



In Vitro Bioactivity Workflows for Enhanced Prioritization and Rapid Screening of Existing Chemicals in Canada: A Journey of Insights and Learnings

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Source: Al-Koshi Cleaning Chemicals

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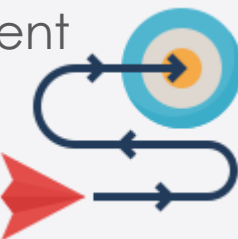


Health and Environmental
Sciences Institute

ESTAR MOLECULAR POINT OF
DEPARTURE PROJECT GROUP

Moving Towards Integrating New Approach Methodologies (NAMs) for Chemicals Management in Canada

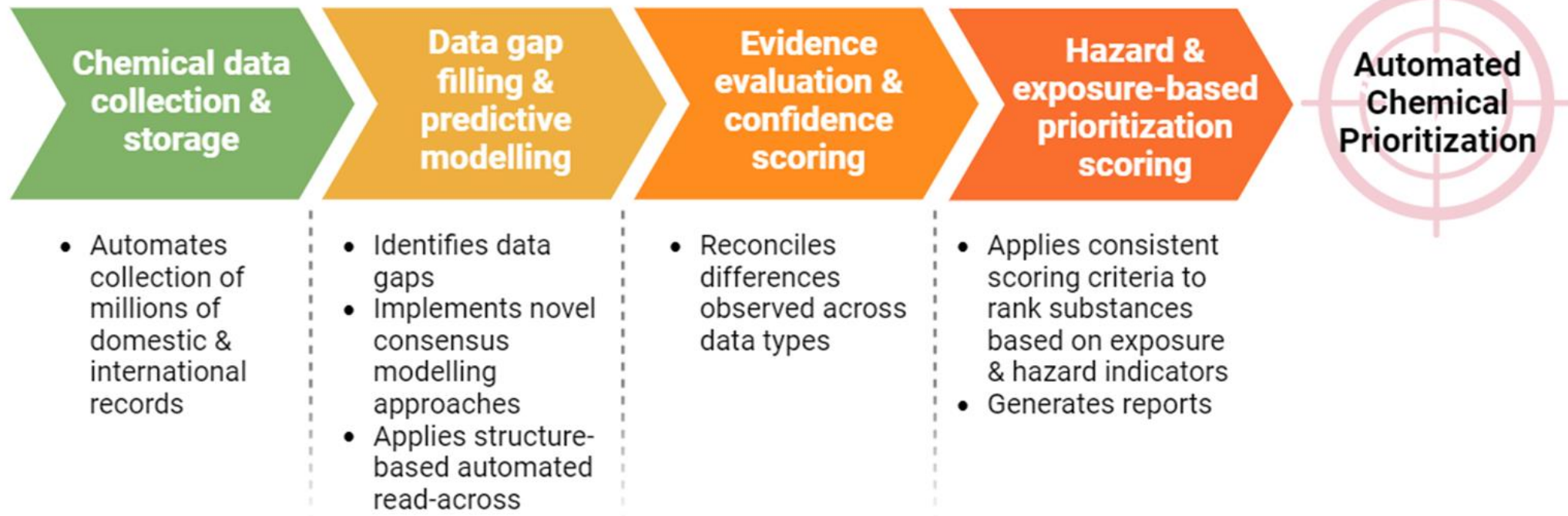
- Transition to NAM as a complete replacement over animals will be a challenging and a slow process
 - The better starting point may be to ask:
 - » What are the near-term opportunities for deploying currently available NAM with respect to prioritization and risk assessment activities?
 - » How can NAM help to focus the burden of assessing chemicals in Canada
- Finding a starting point -> In Canada, the focus of early development and application of NAM has surrounded program opportunities where:
 - NAM is a step above the status quo
 - where the development of fit-for-purpose NAM-based approaches help to make better informed decisions in the absence of other traditional forms of toxicity test data
 - » Addressing data poor substances
 - » Adding mechanistic or mode of action information to better inform risk management



Overarching Workflow for Data Collection, Processing and Interpretation



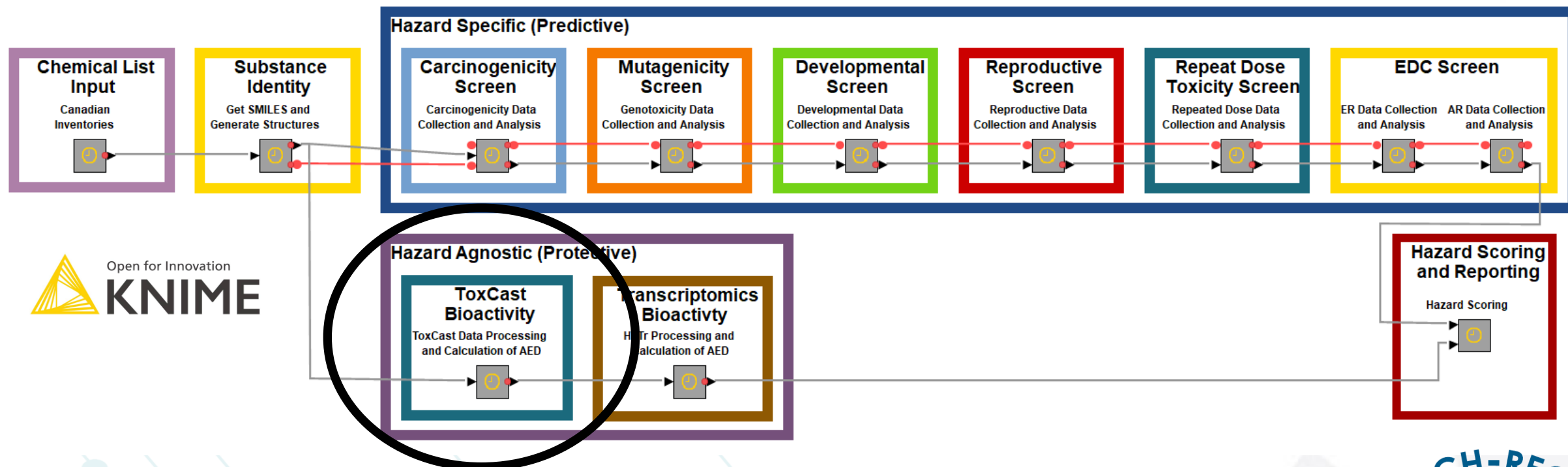
Health Canada Automated Workflow for Prioritization



Modular Design, Flexible and can be Modified as Science and Information Sources Evolve

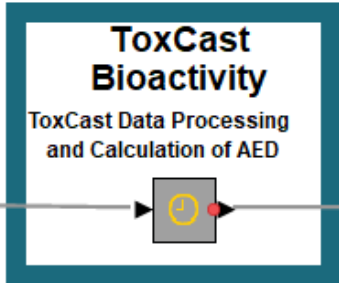


Health Canada Automated Workflow for Prioritization



- Collect and interpret New Approach Methods (NAM) and traditional data streams
- Workflow combines various biological levels into a decision framework
 - IATA structured nodes across toxicity endpoints
- Use best available data and processes to improve efficiency and reproducibility
- Deliver transparent and evidence-based decisions through integrated knowledge





