

# Next Generation risk assessment: From concept to application

**Maria Baltazar, Unilever Safety, Environmental & Regulatory Science, UK**

**SERS**  
Safety, Environmental  
& Regulatory Science

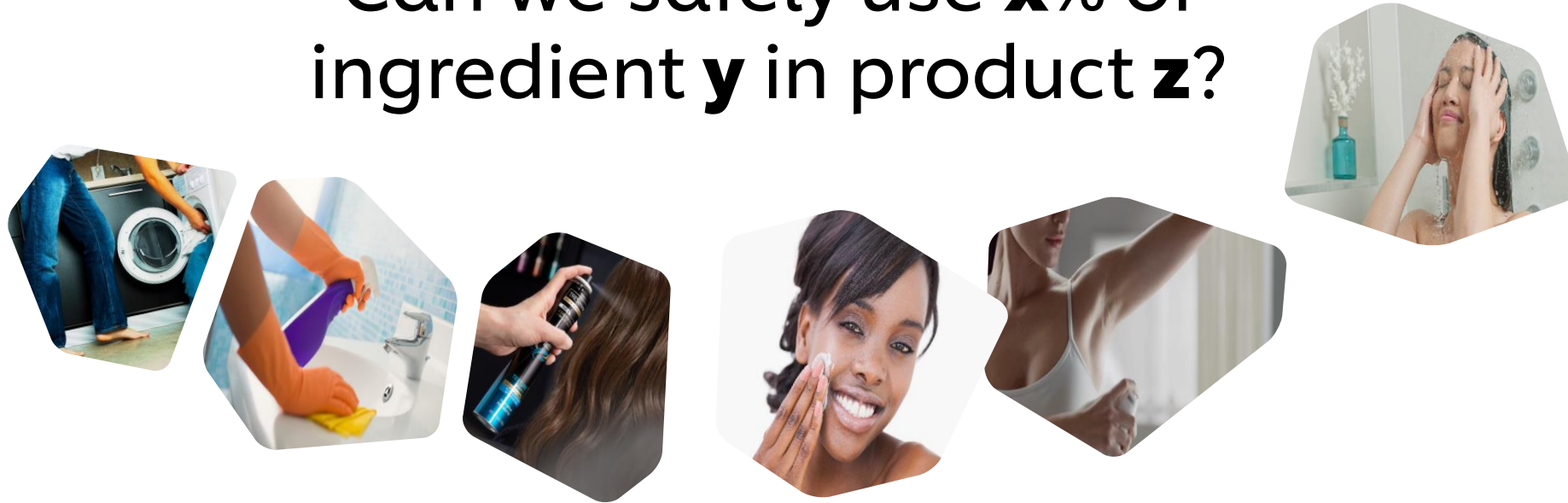


# Outline

- Introduction to Next generation risk assessment (NGRA)
- Ongoing efforts to develop systemic toxicity NGRA approaches
- Unilever approach to developing an early tier NAM-systemic toolbox and workflow
- Application of NGRA principles to a case study with climbazole in a face cream product.

# The objective of a consumer product risk assessment is...

Can we safely use **x%** of ingredient **y** in product **z**?



All safety assessments of cosmetic ingredients are exposure-driven:



## Introduction to 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<sup>1</sup>*

**New approach methodologies (NAMs)<sup>2</sup>** can be defined as any *in vitro*, *in chemico* or computational (*in silico*) method that when used alone, or in concert with others, enables improved chemical safety assessment through more protective and/or relevant models and as a result, contributes to the replacement of animals.

## Principles of NGRA from ICCR

### 4

#### Main overriding principles:

- The overall goal is a human safety risk assessment
- The assessment is exposure led
- The assessment is hypothesis driven
- The assessment is designed to prevent harm

### 3

#### Principles describe how a NGRA should be conducted:

- Following an appropriate appraisal of existing information
- Using a tiered and iterative approach
- Using robust and relevant methods and strategies

### 2

#### Principles for documenting NGRA:

- Sources of uncertainty should be characterized and documented
- The logic of the approach should be transparent and documented

# NGRA: The overall goal is a human safety risk assessment



**Tox21/ToxCast**  
~700 HTS Biological  
Pathways Assays



“Advances in toxicogenomics, bioinformatics, systems biology, and computational toxicology could transform toxicity testing from a system based on whole-animal testing to one founded primarily on *in vitro* methods that evaluate changes in biologic processes using cells, cell lines, or cellular components, preferably of human origin.” 2007

**National Institute of Environmental Health Sciences (NIEHS) / National Toxicology Program (NTP)**

**National Center for Advancing Translational Sciences (NCATS)**

**U.S. Food and Drug Administration (FDA)**

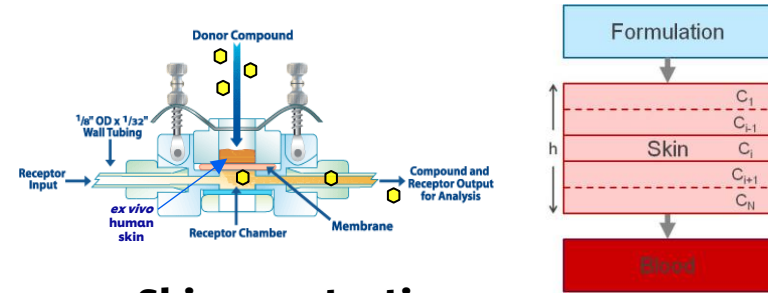
**National Center for Computational Toxicology (EPA)**

# NGRA: The assessment is exposure-led

- Route of exposure
- Consumer use (Habits & Practices)
- Applied dose (external concentration)

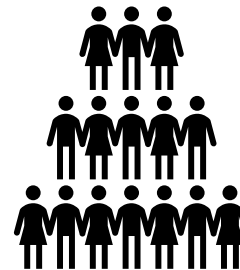


## ADME parameters

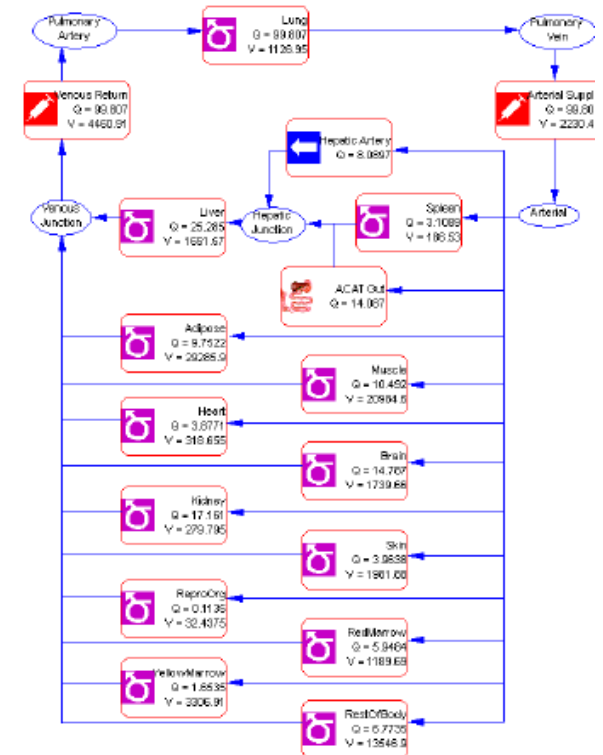


- Skin penetration
- Phys-chem properties
- Hepatic clearance
- Fraction unbound
- blood:plasma ratio

## Uncertainty analysis- Population simulation



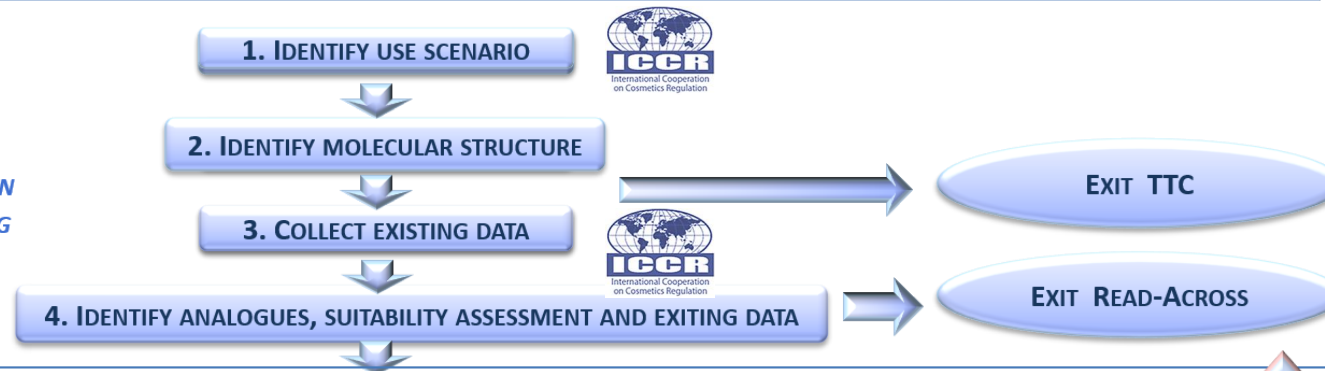
## Physiologically-based kinetic (PBK) modelling - Internal concentration (plasma, urine, organ-level)



# NGRA: The assessment is hypothesis driven & should be conducted Using a tiered and iterative approach



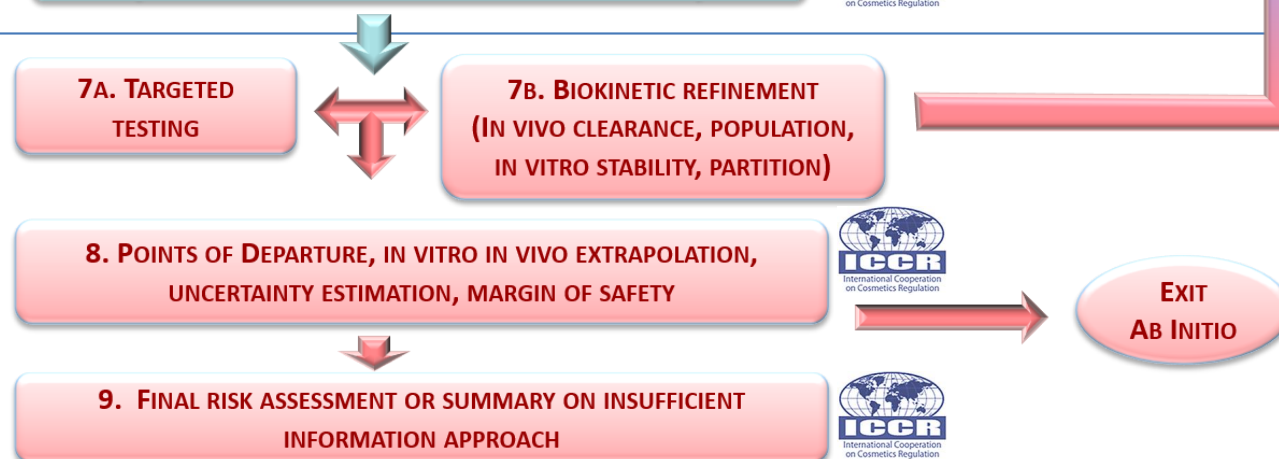
**TIER 0: IDENTIFY**  
USE SCENARIO,  
CHEMICAL OF CONCERN  
AND COLLECT EXISTING  
INFORMATION



**TIER 1: HYPOTHESIS**  
FORMULATION FOR AB  
INITIO APPROACH



**TIER 2:**  
APPLICATION OF AB  
INITIO APPROACH

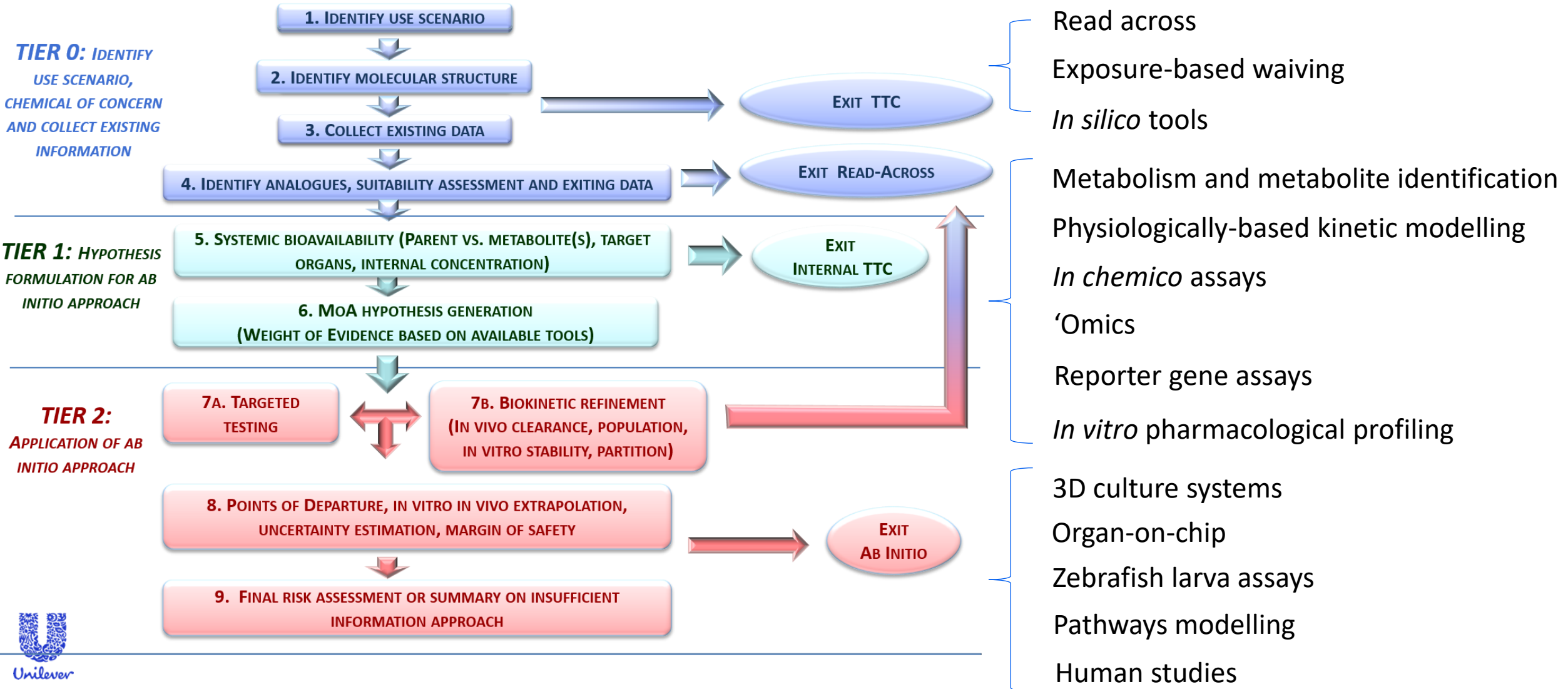


Continue through tiers until enough information to make a decision: assessment may be complete at any tier

Berggren et al., (2017)  
*Computational Toxicology* 4: 31-44.  
<https://doi.org/10.1016/j.comtox.2017.10.001>



# NGRA: Using robust and relevant methods and strategies



# NGRA: Using robust and relevant methods and strategies

Readiness judged by ICCR in 2018:  
 (ICCR IS JWG Part 2 FINAL ([iccr-cosmetics.org](http://iccr-cosmetics.org)))

**TIER 0: IDENTIFY**  
 USE SCENARIO,  
 CHEMICAL OF CONCERN  
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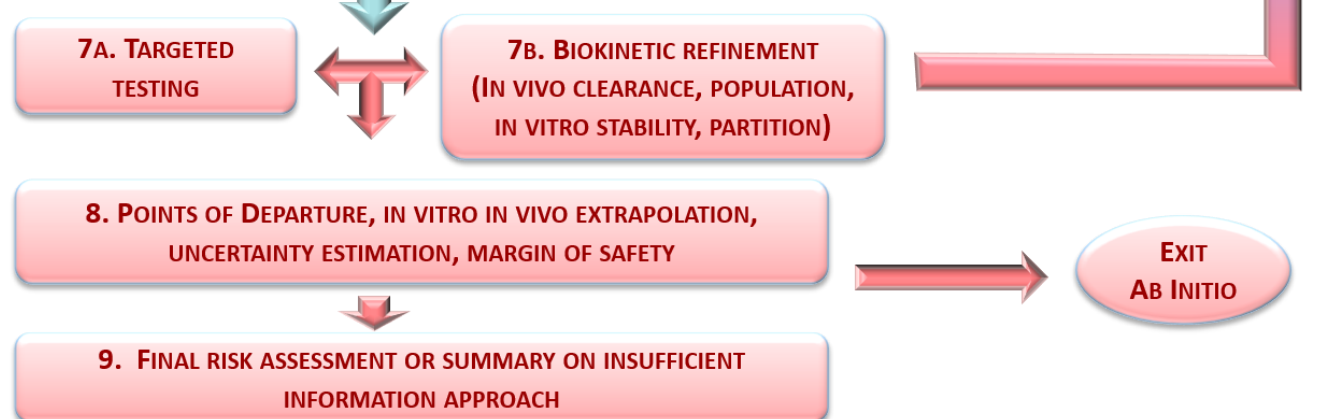
- Read across
- Exposure-based waiving
- In silico* tools

**TIER 1: HYPOTHESIS**  
 FORMULATION FOR AB  
 INITIO APPROACH



- Metabolism and metabolite identification
- Physiologically-based kinetic modelling
- In chemico* assays

**TIER 2:**  
 APPLICATION OF AB  
 INITIO APPROACH



- 'Omics
- Reporter gene assays
- In vitro* pharmacological profiling

- 3D culture systems
- Organ-on-chip
- Zebrafish larva assays
- Pathway modelling
- Human studies

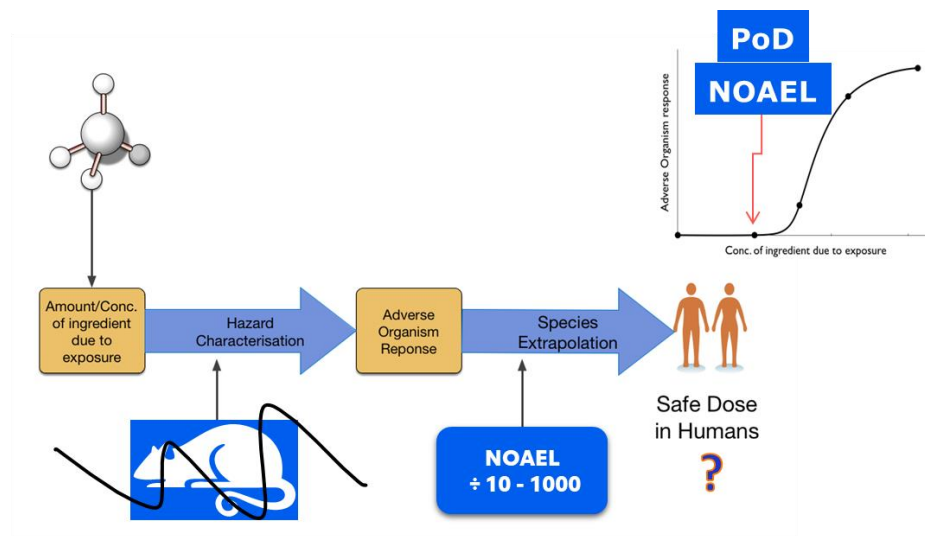
Note - not universally considered a non-animal approach



