

Advancing Next Generation Risk Assessment through AI-Augmented Weight-of-Evidence Frameworks

ITTP Summer School

July 8th, 2026

SERS
Safety, Environmental
& Regulatory Science



Unilever Policy & Approach

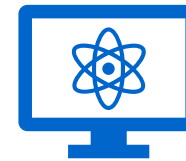
Safe & Sustainable Products without Animal Testing

Unilever

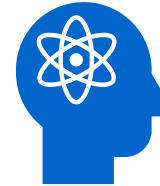
What we believe

- **Every Unilever product must be safe for people and our environment**
- **Animal testing is not needed to assess ingredient & product safety** – there are a wide range of non-animal alternatives grounded in modern science and new technology

How we do it



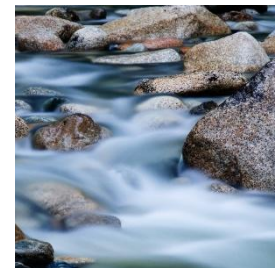
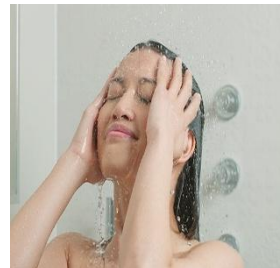
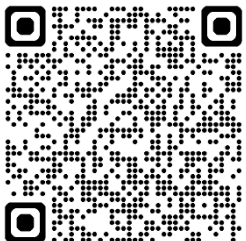
40+ years of non-animal safety science



70+ collaborations

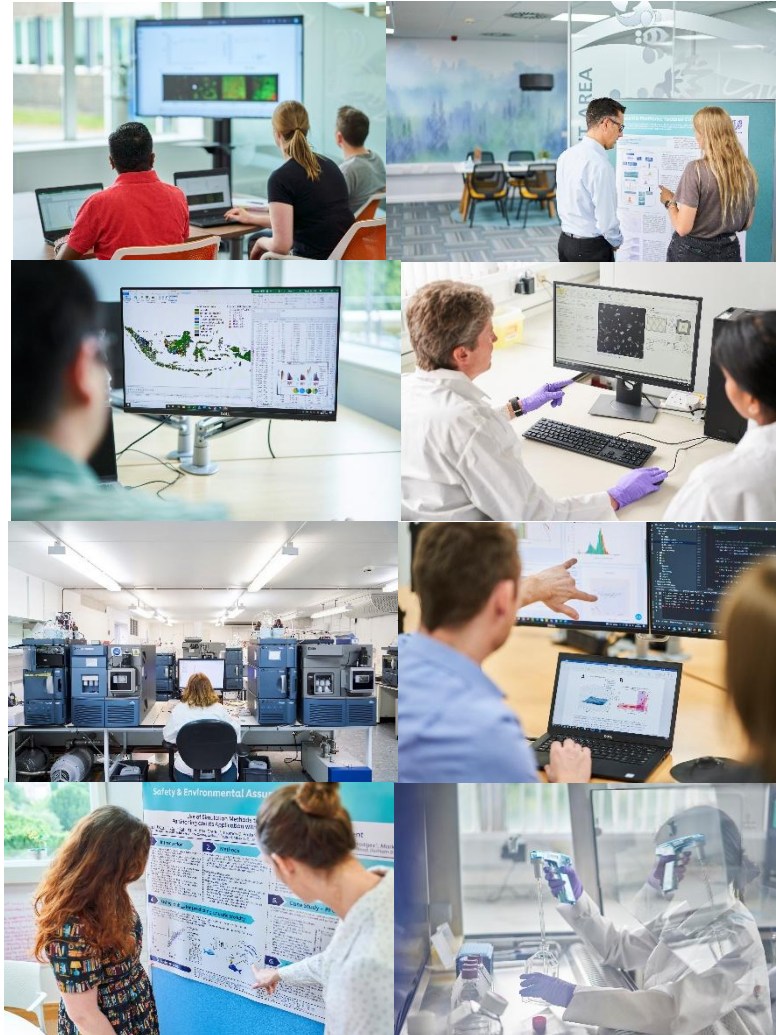


600+ publications



SERS Expertise

Unilever



SERS
Safety, Environmental
& Regulatory Science

SERS is a diverse, multi-disciplinary team of ~180 scientists covering:

- Cell & Molecular Biology
- Chemistry
- Computational Modelling
- Environmental Safety
- Environmental Sustainability
- Exposure Science
- Informatics & Data Science
- Mathematics & Statistics
- Microbiology
- Process Safety
- Regulatory Science (chemical & food safety)
- Toxicology

20+ Nationalities
15+ Languages
9 Countries

- Deploy expertise on higher risk business projects
- Collaborate with leading external research teams to develop & apply new scientific capability
- Leverage science & global networks for consumer trust & freedom to operate

Safety Risk Assessments

- Consumers, Workers, Environment

Life Cycle Assessments

- Environmental Impacts

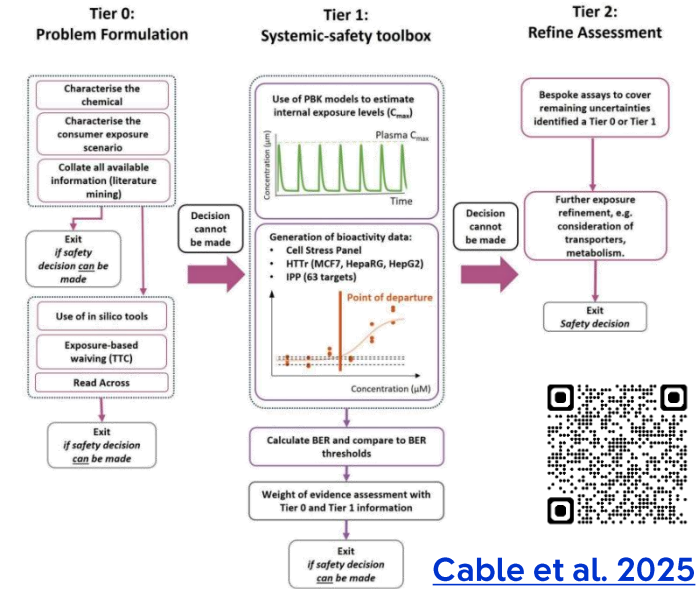
Product Compliance

- Regulatory Data & Dossiers

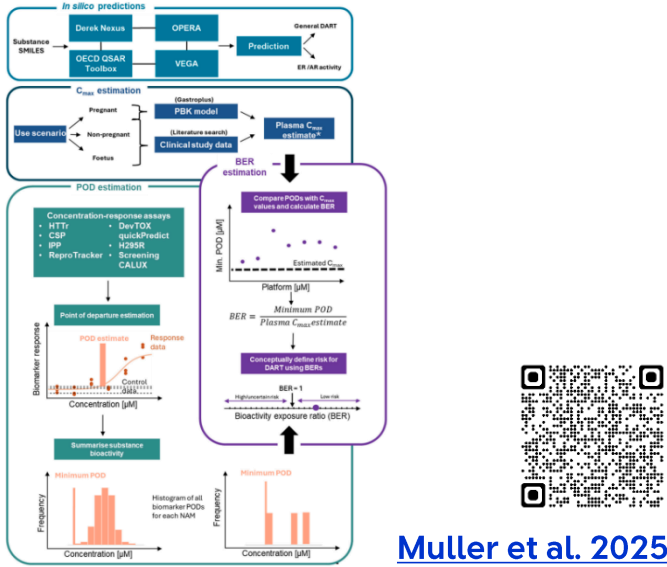
Unilever Next Generation Risk Assessment frameworks



Systemic



Developmental & Reproductive



Ongoing Partnerships

EPA and Unilever Announce Major Research Collaboration to Advance Non-animal Approaches for Chemical Risk Assessment

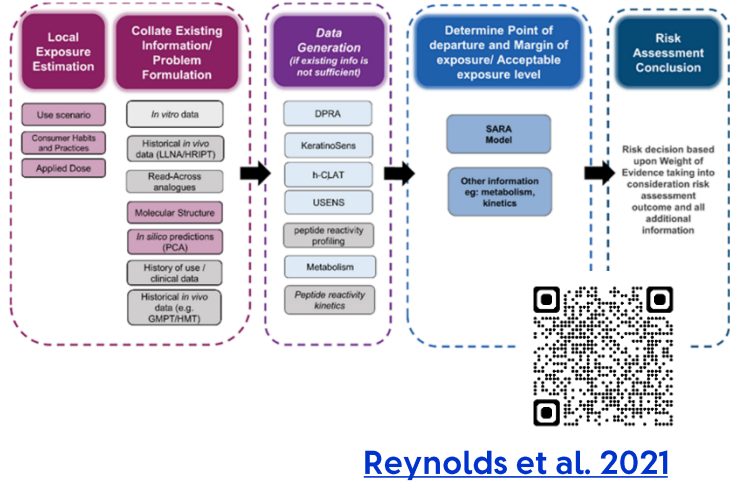
August 19, 2021

NICEATM News - 2021 Issue 25: May 27

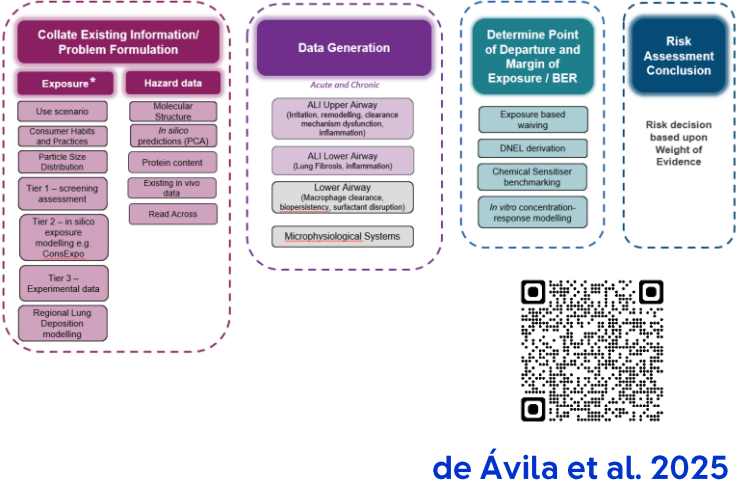
In this Newsletter:

- NICEATM to Collaborate with Unilever on Development of Predictive Model for Skin Sensitization
- NICEATM to Collaborate with Unilever on Development of Predictive Model for Skin Sensitization

Skin Sensitisation



Inhalation



sers.unilever.com

Safe and Sustainable Products without Animal Testing

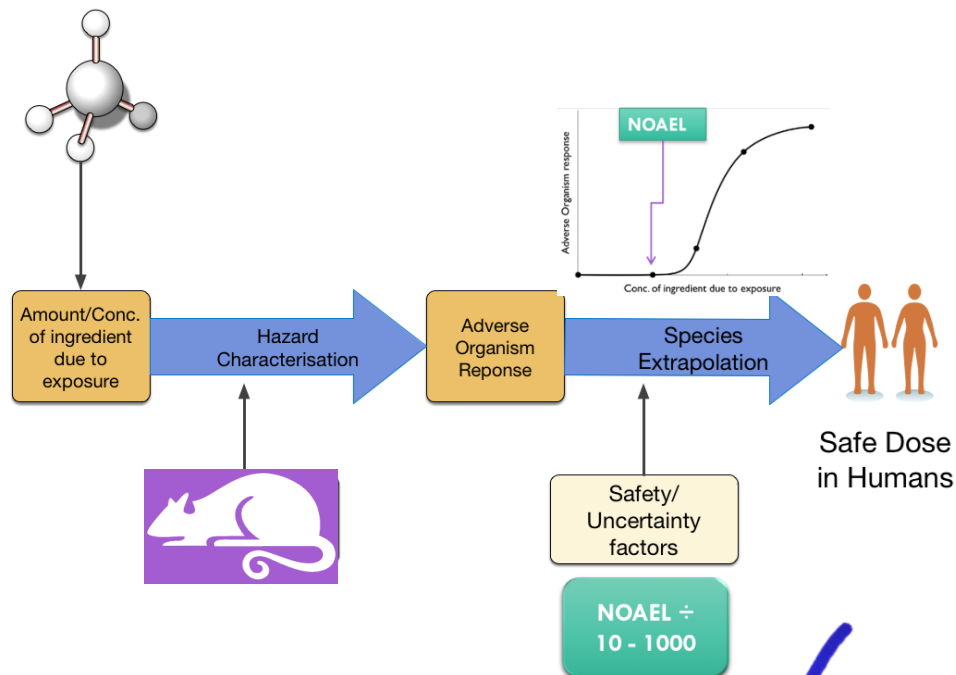
Unilever



How can we ensure that our products are safe?