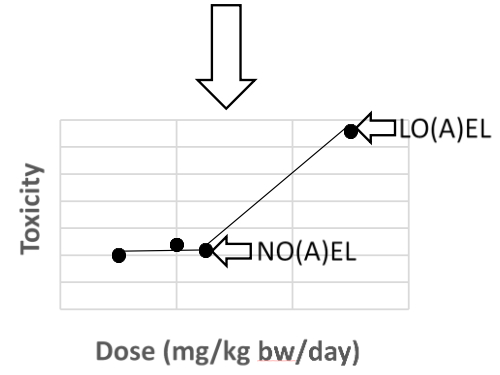
HEALTH CANADA SCIAD PRESENTS COMPARATIVE ANALYSIS

- Applied workflow to 40+ chemicals previously assessed on the Chemicals Management Plan
 - ToxCast data -> $POD_{\text{Bioactivity}}$
- Compared $POD_{\text{Bioactivity}}$ to traditional PODs used in the Screening Assessment Report (SAR)
- BERs derived: $POD_{\text{Bioactivity}}$ compared to Canadian exposure values from:
 - biomonitoring data
 - environmental media
 - consumer products
- BERs were evaluated to assess the utility of bioactivity data in prioritizing chemicals for risk assessment

Traditional Risk Assessment

Repeat Dose, Developmental, and Reproductive Studies

Extract NO(A)ELs and LO(A)ELs from Animal Studies Assessed by Health Canada



Evaluators Identify Most Appropriate NO(A)EL or LO(A)EL for Risk Characterization

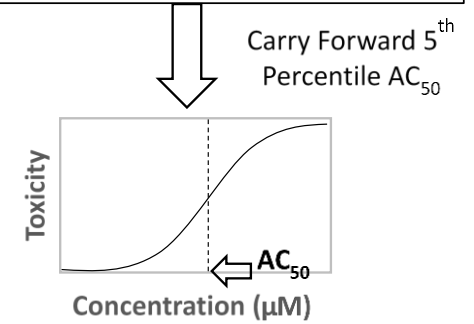
$POD_{\text{Traditional}}$

MOE

BER Approach

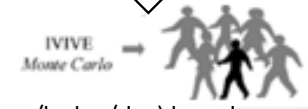
~1,400 ToxCast Endpoints (*in vitro* studies)

Extract AC_{50} values from ToxCast Endpoints



Carry Forward 5th Percentile AC_{50}

in vitro to *in vivo* Extrapolation



AED (mg/kg bw/day) based on upper 95th percentile steady state blood concentration representing a "sensitive" population

Carry Forward AED to Represent $POD_{\text{Bioactivity}}$

$POD_{\text{Bioactivity}}$

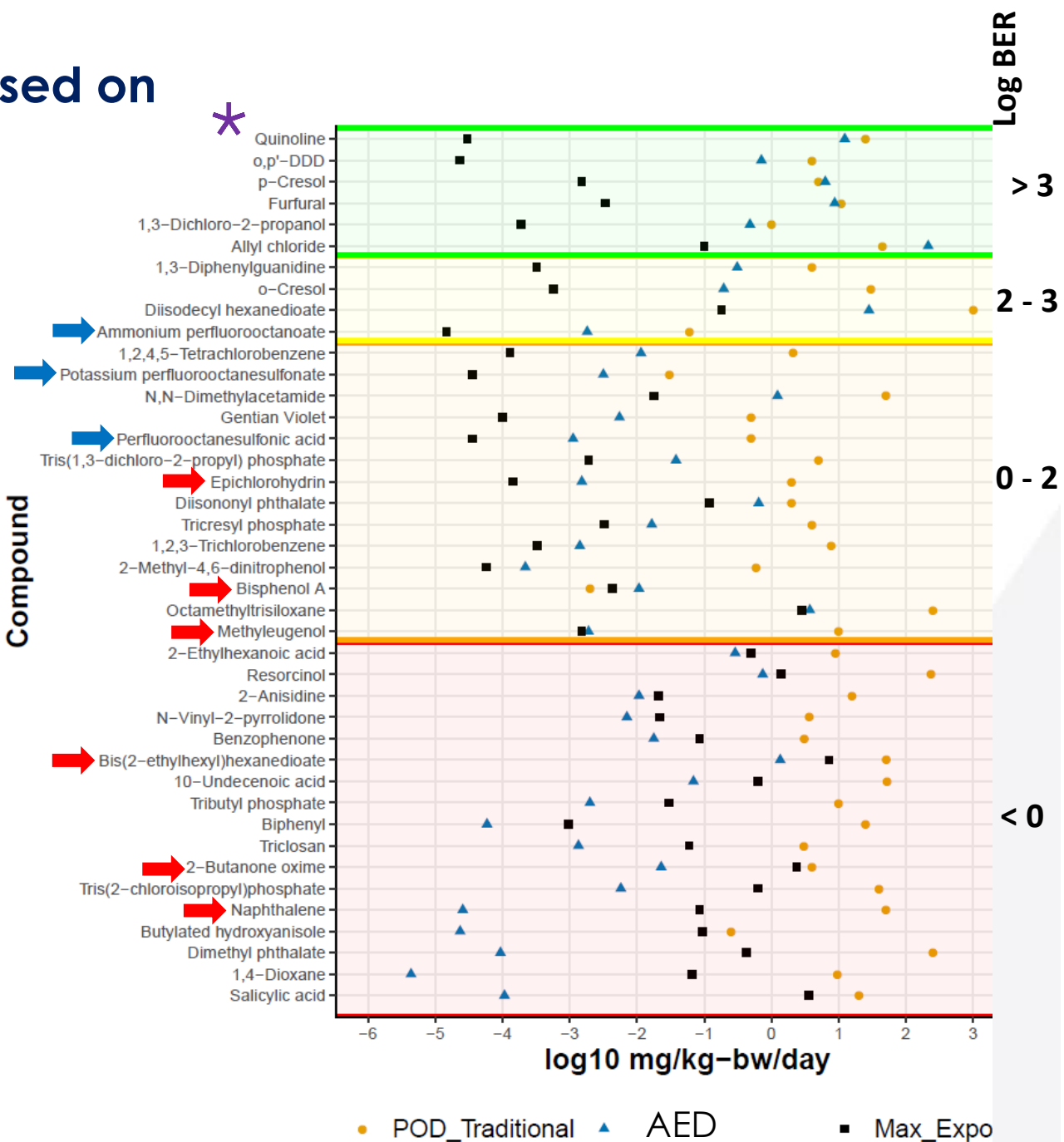
BER

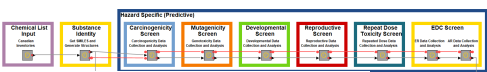
Exposure Level for Canadian Population (mg/kg bw/day)