# NGRA: The assessment is designed to prevent harm

## Focus on protection

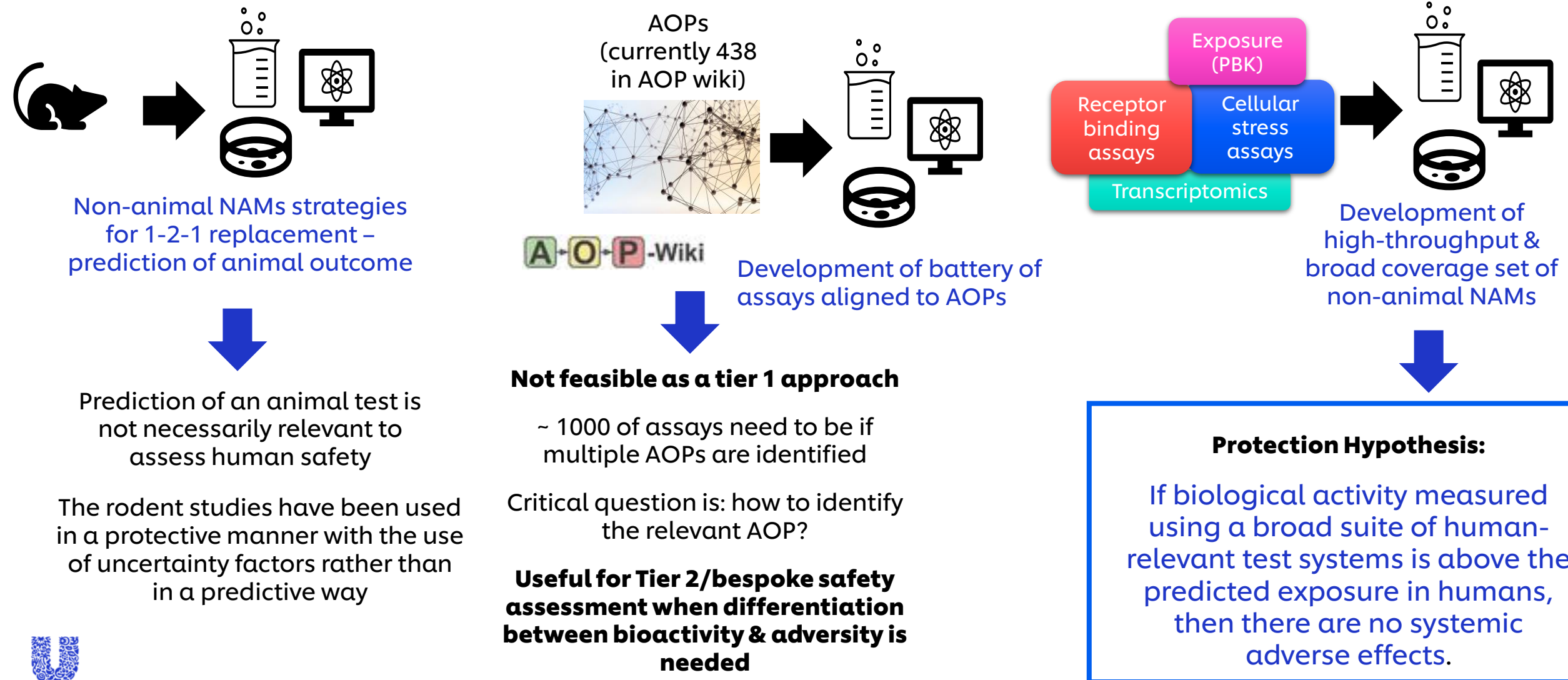
- Non-specific endpoints from in vivo toxicological studies data are often used to derive points of departure (POD) (e.g. no-observed-effect-level or no-observed adverse effect level (NO(A)EL))
- Uncertainty or safety assessment factors are applied to POD to calculate recommended exposure levels that are broadly protective but not necessarily target-specific.



**Are non-animal safety assessments even possible for systemic toxicity?**

Many possible adversities...ADME considerations...Homeostasis

# Yes... but it requires a different way of thinking about the problem



Non-animal NAMs strategies for 1-2-1 replacement – prediction of animal outcome



Prediction of an animal test is not necessarily relevant to assess human safety

The rodent studies have been used in a protective manner with the use of uncertainty factors rather than in a predictive way

AOPs (currently 438 in AOP wiki)



A-O-P-Wiki

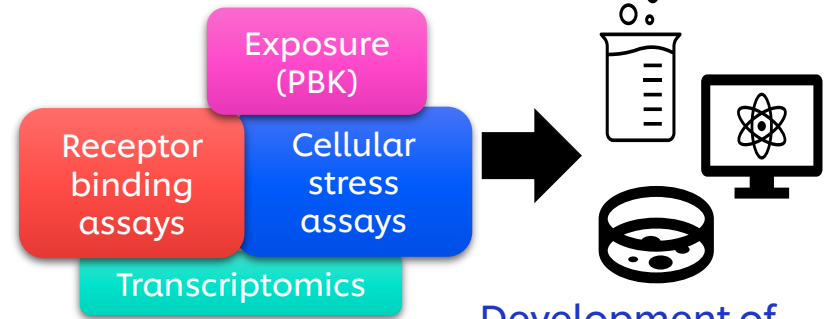


**Not feasible as a tier 1 approach**

~ 1000 of assays need to be if multiple AOPs are identified

Critical question is: how to identify the relevant AOP?

**Useful for Tier 2/bespoke safety assessment when differentiation between bioactivity & adversity is needed**



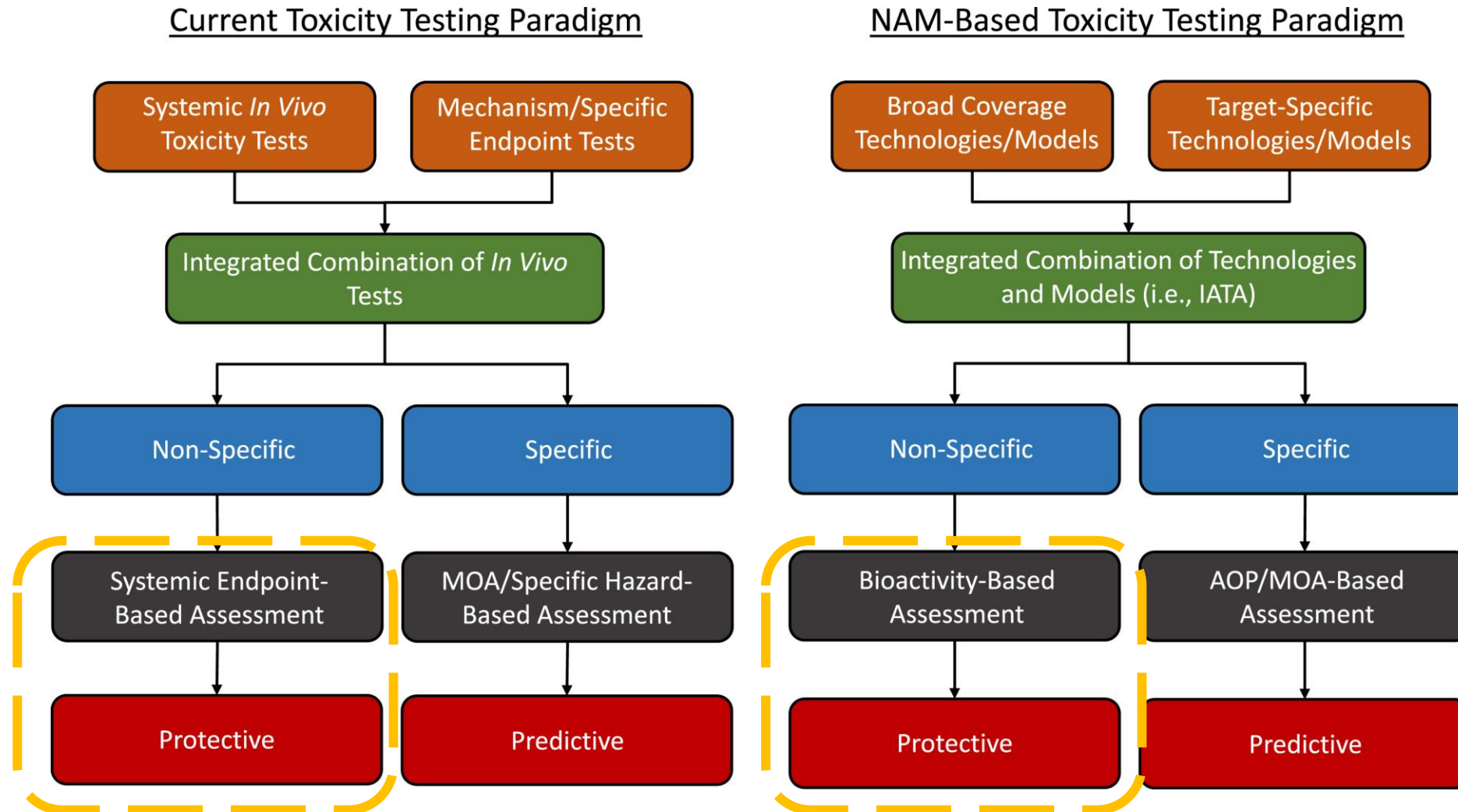
Development of high-throughput & broad coverage set of non-animal NAMs



**Protection Hypothesis:**

If biological activity measured using a broad suite of human-relevant test systems is above the predicted exposure in humans, then there are no systemic adverse effects.

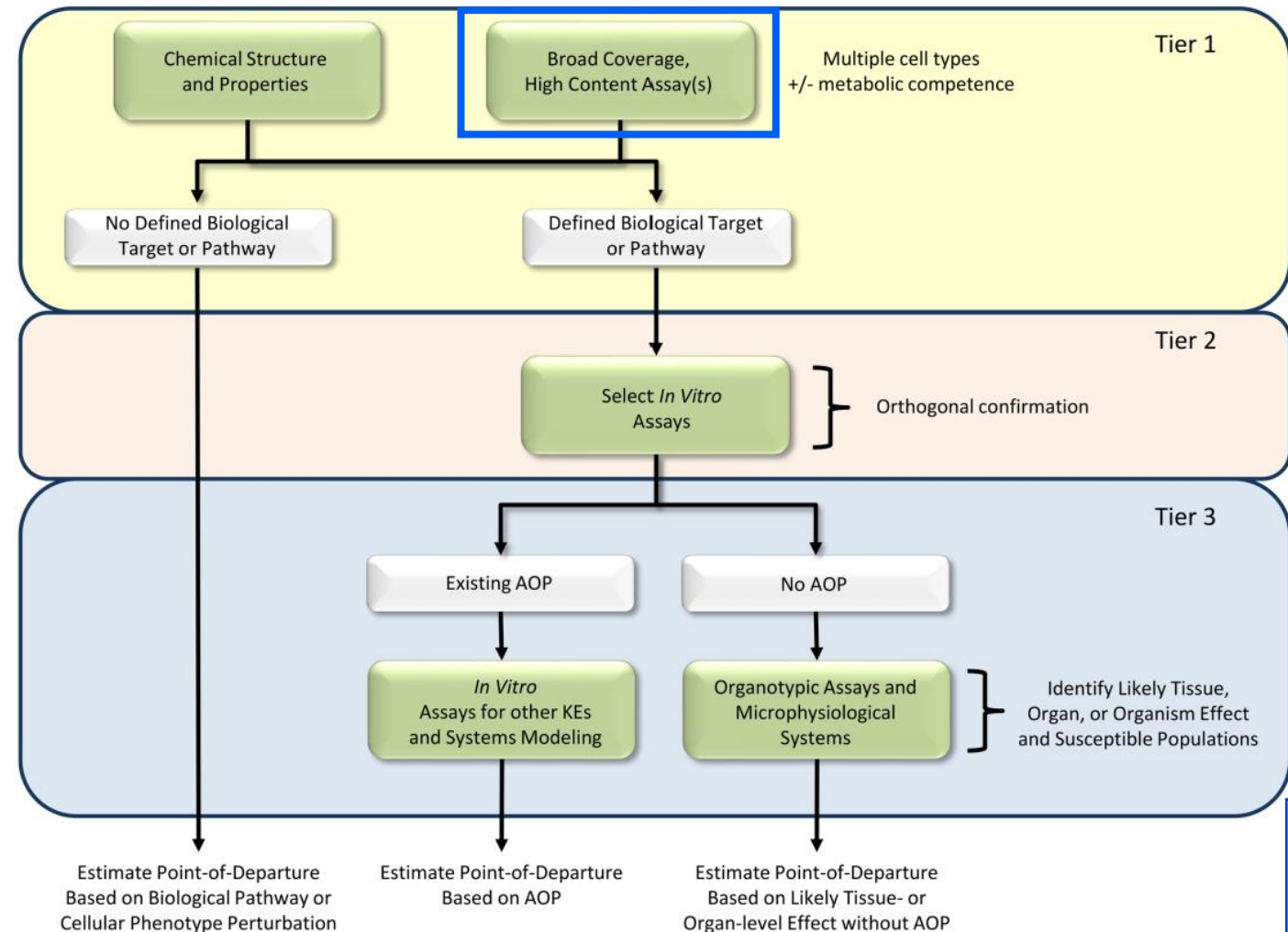
# Current toxicity paradigm & NGRA both designed to prevent harm



# Example from the US EPA framework for deriving protective PoDs

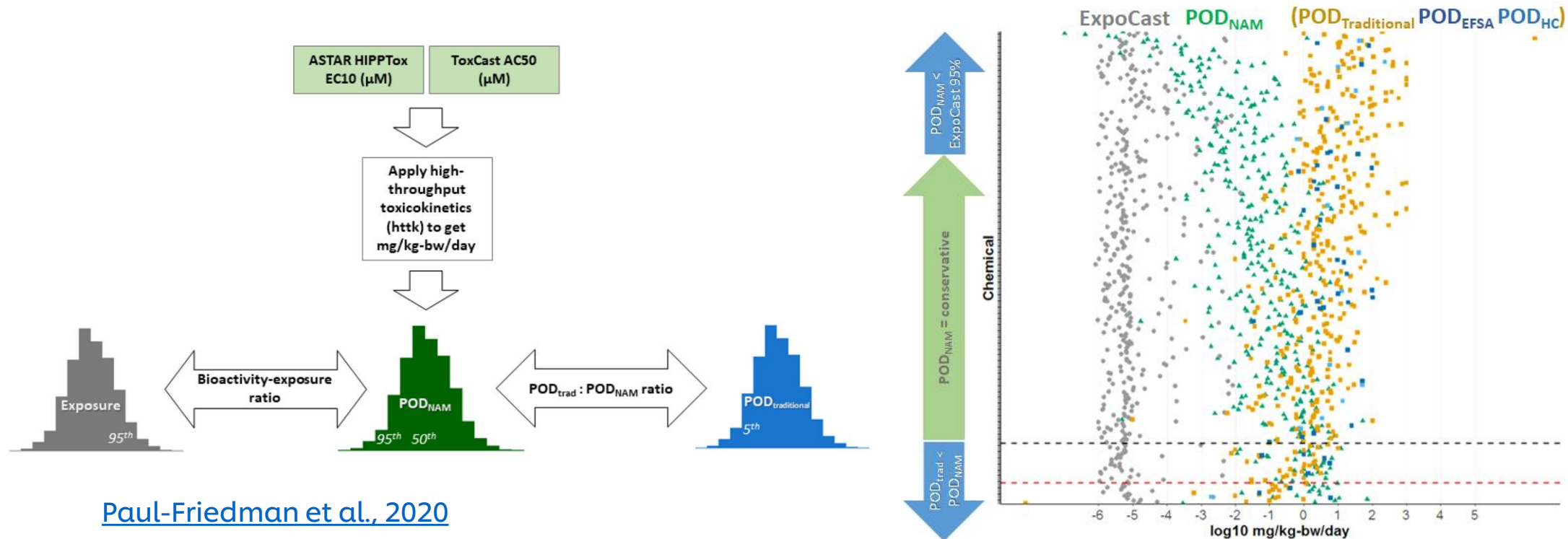
High-throughput transcriptomics (HTTr)<sup>1</sup> and High-throughput phenotypic profiling (HTTP)<sup>2</sup> developed to increase biological coverage

1. Harrill J et al 2019. Considerations for strategic use of high-throughput transcriptomics chemical screening data in regulatory decisions. *Current Opinion in Toxicology* 15, 64-75.
2. Nyffeler J et al 2019. Bioactivity screening of environmental chemicals using imaging-based high-throughput phenotypic profiling. *Toxicol Appl Pharmacol.* 2020;389:114876.



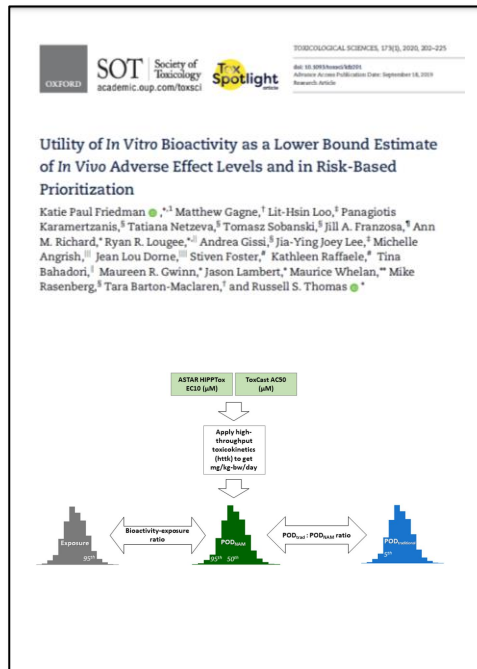
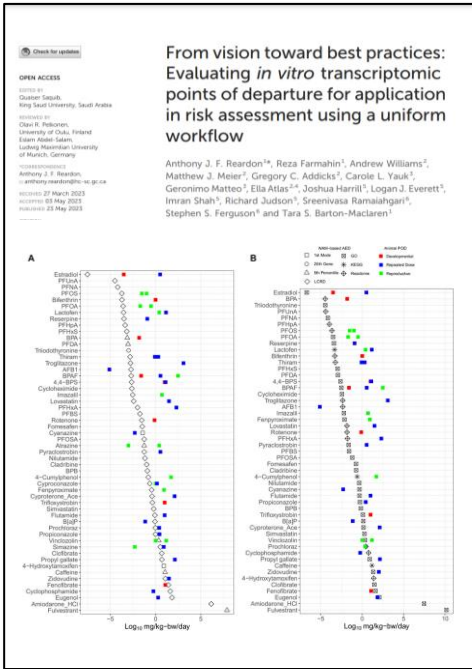
Russell S Thomas et al., 2019. The Next Generation Blueprint of Computational Toxicology at the U.S. Environmental Protection Agency. *Tox Sci* 169(2):317-332.

# Case Study Demonstrating Application of Bioactivity as a Protective POD



- For 89% of the chemicals NAM PoD was more conservative than the traditional POD.
- Bioactivity:exposure ratios (BERs) approach useful for accelerate screening and assessment using NAMs for hazard and exposure.