*Q: Is ingredient **X** safe to use in product **Y** at **Z**%?*

'Traditional' Risk Assessment

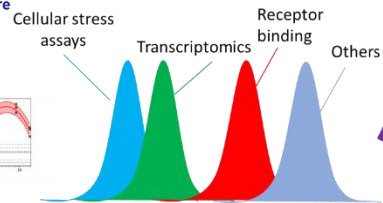
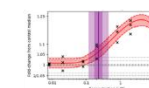


'Next Generation' Risk Assessment

Hazard identification and characterisation of ingredients



Point of departure derived from concentration-response data



Risk Assessment

Calculation of Bioactivity Exposure Ratio (BER)



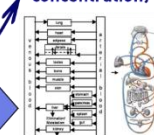
The BER/MoE is defined as the ratio of the PoD and the relevant exposure estimate

Consumer Exposure characterisation



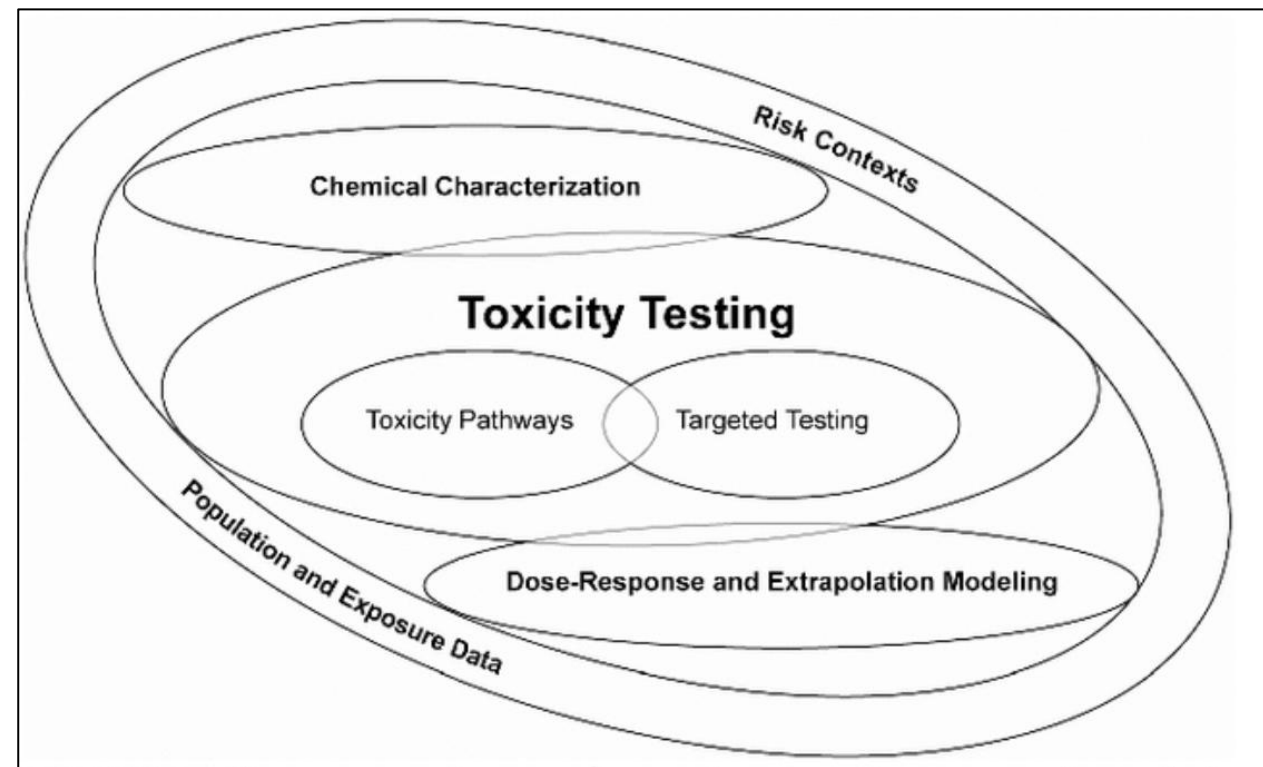
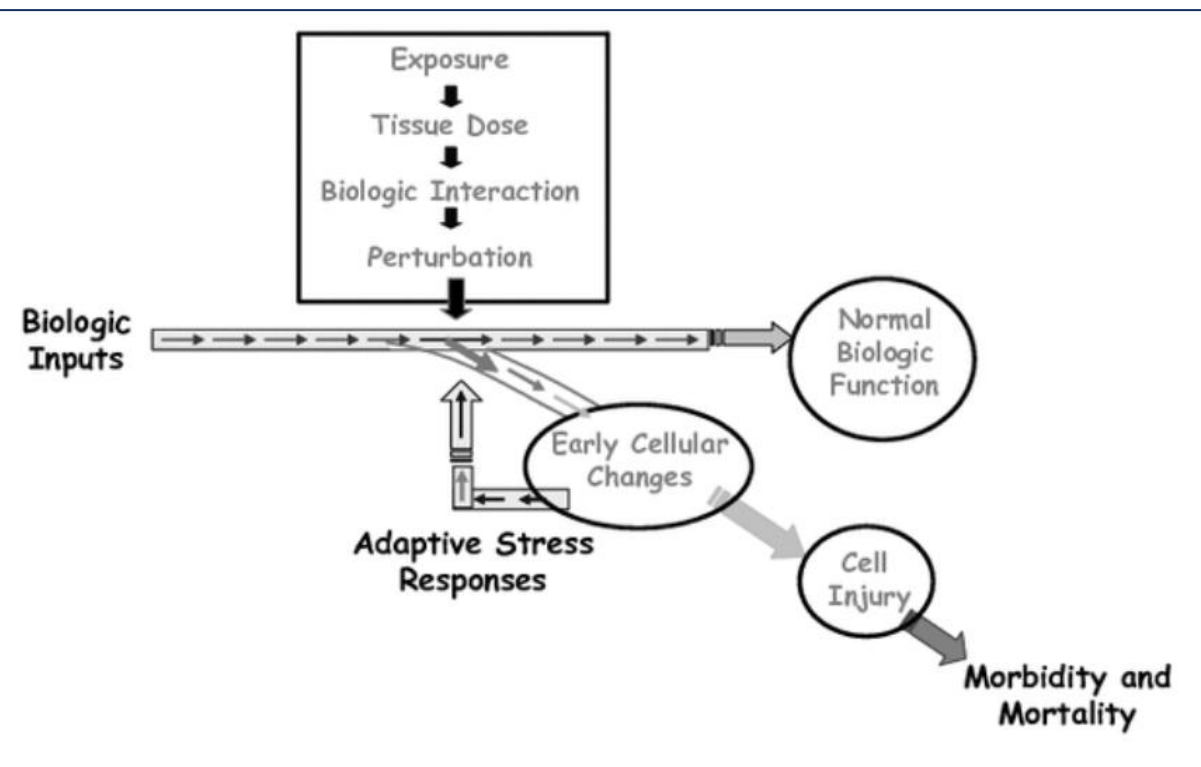
Skin pen

Exposure models (PBK, free/total concentration)



Exposure estimation: Plasma C_{max}

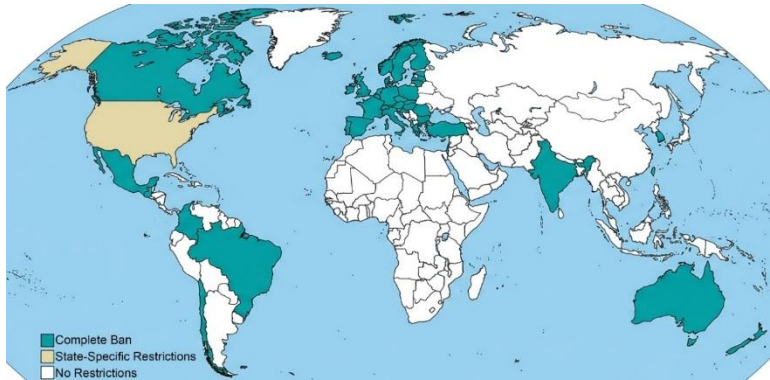




2007



The Need for Implementation of Non-animal Safety Assessments



Rusyn et al, 2026 (submitted)

One Hundred Fourteenth Congress
of the
United States of America
AT THE SECOND SESSION
Begun and held at the City of Washington on Monday,
the fourth day of January, two thousand and sixteen

An Act
To modernize the Toxic Substances Control Act, and for other purposes.
Be it enacted by the Senate and House of Representatives of
the United States of America in Congress assembled,

SECTION 1. SHORT TITLE; TABLE OF CONTENTS.

(a) SHORT TITLE.—This Act may be cited as the “Frank R. Lautenberg Chemical Safety for the 21st Century Act”.

(b) TABLE OF CONTENTS.—The table of contents of this Act is as follows:

TABLE OF CONTENTS

SECTION 1. SHORT TITLE; TABLE OF CONTENTS.

SECTION 2. FINDINGS, POLICY, AND INTENT.

SECTION 3. DEFINITIONS.

SECTION 4. AMENDMENTS TO THE TOXIC SUBSTANCES CONTROL ACT (15 U.S.C. 2601).

SECTION 5. REGULATORY REFORMS.

SECTION 6. RURAL HEALTHCARE CONNECTIVITY.

SECTION 7. RURAL HEALTHCARE CONNECTIVITY.

SECTION 8. RURAL HEALTHCARE CONNECTIVITY.

SECTION 9. RURAL HEALTHCARE CONNECTIVITY.

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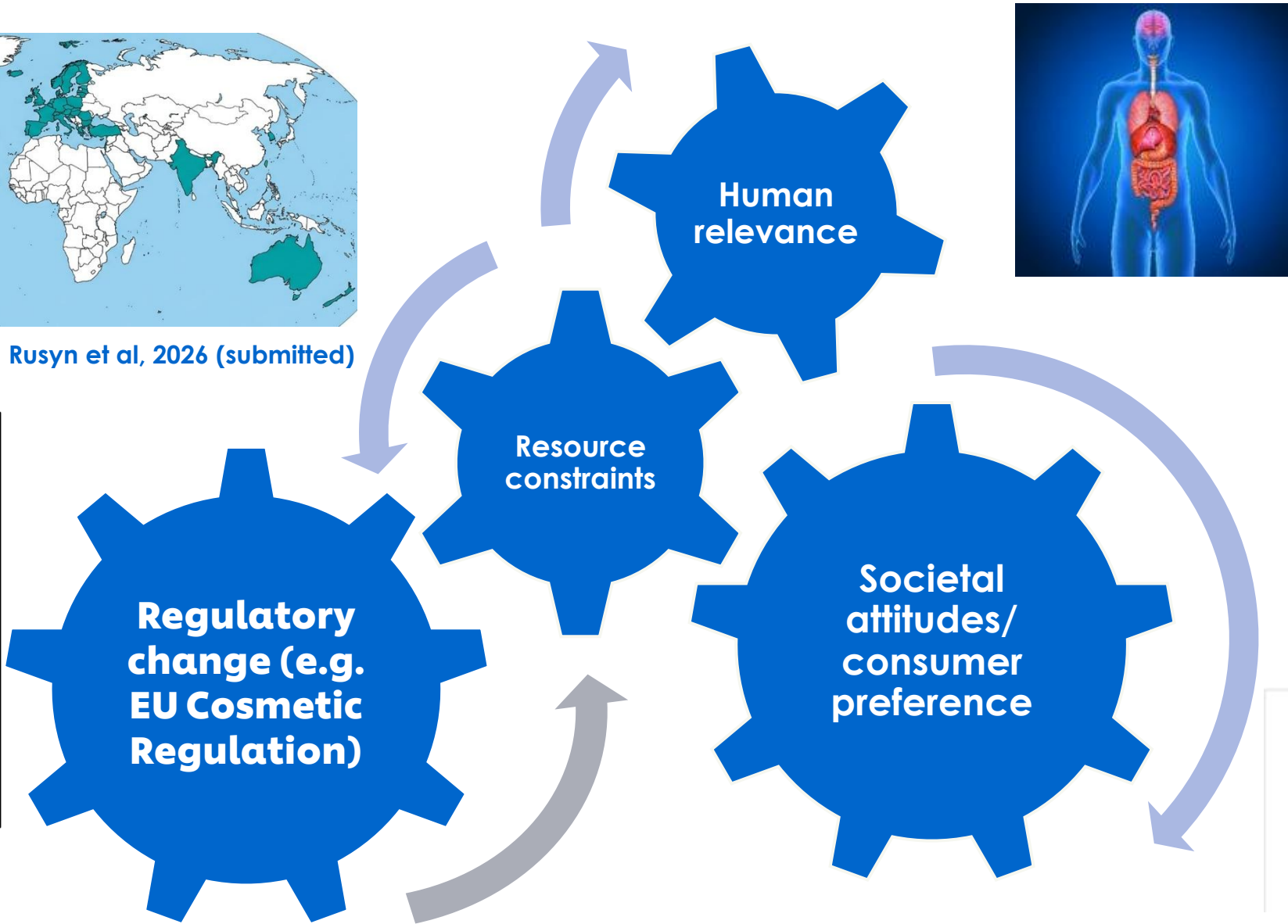
SECTION 96. RURAL HEALTHCARE CONNECTIVITY.

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SECTION 100. RURAL HEALTHCARE CONNECTIVITY.



Next Generation Risk Assessment (NGRA)

Existing Data

ECOTOX Knowledgebase

A→O→P-Wiki

Knowledge Bases



Scientific literature



CompTox Chemicals Dashboard
Search 1,200,063 Chemicals

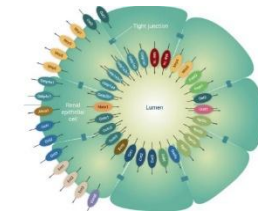
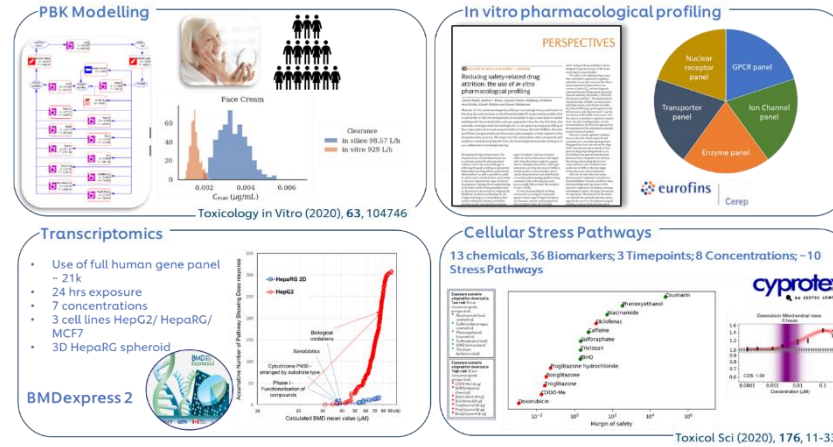