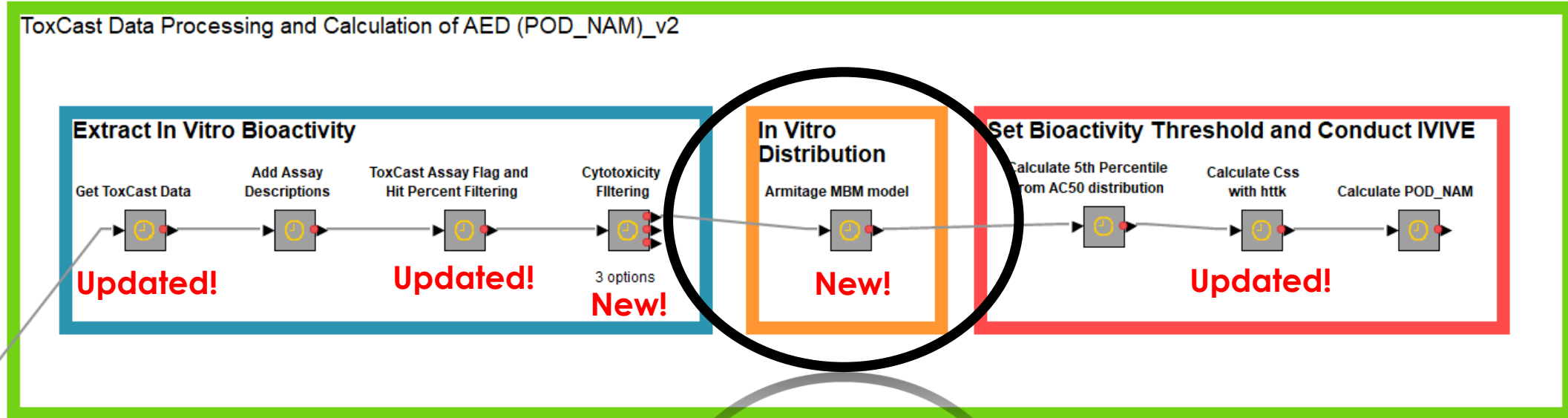
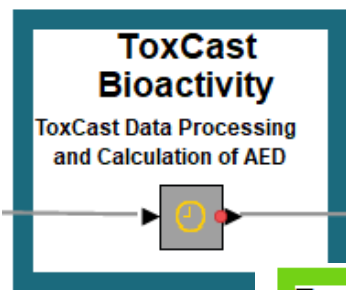
Bioactivity Exposure Ratio (BER) Based on Canadian Exposure Levels

- NAM-derived AED lower than $POD_{Traditional}$ for 38 out of 41 chemicals assessed previously
- All non-genotoxic compounds assessed as toxic to human health (red arrows) had a BER < 100
- All non-genotoxic compounds assessed as ecotoxic (blue arrows) had a BER 100 -1000
- One toxic chemical (**Quinoline**; purple star), assessed as potentially genotoxic, was identified as low priority using this approach
 - Only 5 ToxCast assays measure DNA damage or stalled replication and these have low sensitivity
- Approach for genotoxicity that builds on these experiences is needed**





Ongoing Refinements as Science Evolves

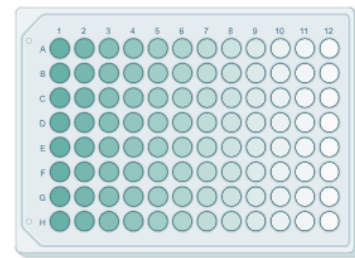


Can we improve our AED estimates for these data?

- In vitro disposition models

In Vitro Mass Balance Modelling (IV-MBM) Adjusts for In Vitro Toxicokinetics and In Vitro/In Vivo Bioavailability

- When doing HTK-IVIVE we typically assume that the in vitro BMC is equivalent to:
 - the freely dissolved concentration
 - the concentration in serum/blood that exerts a toxic effect
- IV-MBM modelling uses assay & chemical specific information to adjust the NAM BMC (nominal in vitro POD) to reflect these assumptions



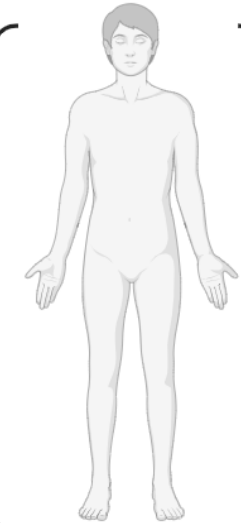
NAM BMC (uM) =
Nominal in vitro POD

1. Assumed equal

2. Adjusted IV-MBM



In vitro
POD (uM)



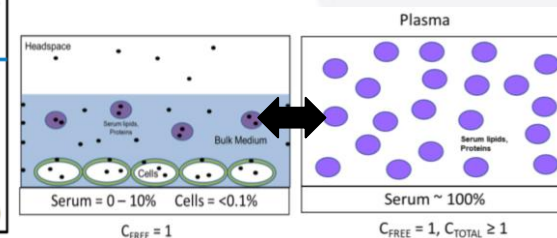
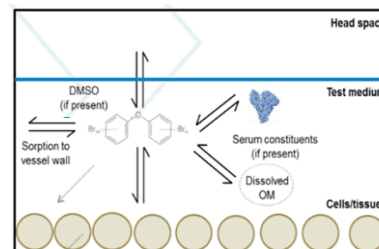
Administered
Equivalent Dose
(mg/kg BW/day)

In vitro toxicokinetics

- calculates the freely dissolved concentration

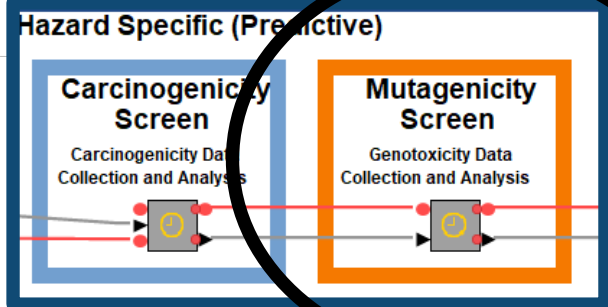
In vitro/in vivo bioavailability

- corrects differences b/w exposure medium and serum



EXPLORING IN VITRO GENOTOXICITY DATA TO ESTIMATE POINTS OF DEPARTURE

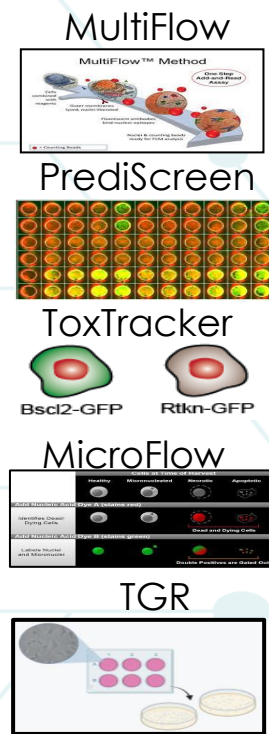
- Develop parallel approach that addresses gaps identified in the BER SciAD (2021)
- Evaluate impact of Httk refinements