# Examples of ongoing or completed case studies for NAM/NGRA BER based risk assessment or prioritisation



**Science Approach Document**

**Bioactivity Exposure Ratio: Application in Priority Setting and Risk Assessment**

Health Canada

March 2021

<https://www.canada.ca/en/environment-climate-change/services/evaluating-existing-substances/science-approach-document-bioactivity-exposure-ratio-application-priority-setting-risk-assessment.html>

OECD  
Organization 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 Phenylethanol when included at 1% in a body lotion

Series on Testing and Assessment,  
No. 349

JT0483903

This document, as well as any data and any included therein, are without prejudice to the status of or sovereignty over any territory, or the delimitation of international frontiers and boundaries and to the name of any territory, city or area.

**EUTOXRISK**  
EU-ToxRisk  
An Integrated European 'Flagship' Program  
Driving Mechanism-based Toxicity Testing and Risk Assessment  
for the 21<sup>st</sup> Century

**Case Study 16 Reporting Template**

**Team: 2**

Team Members: Barira Islam; Ugis Sarkans; Marcel Leist Alessandra Roncaglioni; Jukka Sund; Andrew White,

Compound ID: CS\_16-02  
Compound Name: (4-Hydroxy-2,2,6,6-tetramethylpiperidin-1-yl)iodoacetate; TEMPOL

Structure: CC1(C)C(C(C1)O)C(=O)N

**Ab Initio Case Study Objectives**

**Search Objectives**

- Establish ability of Ab Initio to establish a priori alternatives for risk assessment
- Compare Ab Initio to existing data for the same chemical to determine if Ab Initio is a viable alternative to existing data for risk assessment
- Establish whether Ab Initio is a viable alternative to existing data for risk assessment
- Determine the extent to which the use of Ab Initio can be identified as a viable alternative to existing data for risk assessment

**Project Objectives**

- Establish the assessment approach to be used in Ab Initio, taking into account the available data and the objectives of the study
- Establish the Ab Initio workflow and the data to be used
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Other Identifiers: CAS ID 2226-96-2; CHE

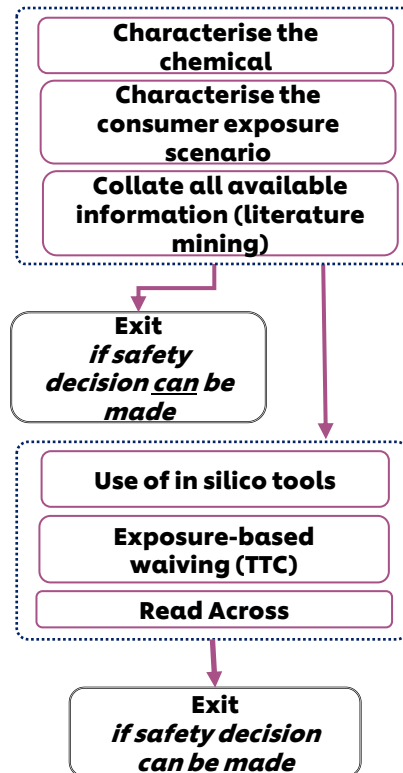




# Unilever development of a systemic toolbox

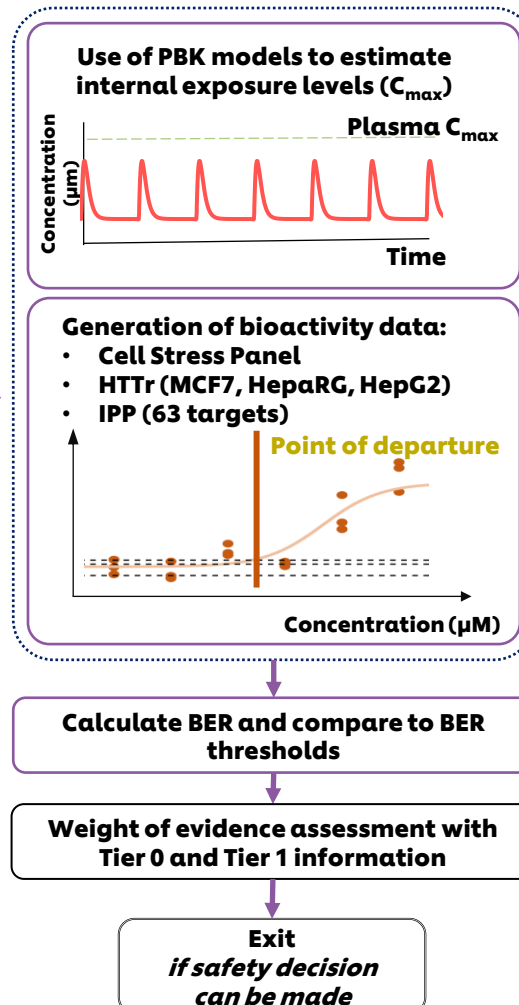
# A NAMs/NGRA Tiered Framework Approach: The overall goal is a human safety risk assessment

## Tier 0: Problem Formulation



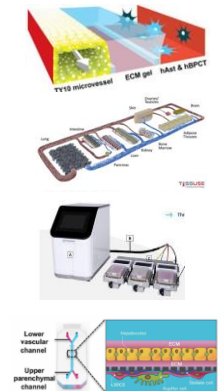
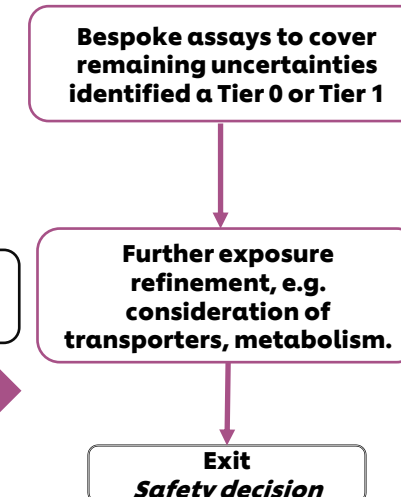
Decision cannot be made

## Tier 1: Systemic-safety toolbox



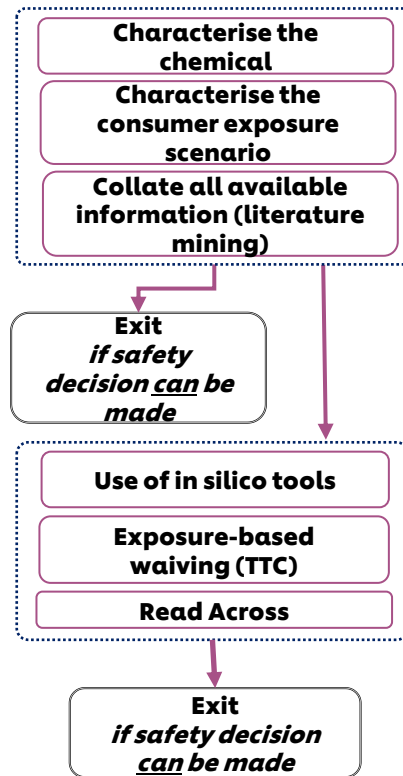
Decision cannot be made

## Tier 2: Refine Assessment

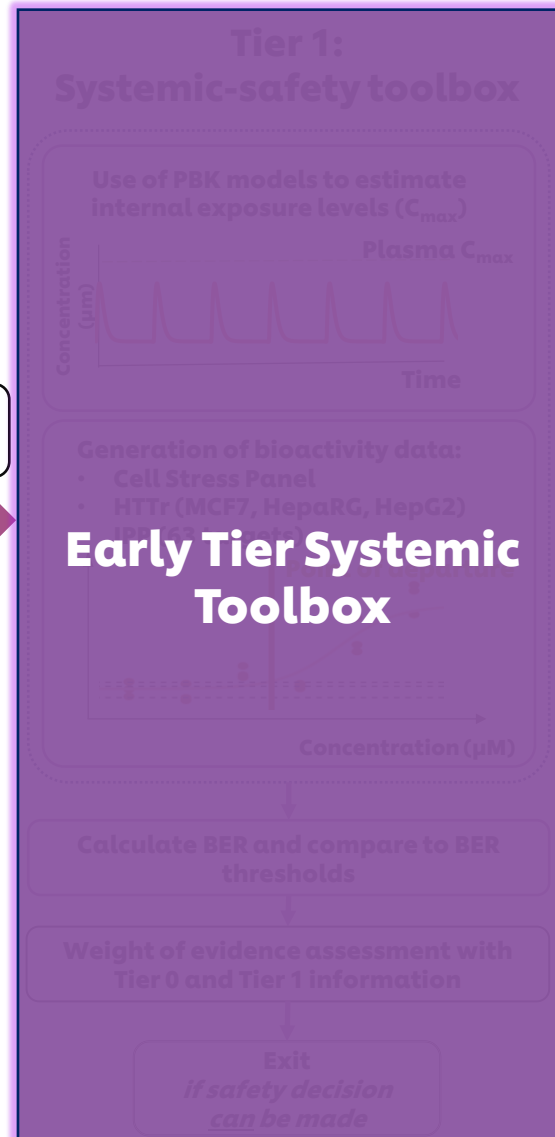


# A NAMs/NGRA Tiered Framework Approach: The overall goal is a human safety risk assessment

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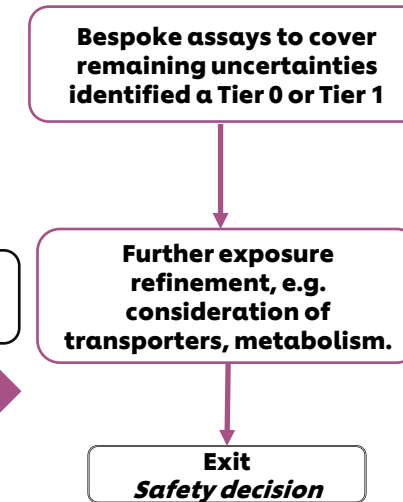


Decision cannot be made



Decision cannot be made

## Tier 2: Refine Assessment



# Evaluation/“Validation” of an Early Tier Toolbox for Systemic Safety

**AIM:** Use NAMs to ensure the protection of consumers: can the approach be used to confidently identify high/low risk chemical exposure scenarios?

- 1. Define the toolbox components** Choose and evaluate a set of NAMs covering exposure modelling and bioactivity investigations
- 2. Select test chemicals** Choose as many as practicable to maximise coverage of different chemistries and biological effects/toxicity
- 3. Set performance criteria** Define the ‘truth’ that the performance of the toolbox will be compared to

## Evaluation split into 2 stages: Pilot and extended evaluation

### Pilot study (Middleton et al., 2022)<sup>1</sup>



Define what the toolbox contains (which NAMs) and the workflow & protocols



Set performance criteria



10 chemicals selected by experts  
24 exposure scenarios



Define prototype decision model for determining protective BER threshold

### Extended evaluation (Cable et al., 2025)<sup>2</sup>



Repeat protocols/workflow as established in Middleton et al. 2022



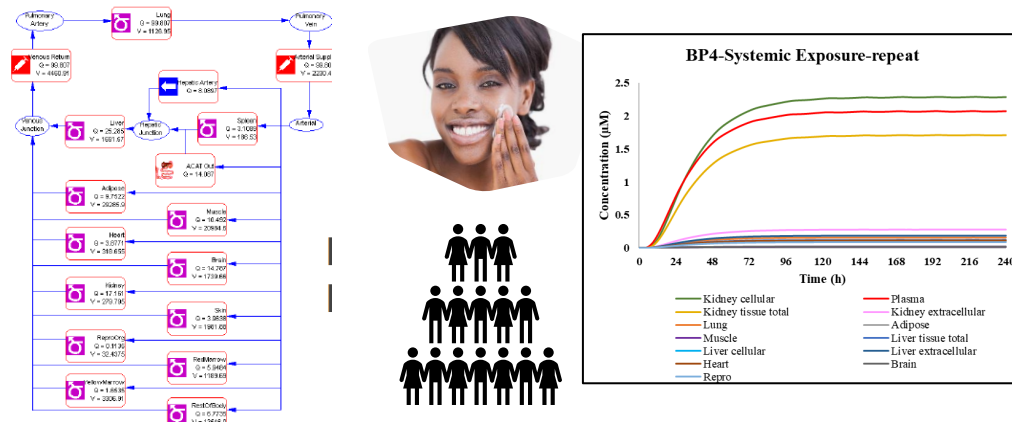
38 chemicals selected semi-randomly  
70 exposure scenarios



Apply the BER threshold

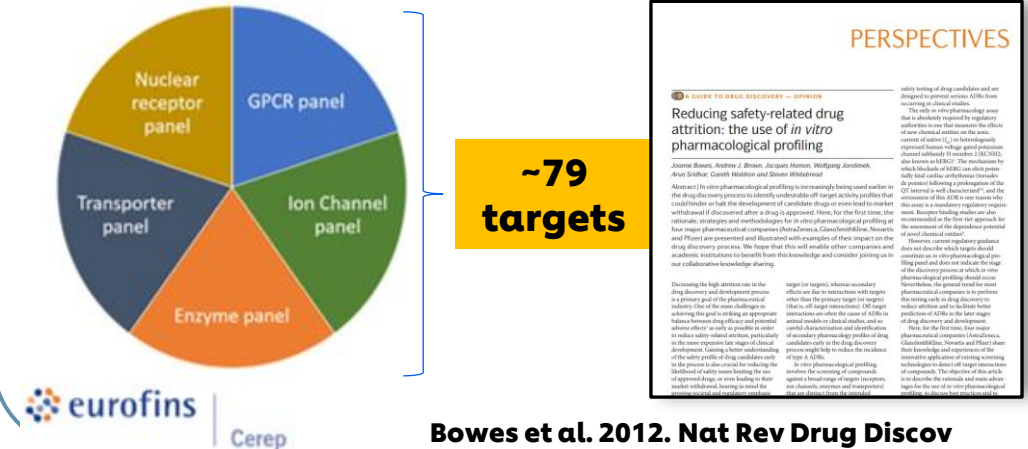
# Our Key NAMs

## Internal exposure - PBK modelling



Moxon TE et al., 2020. *Toxicology In Vitro*, 63, 104746

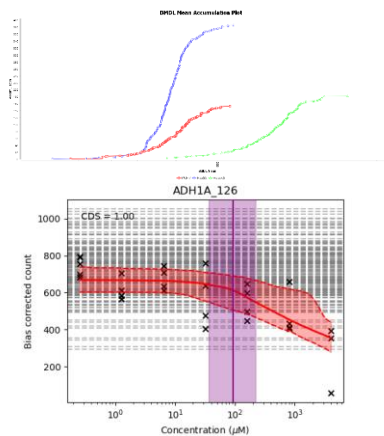
## In vitro pharmacological profiling



Bowes et al. 2012. *Nat Rev Drug Discov* 11(12): 909-22

## High-Throughput transcriptomics (HTTr)

- TempO-seq technology – full gene panel
- 24hr exposure
- 7 concentrations
- Various cell models (e.g. HepG2, MCF7, HepaRG)
- Dose-response analysis using BMDExpress2 and BIFROST model



Reynolds et al. 2020. *Comp Tox* 16: 100138  
Baltazar et al. 2020. *Toxicol Sci* 176(1): 236-252

## Cell stress panel (CSP)

- 36 biomarkers covering 10 cell stress pathways
- HepG2
- 24hr exposure
- 8 concentrations
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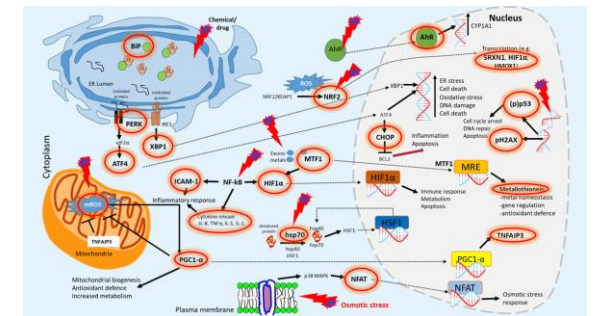


Image kindly provided by Paul Walker (Cypotex)

Hatherell et al. 2020. *Toxicol Sci* 176(1): 11-33

# Defining the toolbox components

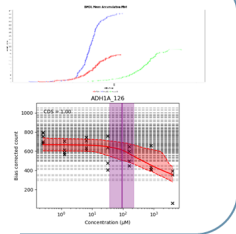
Pilot study  
(Middleton  
et al., 2022)

## Point of Departure determination from Bioactivity assays

### Non-specific effects

#### High-Throughput transcriptomics (HTTr)

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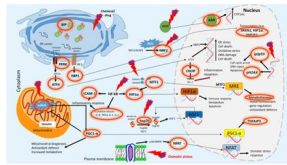


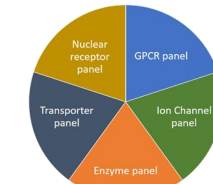
Image kindly provided by Paul Walker (Cyprotex)

Hatherell et al. 2020. Toxicol Sci 176(1): 11-33

### Specific effects

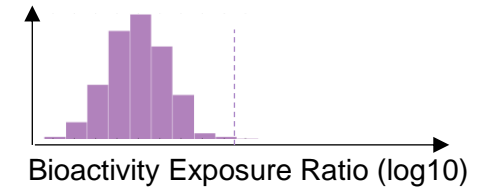
#### In vitro pharmacological profiling

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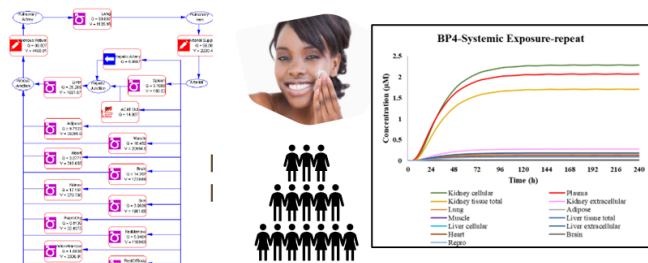


eurofins | Cerep

### Bioactivity Exposure Ratio Distribution



### Internal exposure - PBK modelling



Moxon TH et al., 2020. Toxicology In Vitro, 63, 104746

Plasma  
 $C_{max}$   
estimate

$C_{max}$  Error  
Distribution  
model (CMED)  
(Bayesian model)

# Standardisation of experimental design & computational pipelines

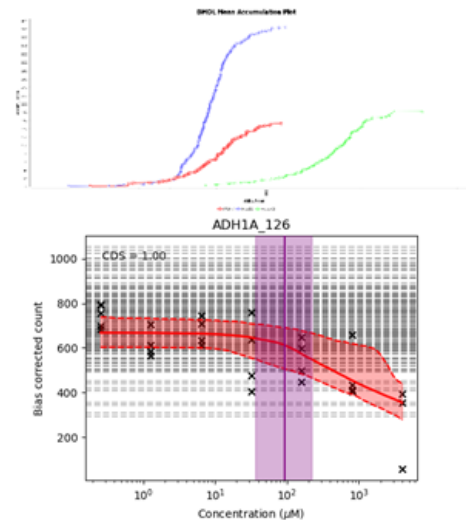
Pilot study  
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## Point of Departure determination

### Non-specific effects

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- Various cell models (e.g. HepG2, MCF7, HepaRG)
- Dose-response analysis using BMDExpress2 and BIFROST model



#### PODs obtained from Transcriptomics for each cell line, MCF7, HepG2, HepaRG 2D:

- **The minimum pathway BMDL** from the transcriptomics platform estimated using BMDExpress.
- **The global POD** from the transcriptomics platform estimated using BIFROST. The global POD represents an estimate of the minimum effect concentration across all genes. The method quantifies uncertainty in the POD as a probability distribution for each gene.