External Data

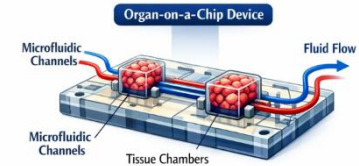


Regulatory Guidance

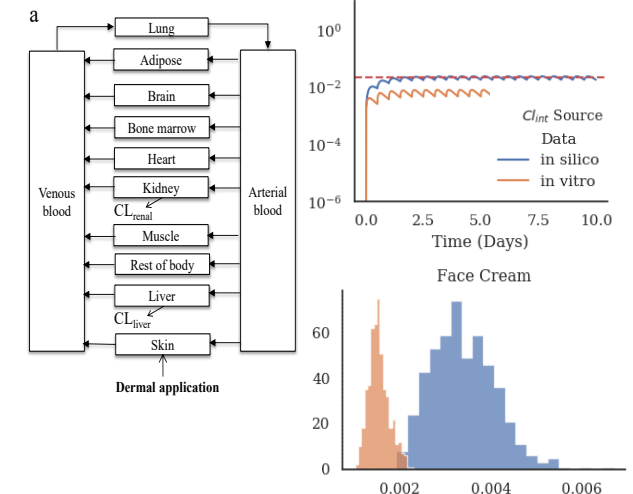
New (NAM) in vitro data



Credit:
Newcells aProxi
mate™ platform



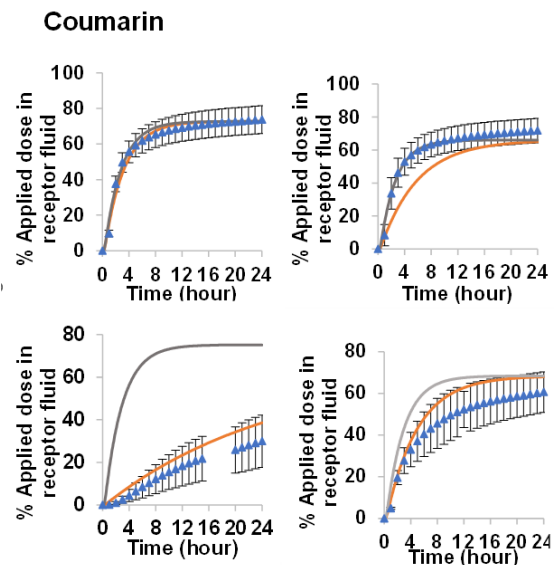
Exposure modelling



Computational Modelling in NGRA

Exposure modelling

Dermal

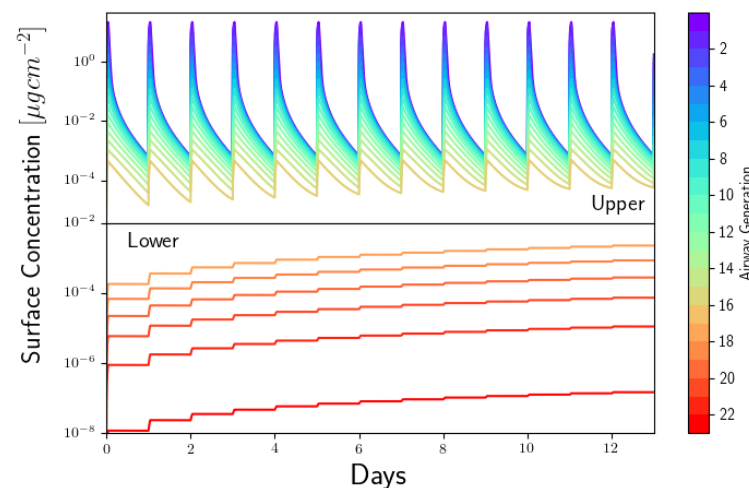


- Diffusion through skin
- Partitioning from complex mixtures into stratum corneum

SERS

Safety, Environmental
& Regulatory Science

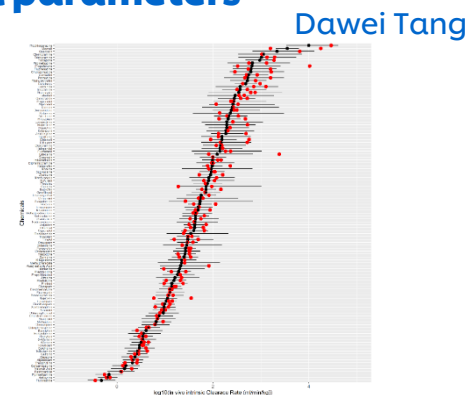
Respiratory



- Particle deposition modelling
- Clearance mechanism modelling; mucous transport, trans-epithelial diffusion

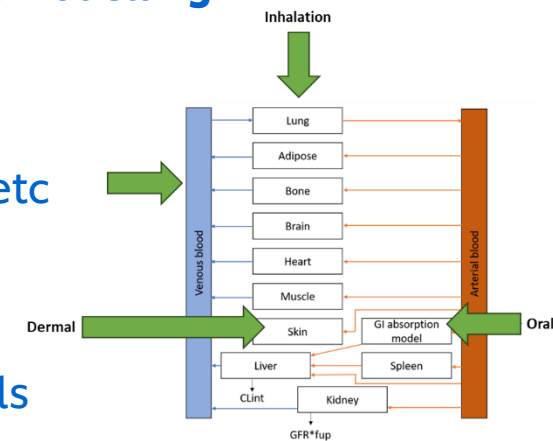
Estimation of key model parameters

- Estimation of e.g., intrinsic clearance.
- Hierarchical Bayesian statistical model



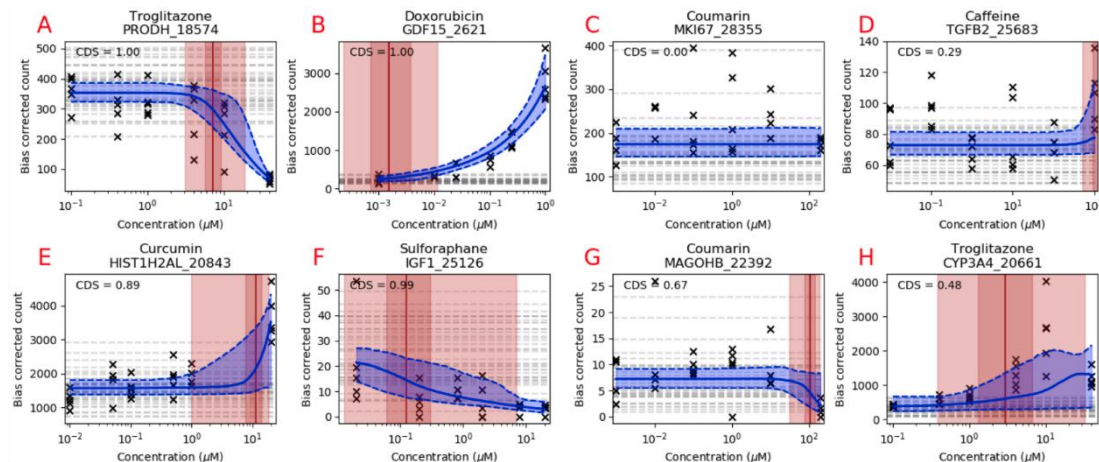
Systemic exposure modelling

- Applicable to humans, fish etc
- ODE-based models
- Bespoke vs generic models



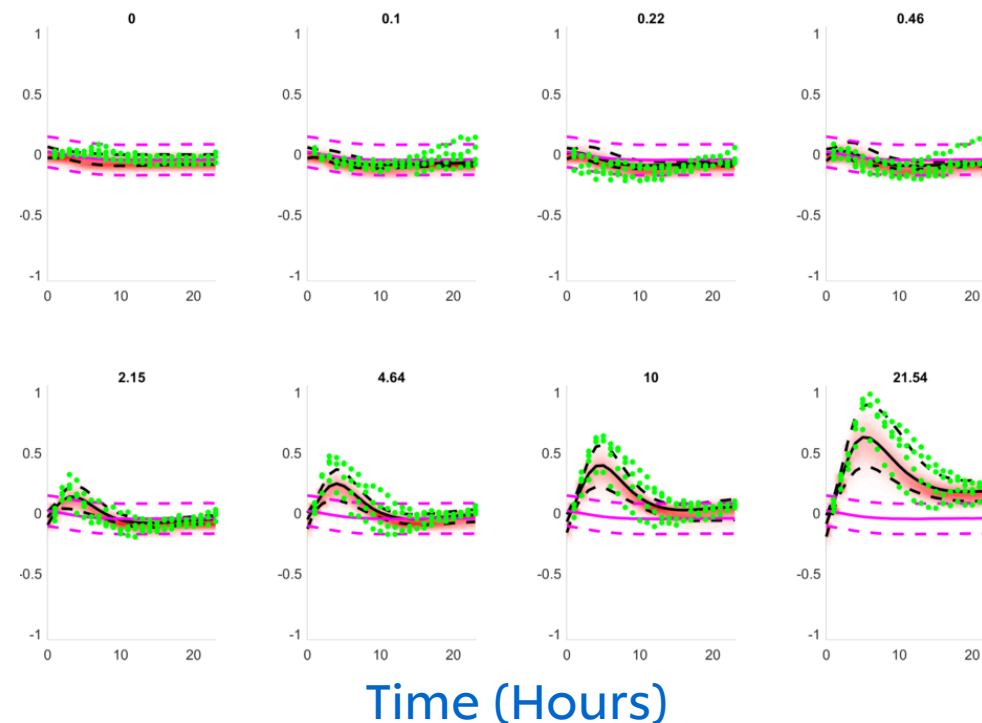
Computational Modelling in NGRA

Hazard modelling



- Modelling of concentration response data
- Estimation of point of departure (PODs)
- Various approaches being used:
 - Maximum likelihood/parametric models
 - Bayesian non-parametric models

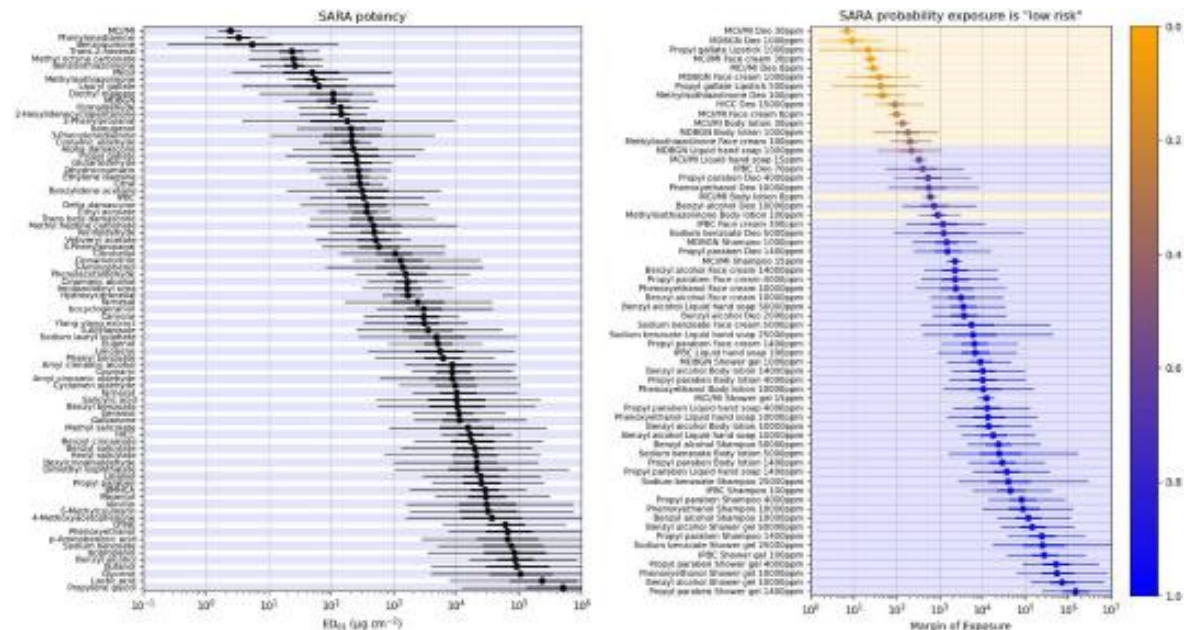
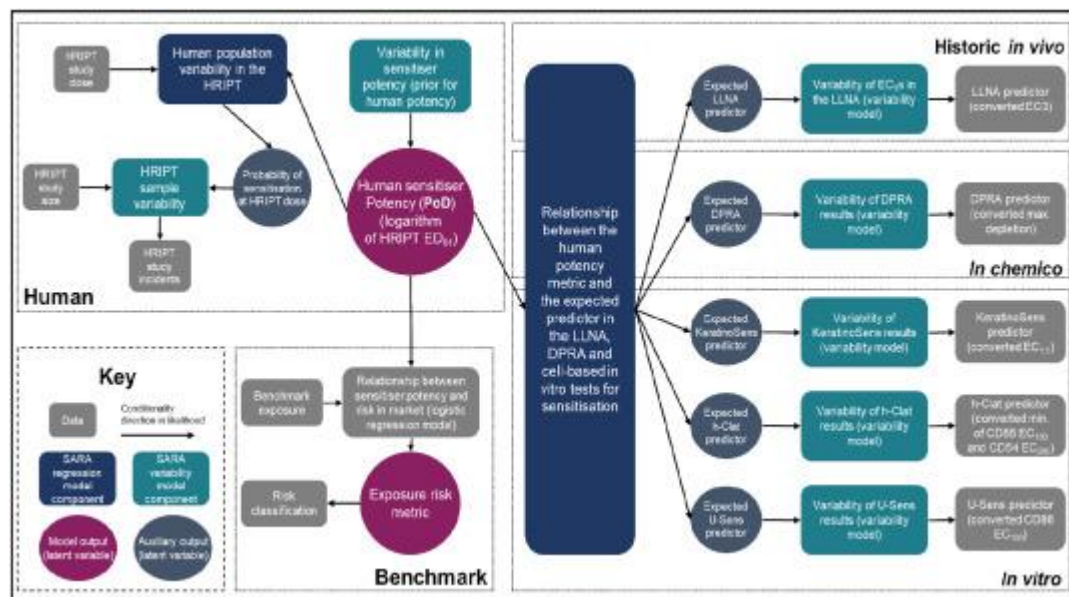
NRF2 response



- Time series analysis of biological response
- Using mechanistic ODE models
- Gaussian process state space models