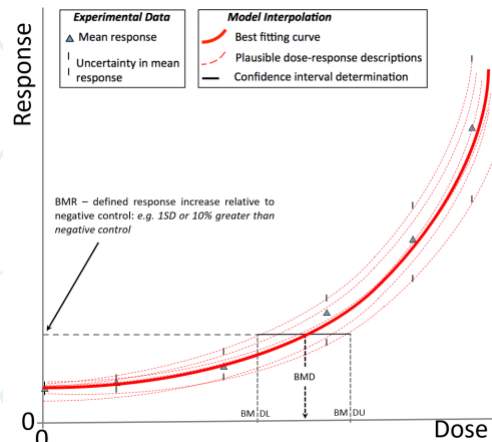
Case Study Part 1

31 compounds with:

- in vivo genetic toxicity PODs
- NAM genetic toxicity data



Genetic toxicity NAMs



Benchmark Concentration Modeling

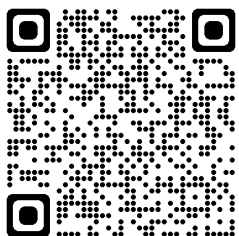


in vitro to in vivo
Extrapolation (IVIVE)
using htk in R

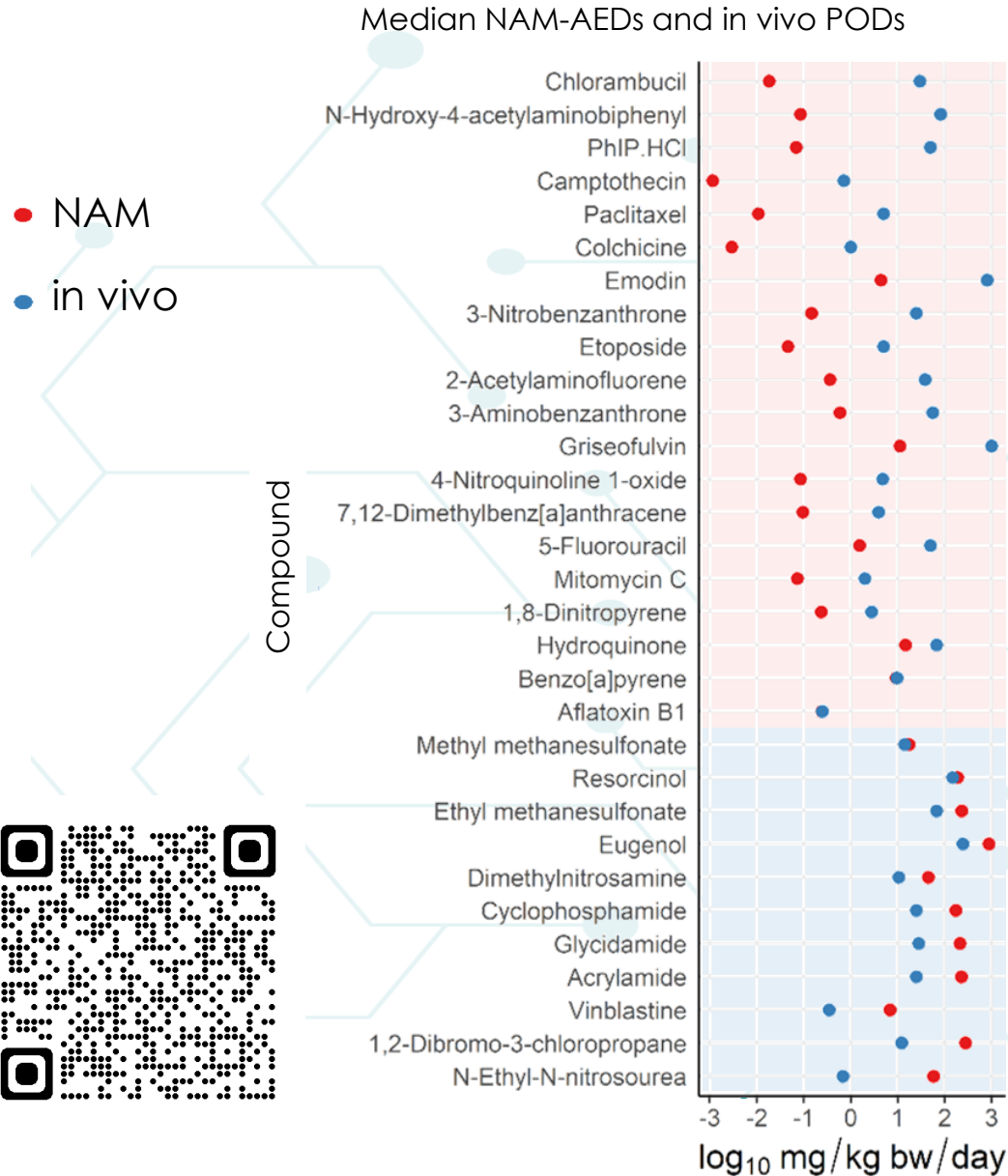
Pearce et al. 2017. J Stat Softw



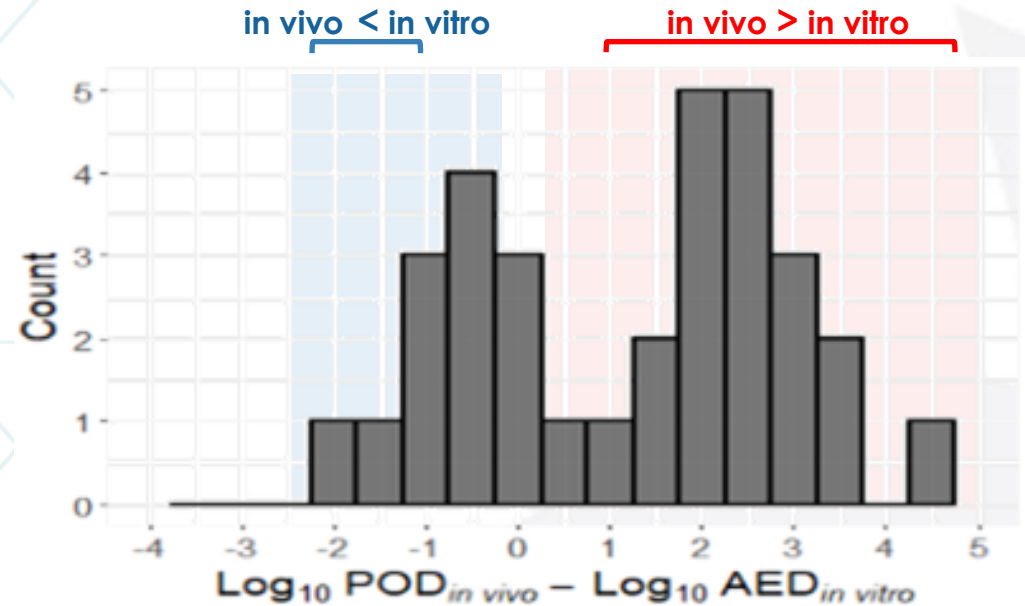
Human Administered Equivalent Dose (AED; mg/kg bw/day)