Reynolds et al. 2020. Comp Tox 16: 100138  
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# Standardisation of experimental design & computational pipelines

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## Point of Departure determination

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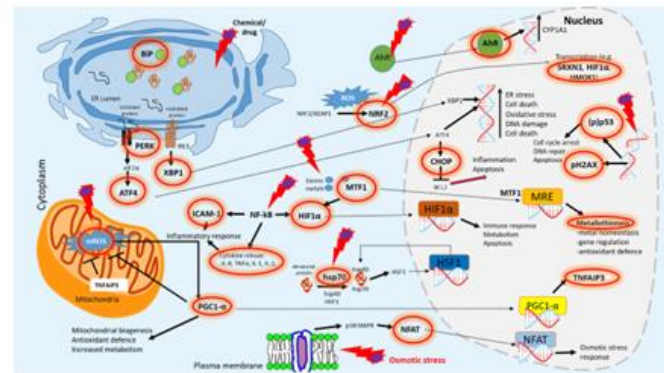


Image kindly provided by Paul Walker (Cyprotex)

Hatherell et al. 2020. Toxicol Sci 176(1): 11-33

- **The CSP global POD**, as estimated using BIFROST (i.e. minimum across all 36 biomarkers)
- The global POD represents an estimate of the minimum effect concentration across all biomarkers. The method quantifies uncertainty in the POD as a probability distribution for each biomarker.

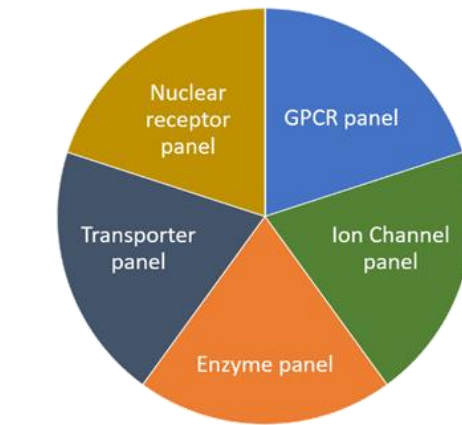
# Standardisation of experimental design & computational pipelines

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## Point of Departure determination

### Specific effects

### In vitro pharmacological profiling



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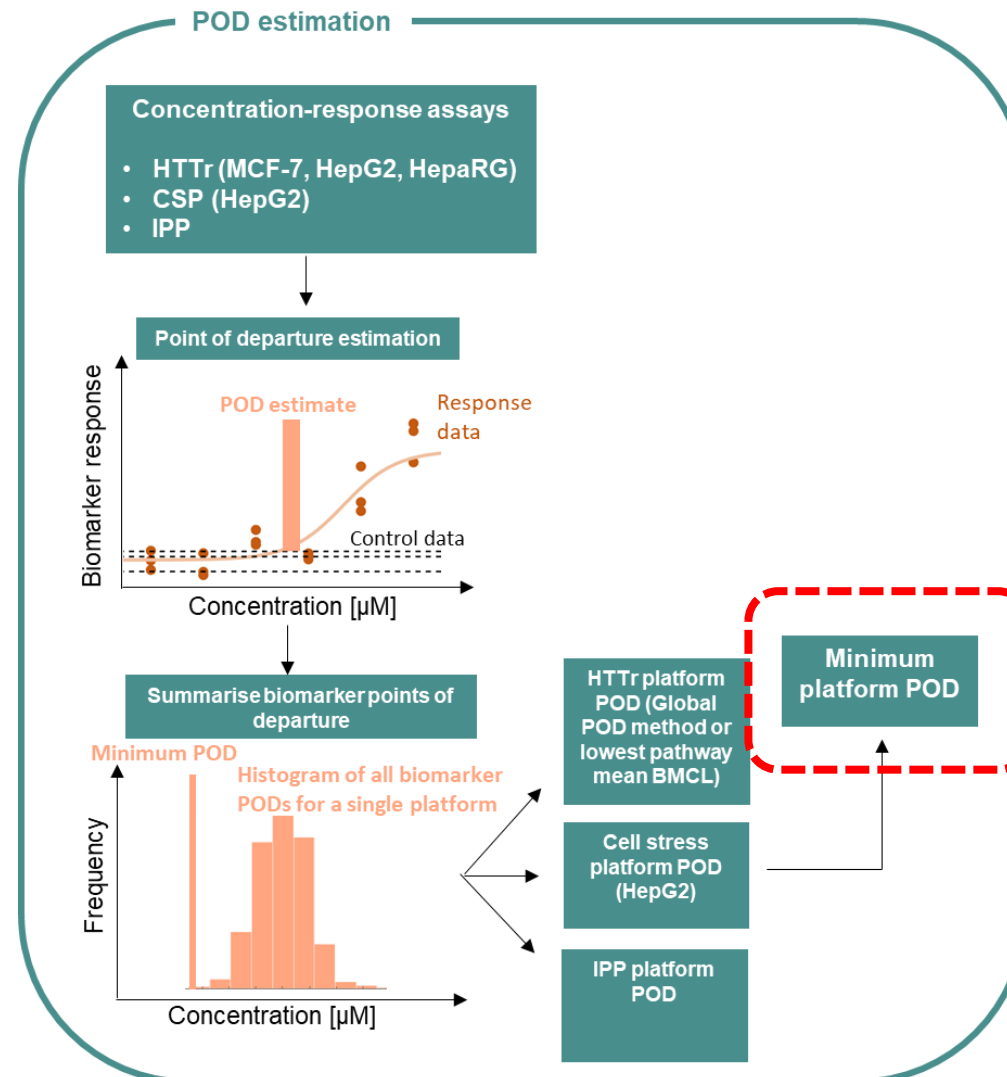
1. Bowes J et al 2012. Nat Rev Drug Discov;11(12):909-22.
2. Lynch JJ et al., 2017 Pharmacol Toxicol Methods;87:108-126.
3. Smit IA et al., 2021 Chem Res Toxicol;34(2):365-384.
4. Letswaart R et al., 2020 EBioMedicine;57:102837

- Panel developed by the pharmaceutical industry and used during early drug discovery to predict, assess and minimise/avoid risk of potential off-target adverse drug reactions.

- Initial panel of 44 targets identified to be related to adverse health outcomes
- Extended to 63 targets to include extra nuclear receptors
- Experiment in 2 phases:
- Screening at a fixed concentration (10 or 100  $\mu\text{M}$ )
- **Dose-response assays on positive hits to identify a point of departure (PoD) expressed as an IC50 value**

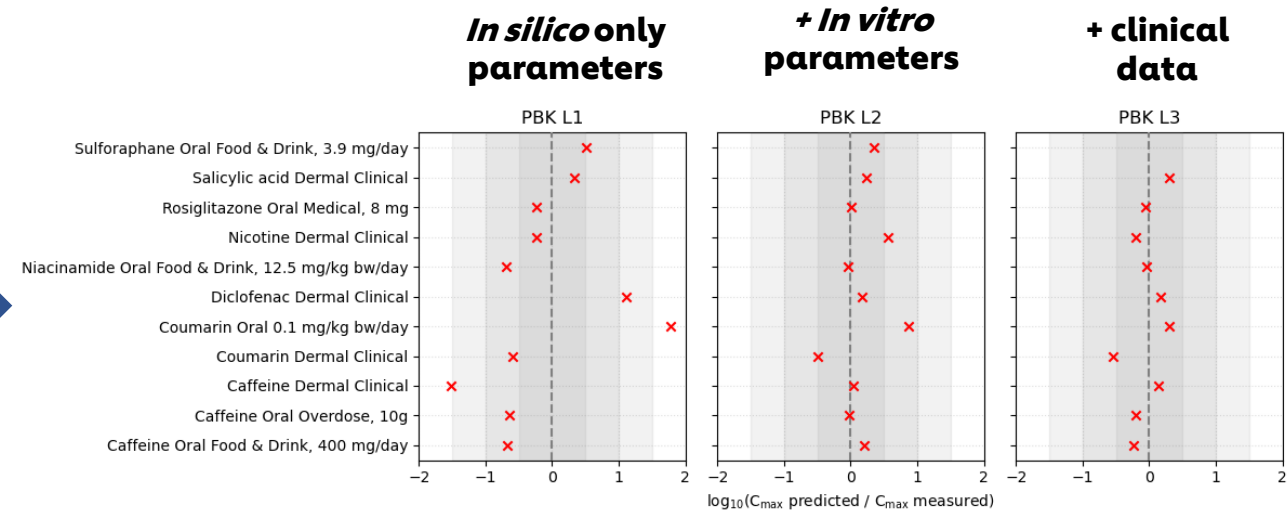
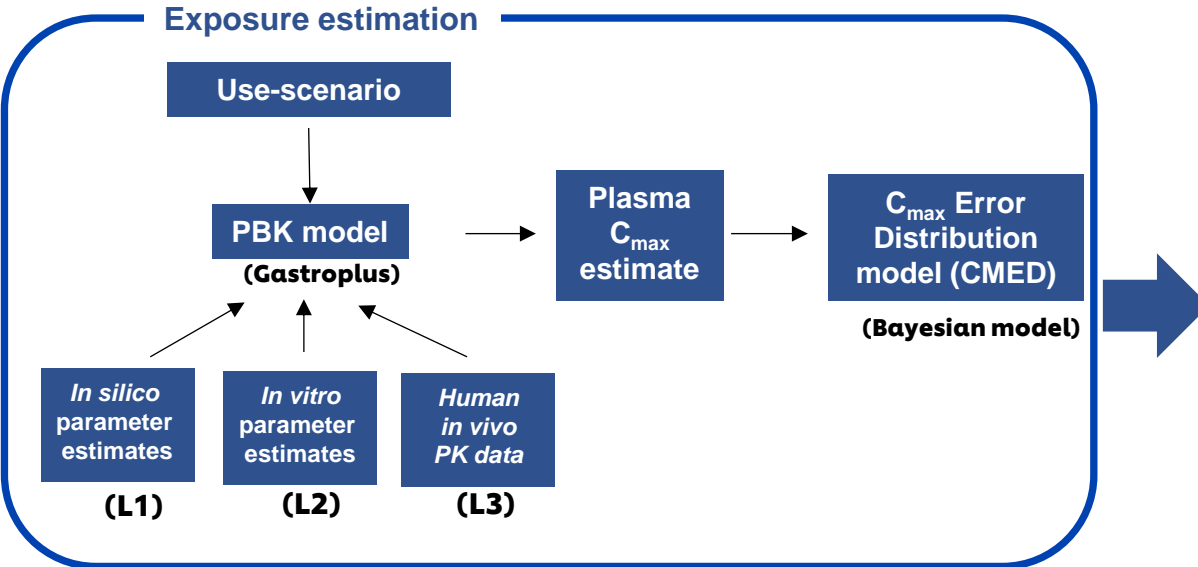
# Estimating PODs from bioactivity platforms- Minimum POD is selected for calculating a BER

Pilot study  
(Middleton  
et al., 2022)



# Estimation of plasma $C_{max}$ using Physiologically-based kinetic modelling: workflow & uncertainty analysis

Pilot study  
(Middleton  
et al., 2022)



- The PBK prediction error decreases as we go 'up' parameterisation levels
- Developed a Bayesian statistical model to quantify the error for a novel chemical
- **Output: Plasma  $C_{max}$  distribution at each PBK level**

# Defining the toolbox components

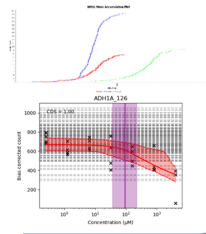
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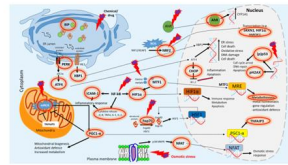


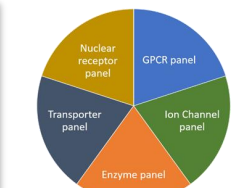
Image kindly provided by Paul Walker (Cyprotex)

Hatherell et al. 2020. Toxicol Sci 176(1): 11-33

### Specific effects

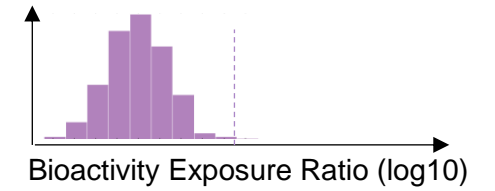
#### In vitro pharmacological profiling

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



eurolins | Cerep

### Bioactivity Exposure Ratio Distribution



### PBK Modelling

Face Cream

Clearance  
in silico 98.57 L/h  
in vitro 929 L/h

Toxicology in Vitro (2020), 63, 104746

Plasma  $C_{max}$  estimate →  $C_{max}$  Error Distribution model (CMED) (Bayesian model)

## Set performance criteria

- Assuming the current risk assessments are protective for human health:
  - The performance of the NAM toolbox is assessed against historical safety decisions
  - Benchmark chemical-exposure scenarios with known outcomes, low and high risk to define a safe BER threshold

**What we are trying to test:** Are the decisions made with a Tier 1 toolbox equivalent or better than the decisions we have been making with animal data?

**What we are not trying to test:** is the toolbox predictive of all possible adverse effects for a given chemical?

# Set performance criteria for evaluating the protectiveness and utility of the toolbox

Pilot study  
(Middleton  
et al., 2022)

## Benchmarking using chemical-exposure scenarios

- Chemicals with well-defined human exposures
- Traditional safety assessment available (e.g. regulatory opinions)
- Risk benchmarked to acceptability in a consumer product context

### Protectiveness

How many of the high risk exposure scenarios are identified as uncertain/high risk  
**(i.e. BER < threshold)**

### Utility

How many of the low risk scenarios are identified as low risk at this early tier stage in a risk assessment framework  
**(i.e. BER > threshold)**

# Select test chemicals with known human exposure and associated risk assessments

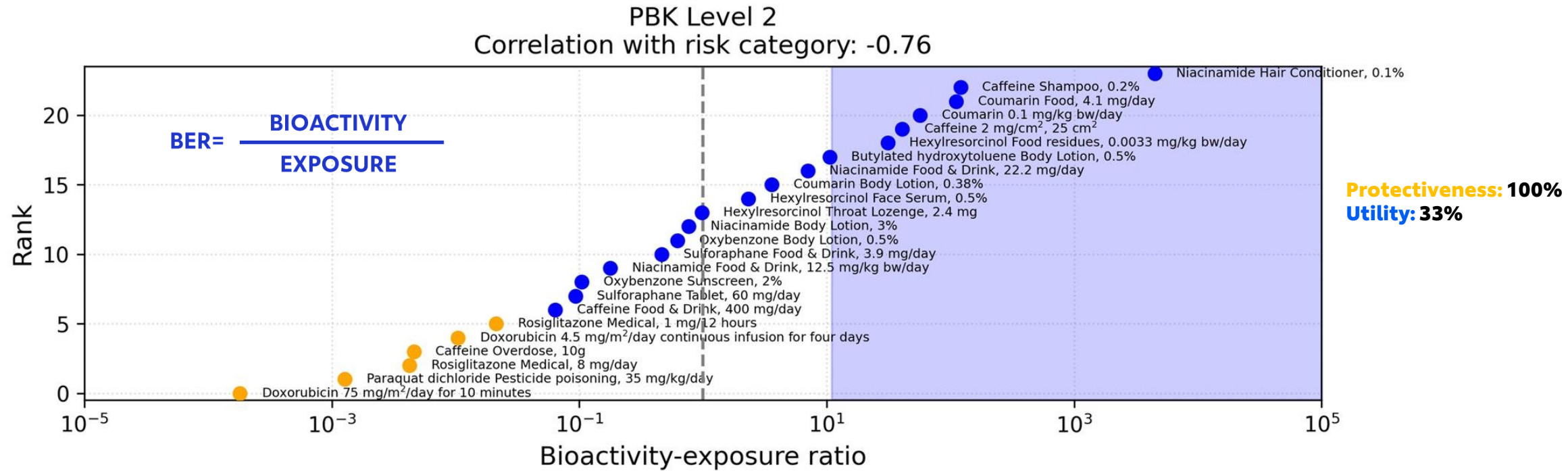
Pilot study  
(Middleton  
et al., 2022)

Chemical	Exposure scenario	Risk classification
<b>Oxybenzone</b>	<b>2 scenarios:</b> 0.5%; 2% sunscreen	Low risk
<b>Caffeine</b>	<b>2 scenarios:</b> 0.2% shampoo & coffee oral consumption 400 mg/day	Low risk
<b>Caffeine</b>	10g – fatal case reports	High risk
<b>Coumarin</b>	<b>3 scenarios:</b> 4 mg/d oral consumption; 1.6% body lotion (dermal); TDI 0.1 mg/kg oral	Low risk
<b>Hexylresorcinol</b>	<b>3 scenarios:</b> Food residues (3.3 ug/kg); 0.4% face cream; throat lozenge 2.4 mg	Low risk
<b>BHT</b>	Body lotion 0.5%	Low risk
<b>Sulforaphane</b>	<b>2 scenarios:</b> Tablet 60 mg/day; food 4.1-9.2 mg/day	Low risk
<b>Niacinamide</b>	<b>4 scenarios:</b> oral 12.5-22 mg/kg; dermal 3% body lotion and 0.1 % hair condition	Low risk
<b>Doxorubicin</b>	75 mg/m <sup>2</sup> IV bolus 10 min; 21 days cycles; 8 cycles	High risk
<b>Rosiglitazone</b>	8 mg oral tablet	High risk
<b>Paraquat</b>	Accidental ingestion 35 mg/kg	High risk



# NAM Systemic toolbox 100% protective for high-risk chemical exposure scenarios

Pilot study (Middleton et al., 2022)



Chemical-exposure scenarios with a BER point estimate outside the blue-shaded region would be identified as “uncertain” risk under this decision model. The grey-dashed line corresponds to BER = 1. Blue shaded region BER > 11 corresponding to threshold BER for PBK level 2 above which an exposure would be considered low risk. Blue circles: low risk chemical-exposure scenario; Yellow circles: high risk chemical-exposure scenario

## Threshold values of the BER point estimates for determining whether an exposure is low risk are dependent on the confidence on the PBK model

Pilot study  
(Middleton  
et al., 2022)

PBK Level	Threshold BER Required for Exposure to Be Identified as Low Risk	Confidence Threshold ( $p_{\text{threshold}}$ ) Required for Exposure Scenario to Be Identified as Low Risk
1	110	.98
2	11	.97
3	2.5	.95