Computational Modelling in NGRA

Quantifying risk

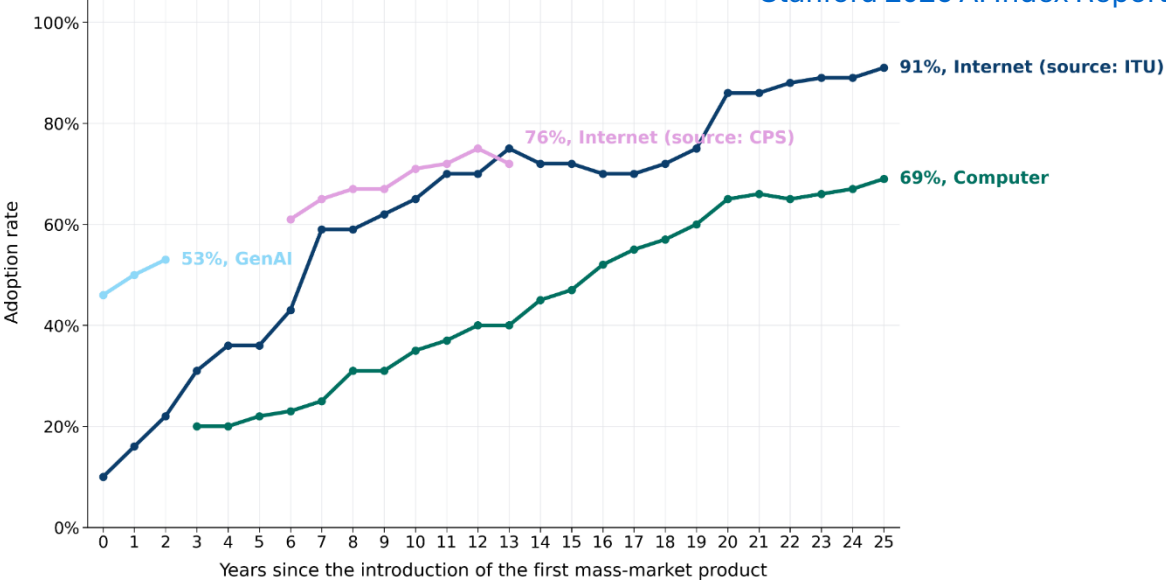


- Example: Skin Allergy Risk Assessment (SARA) model
- Hierarchical Bayesian statistical model
- Uses training data from benchmark exposure scenarios and safety decisions.
- Variant of the model being developed in collaboration with US federal agency (NICEATM)



AI in Safety Critical Systems

Speed of AI adoption by technology
Stanford 2026 AI Index Report



FDA NEWS RELEASE

FDA Expands Artificial Intelligence Capabilities with Agentic AI Deployment

Considerations for the Use of Artificial Intelligence to Support Regulatory Decision-Making for Drug and Biological Products Guidance for Industry and Other Interested Parties

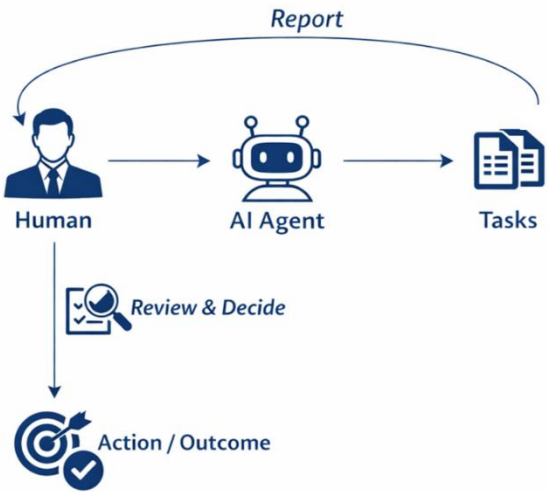
DRAFT GUIDANCE

This guidance document is being distributed for comment purposes only. Comments and suggestions regarding this draft document should be submitted within 90 days of publication in the Federal Register of the notice announcing the availability of the draft guidance. Submit electronic comments to <https://www.regulations.gov>. Submit written comments to the Division of Regulatory Management Staff (FDA-305), Food and Drug Administration, 5630 Fishers Lane, Rm. 1061, Rockville, MD 20852. All comments should be identified with the docket number listed in the notice of availability that publishes in the Federal Register.

For questions regarding this draft document, contact (CDER) Tala Fakhouri, 301-437-7407; (CDER) Office of Communication, Outreach and Development, 800-437-4709 or 240-402-8010; or (CDER) Digital Health Center of Excellence, digitalhealth@fda.hhs.gov.

U.S. Department of Health and Human Services
Food and Drug Administration
Center for Drug Evaluation and Research (CDER)
Center for Biologics Evaluation and Research (CBER)
Center for Device and Radiological Health (CDRH)
Center for Veterinary Medicine (CVM)
Office of Combination Products (OCP)
Office of Inspection and Investigations (OI)

January 2025
Artificial Intelligence



Moving towards AI-augmented safety frameworks

Traditional Toxicology =
Empirical science focused on
observations from animal
studies



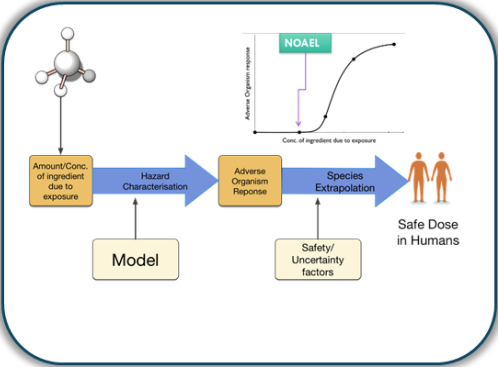
**Increased focus on the use of exposure
science and understanding of human
biology (NAMs, PBK, DAs, IATAs,
protection of human health, AOPs...)**



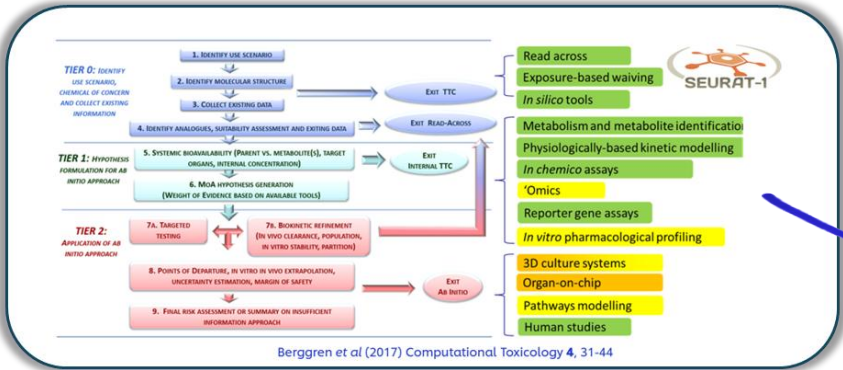
**Augmentation of Next Generation Risk
Assessment using AI-based technologies
and approaches**



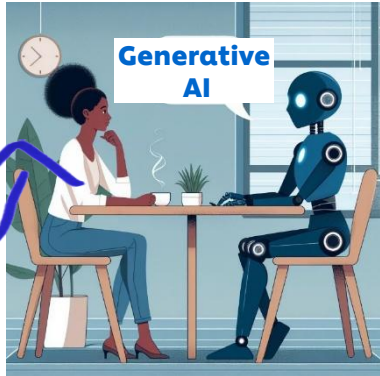
'Traditional' Risk Assessment



Next Generation Risk Assessment (NGRA)

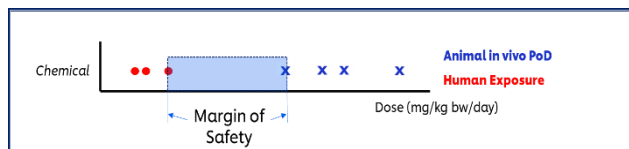


AI-augmented NGRA



Unilever

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

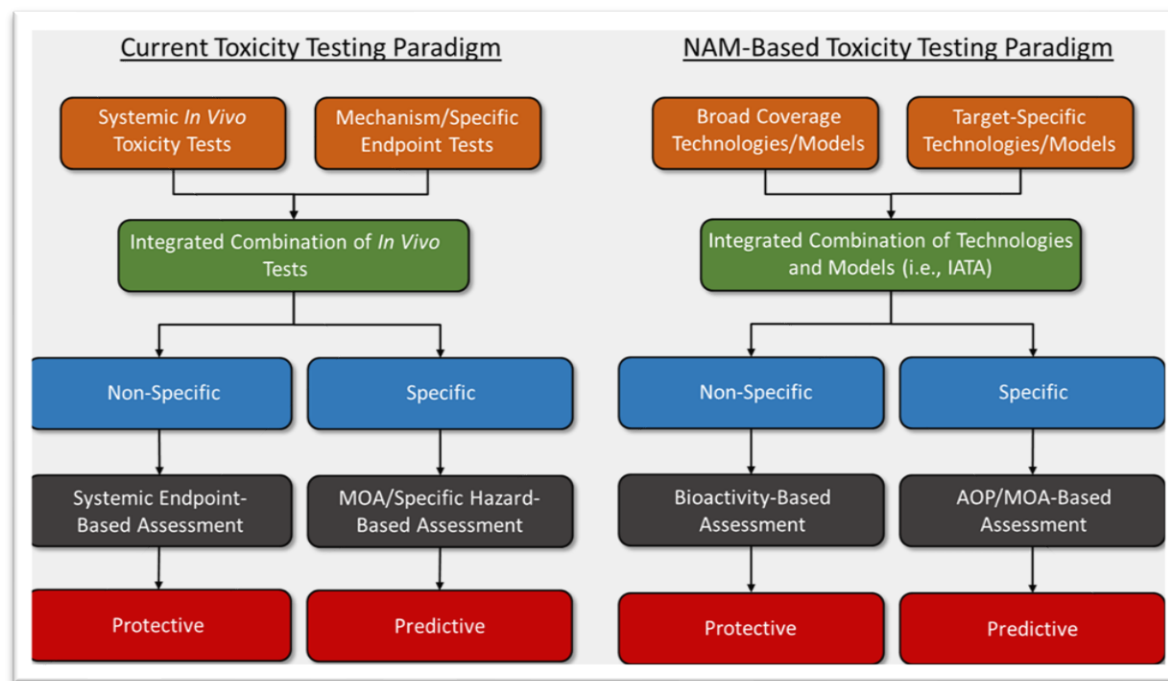


Useful for Tier 2/bespoke safety assessment when differentiation between bioactivity & adversity is needed

The diagram illustrates the workflow of the PBK assay. It begins with four interconnected assay components: 'Receptor binding assays' (orange box), 'Cellular stress assays' (dark blue box), 'Transcriptomics' (teal box), and 'Exposure (PBK)' (pink box). A large blue arrow points from these components to a set of three icons: a beaker with bubbles, a petri dish, and a computer monitor displaying a molecular structure. A second large blue arrow points downwards from this icon set, indicating the progression to the next stage of the assay.

If biological activity measured using a broad suite of human-relevant test systems is above the predicted exposure in humans, then systemic adverse effects are highly unlikely

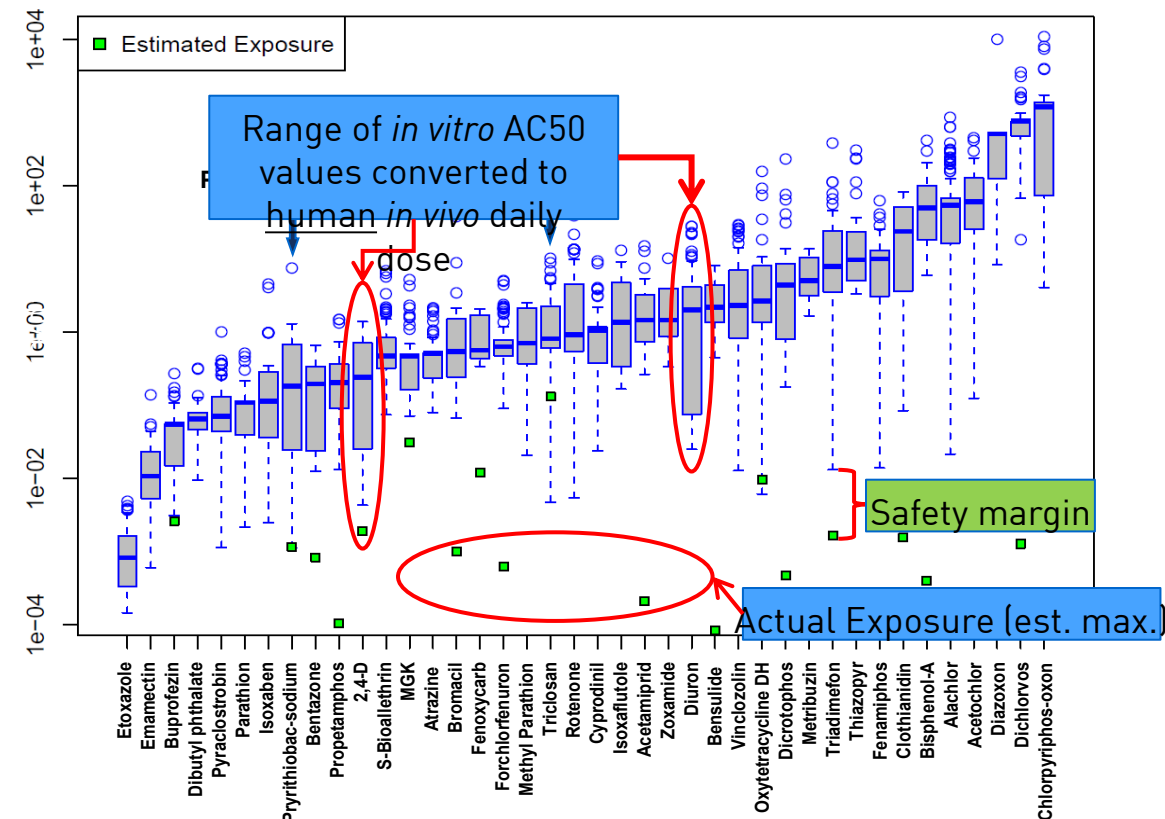
Next Generation Risk Assessment (NGRA) – Protection not Prediction



Adverse effects in traditional and alternative toxicity tests

Patience Browne^{a,c}, Katie Paul Friedman^b, Kim Boekelheide^c, Russell S. Thomas^b

Distributions of Oral Equivalent Values and Predicted Chronic Exposures

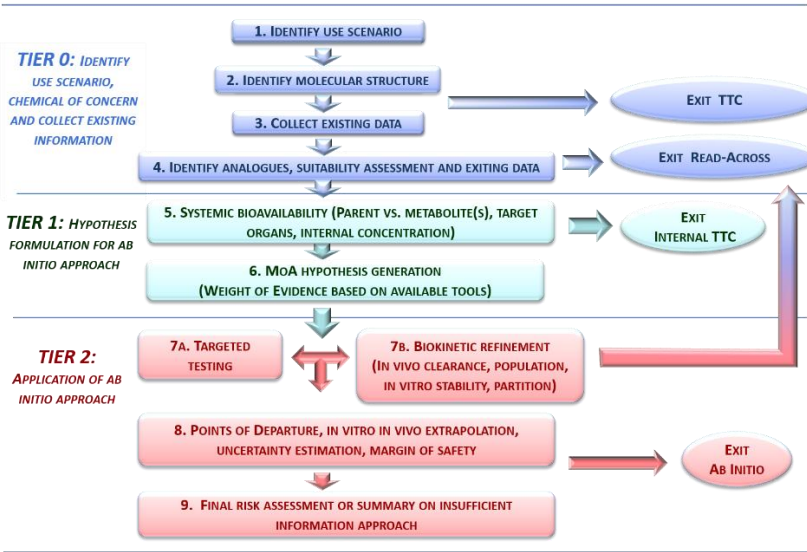


Thomas RS et al., 2019. Tox Sci. 1;169(2):317-332.

