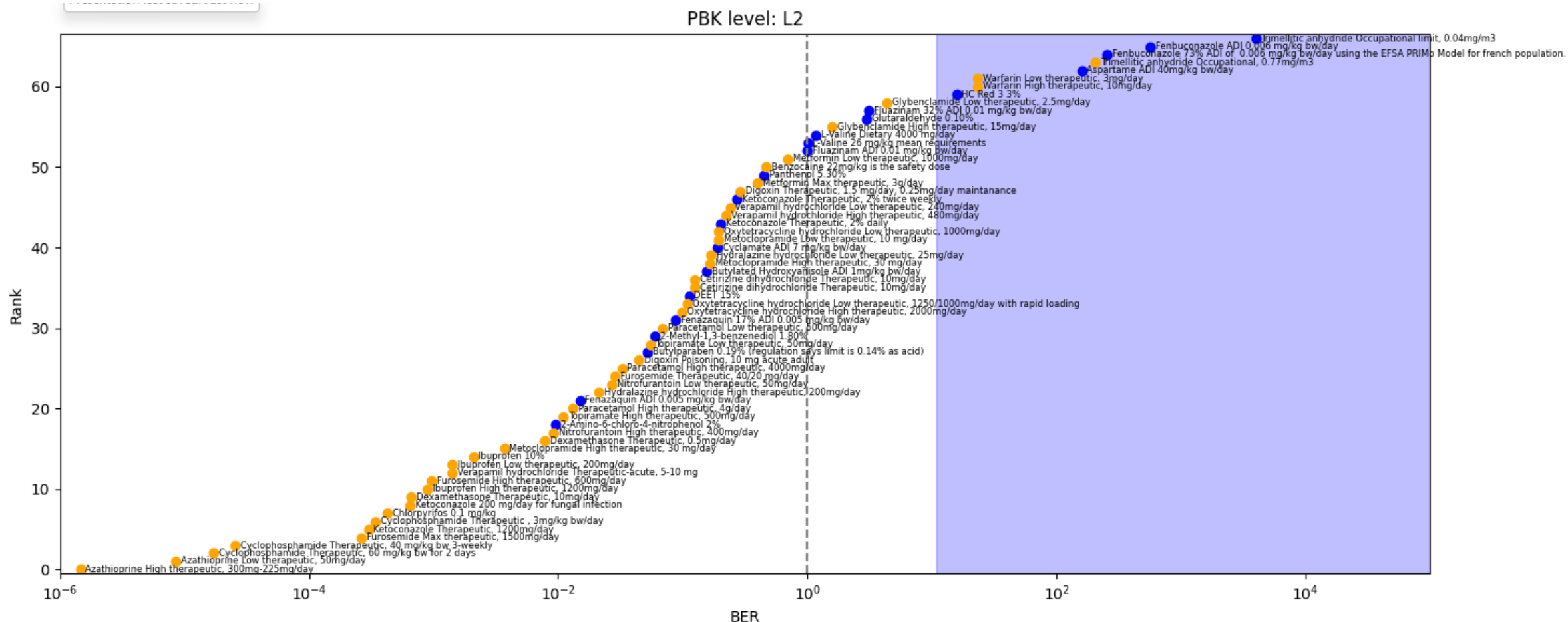
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Expanding the set of benchmark chemical exposure scenarios



- Manual chemical or exposure scenario selection may result in strong biases.
- Therefore, for the extended evaluation, benchmarks were selected using a semi-randomised process.
- The final set of benchmarks represented a wide range of different potencies, chemotypes and potential toxicity mechanisms.