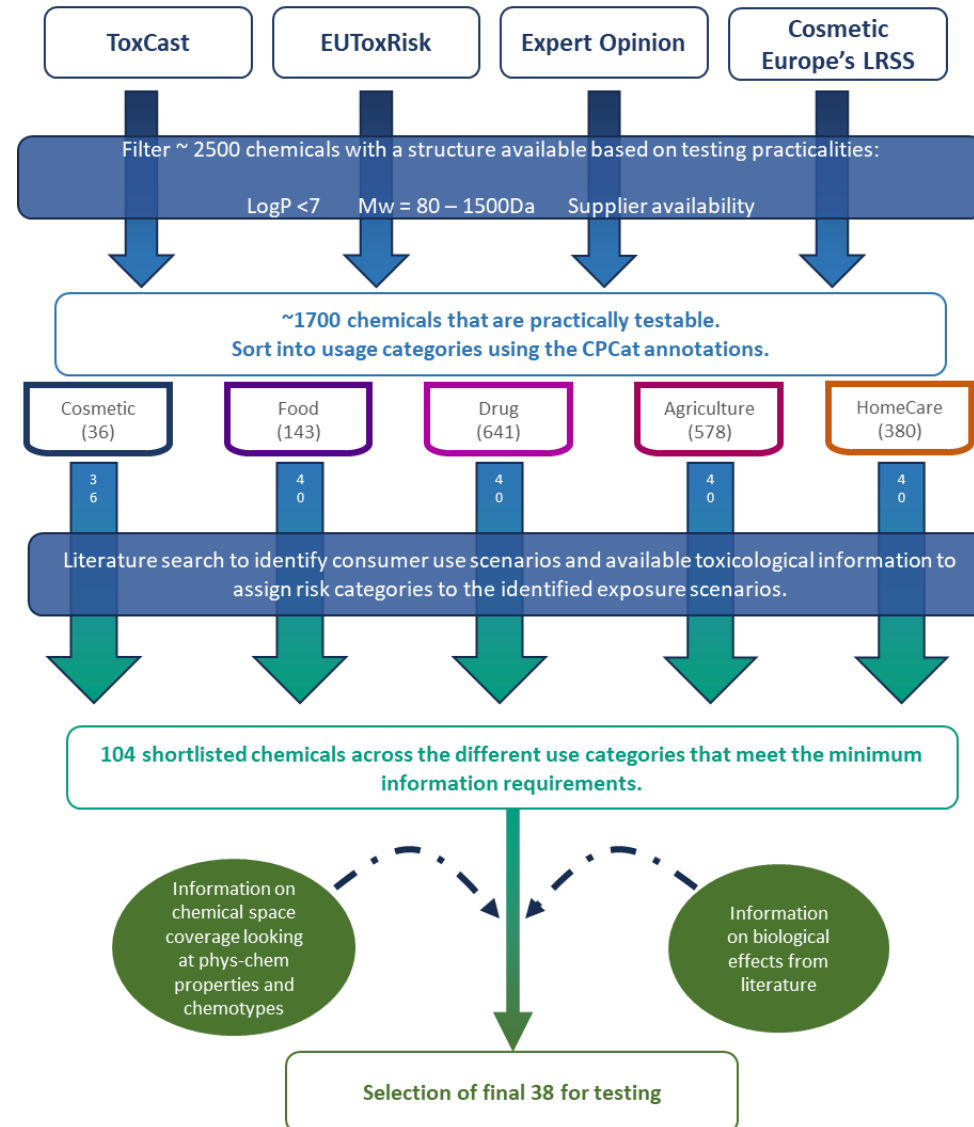


**Are these thresholds still protective if we increase the number and diversity of chemicals?**

Extended evaluation  
(Cable et al., in  
preparation)

# Semi-random selection of the 38 chemicals covering multiple use categories and chemistry

Extended evaluation  
(Cable et al., 2025)



## Semi-random selection of the 38 chemicals covering multiple use categories and chemistry

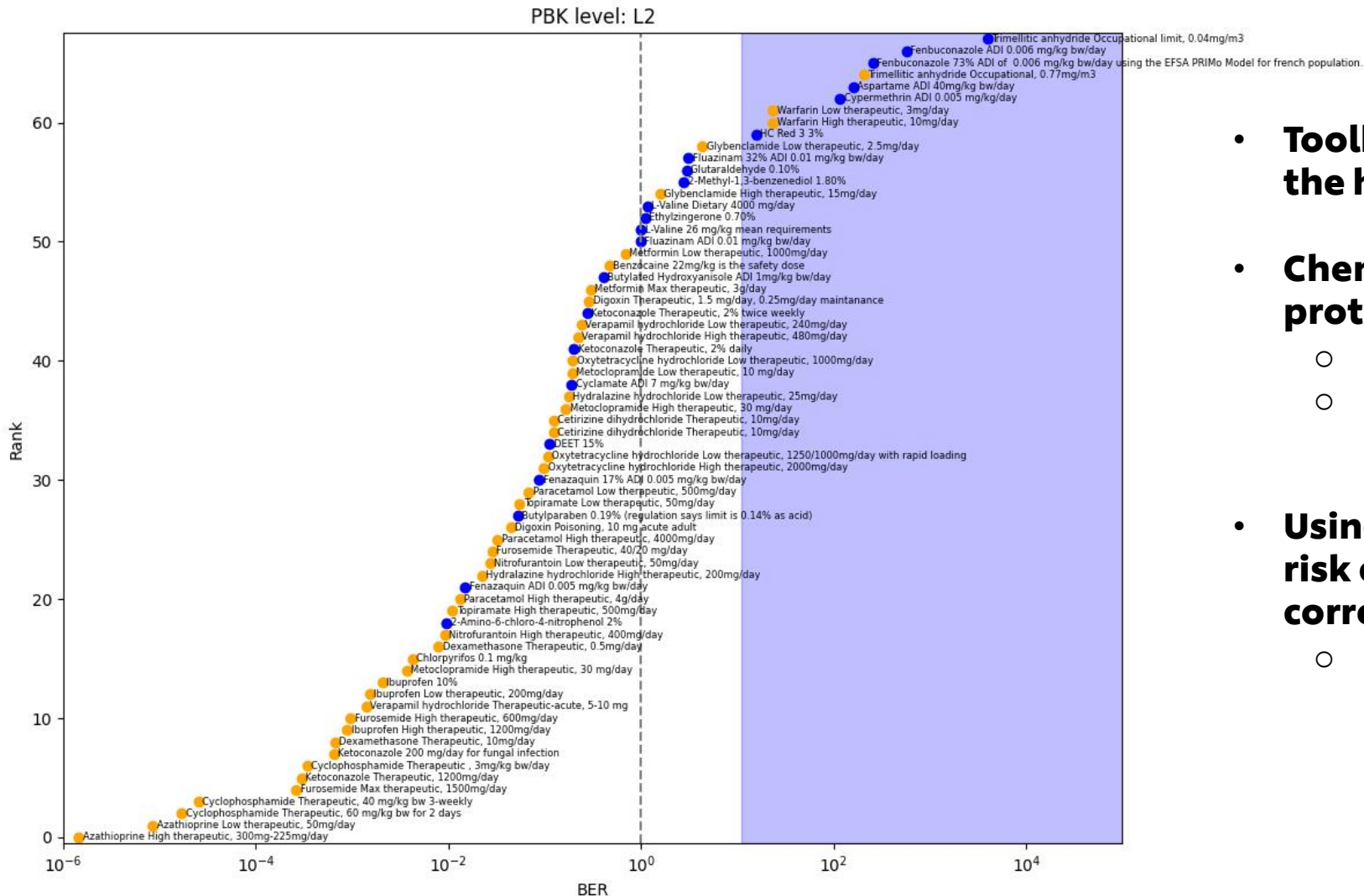
Extended evaluation  
(Cable et al., 2025)

### 38 test chemicals

- 9 cosmetic ingredients, 21 drugs, 3 food additives, 5 agricultural chemicals, 1 industrial chemical
  - Oral, dermal, IV and inhalation exposure scenarios
- Organ toxicities, CNS disruptions, immune system dysregulation, non-specific effects, blood-based disorders etc...

Extended evaluation  
(Cable et al., 2025)

## NAM Systemic toolbox remains protective (93%) when 38 additional chemicals and 70 exposure scenarios were tested

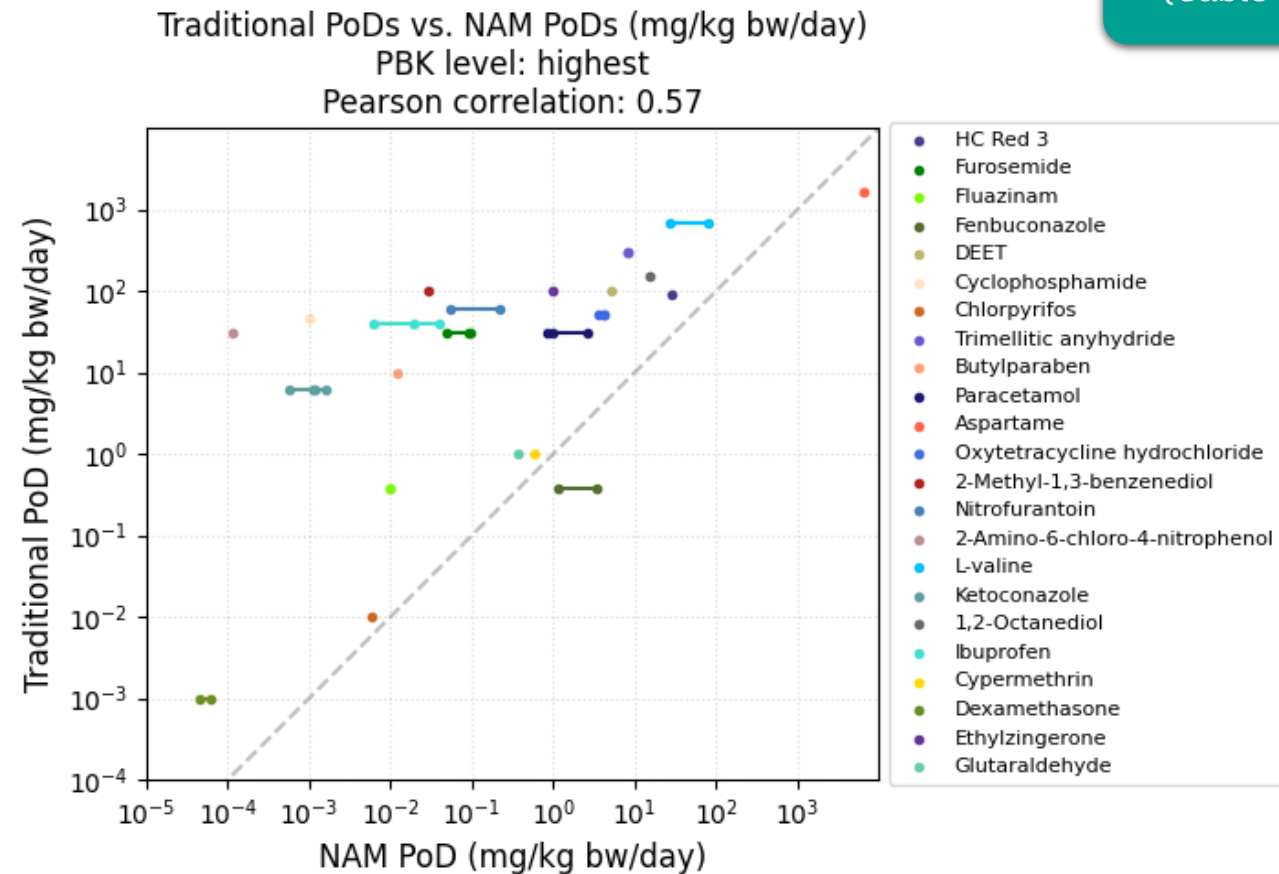


- **Toolbox not protective for 3/46 of the high-risk exposure scenarios**
- **Chemical- Exposure scenarios not protective for:**
  - Warfarin therapeutic oral dose
  - Trimellitic anhydride inhalation exposure
- **Using BER > 11, only 27% of the low-risk chemical-scenarios would be correctly identified as such**
  - For the other 73%, refinement is needed (i.e. Approaches to distinguish bioactivity from adversity; refine exposure estimates etc.).

## NAM PoDs are more conservative (i.e. lower) than the minimum in vivo PoD

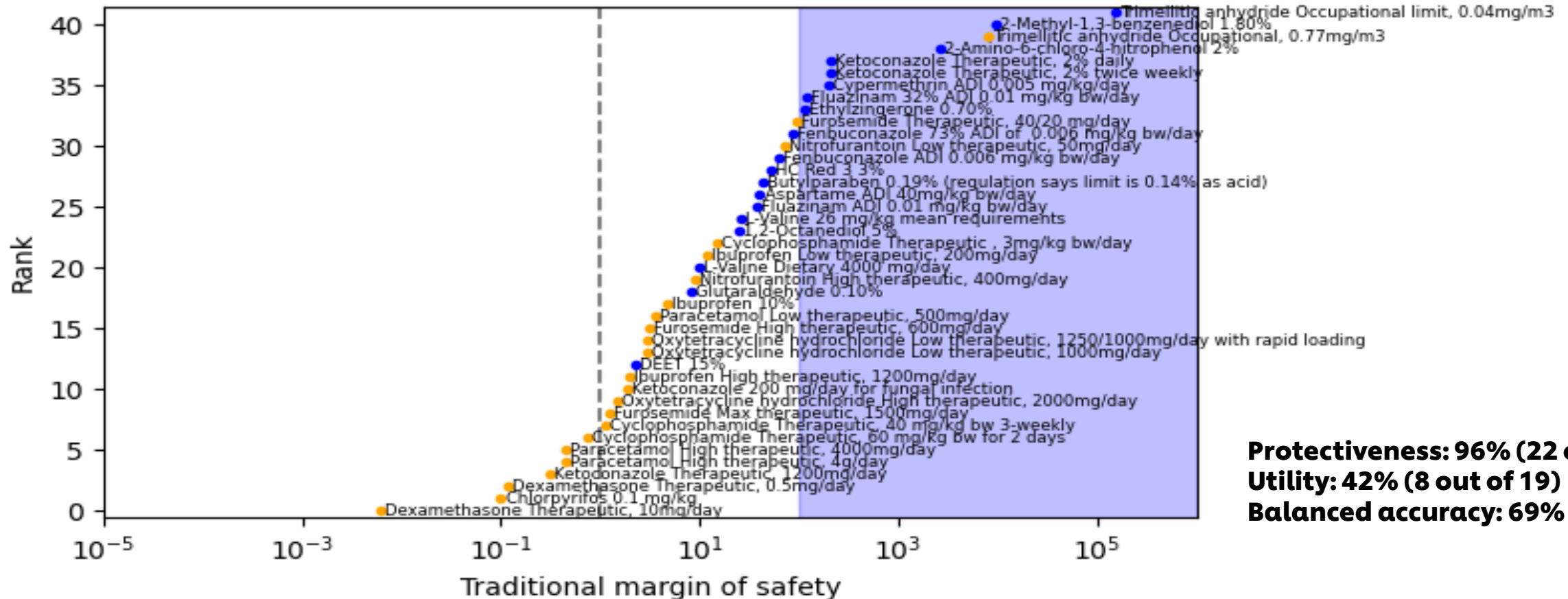
Extended evaluation  
(Cable et al., 2025)

- For 25 chemicals the lowest in vivo NOAEL or NOEL was identified from three sources: ToxRefDB, the supplementary material of Paul-Friedman et al (2020) and published regulatory opinions
- Reverse dosimetry was performed to transform PODNAM in  $\mu\text{M}$  to an external dose in mg/kg/day
- The range reflects that for some chemicals more than 1 exposure scenario was assigned



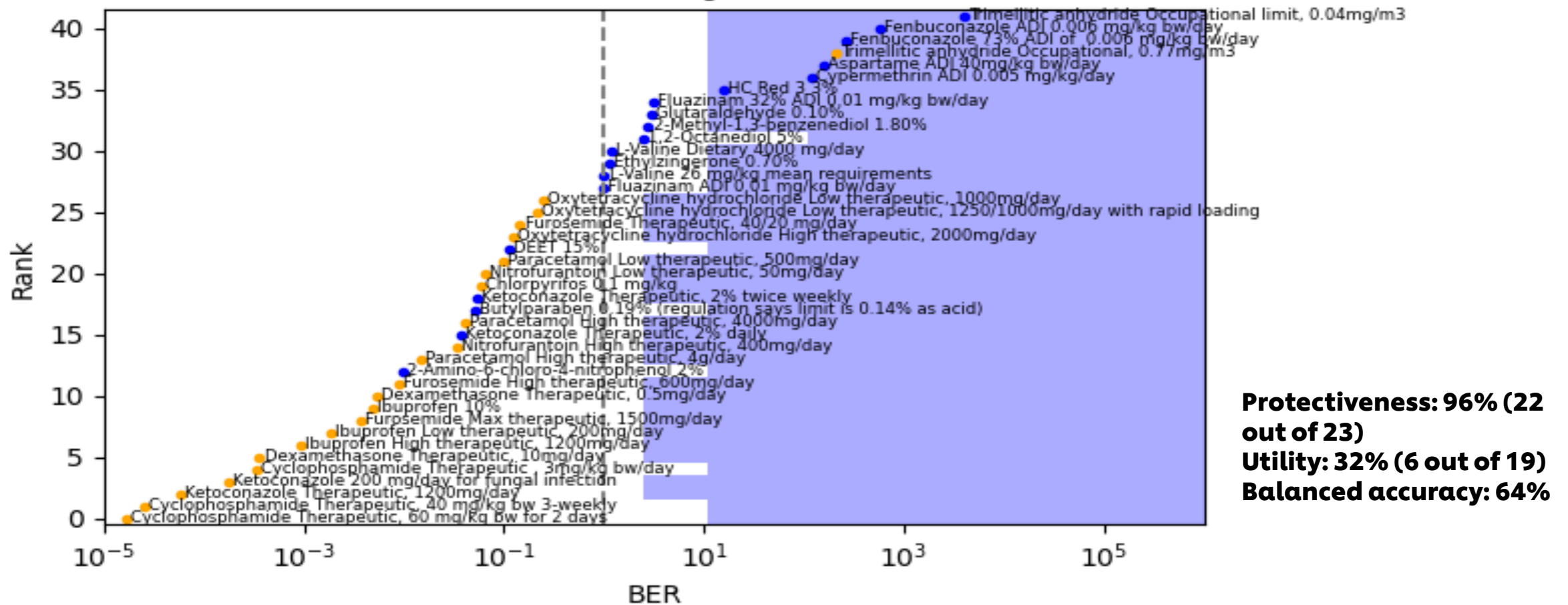
# The protectiveness and utility of the traditional approach was calculated to be 97% and 42% when using the lowest in vivo NOEL/NOAEL and a Margin of Safety of 100

Comparison of traditional margins of safety and benchmark risk classifications



For the same chemicals, the performance of the NAM-based toolbox was equivalent (96% protectiveness and 32% utility)

Comparison of BERs and benchmark risk classifications  
PBK level: highest





## Key findings from the evaluations

- The toolbox is protective for a wide range of chemicals and could be used within a weight of evidence risk assessment framework.
- PODNAM are conservative for most of the chemicals.
- For majority of the chemicals, the lowest PoD was obtained from the transcriptomics when using the gene-level PoD, followed by IPP.
- Systemic Toolbox is protective for high-risk chemicals despite not always capturing the MoA.
- For chemicals with a specific MoA, IPP is able to detect if the target is present in the panel.
- Generic PBK models might be insufficient to provide more accurate predictions for chemicals which are substrates of transporters.
- Dose-response modelling for transcriptomics leads to a high number of false positives – i.e. low risk exposure scenario being classified as uncertain risk.