20/31 Chemicals have Median AEDs Lower than Animal-Based PODs



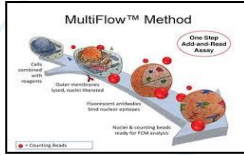
- Across the 31 chemicals, 198 NAM-AEDs derived from **in vitro genotoxicity data** were compared to 321 PODs from **in vivo genotoxicity data**
- The NAM-derived AEDs were typically protective of human health
 - i.e., lower than animal-based PODs



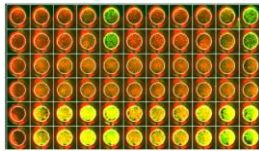
HTTK IVIVE Approach for GTTC Case Study - Part 2

- 31 compounds**
- in vivo genetic toxicity PODs
 - NAM genetic toxicity data

MultiFlow



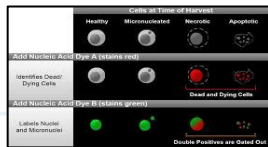
PrediScreen



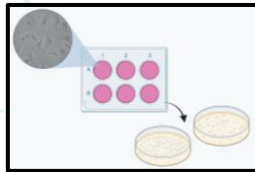
ToxTracker



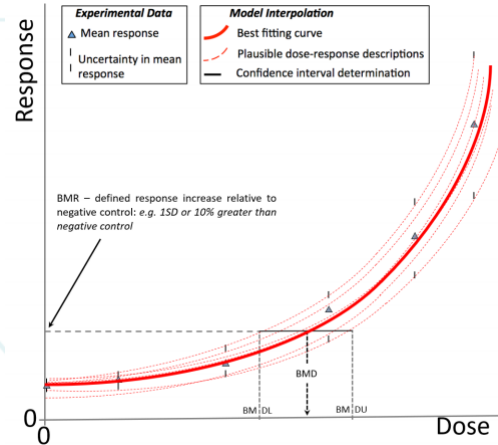
MicroFlow



TGR



Genetic toxicity NAMs



Benchmark Concentration Modeling (Nominal in vitro PoD)

Nominal in vitro POD



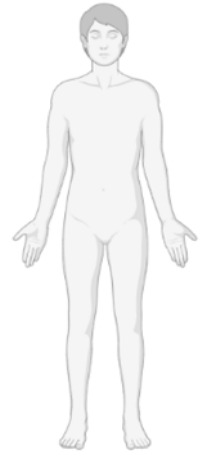
Adjusted in vitro POD

In vitro mass balance model (IV-MBM) to account for in vitro disposition
Armitage et al. (2021)



in vitro to in vivo Extrapolation (IVIVE) using htkk in R

Pearce et al. (2017)



Human Administered Equivalent Dose (AED; mg/kg bw/day)

In vitro mass balance model (IV-MBM) scaling factor reduces in vitro BMC

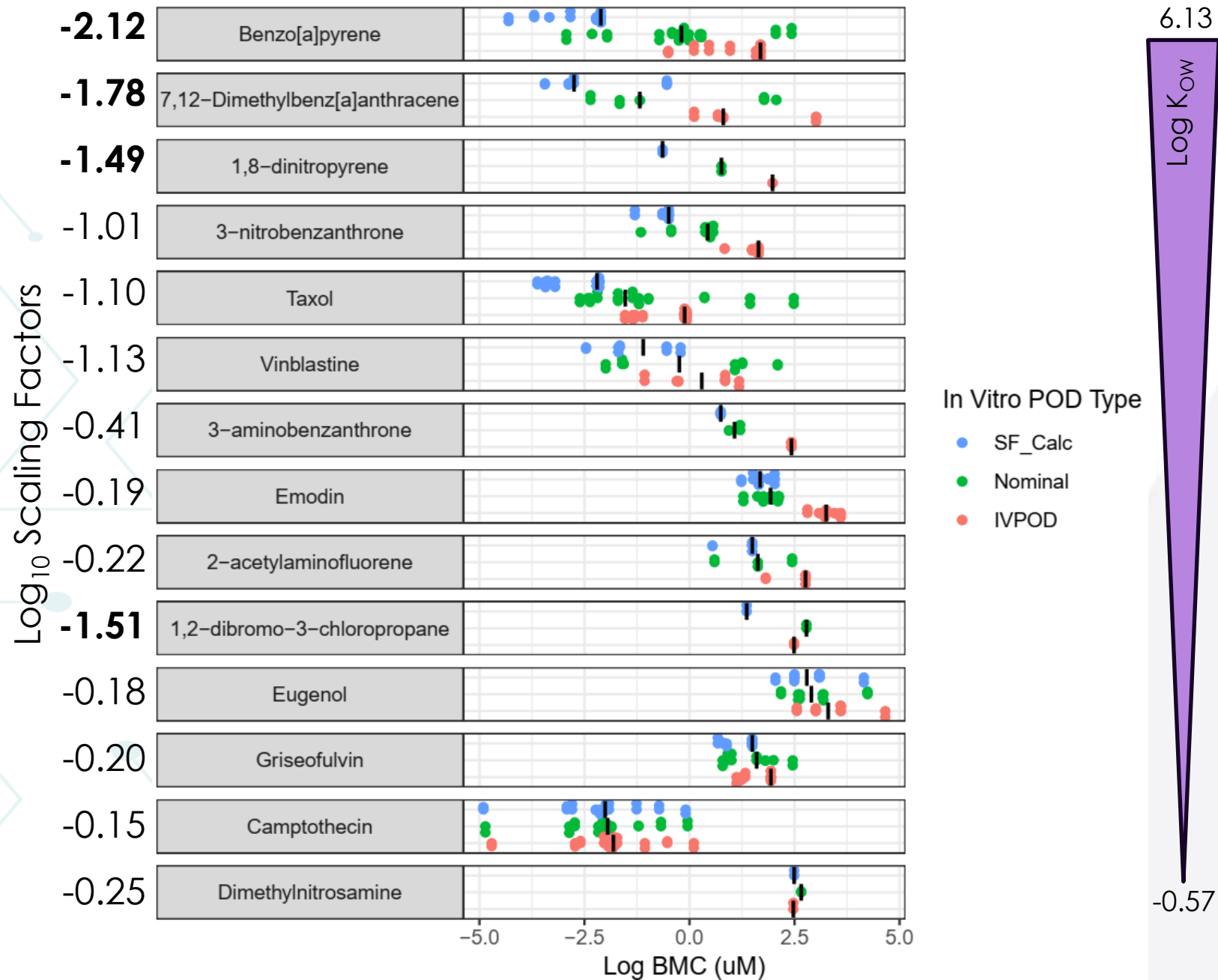
In vitro mass balance model (IV-MBM) to adjust in vitro PODs

Nominal = unadjusted BMC

Scaling Factor (SF) to correct **in vitro bioavailability** reduces the in vitro BMC

Media-plasma bioavailability differences increases the in vitro POD (IVPOD)

Bioavailability in plasma can be greatly reduced in comparison with serum-free or low serum exposure media conditions