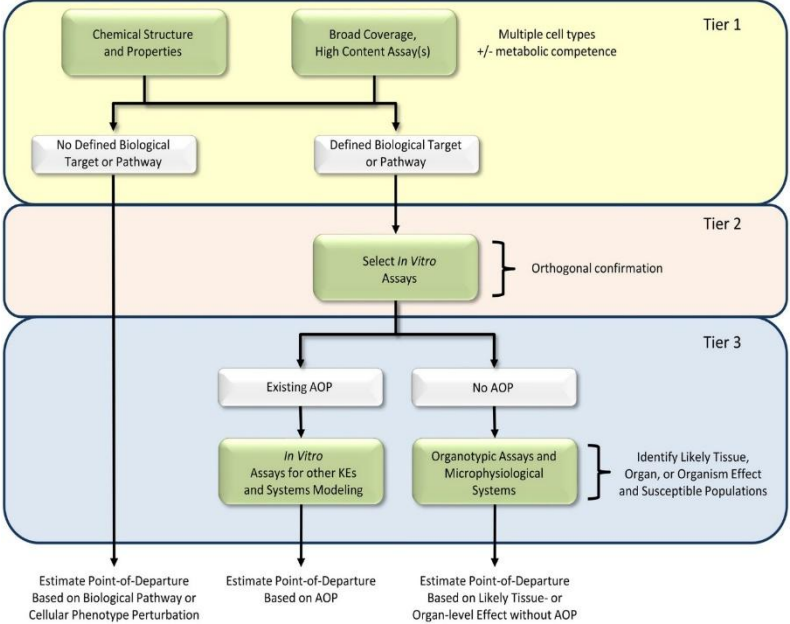
Decision focuses on **Protection of human health**

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.**

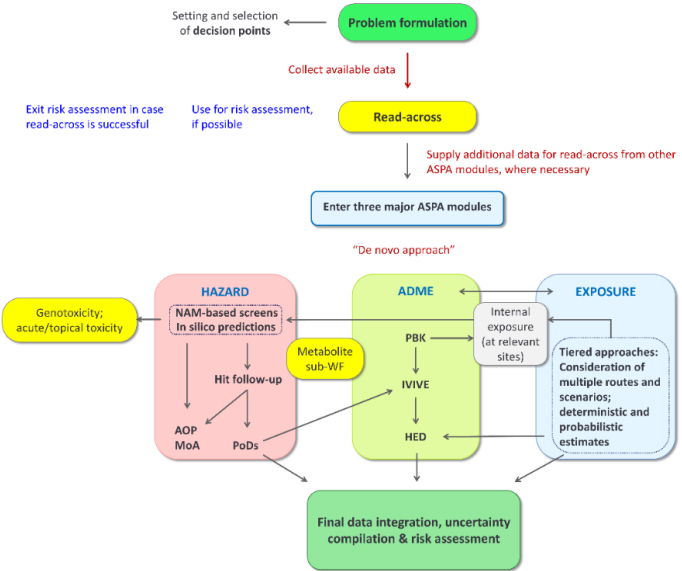
Decision frameworks in NGRA – Tiered approaches



Seurat-1 Berggren et al., 2017



EPA Blueprint Thomas et al., 2019

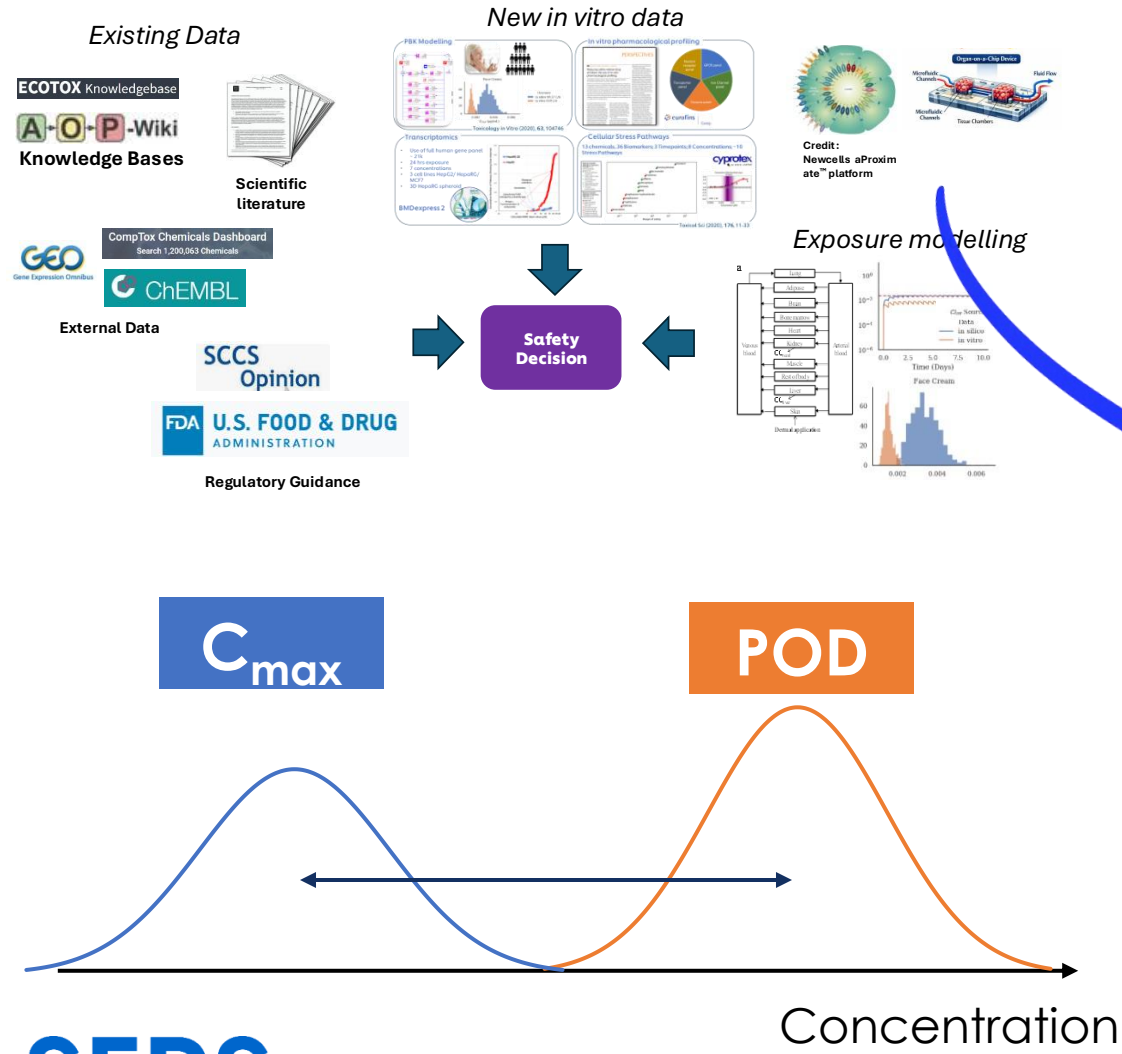


ASPA Leist et al 2025

Safety Decision

Decision frameworks in NGRA – Putting theory into practice

Unilever



Problem formulation – Tier 0

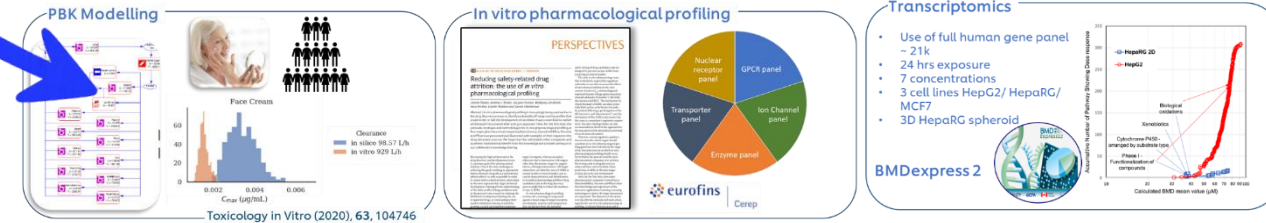
ECOTOX Knowledgebase

A-O-P-Wiki

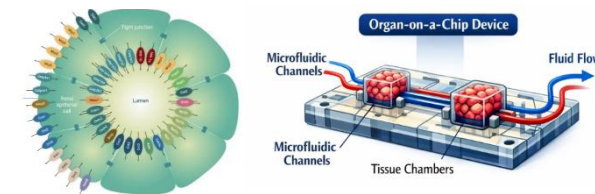
CompTox Chemicals Dashboard
Search 1,200,063 Chemicals

ChEMBL

Bioactivity : Exposure Ratio – Tier 1



Refine Bioactivity:Exposure– Tier 2



Safety Decision

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

Archives of Toxicology (2022) 96:2865–2879
<https://doi.org/10.1007/s00204-022-03365-4>

Received: 17 May 2022 / Accepted: 11 August 2022 / Published online: 20 August 2022

REVIEW ARTICLE



Evaluation of an Early Tier Toolbox for Systemic Safety

AIM: Use robust/reproducible NAMs to ensure the protection of consumers: can the approach be used to confidently identify 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

Performance Criteria

Benchmarking using chemical-exposure scenarios

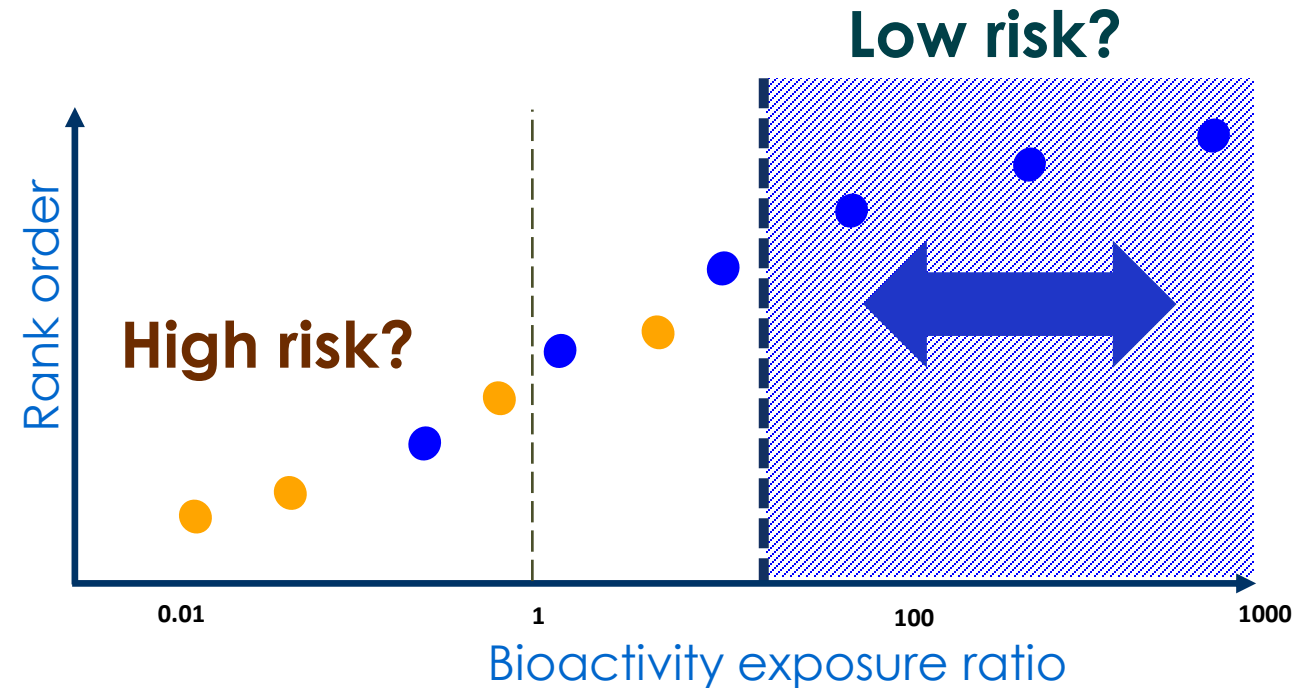
- Chemicals with well-defined human exposures
- Traditional safety assessment available
- High certainty in the risk classification for each chemical-exposure scenario from a consumer goods perspective
- Risk class is relative to consumer health (N.B. drugs = high-risk)



'Low' risk for consumers from systemic perspective



'High' risk for consumers from systemic perspective



Protectiveness

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

Utility

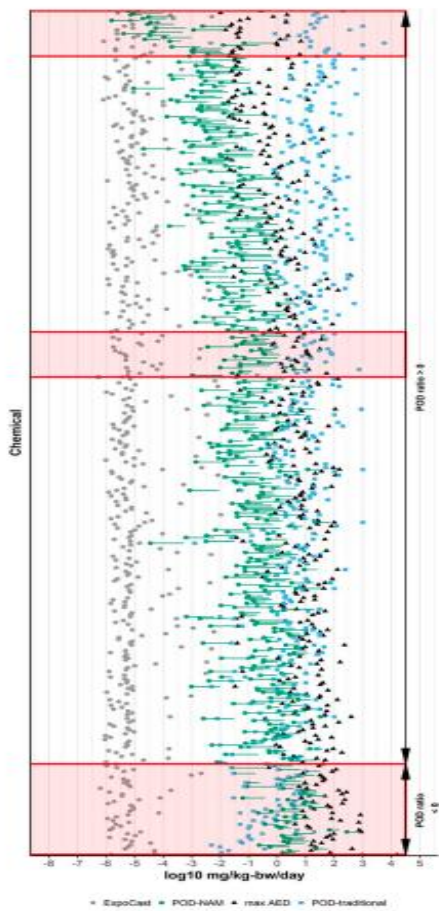
How many of the low-risk scenarios are identified as low-risk at this early tier stage in a risk assessment framework?
(i.e. **BER > threshold**)

In vitro cell assay (NAM) data are generally protective...