# Application of NGRA to the evaluation of Climbazole as a cosmetic ingredient

## Climbazole: Objectives and Approach

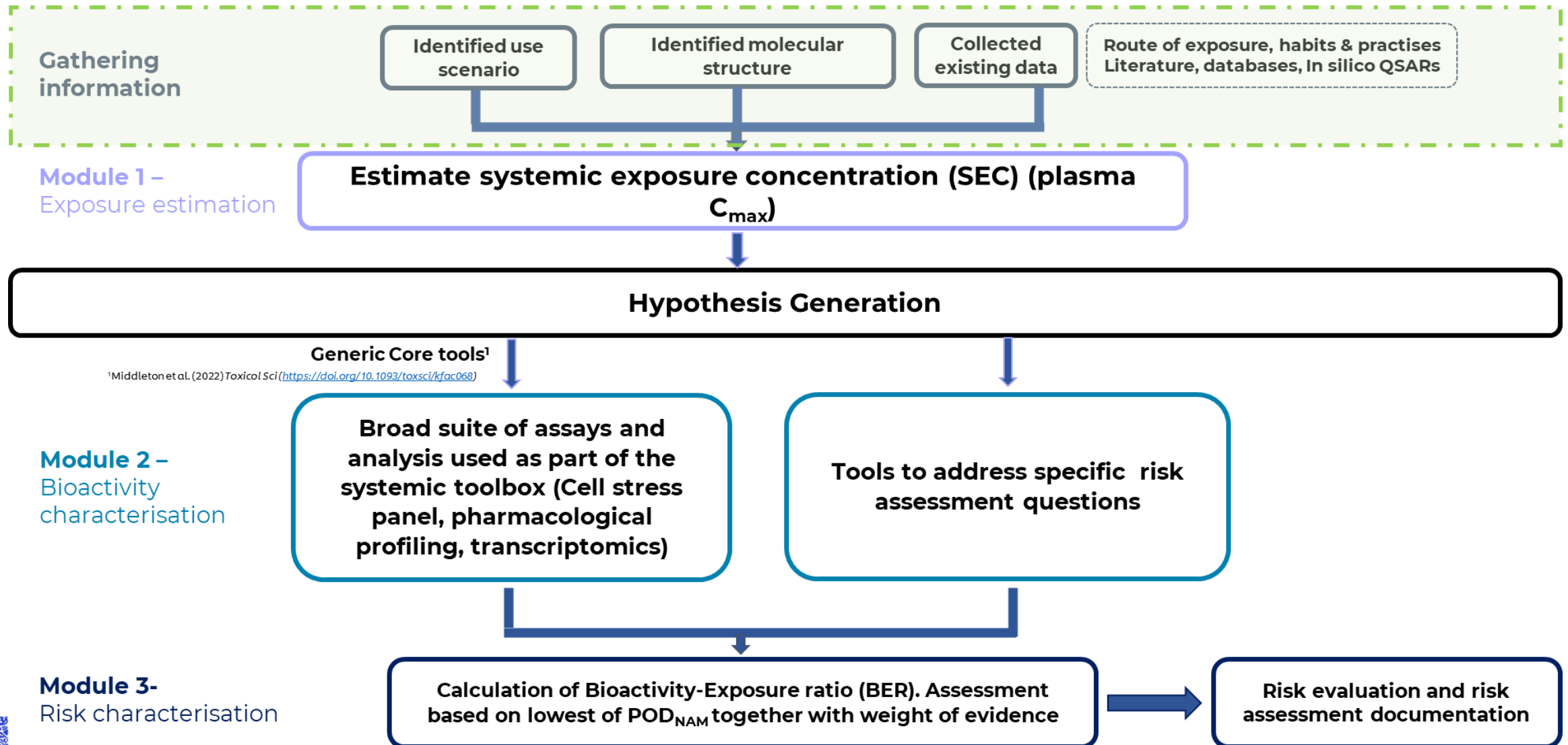
- Climbazole is an active ingredient used in several consumer products. We know that bioactivity-based NGRA can result in very conservative safety decisions, so the objective of this case study was to:
- Assess whether a tiered NGRA approach is sufficiently protective *and also useful* to assess the safety of a regulated cosmetic ingredient

**Is Climbazole safe when used at  
0.2% in a face cream?**

## Climbazole: Rules and Assumptions

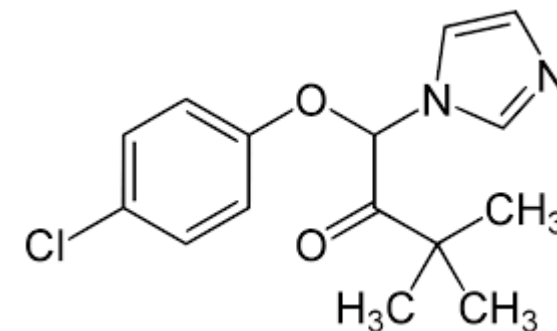
- **For the purposes of this exercise, it has been assumed that no *in vivo* animal data exist on the ingredient**
- **Focus on systemic toxicity**
- **Stand-alone illustration of how to assess systemic toxicity effects (not including genetic toxicity) using NAMs**

# Climbazole: Overall approach



## Climbazole: Use Scenario and Molecular Structure

- Climbazole (CAS 38083-17-9) has been used in Europe in cosmetics for decades as an anti-dandruff agent or preservative. It is currently regulated under Annex V of the Cosmetic Regulation and approved for use at up to 0.2% as a preservative in leave-on cosmetics.
- It is also approved for at up to 2% in rinse-off shampoo formulations.
- **The specific use scenario** of this case study is for **dermal application of a leave-on face cream formulation containing Climbazole at 0.2% w/w**



### Daily use of face cream:

- **Amount applied = 1.54 g/day** \*derived from the SCCS Notes of Guidance 2023
- **Concentration in the finished product = 0.2%**

## Climbazole: Alerts from *in silico* tools

- **DEREK Nexus**  **likely toxicity based on chemical structure**
- **METEOR Nexus**  **possible biotransformation based on chemical structure**
- **OECD QSAR Toolbox.**  **possible mechanisms of action**
- **TIMES** **likelihood of skin sensitisation of the parent and metabolites**
- **OPERA**  **physchem, environmental fate, range of human-relevant toxicity endpoints**
- **VEGA**  **physchem, human-relevant toxicity endpoints**

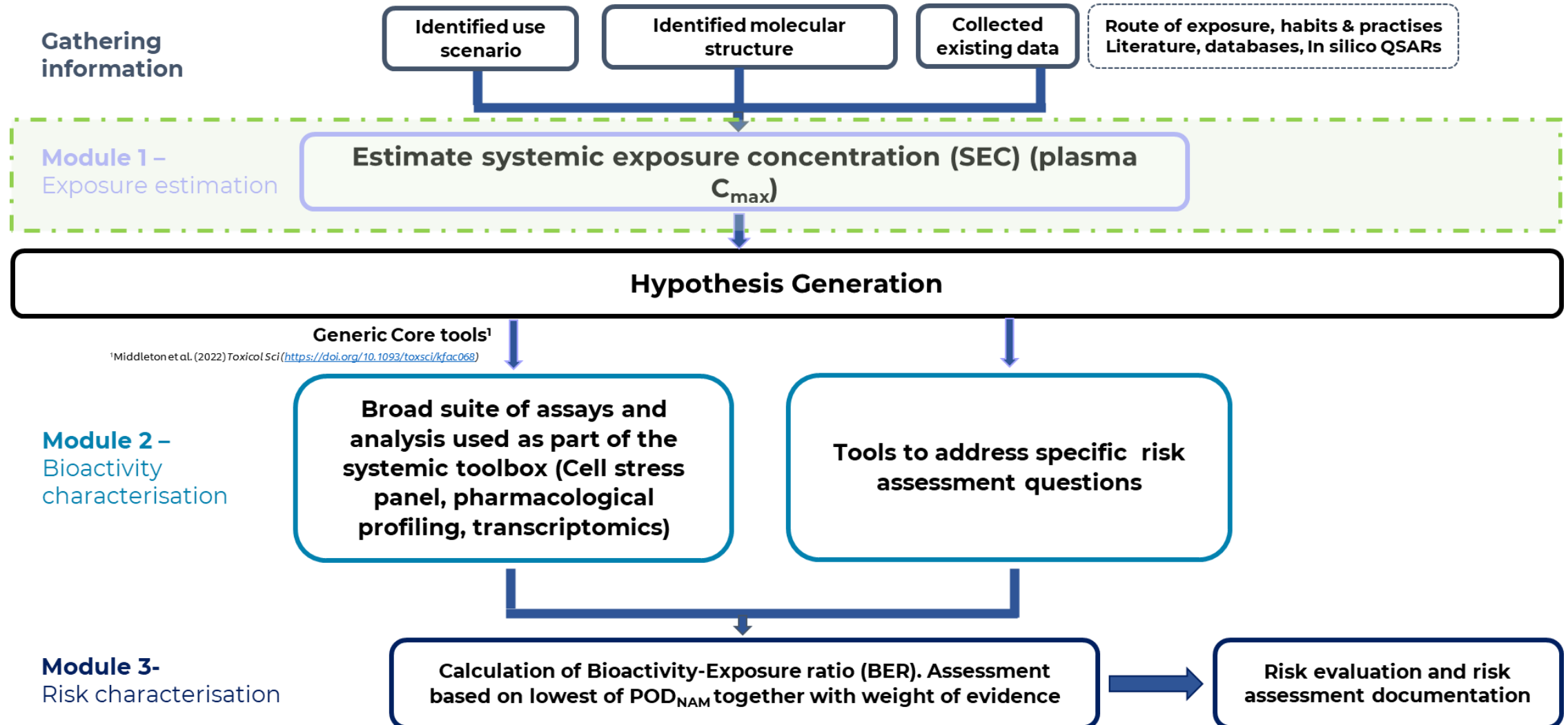
**AFSA training on predictive chemistry:** <https://youtu.be/rLWaSgGFGCI>

## Climbazole: Alerts from *in silico* tools

- Climbazole was within the domain of all models used.
- Climbazole was predicted to have a high order of toxicity, Cramer Class III
- Alerts for hepatotoxicity and protein binding were flagged across DEREK Nexus and the OECD QSAR Toolbox
- There were no alerts for mutagenicity or genotoxicity, however DEREK Nexus and the ISS model within the OECD QSAR Toolbox flagged alerts for carcinogenicity.
- There were no alerts for binding to either ER or AR.
- Climbazole flagged alerts for adrenal gland toxicity and reproductive and developmental toxicity.



# Climbazole: Overall approach



# Climbazole: Exposure Estimation

Exposure scenario and target individual/population

<b>Product type:</b>	<b>Face cream</b>
<b>Amount used per day (g/day):</b>	1.54
<b>Frequency of use:</b>	2.14 times per day
<b>Ingredient inclusion level:</b>	0.2%
<b>Application site:</b>	½ area head
<b>Skin surface area (cm<sup>2</sup>):</b>	565
<b>Target individual</b>	<b>60 kg European female</b>
<b>Leave on or rinse off:</b>	Leave on
<b>Amount of ingredient in contact with skin per occasion (mg):</b>	3.08

External applied dose = **0.0513 mg/kg bw/day**  
 $(A \times 1000 \text{ mg/kg} \times C/100) / 60 = \text{mg/kg bw/day}$

# Climbazole: Exposure Estimation

## From applied dose to internal concentrations

### External dose

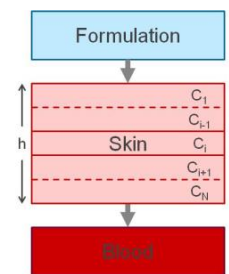
- **Route of exposure**
- **Consumer use (Habits & Practices)**
- **Applied dose (external concentration)**



### ADME parameters

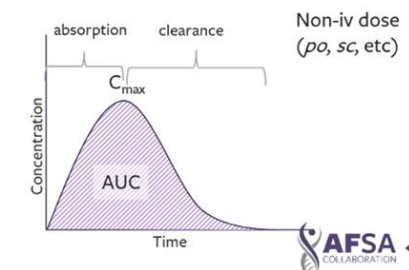
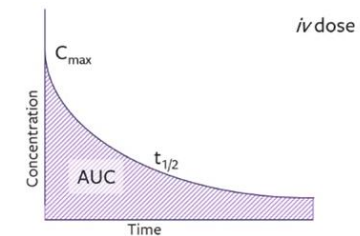
**A**bsorption  
**D**istribution  
**M**etabolism  
**E**limination

- **Skin penetration**
- **Phys-chem properties**
- **Hepatic clearance**
- **Fraction unbound**
- **Blood:plasma ratio**



### Kinetic profile of chemical

**Physiologically-based kinetic (PBK) modelling**  
– **Internal concentration (plasma, urine, organ-level)**



Images from: AFSA training module  
"Dosimetry (Internal Exposure)", 2022

# Climbazole: Exposure Estimation

## What is PBK modelling?

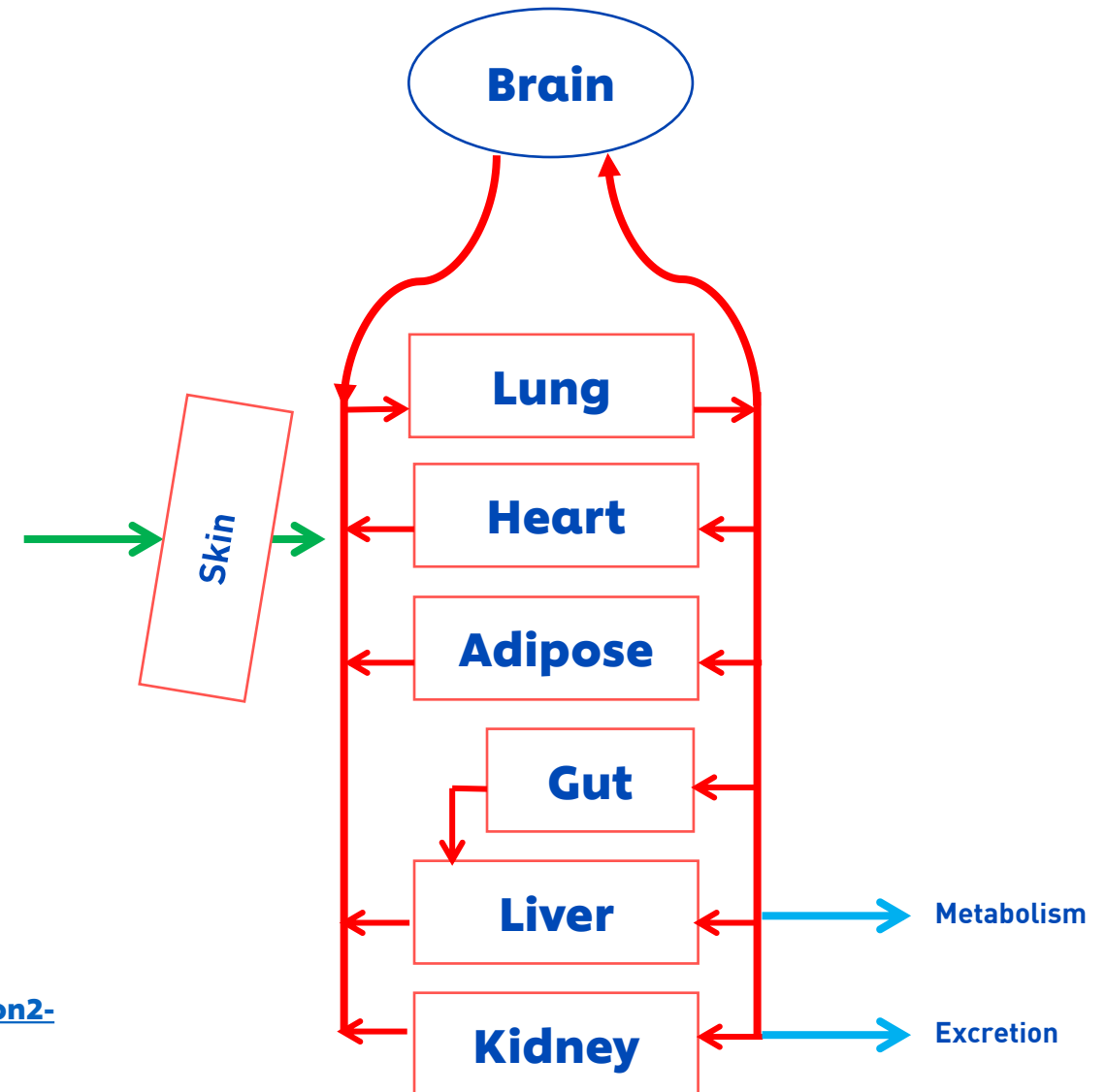
- **Mathematical description of interconnected compartments representing the human body**
- **Describe ADME (Absorption, Distribution, Metabolism, and Excretion) properties of a chemical within the body**
- **Prediction of concentration in blood, plasma, and tissues over time**
- **Can model an individual or a population**

Links to training materials on PBK modelling:

NURA Dynamic discussions: [https://pcrm.widen.net/view/video/xr5ojwu8vo/Session2-DyNAMic-Discussions-2023?x.share=true&x.portal\\_shortcode\\_generated=a7lwj1xi&x.app=portals](https://pcrm.widen.net/view/video/xr5ojwu8vo/Session2-DyNAMic-Discussions-2023?x.share=true&x.portal_shortcode_generated=a7lwj1xi&x.app=portals)

AFSA: <https://youtu.be/UGKEMS6DPRo>

AFSA: <https://youtu.be/UGKEMS6DPRo>



# Climbazole: Exposure Estimation

## ADME Data Generation

- **In silico tools exist to predict ADME properties from structure (ADMET predictor withing GastroPlus)**
- **The most important ADME properties were generated through in vitro testing:**
  - ***Dermal absorption:*** used to derive kinetic parameters for chemical partitioning in the skin layers and absorption through systemic circulation ***Low dermal penetration in vitro***
  - ***Blood to plasma ratio:*** determines the concentration of the chemical in whole blood compared to plasma and provides an indication of chemical binding to erythrocytes. ***Binds RBCs***
  - ***Plasma protein binding:*** the degree of binding determines the free available concentration of the chemical in plasma. ***High binding to human plasma proteins (97.09%)***
  - ***Metabolic stability:*** evaluated using plated primary hepatocytes and it is used to understand the route of elimination of a chemical and derive values for intrinsic hepatic clearance and half-life. ***High clearance in the assay***

# Climbazole: Exposure Estimation

## ADME Data Generation

	Source
<b>Molecular weight</b>	<b>292.76 g/mol</b>
<b>Log P</b>	<b>ADMET predictor</b>
<b>pKa</b>	<b>ADMET predictor</b>
<b>Fraction unbound in plasma (<math>f_{up}</math>)</b>	<b>Measured</b>
<b>Blood: plasma ratio</b>	<b>Measured</b>
<b>Hepatic intrinsic clearance (L/h)</b>	<b>Measured</b>
<b>ECCS classification</b>	<b><i>Varma et al., 2015</i></b>
<b>Renal excretion</b>	<b>GFR*Fup</b>
<b>Dermal absorption parameters: Partition coefficient and diffusivity in skin layers</b>	<b>Measured, Eurofins, <i>Ex vivo</i> skin penetration study designed according to <i>Davis et al. 2011</i> meeting OECD TG 428 and SCCS guidance</b>