Toolbox performance: PBK L2 exposure estimates



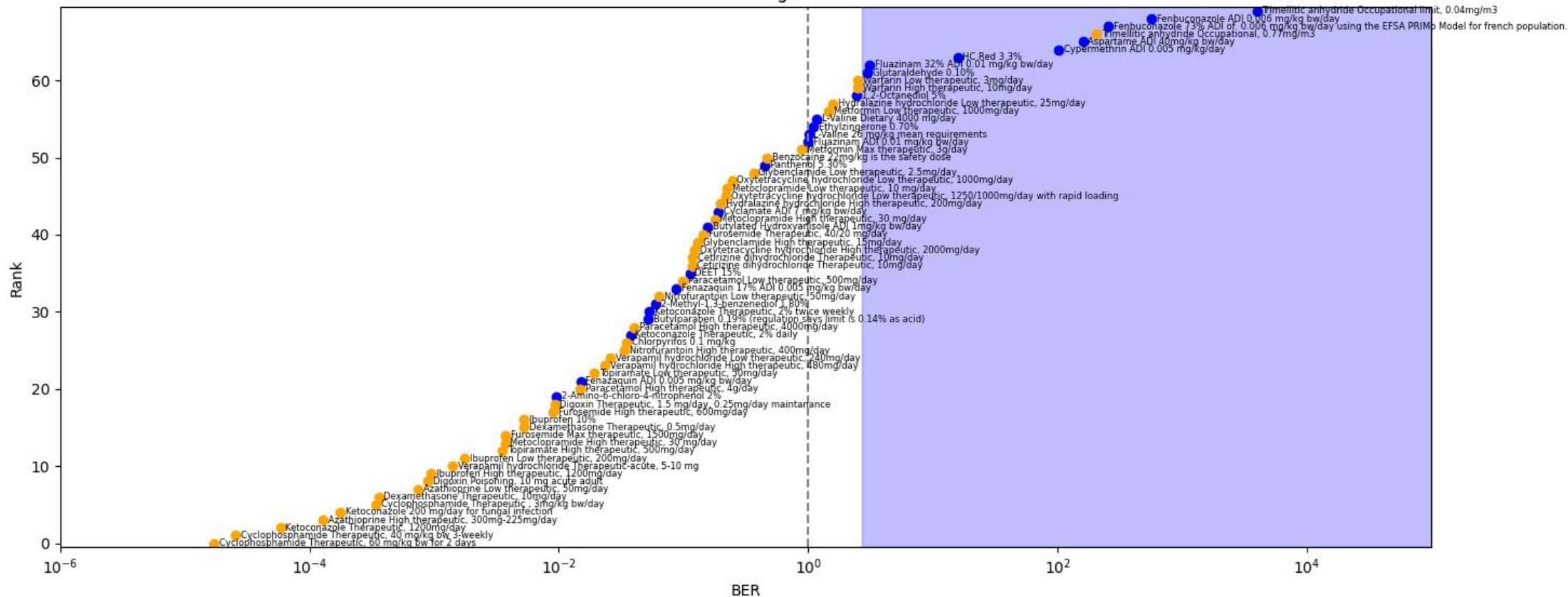
Protectiveness: 93% (43 out of 46)

Utility: 24% (5 out of 21)

Balanced accuracy: 59%

Toolbox performance: Highest available PBK level

PBK level: highest



Protectiveness: 98% (45 out of 46)

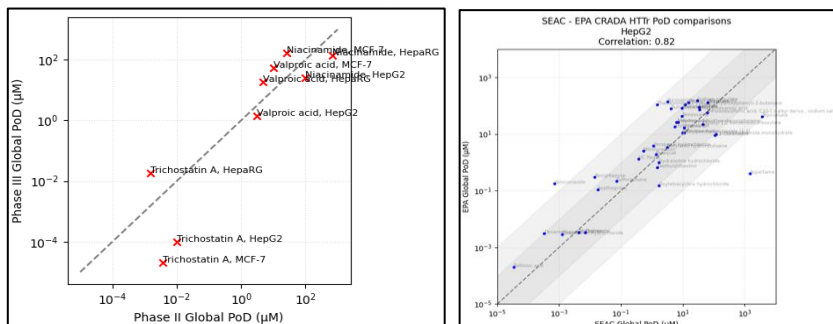
Utility: 33% (8 out of 24)

Balanced accuracy: 66%

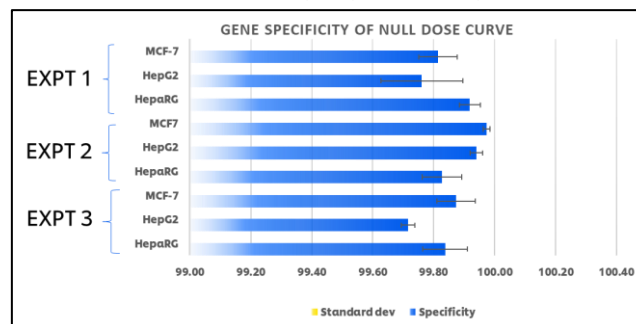
Discussion and next steps

- We have now extended the evaluation to 38 chemicals and 70 exposure scenarios. Protective for 93-98% of scenarios (depending on PBK level).
- Unilever-EPA CRADA:** Generating data for 10 cell lines, using high-throughput transcriptomics and phenotypic profiling.
- We are continuing to further establishing scientific confidence through a range of activities.