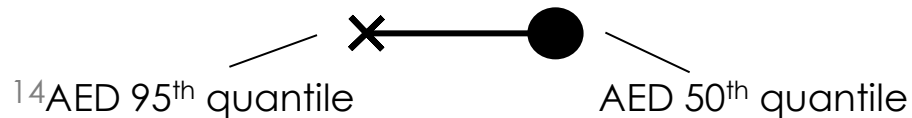


Comparing Case Study Results: Nominal & IV-MBM AEDs

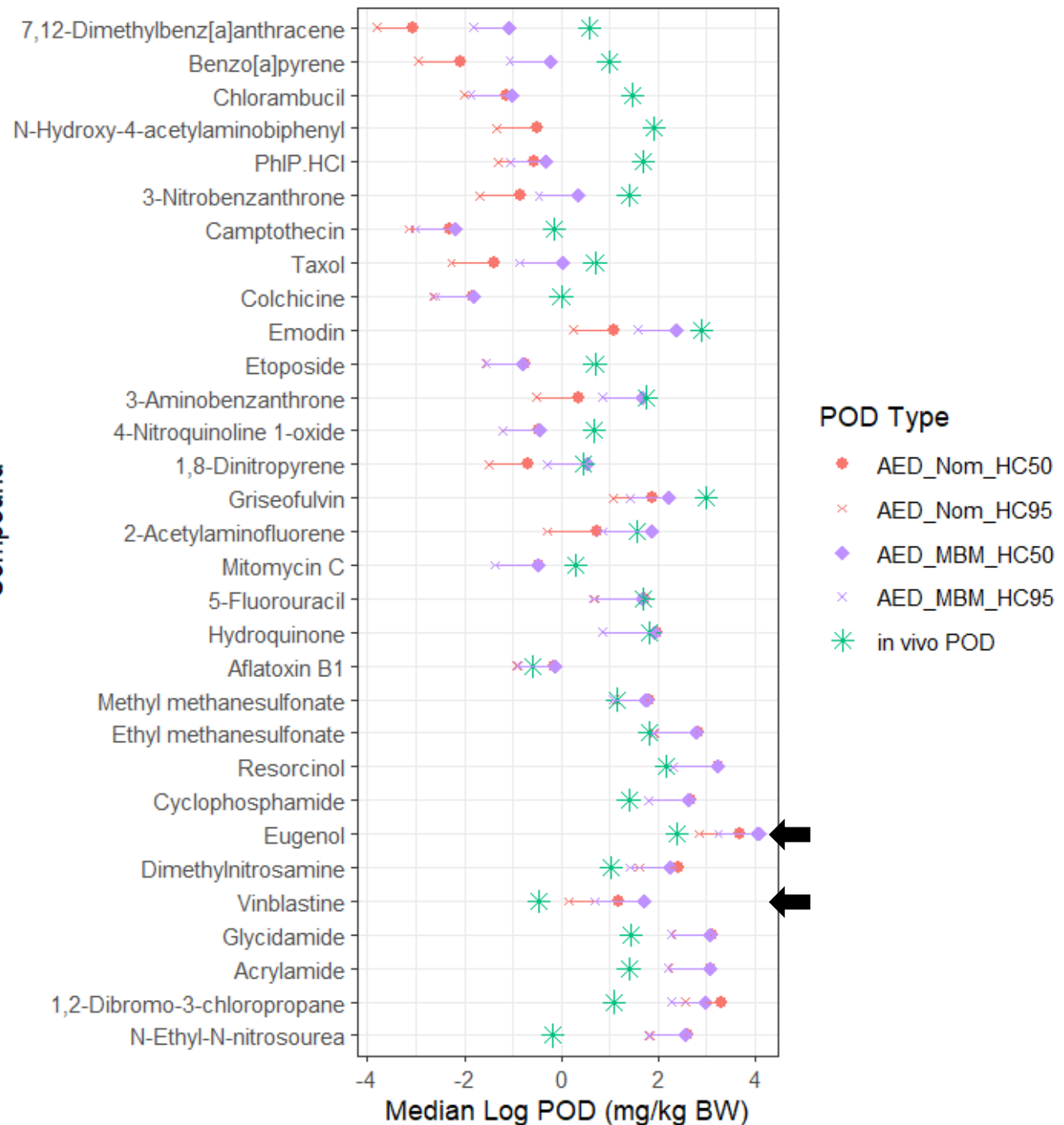
With IV-MBM:

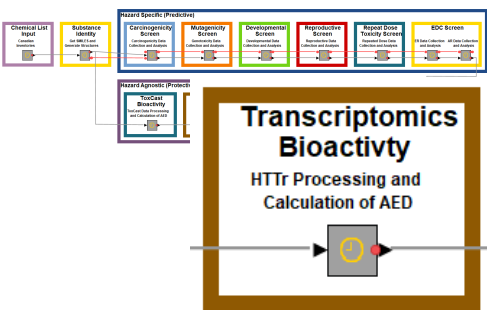
- IV-MBM improved AED estimation when applied to the ORD httk AEDs
 - AEDs closer to in vivo PODs: **11**
 - AEDs same distance from in vivo PODs: 17
 - AEDs farther from in vivo PODs: 2
- IV-MBM tends to have greater impact on compounds with higher LogK_{OW}