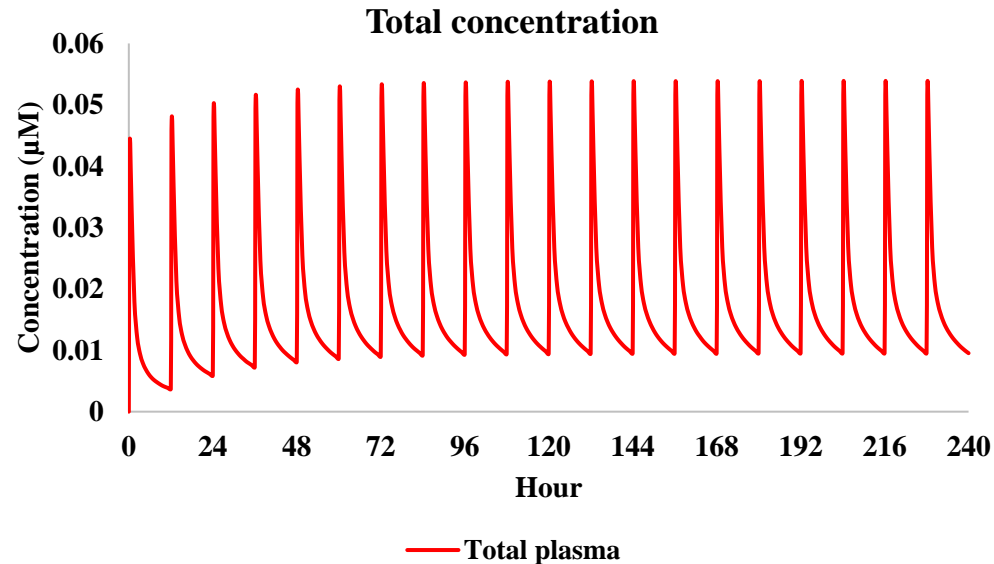
### Main observations:

- **Very low skin penetration (~1.5% over 24h)**
- **Climbazole was readily cleared in the plateable hepatocyte assay**

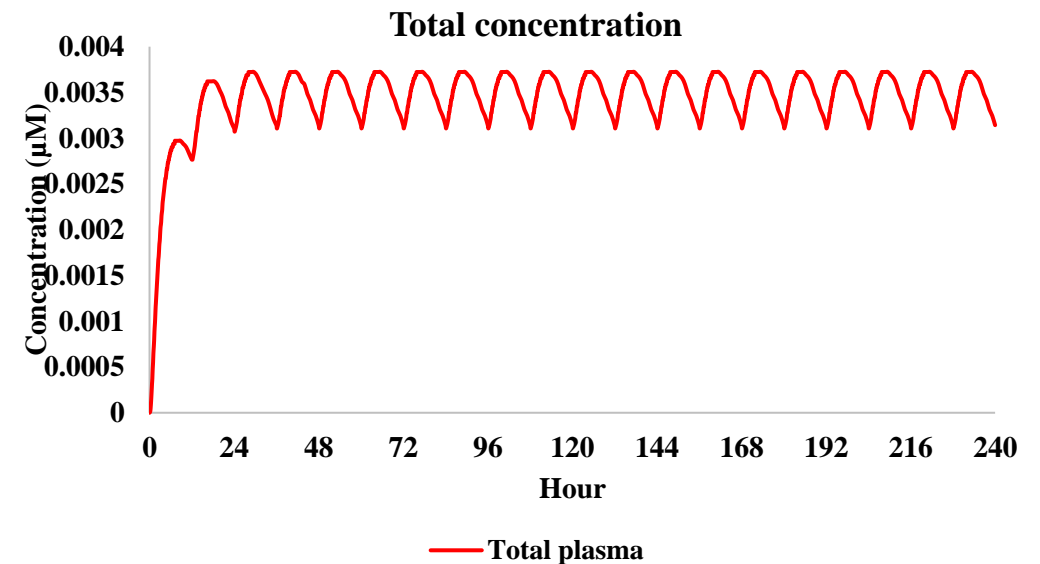
# Climbazole: Exposure Estimation

## From applied dose to internal concentration

### L1 - in silico parameters

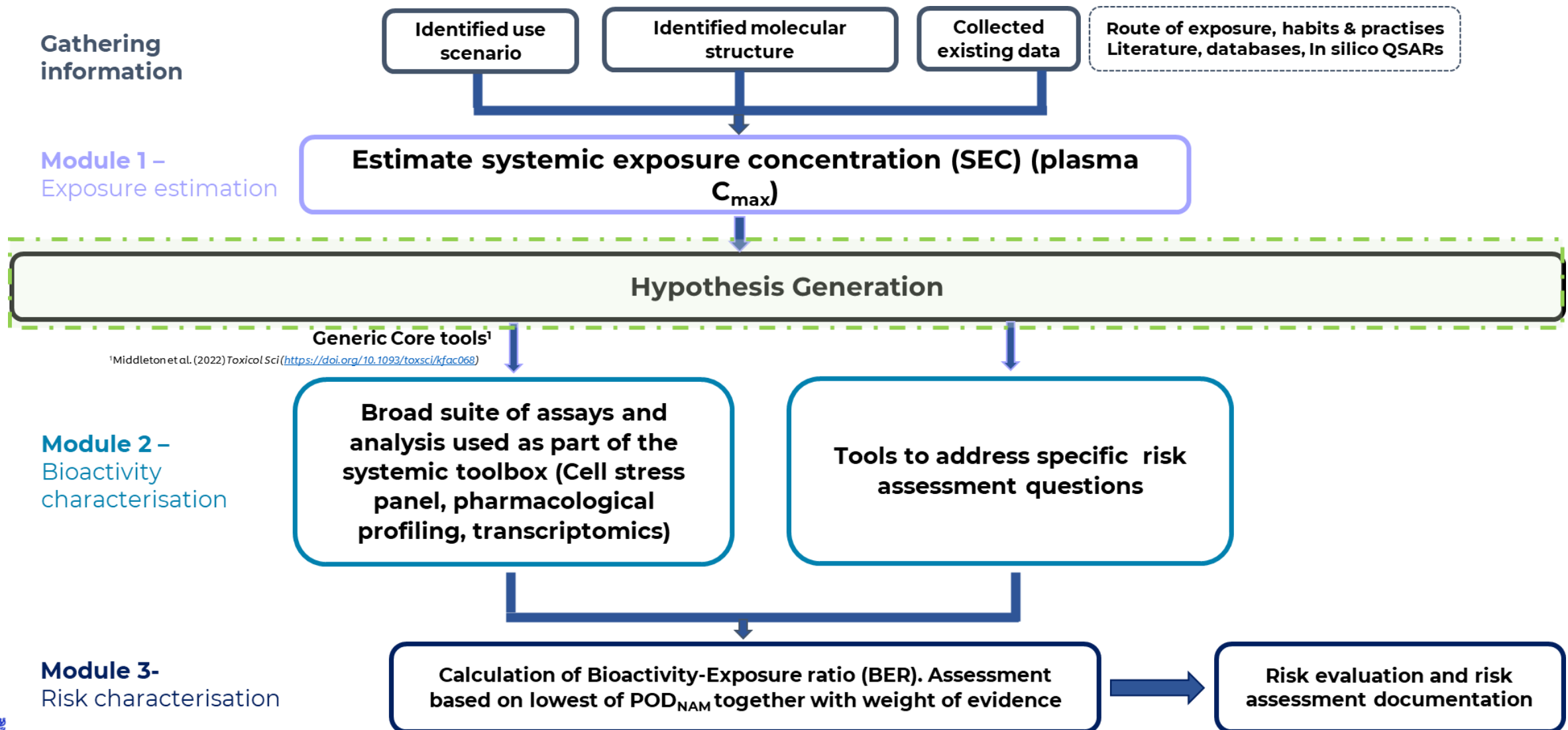


### L2 - key in vitro parameters



- Overall, upon dermal absorption only a small amount of Climbazole enters systemic circulation, after which the most likely route of elimination is liver clearance of climbazole.
- Following twice daily application of a face cream, climbazole does not appear to accumulate with the PBK model run up to 10 days.
- Refining key ADME parameters using experimental data led to a reduction in the predicted plasma  $C_{\text{max}}$

# Climbazole: Overall approach



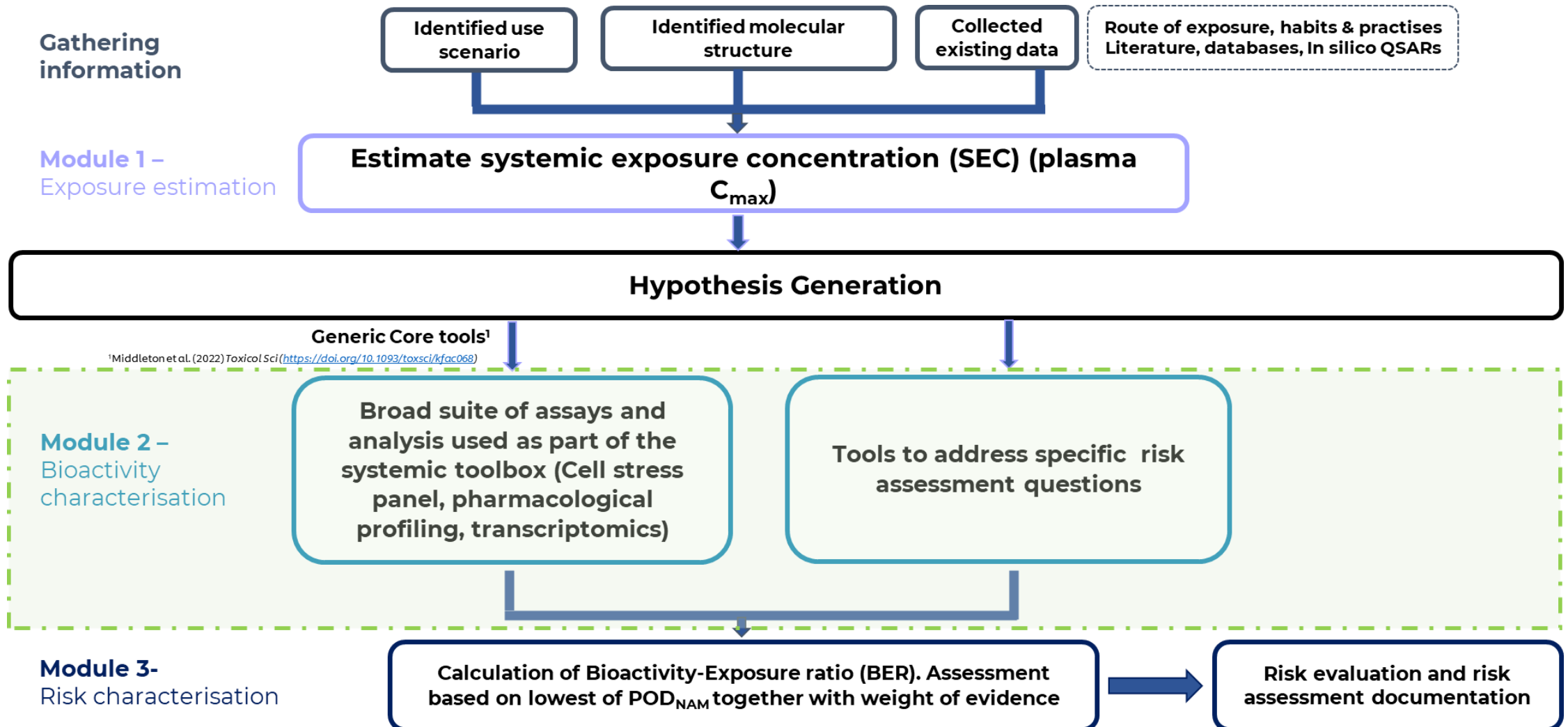


# Climbazole: Hypothesis Generation

## Hypothesis Generation

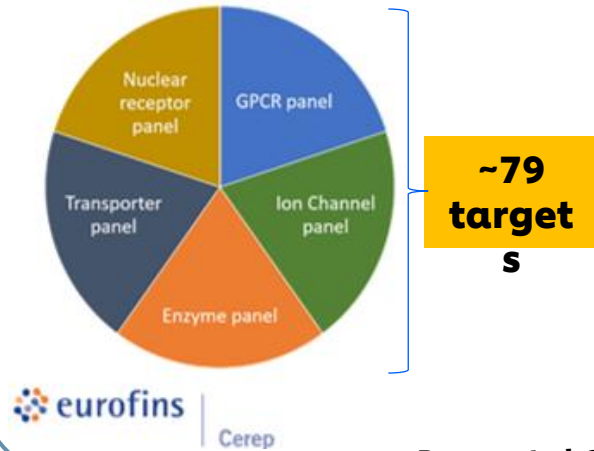
- In silico **alerts for hepatotoxicity** → covered by cell lines, no specific method so broad screening of activity?
- In silico **alerts for reprotox and adrenal gland toxicity** → Imidazole derivatives are known to inhibit ergosterol synthesis and this is the primary mechanism of action for the anti-fungal efficacy displayed by azole fungicides (other imidazole derivatives, including ketoconazole). Likely to explain the efficacy of climbazole as a preservative and therefore needs investigating.

# Climbazole: Overall approach



# Climbazole: Broad suite of bioactivity assays

## In vitro pharmacological profiling



**PERSPECTIVES**

**GUIDE TO DRUG DISCOVERY — OPINION**  
**Reducing safety-related drug attrition: the use of *in vitro* pharmacological profiling**

Jonathan Bowes, Andrew J. Brown, Jacques Hamon, Wolfgang Jorntrup, Alex Siskin, Gareth Wallace and David Wilkinson

**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 necessitate their 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 in our collaborative knowledge sharing.

During the high attrition rate in the drug discovery and development process is a primary goal of pharmaceutical research is to identify and develop drugs that are safe and effective. One of the main challenges in achieving this goal is understanding and predicting adverse effects on target tissues, cell signal transduction, off-target interactions and the safety profile of a drug candidate. This can be achieved through a range of *in vitro* assays that can be used to identify potential off-target interactions and to understand the underlying mechanisms of drug-induced toxicity. *In vitro* pharmacological profiling involves the screening of compounds against a broad range of target proteins, including receptors, ion channels, transporters and enzymes, to identify potential off-target interactions that could lead to adverse effects.

Bowes et al. 2012. *Nat Rev Drug Discov* 11(12): 909-22

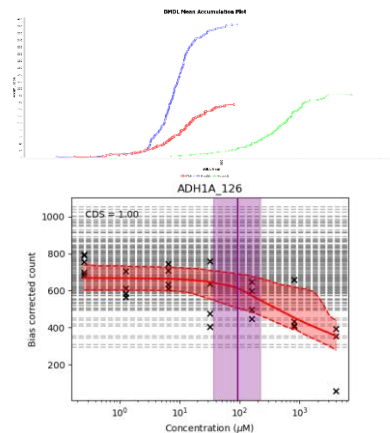
**To investigate specific biological activity with 44 key targets involved in drug attrition (Pharma) and additional targets relevant to exposure to cosmetics—now expanded to 79 targets**

**To characterize non-specific biological activity which is not mediated via a specific protein/receptor interaction**

**Transcriptomics was applied as a broad nontargeted biological screen**

## High-Throughput transcriptomics (HTTr)

- TempO-seq technology: full gene panel
- 24hr exposure
- 7 concentrations
- Various cell models (e.g. HepG2, MCF7, HepaRG)
- Dose-response analysis using BMDExpress2 and BIFROST model



Reynolds et al. 2020. *Comp Tox* 16: 100138  
 Baltazar et al. 2020. *Toxicol Sci* 176(1): 236-252

## Cell stress panel (CSP)

- 36 biomarkers covering 10 cell stress pathways
- HepG2
- 24hr exposure
- 8 concentrations
- Dose-response analysis using BIFROST model

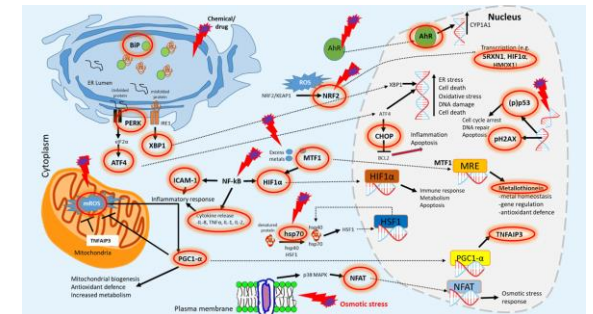


Image kindly provided by Paul Walker (Cyprotex)

Hatherell et al. 2020. *Toxicol Sci* 176(1): 11-33

# Climbazole: Results from key NAMs

## Deriving Points of Departure (PoDs)

### HTTr (HepG2, HepaRG, MCF-7)

- Effects on the transcriptome were noticed in all 3 cell lines tested.
- PoDs were calculated using changes at the gene level and changes at a pathway level
- MCF-7 cells were the most sensitive cell line with a gene level PoD of 0.094  $\mu\text{M}$  and a pathway level PoD of 12.9  $\mu\text{M}$

### Cell Stress Panel

- Climbazole was active across all but 3 measured biomarkers.
- PoDs for stress biomarkers correlated with PoDs for cell health biomarkers measured in the same assay and were all around the highest tested dose. Indicating that all effects measured are potentially indicative of the start of cytotoxicity.
- Global PoD calculated to be 12  $\mu\text{M}$

### In vitro Pharmacological profiling

- Screening performed at 10  $\mu\text{M}$
- ~79 targets compiled by Cosmetics Europe Safety pharmacology WG
- 3 hits: Aromatase, PXR and SLC6A3
  - Dopamine transporter, SLC6A3 **IC50 calculated to be 0.073  $\mu\text{M}$**
  - **Aromatase IC50 calculated to be 0.091  $\mu\text{M}$**
  - Pregnancy X Receptor IC50 calculated to be 4.9  $\mu\text{M}$

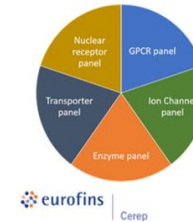
# Climbazole: Tools to address specific questions



## Tools to address specific risk assessment questions



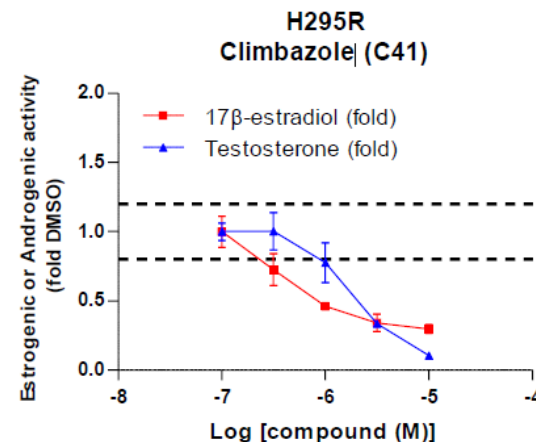
Structural alerts for reprotoxicity and structural similarity to compound known to interfere with sterol biosynthesis



Aromatase inhibition known to affect natural biosynthesis of hormones  
Climbazole IC50 : 0.091  
Prochloraz IC50: 0.021