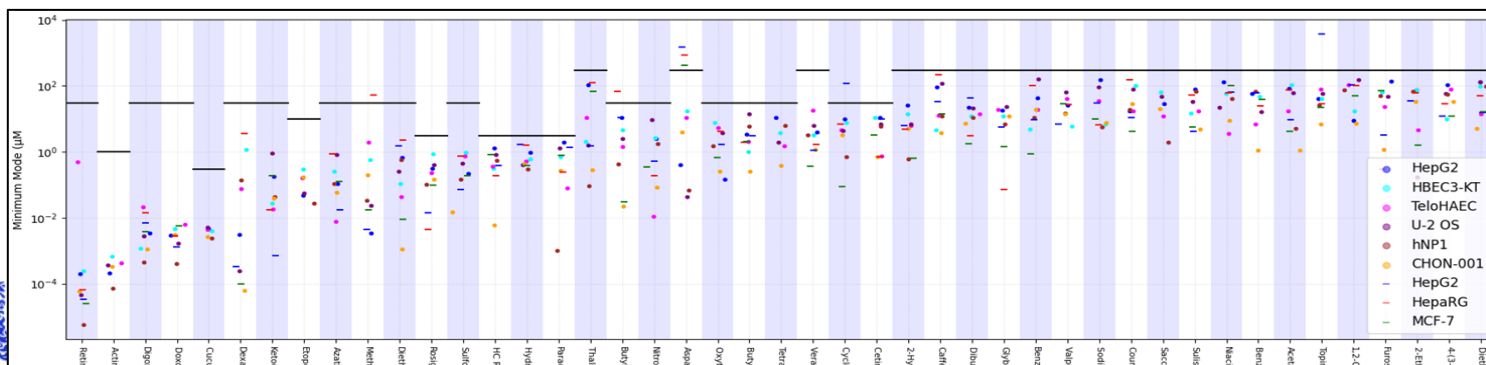
POD reproducibility



Evaluating different POD methods (sensitivity, specificity etc)



Cell line selection and POD diversity

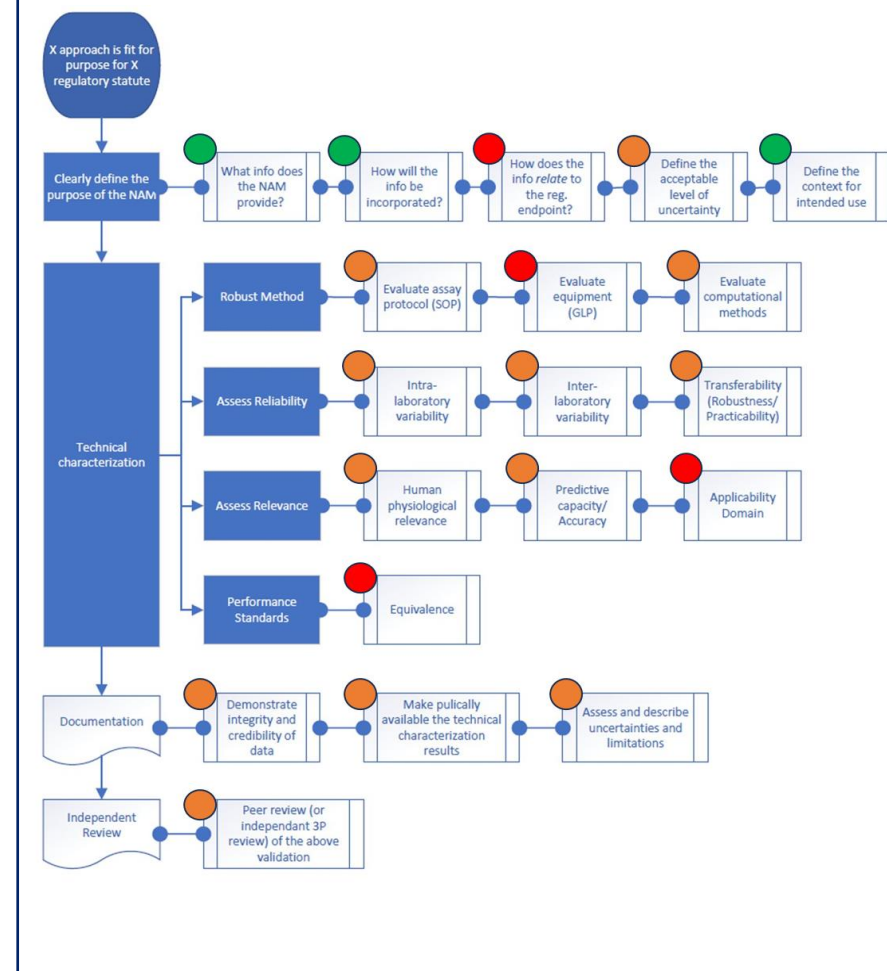


A framework for establishing scientific confidence in new approach methodologies

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Thank You



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