30 compounds with full data set



Compound



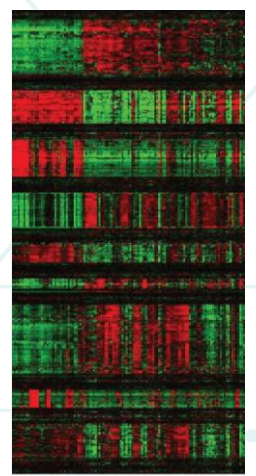


Gaining Experience in Deriving In Vitro Transcriptomic Points of Departure

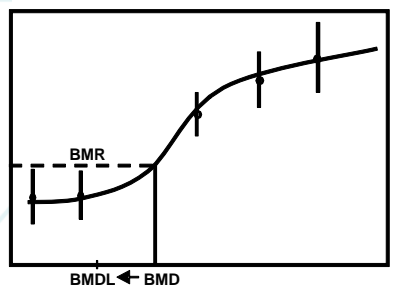
**Data generation:
Gene Lists**

Assay Profile	Accession Number	Gene Symbol	25mg/kg/day		5mg/kg/day		1mg/kg/day	
			ED50 p-value	fold change	ED50 p-value	fold change	ED50 p-value	fold change
A_55_P1963017	NM_00102343	Itih1	0.89	2	0	0	0	0
A_55_P133397	NM_145124	Cttnb1	0.97	1.2	0	0	0	0
A_55_P2114187	XM_884904	LCKCN20315	0.90	1.4	0.05	3.7	0	29.2
A_55_P044703	NM_021443	Cttn	0.46	0	0	0	0	16.7
A_55_P139618	NM_00102343	Itih2	0.59	2.2	0	4.4	0	15.1
A_55_P2087798	NM_008470	Krt16	0.43	2.3	0	7.3	0	13.3
A_55_P398925	NM_173849	Itih2l1	0.91	1.6	0.02	2.8	0	11.4
A_55_P2028499	NM_022800	Ddit4b-ctd	0.96	1.2	0.07	2.3	0	11.4
A_55_P1998451	NM_145211	Cttnb1	0.19	2.4	0	5.1	0	8.9
A_55_P2103098	NM_015783	Itih15	0.53	2.1	0	4.6	0	8.8
A_55_P2151433	NM_009265	Serp3b	0.86	1.8	0	3.7	0	8.5
A_55_P386870	NM_011472	Serp2f	0.94	1.5	0.04	3	0	7.2
A_55_P113996	NM_024987	Itih2	0.53	4.3	0.5	3	0.001	8.8
A_55_P187282	NM_009892	Cttnb3	0.98	1.2	0.84	1.4	0	6.4
A_55_P114206	NM_017796	Ddit3	0.96	1.2	0.3	1.7	0	5.7
A_55_P2062486	NM_00149164	Tnfr2	0.84	2.2	0	4.4	0	5.6
A_55_P128337	NM_015783	Itih15	0.84	1.5	0	2.7	0	3.4
A_55_P1994671	NM_009114	Serp3a	0.57	2.1	0	6.1	0	5.2
A_55_P1978463	NM_021124	Itih-2f	0.52	1.6	0.03	3	0	5.1
A_55_P134662	NM_011474	Serp3b	0.97	1.2	0.85	2.3	0	5
A_55_P304170	NM_023386	Rgs4	0.75	1.6	0	3	0	3
A_55_P2113064	NM_008204	Itih-2d2	0.76	1.7	0.01	2.3	0	3
A_55_P1978628	NM_001644	Serp3a	0.59	1.5	0	2.7	0	4.6
A_55_P2112093	NM_009362	Tnfr1	0.55	3.1	0.04	4.3	0	4.6
A_55_P1994919	NM_023286	Rgs3	1	1	0.41	1.9	0	4.6
A_55_P195372	NM_011477	Serp3b	0.66	1.7	0	2.3	0	4.5
A_55_P194808	NM_008476	Krt6a	0.86	1.4	0.02	3.2	0	4.5
A_55_P101935	NM_010649	Krt6b	0.82	1.5	0.01	2.7	0	4.5
A_55_P1993180	NM_007462	Avp1	0.81	1.7	0	3.3	0	4.5
A_55_P126272	NM_010649	Krt6b	0.82	1.5	0.01	2.9	0	4.5
A_55_P112355	NM_018718	Itih1	0.68	2	0	4.1	0	4.4
A_55_P2021383	NM_009362	Tnfr1	0.5	3.2	0.01	4.5	0.001	4.3
A_55_P2014462	NM_021274	Cxcl10	0.89	1.6	0	3	0	4.3
A_55_P198997	NM_201363	Serp3b-ctd	0.86	2	0	3.1	0	4.3
A_55_P1970144	AK049486	Itih1a14	0.97	1.2	0.17	1.9	0	4.3
A_55_P06883	NM_185026	Ddit3-ctd	0.26	2.1	0	3.6	0	4
A_55_P1963001	NM_013265	Cttnf2	0.9	1.3	0.04	1.8	0	4
A_55_P2014634	NM_00103307	Ndc5	0	2.2	0	2.7	0	3.8
A_55_P0619387	NM_009362	Tnfr1	0.93	1.3	0	2.1	0	3.8
A_55_P114244	XM_00178461	Cttnb14	0.96	1.3	0.05	2.5	0	3.8
A_55_P194824	NM_00101092	Itih1a1	0.73	1.6	0.01	2.4	0	3.7
A_55_P2001748	NM_009126	Serp3b-ctd	0.74	1.5	0	2.3	0	3.6
A_55_P2410304	NM_001146273	Itih1	0.99	1.8	0	3.1	0	3.6
A_55_P2424243	NM_018714	Cttnf1	0.87	1.5	0	2.9	0	3.6

**Extraction:
predictive signatures and
pathways**



**Dose-response
modeling**



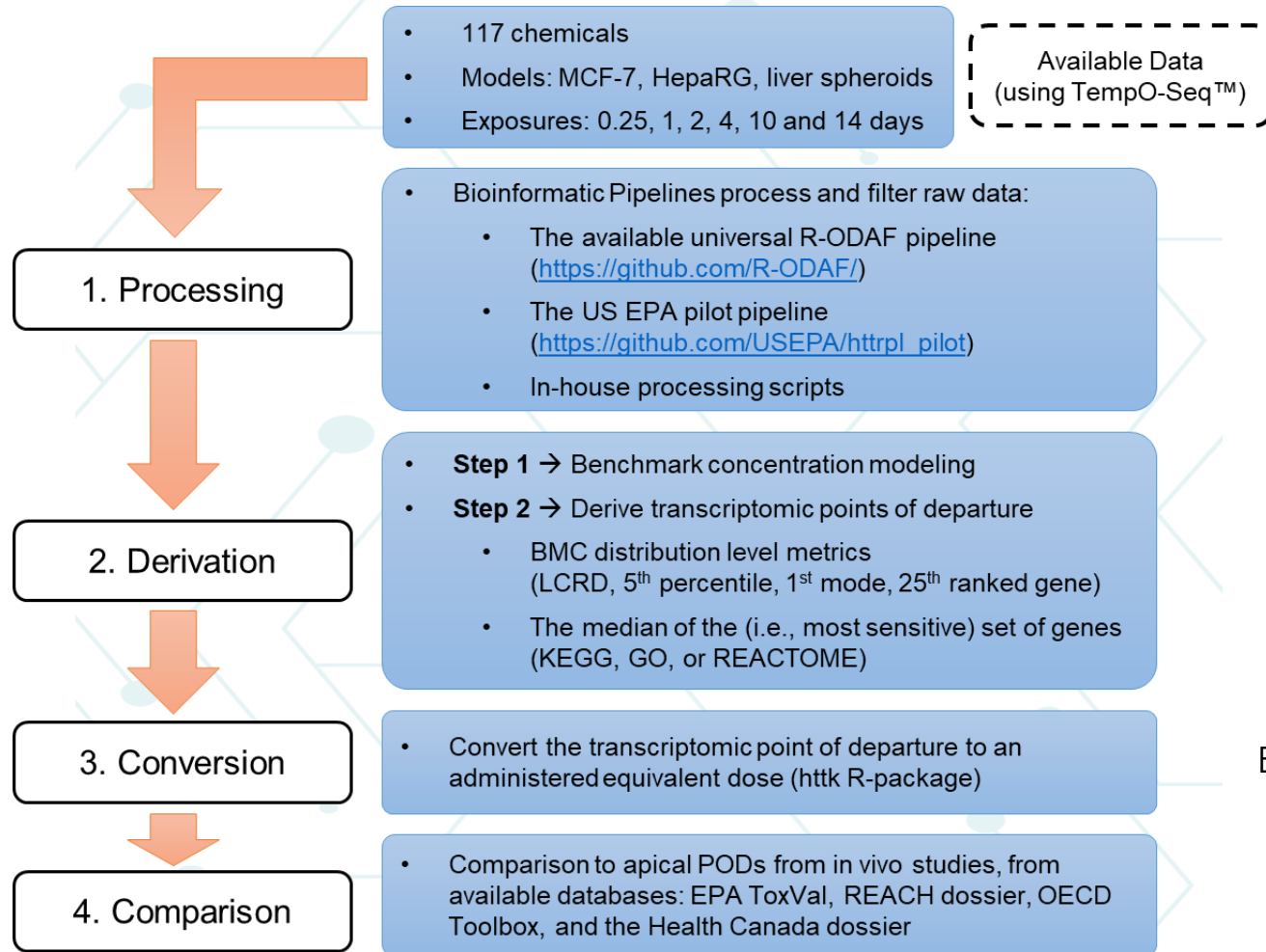
**Establish Points of
Departure (PODs)**