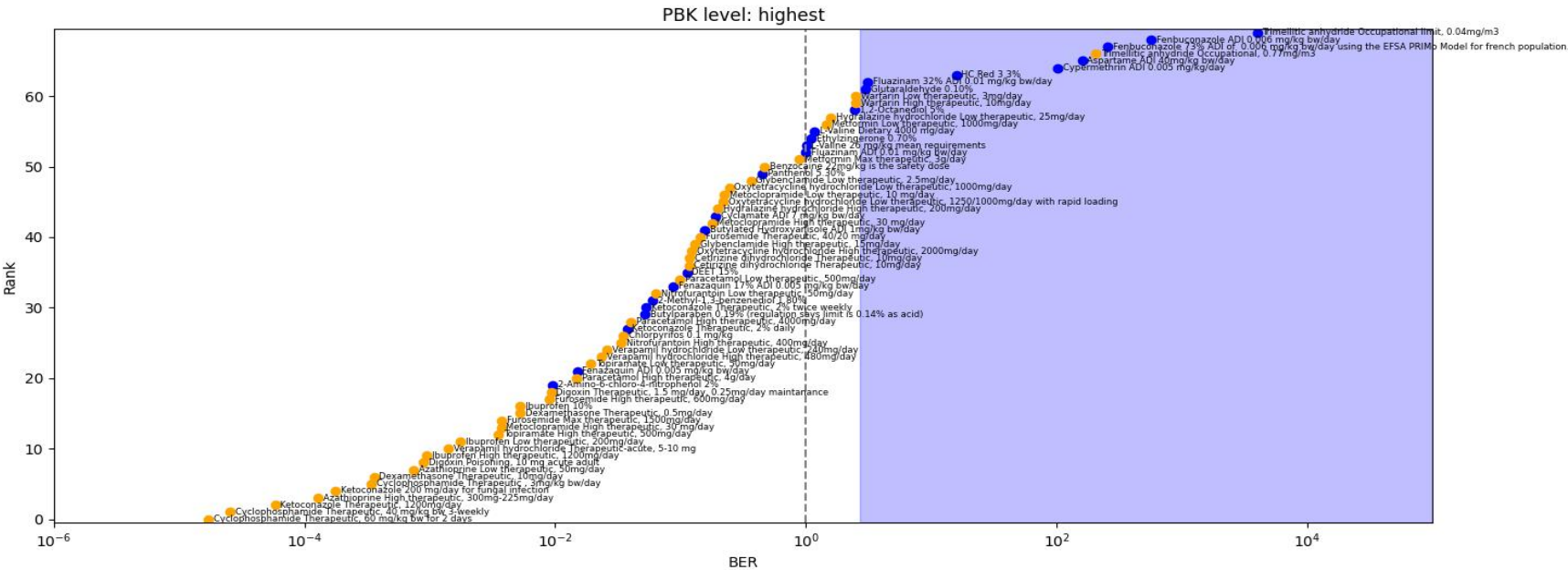
Unilever

Protectiveness98% (45 out of 46)

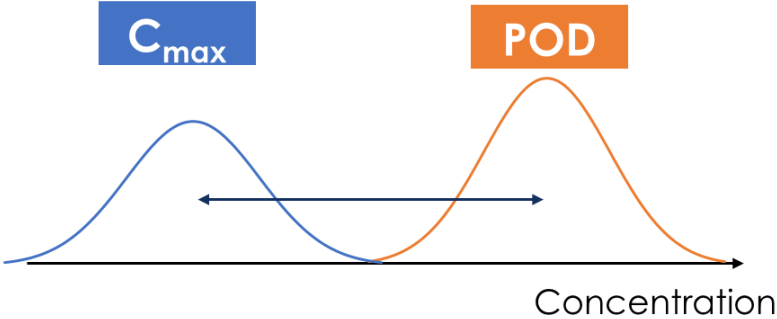
Utility33% (8 out of 24)



Paul-Friedmann et al 2020



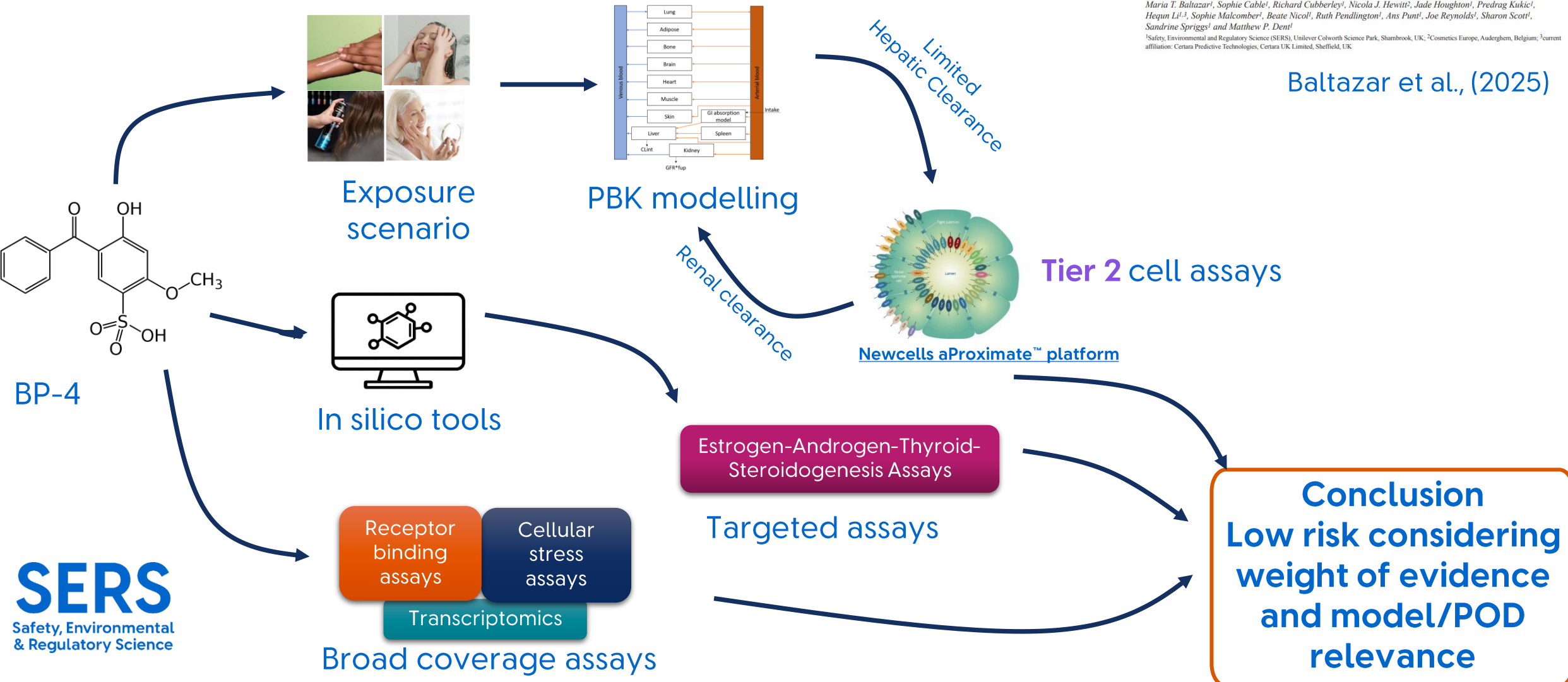
Middleton et al., 2022; Cable et al., 2025



- High protectiveness (>90%, i.e., conservative)
- In many cases, need to go to Tier 2 if ingredients are active or exposures are high

Case Study: BP-4 as an ingredient in sunscreens

Is Benzophenone-4 safe in a sunscreen product at the maximum approved level of 5%?



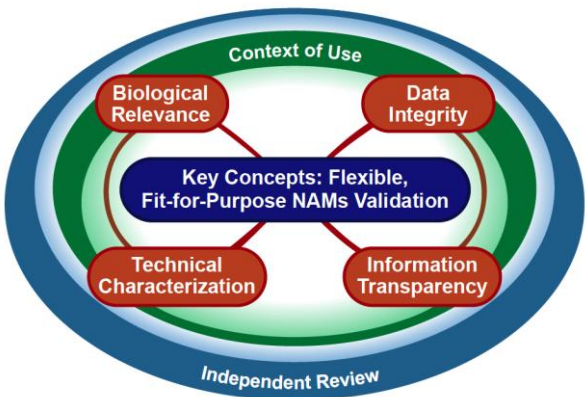
Research Article
Making Safety Decisions for a Sunscreen Active Ingredient Using Next-Generation Risk Assessment: Benzophenone-4 Case Study
Maria T. Baltazar¹, Sophie Cable¹, Richard Cubberley¹, Nicola J. Hewitt², Jade Houghton¹, Predrag Kukic¹, Hequn Li^{1,3}, Sophie Malcomber¹, Beate Nicol¹, Ruth Pendlington¹, Ans Punt¹, Joe Reynolds¹, Sharon Scott¹, Sandrine Spriggs¹ and Matthew P. Dent¹
¹Safety, Environmental and Regulatory Science (SERS), Unilever Colworth Science Park, Sharnbrook, UK; ²Cosmetics Europe, Auderghem, Belgium; ³current affiliation: Certara Predictive Technologies, Certara UK Limited, Sheffield, UK

Baltazar et al., (2025)

Evaluating non-animal toolboxes and workflows

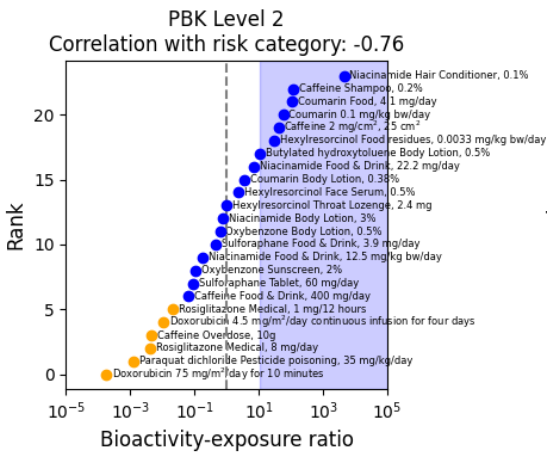
Unilever

Validation Frameworks



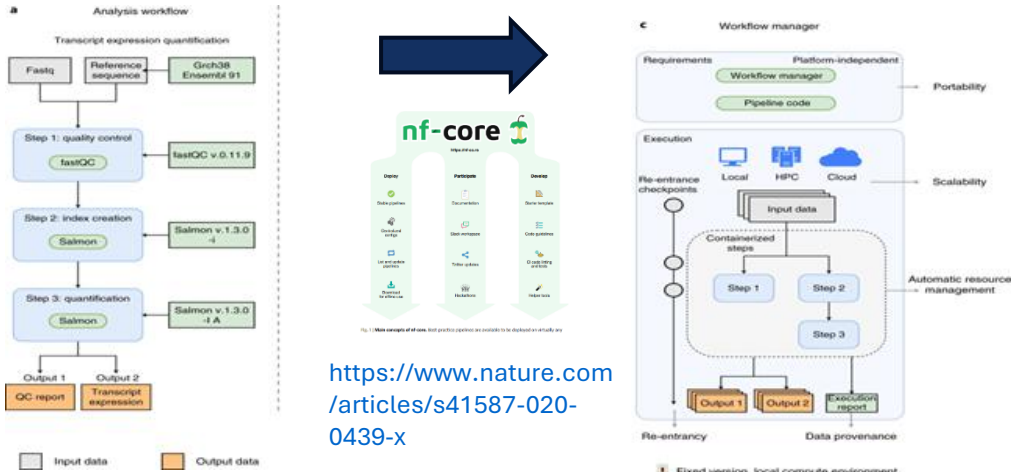
ICCVAM Validation Report, Figure 1
(adapted from van der Zalm et al. 2022 Arch Tox)