**Is Climbazole likely to interfere with hormone synthesis in vivo; and if so at what concentration?**

**Generation of data in OECD 456, H295R steroidogenesis assay coupled to ER $\alpha$  and AR-CALUX**

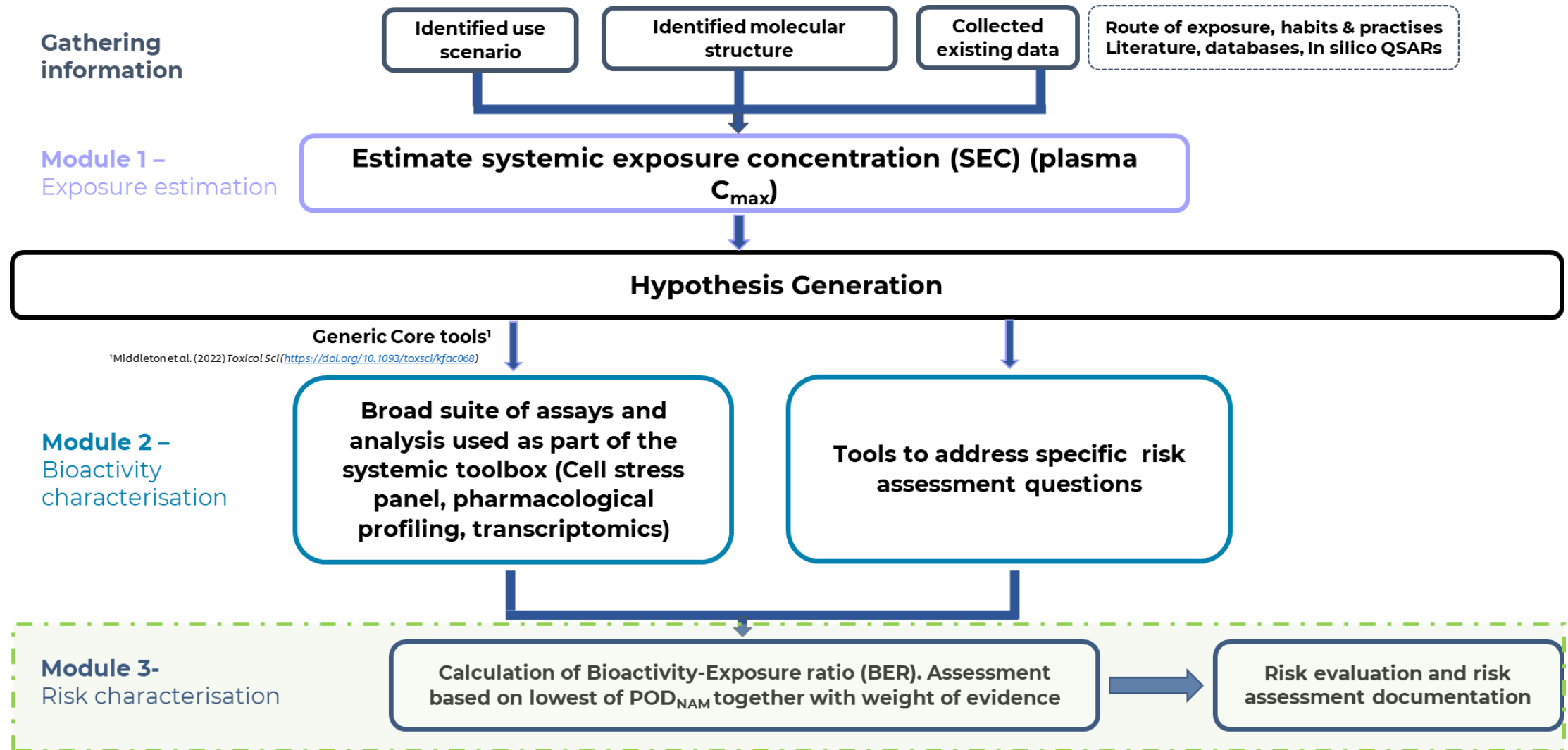


**Climbazole demonstrated preferential interference with estradiol production, fits with initial alerts.**

**Climbazole LOEC = 0.3  $\mu$ M (ER $\alpha$  CALUX)**

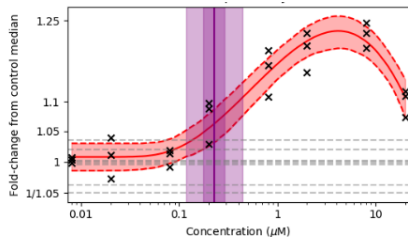
**Prochloraz LOEC = 0.0001 – 0.1  $\mu$ M (Hecker et al., 2018)**

# Climbazole: Overall approach



# Climbazole: Calculation of the Bioactivity Exposure Ratio (BER)

**Point of departure (POD) derived from concentration-response data**



**Systemic toolbox of assays (NAMs) which cover a broad biological space – measurements of bioactivity**

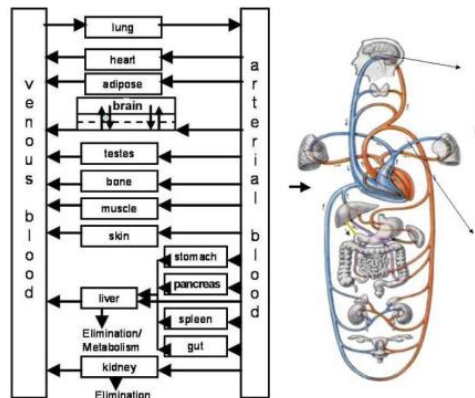


**Calculation of Bioactivity exposure ratio (BER)**

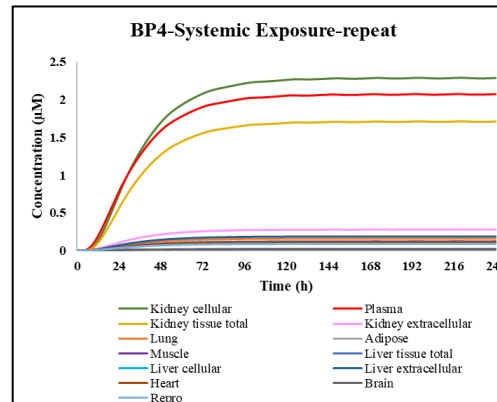
**Exposure models (PBK, free/total concentration)**



**Skin pen**



**Exposure estimation: Plasma C<sub>max</sub>, organ distribution, AUC**



**The BER is defined as the ratio between the POD and the relevant exposure metric**

## Climbazole: Calculation of the Bioactivity Exposure Ratio (BER)

<b>NAM</b>	<b>PoD (µM)</b>	<b>BER (using L1 C<sub>max</sub>)</b>	<b>BER (using L2 C<sub>max</sub>)</b>
<b>In Vitro Pharmacological Profiling</b>	<b>0.073</b>	<b>1.4</b>	<b>19.7</b>
<b>Cell Stress Panel</b>	<b>12</b>	<b>222.2</b>	<b>3243.2</b>
<b>BIFROST HTTr MCF-7</b>	<b>0.094</b>	<b>1.7</b>	<b>25.4</b>
<b>BIFROST HTTr HepG2</b>	<b>0.72</b>	<b>13.3</b>	<b>194.6</b>
<b>BIFROST HTTr HepaRG</b>	<b>0.34</b>	<b>6.3</b>	<b>91.9</b>
<b>BMD Pathway HTTr MCF-7</b>	<b>12.9</b>	<b>238.9</b>	<b>3486.5</b>
<b>BMD Pathway HTTr HepG2</b>	<b>48.4</b>	<b>896.3</b>	<b>13081.1</b>
<b>BMD Pathway HTTr HepaRG</b>	<b>48.1</b>	<b>890.7</b>	<b>13000.0</b>
<b>H295R ER-CALUX LOEC</b>	<b>0.3</b>	<b>5.6</b>	<b>81.1</b>
<b>H295R AR-CALUX LOEC</b>	<b>1.0</b>	<b>18.5</b>	<b>270.3</b>



# Climbazole: Risk assessment conclusion

## Qualitative assessment of uncertainties

Area	Level of certainty (rationale)	Is value likely to be an over- or under-estimate (rationale)	Impact on risk assessment decision
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## Areas

- **Consumer exposure (applied dose)**
- **Identification of metabolites**
- **Consumer exposure (Internal dose)**
- **Range of biomarkers assessed**
- **Use of short-term tests *in vitro* to inform about risks of long-term human exposure**
- **Point of departure selection**

Similar approach to OECD (2021): IATA for Phenoxyethanol

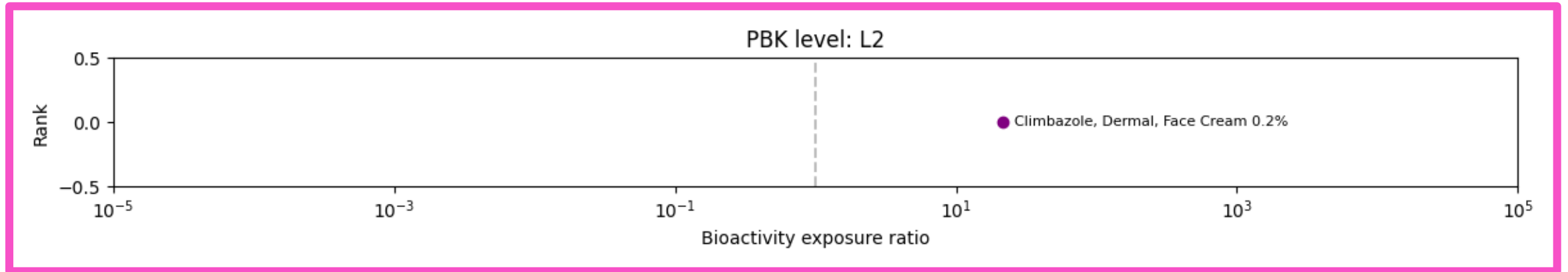
# Climbazole: Risk assessment conclusion

## Qualitative assessment of uncertainties- an example

Area	Level of certainty (rationale)	Is value likely to be an over- or under-estimate (rationale)	Impact on risk assessment decision
<b>Range of biomarkers assessed</b>	<b>Moderate</b> There is increasing evidence that $POD_{NAM}$ obtained from the core NAMs, IPP, CSP and HTTr are protective for a range of chemicals (Middleton <i>et al.</i> , 2022) and previous case studies (Baltazar <i>et al.</i> , 2020, OECD phenoxyethanol). The hypothesis and exposure driven approach led to the inclusion of an additional NAM to investigate the steroidogenic activity and benchmark the potency of the response.	<b>Climbazole showed potential for specific activity through the structural alerts flagged at the in silico stage and the specificity of some of the bioactivity results. This was covered in the NAMs used and a PoD derived. Broad spectrum NAMs showed overall high activity for climbazole in the test systems with leading PoDs derived from gene level changes in the HTTr, which is likely conservative given the low number of genes changing at low concentrations.</b>	There are remaining uncertainties regarding the protectiveness of the tools utilised for a broader range of chemistries. <b>Confidence could be increased by assessing how protective the range of biomarkers are for many more compounds</b> and whether different biomarkers are needed to ensure the <i>in vitro</i> PoD is protective compared with the <i>in vivo</i> PoD.

# Climbazole: Risk assessment conclusion

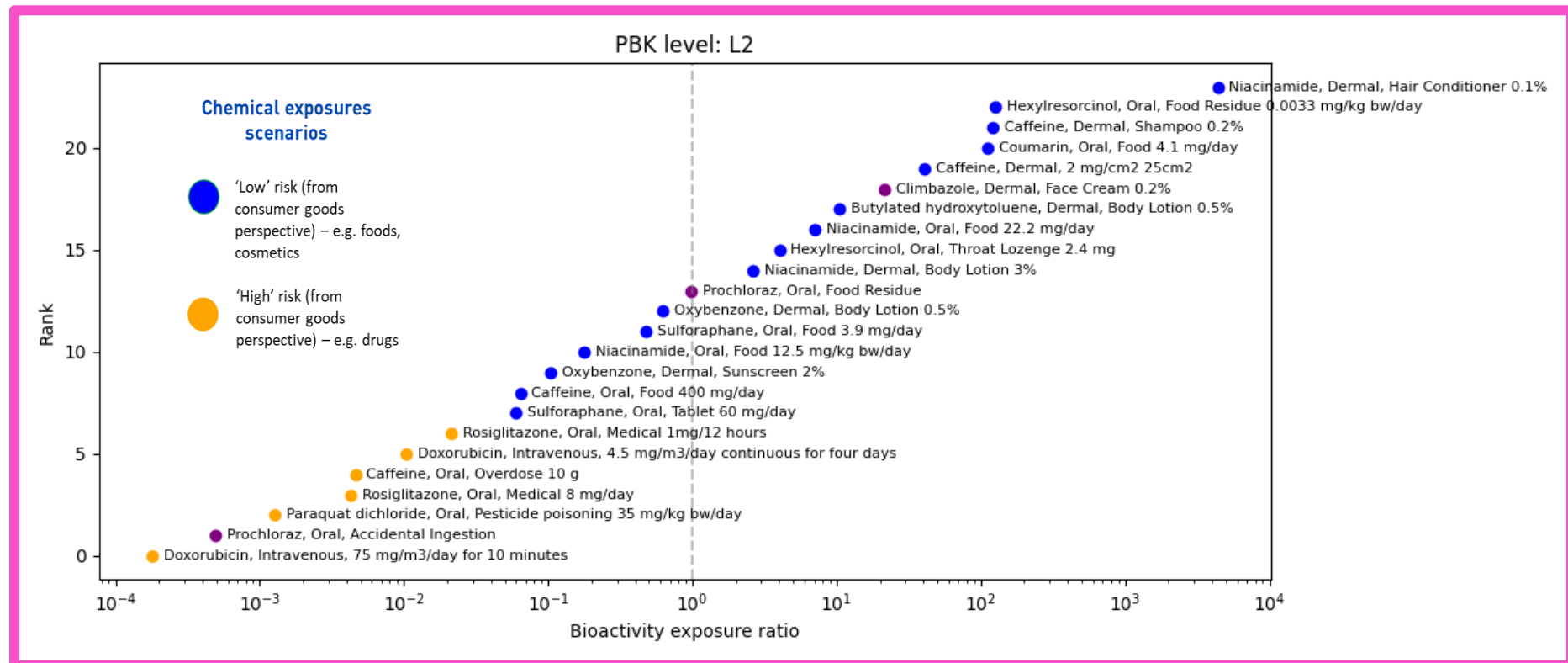
## Interpreting the BER using the lowest $PoD_{NAM}$ and the deterministic BER



# Climbazole: Risk assessment conclusion

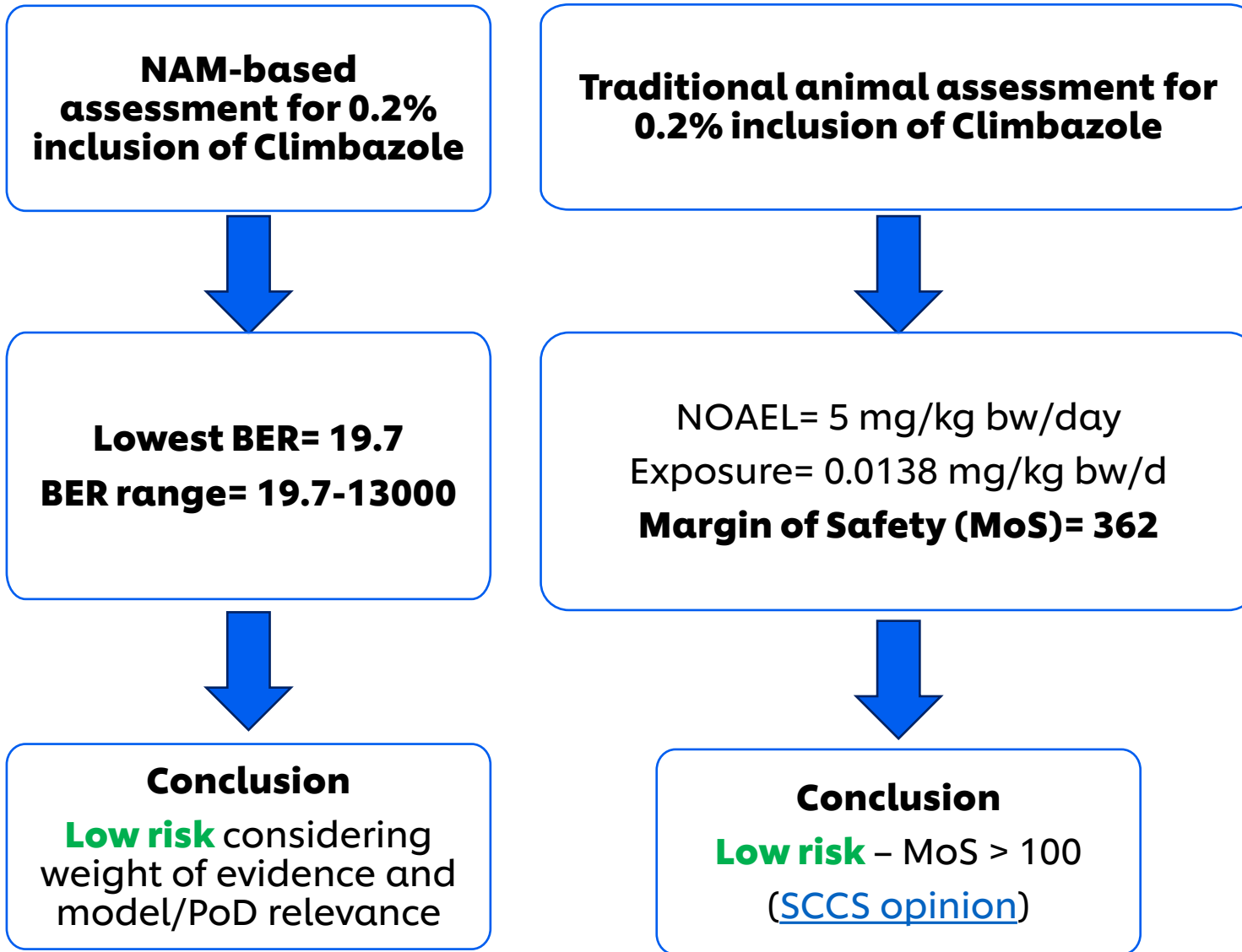
## Interpreting the BER using the lowest $PoD_{NAM}$ and the deterministic BER

**What if the same approach was applied to other chemicals with varying risk classifications?**



**Note: Low risk is different than low toxicity; it is all about integrating exposure.**

## Conclusions & reflections



### NAM-based risk assessments are in generally more conservative than traditional approaches

- Middleton et al. (2022) *Toxicol Sci* (<https://doi.org/10.1093/toxsci/kfac068>)
- Reardon A et al., 2023 <https://doi.org/10.3389/ftox.2023.1194895>
- Zobl et al., 2023 <http://dx.doi.org/10.14573/altex.2309081>
- Paul-Friedman K et al., 2020: <https://doi.org/10.1093%2Ftoxsci%2Fkfkz201>
- Baltazar MT et al., 2020: <http://dx.doi.org/10.1093/toxsci/kfaa048>
- Ebmeyer et al., 2024: <https://doi.org/10.3389/fphar.2024.1345992>
- Cable et al., 2025: <https://doi.org/10.1093/toxsci/kfae159>

## Conclusions and Reflections

- Showcased a range of in silico and in vitro NAMs that can be used for safety decision making for systemic toxicity
- The method is exposure-led and follows a tiered approach for both exposure and bioactivity
- Bespoke NAMs can be added to the NGRA to fill gaps identified along the process
- 'Early tier' in vitro screening tools show promise for use in a protective rather than predictive capacity.
- NGRA requires a mindset shift and a multidisciplinary team

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