Protectiveness & Utility



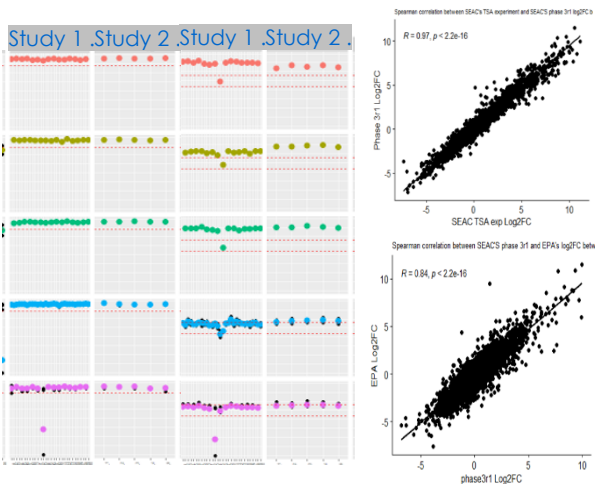
Middleton et al., 2022

Computational reproducibility/portability

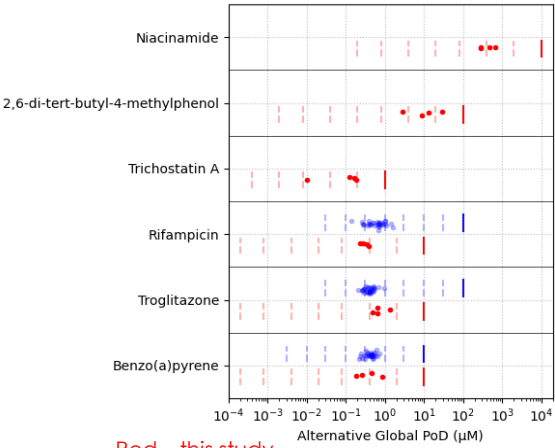


<https://www.nature.com/articles/s41587-020-0439-x>

Assess Quality

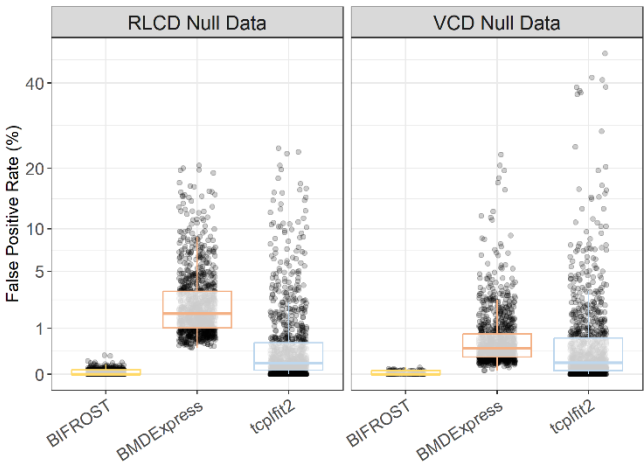


Assess reproducibility



Red – this study
Blue – EPA HepaRG screen
(Rogers et al. EHP 2025 - GSE284321)

Assess Specificity



Factors limiting broad uptake of NGRA approaches *Unilever*

Problem formulation– Tier 0

ECOTOX Knowledgebase

A→O→P-Wiki



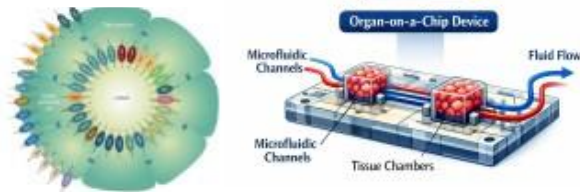
CompTox Chemicals Dashboard
Search 1,200,063 Chemicals

ChEMBL

Bioactivity : Exposure Ratio– Tier 1



Refine Bioactivity:Exposure- Tier 2



Safety Decision

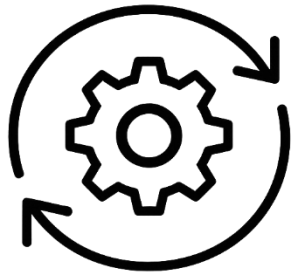
- Requires **multidisciplinary** teams, covering:
 - Risk assessment & business partnering
 - Toxicology
 - Chemistry
 - Statistical modelling
 - Bioinformatics
 - Data Engineer
 - PBK modellers
- Current toolboxes are **protective of human health**, but assessments can be **overly conservative**.
- **Mechanistic interpretation** of data and distinguishing between bioactivity and adversity is an **on-going challenge**.
- **Limited integration and interpretation** of different data types and **sources of information**.

Where can AI add value to NGRA?

Unilever

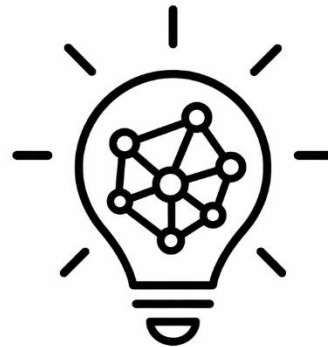
Standardisation and automation

- Automation of **existing processes and workflows**
- Focus: saving time performing routine tasks associated with a risk assessment.
- Example: agentic aids for e.g., REACH submission.



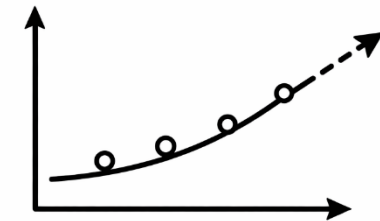
Knowledge and insights

- Extracting **new insights** from multiple data sources
- Focus: supporting risk assessors to **generate structured hypothesis** or **mechanistic interpretations** of the available data.



Predictive modelling

- **Generating predictions** based on available information.
- Focus: Modelling and simulation of **hazard** and **exposure** data.
- Examples: **quantifying risk**, estimating uncertainty, generating **synthetic data** (filling data gaps).



Application of AI in Safety Science: **Agentic workflows**

Unilever

Problem formulation– Tier 0

ECOTOX Knowledgebase

A-O-P-Wiki



CompTox Chemicals Dashboard
Search 1,200,063 Chemicals

ChEMBL

Bioactivity : Exposure Ratio– Tier 1



Refine Bioactivity:Exposure– Tier 2



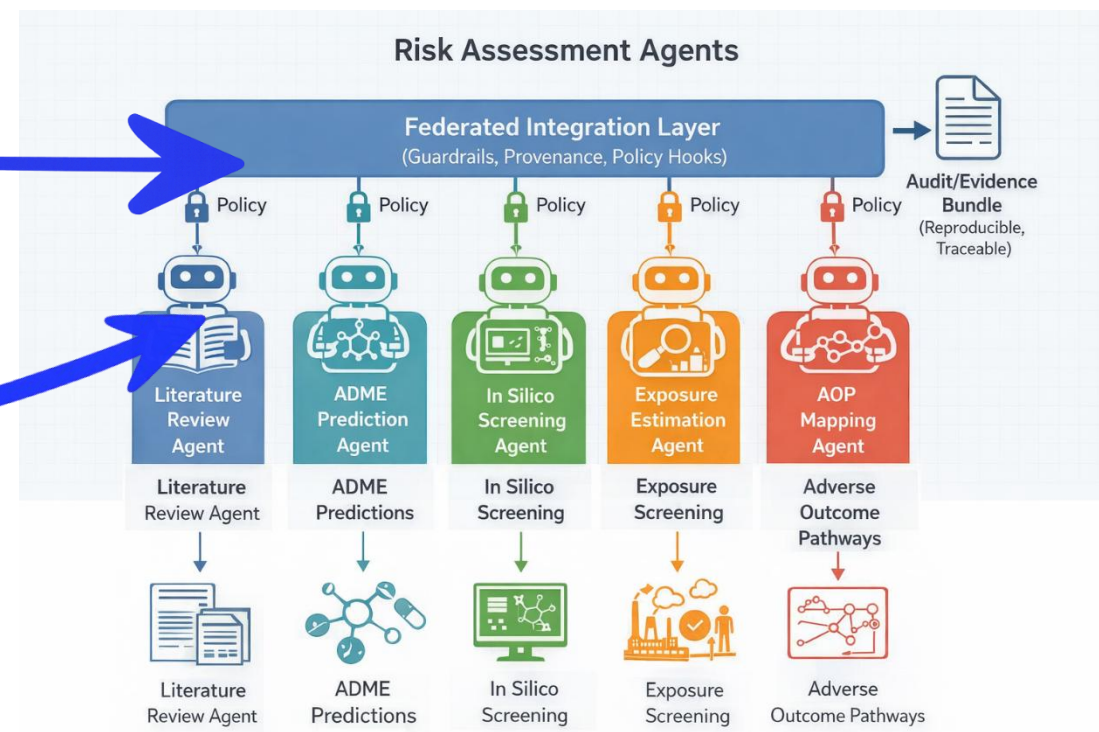
Safety Decision

SERS
Safety, Environmental
& Regulatory Science



Keshav Gubbi

Automation of Tier 0 and Tier 1 workflows



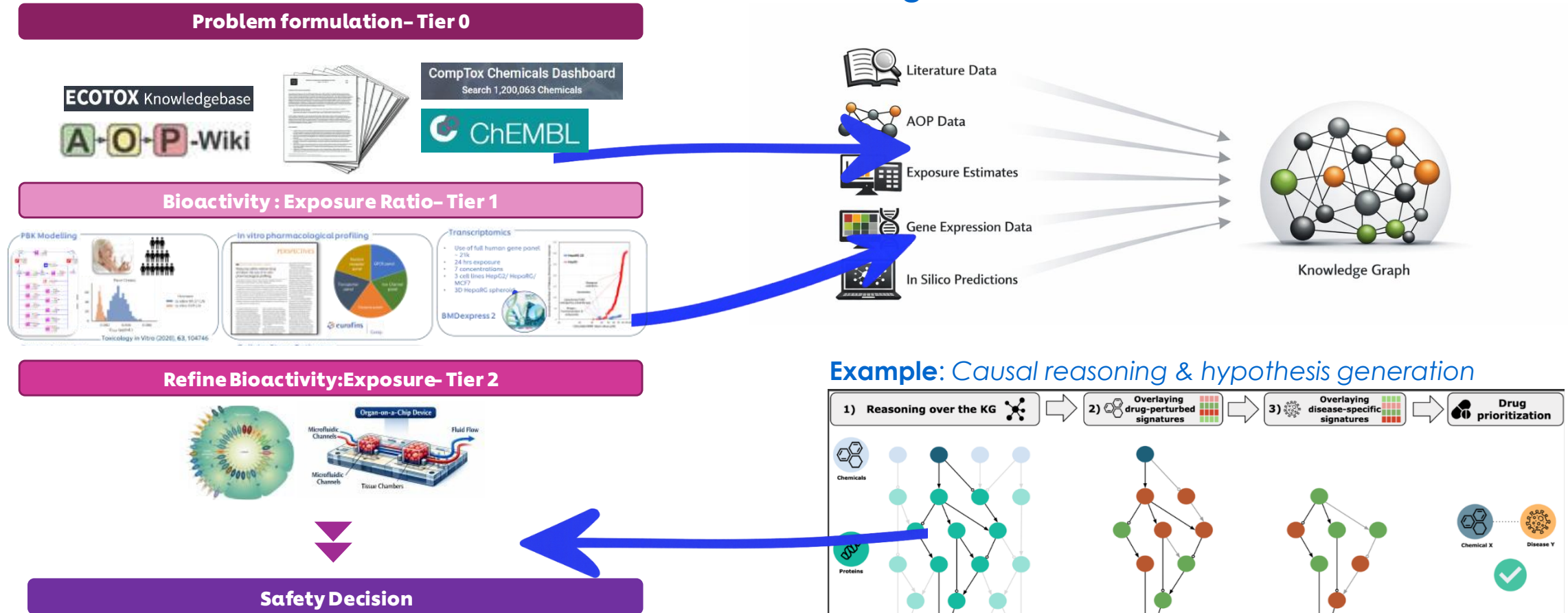
Based on Ivo Djidrovski, 2026

- Increased productivity
- Holistic integration across expertise teams
- Consistency and tractability of outcomes & decisions

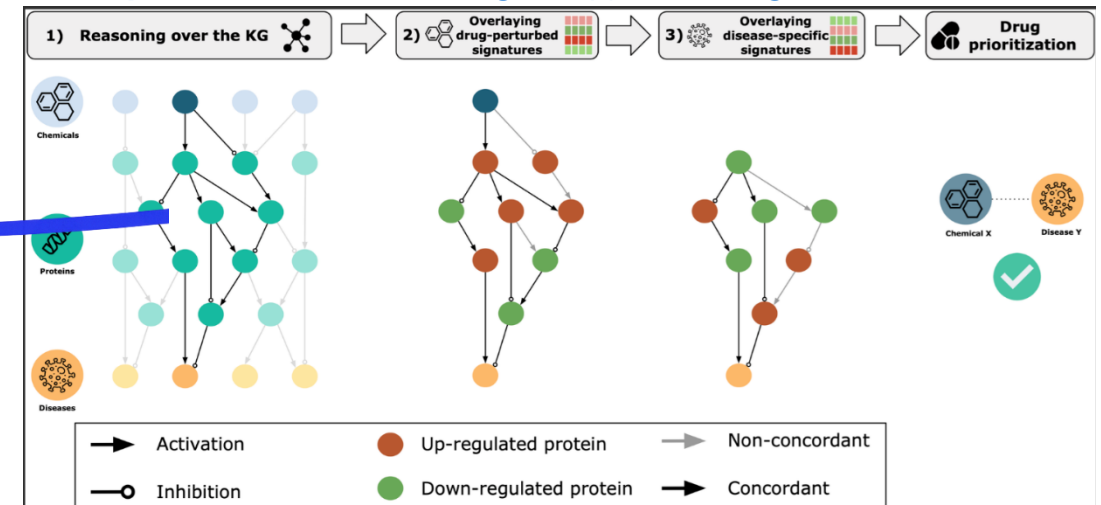
Application of AI in Safety Science: Integrating knowledge across data modalities

Unilever

Integration of structured and unstructured data



Example: Causal reasoning & hypothesis generation



Daniel Domingo-Fernández et al, 2022



Alberto Locca



Keshav Gubbi

Application of AI in Safety Science: Quantifying biological similarity at scale

Unilever

Problem formulation- Tier 0

ECOTOX Knowledgebase

A → Q → P -Wiki



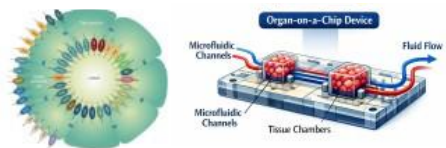
CompTox Chemicals Dashboard
Search 1,200,063 Chemicals

ChEMBL

Bioactivity : Exposure Ratio- Tier 1

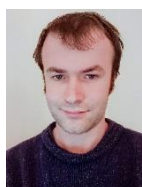


Refine Bioactivity:Exposure- Tier 2

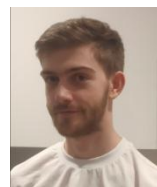


Safety Decision

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& Regulatory Science

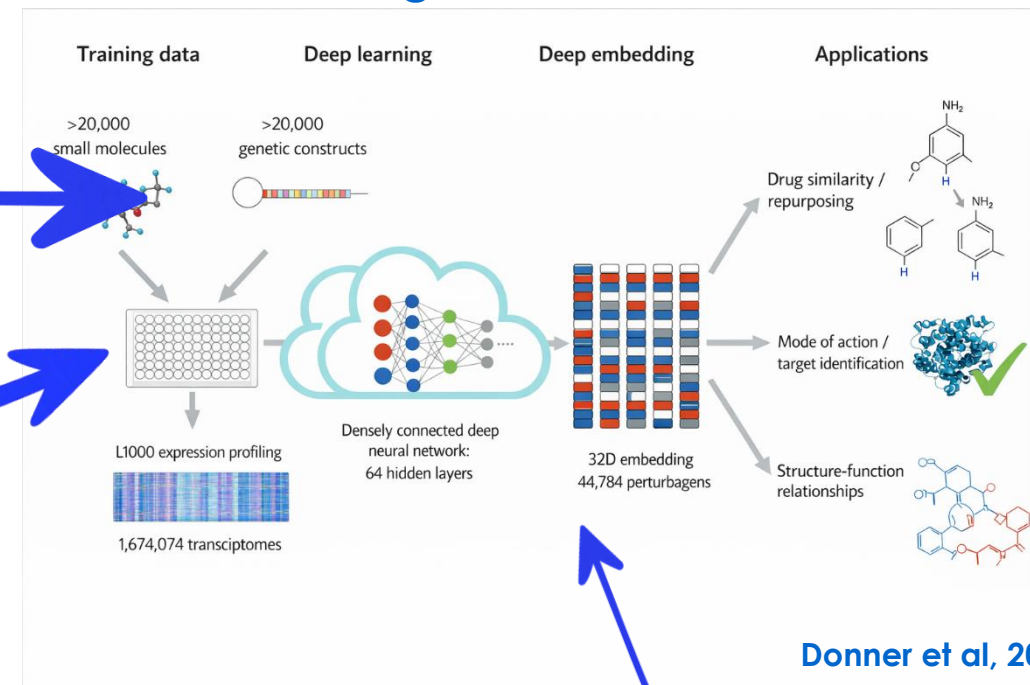


Hugh Barlow



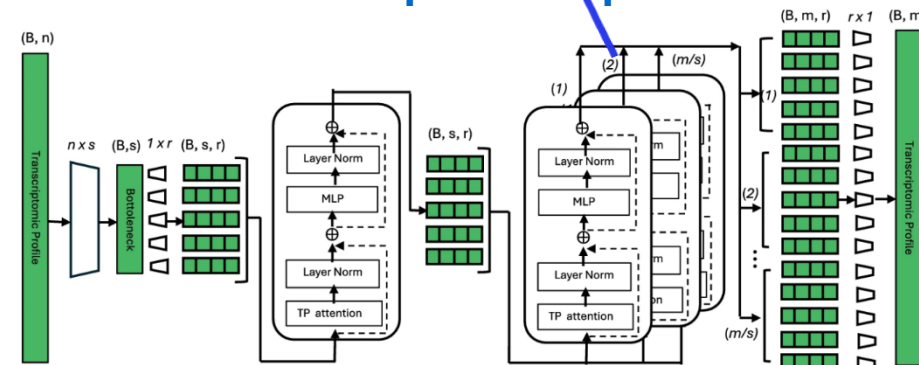
Patrik Engi

Metric learning ...



Donner et al, 2018

... across transcriptomics platforms



Cong et al, 2026

Application of AI in Safety Science: Synthetic data & Foundation Models

Unilever

Problem formulation- Tier 0

ECOTOX Knowledgebase

A-O-P-Wiki



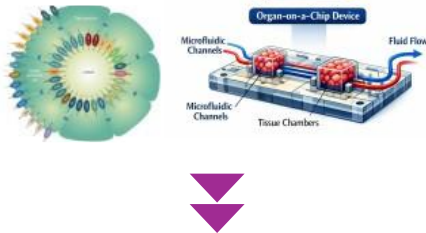
CompTox Chemicals Dashboard
Search 1,200,063 Chemicals

ChEMBL

Bioactivity : Exposure Ratio- Tier 1

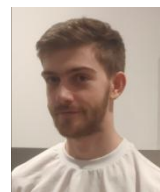


Refine Bioactivity:Exposure- Tier 2



Safety Decision

SERS
Safety, Environmental
& Regulatory Science

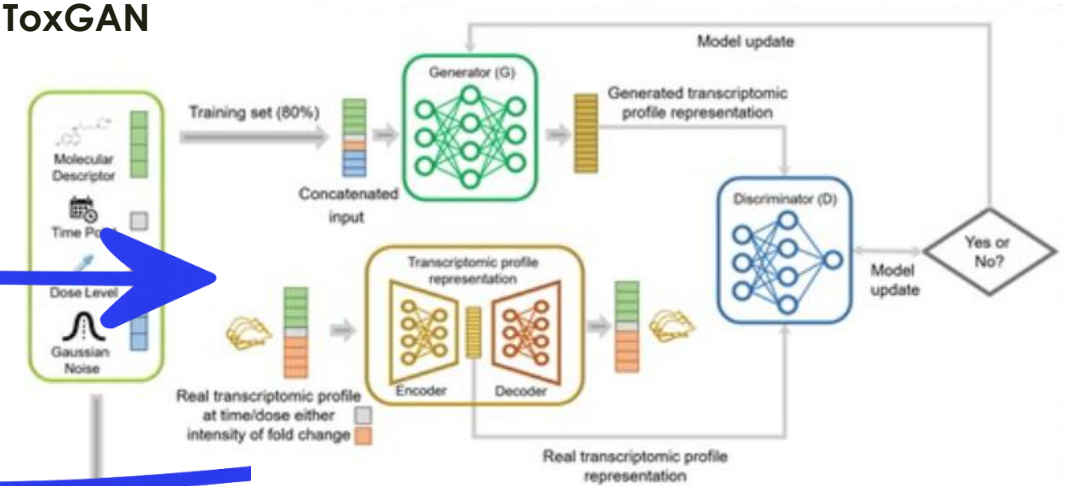


Patrik Engi



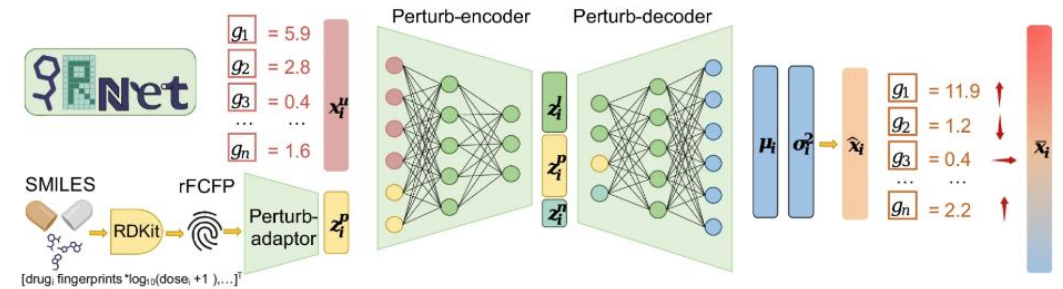
Hugh Barlow

ToxGAN



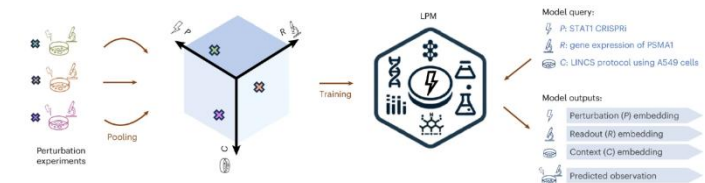
Chen et al, 2022

PRNet



Qi et al, 2024

Large-perturbation model



Building confidence in AI augmented systems

Proportionate Governance

Adoption of AI requires proportional validation and governance based on potential scientific and regulatory impact.

Traceability

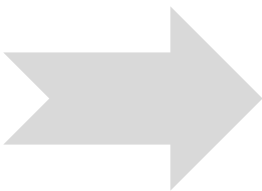
AI-supported outputs should remain traceable to underlying evidence.

Reproducibility and Change control

AI outputs should remain governable over time and variability should be monitored and controlled.

Human Accountability

AI should assist decisions with human oversight, clear responsibilities, and communication of uncertainty.



Use **case studies** to build shared understanding and **practical guidance** on how to evaluate AI-augmented frameworks.

Vadlapati et al, 2026

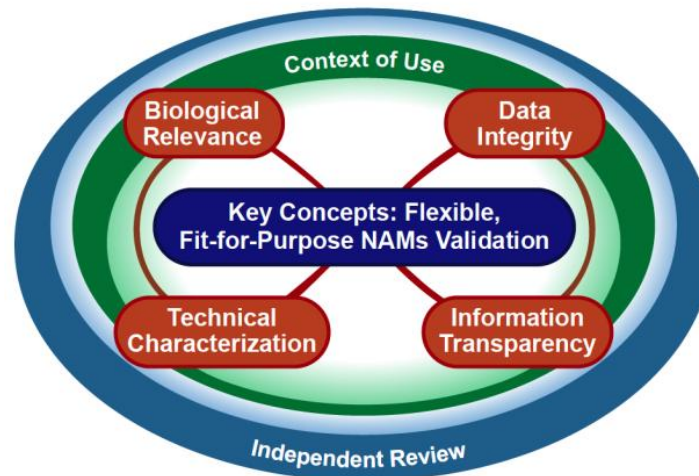
Failure Class	Description	Critical Impact
Epistemic Hallucination	Fabrication of plausible but factually incorrect content	Erroneous medical or military decisions based on false information
Overconfidence Failure	High-confidence presentation of incorrect outputs	Human operators accept errors without scrutiny
Abstention Failure	Inappropriate refusal to respond when a response is needed	System paralysis at critical decision points
Prompt Fragility	Output instability across semantically equivalent phrasings	Unpredictable system behavior in diverse operational conditions
Temporal Drift	Performance changes across model versions and time	Undetected reliability regression following post-deployment updates
Reasoning Collapse	Incoherence, repetition, or truncation under complex reasoning demands	Failure during time-critical operational tasks
Agentic Escalation	Autonomous execution of unsafe or unintended actions	Irreversible operational or physical consequences
Adversarial Manipulation	Prompt injection causing deviation from intended instructions	Command hijacking by deliberate adversaries

Towards AI-augmented safety frameworks: Build Confidence Step by Step

Increased industry: academia: regulatory exchange to amplify positive impact of AI on future toxicology, while also managing governance challenges.



- Leverage scientific and technological progress to advance exposure science, understanding of human biology and risk assessment.



- Guidance and guidelines that will pave the way for regulatory acceptance.

*ICCVAM Validation Report, Figure 1
(adapted from van der Zalm et al. 2022 Arch Tox)*



- Design educational programmes.
- Training and access to information that will support confidence in the use of AI.

Application of AI in Safety Science: Future Trends

Unilever

- Increased access to chemical, biological, exposure and toxicological data that are FAIR (expected exponential increase in global data volumes*).

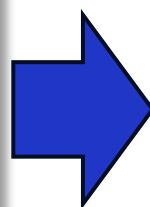


- Access to agentic AI workflows will increase uptake of NGRA-based approaches by making them more accessible and transparent



- Data-driven approaches that strengthen and enhance expert based decision making

* Source: Statista, Bernard Marr & Co



Bioactivity
vs adversity



ECOTOX Knowledgebase

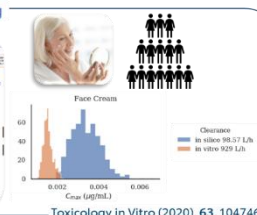
A → O → P -Wiki



CompTox Chemicals Dashboard
Search 1,200,063 Chemicals

ChEMBL

PBK Modelling



In vitro pharmacological profiling

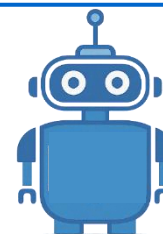


Transcriptomics

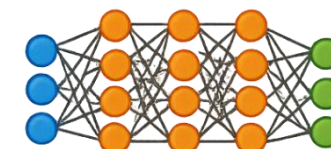
- Use of full human gene panel ~ 21k
- 24 hrs exposure
- 7 concentrations
- 3 cell lines HepG2/ HepaRG/ MCF7
- 3D HepaRG spheroid



AI-adjacent approaches



Agentic AI



Representation learning



Knowledge Graphs



Risk Quantification & Uncertainty analysis



Key drivers & pathways



Next steps (e.g., hypothesis generation)



Weighing and scoring different sources of information

**Traditional Toxicology =
Empirical science focused on
observations from animal
studies**



**Increased focus on the use of exposure
science and understanding of human
biology (NAMs, PBK, DAs, IATAs,
protection of human health, AOPs...)**



**Augmentation of Next
Generation Risk Assessment
using AI-based technologies
and approaches**



Conclusions

1. Leveraging human-in-the-loop AI-driven approaches in future will be crucial to further advance the use of Next Generation Risk Assessment.
2. Increased industry: academia: regulatory exchange is necessary in future to amplify positive impact of AI on toxicology.
3. Managing governance challenges early in the process will be needed, including FAIR use of NAM data, democratisation of AI models, increased international collaboration, including strategic plans, guidance & guidelines.
4. AI will further democratise Next Generation Risk Assessment as multidisciplinary science that brings together toxicologists and scientists with engineers, statisticians and mathematicians.

Acknowledgements

Unilever

Predrag Kukic, Andrew White, Hugh Barlow, Patrik Engi, Mark Liddell, Alberto Locca, Leonardo Contreas, Katie Przybylak, Ramyarrooban Kanapathywasam, Cameron Mackay, Gavin Maxwell



40+ years of developing
non-animal safety
science



70+ collaborations



600+ publications



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