- How do we define the optimal POD?**
- 5th percentile
 - 25th ranked gene
 - 1st mode BMC
 - Lowest median gene set (KEGG, GO, REACTOME)

Thousands of genes per chemical and each exposure

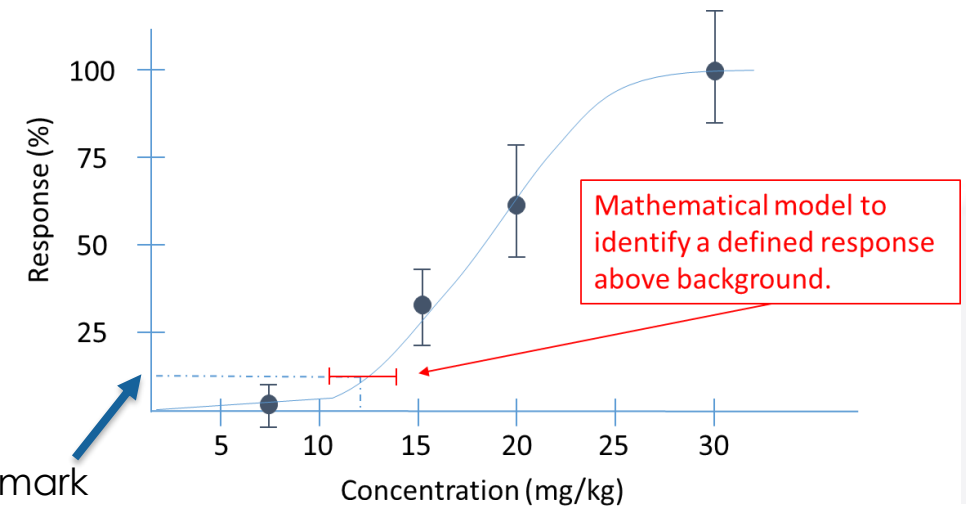
Applied parameters identifying genes with a concentration-response

Evaluating In Vitro Transcriptomic PODs Using a Uniform Workflow: A Meta-analysis of Existing Datasets

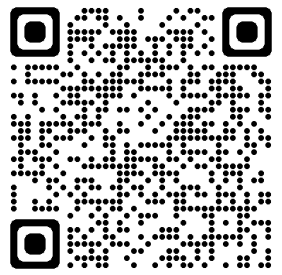


- Applied a uniform analysis across a diverse chemical space using different models and exposure conditions

Benchmark Concentration (BMC) modeling



Benchmark response (BMR)



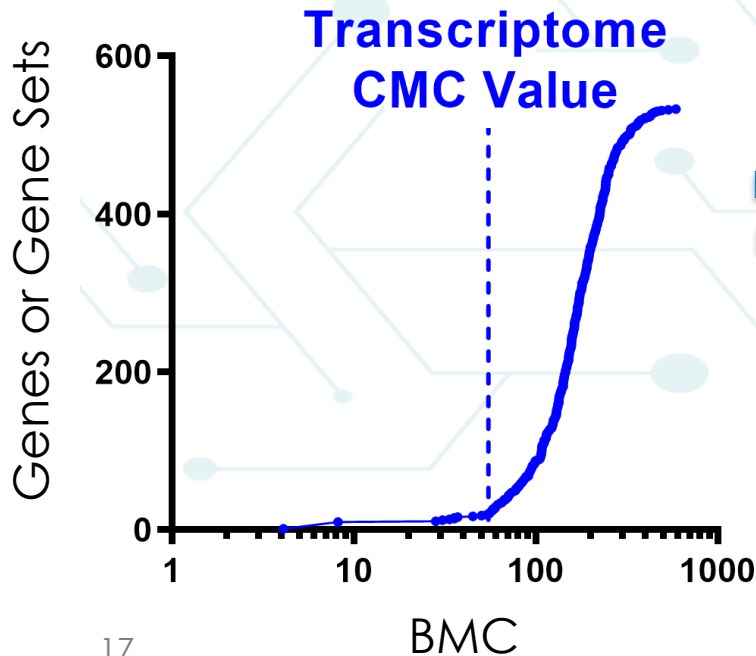
2. Derivation

- **Step 1** → Benchmark concentration modeling
- **Step 2** → Derive transcriptomic points of departure
 - BMC distribution level metrics (LCRD, 5th percentile, 1st mode, 25th ranked gene)
 - The median of the (i.e., most sensitive) set of genes (KEGG, GO, or REACTOME)

- Numerous approaches were considered to derive **tPODs** to define the point of concerted molecular change (**CMC**)

Gene Accumulation Curves

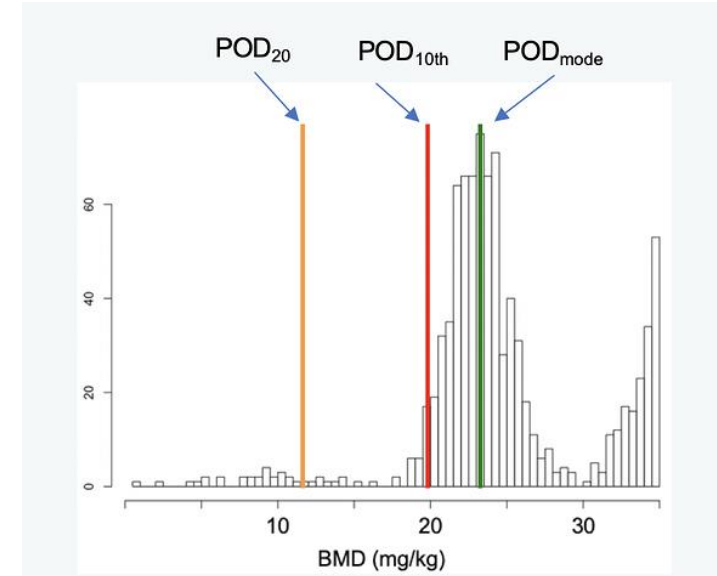
- Each chemical is depicted as a distribution of genes with BMCs to derive tPODs



1 Using distributions of BMCs to derive tPODs

- $tPOD_{20}$ (20th lowest BMD value)
- $tPOD_{10th}$ (10th % of gene BMD values)
- $tPOD_{mode}$ (mode of the first peak)

**We used slightly different metrics*



<https://omicsforum.ca/t/what-is-a-transcriptomic-pod-tpod-and-how-is-it-calculated/195>

2 Using gene sets to derive tPODs

Assign all genes with BMCs to gene sets (eg, pathways)

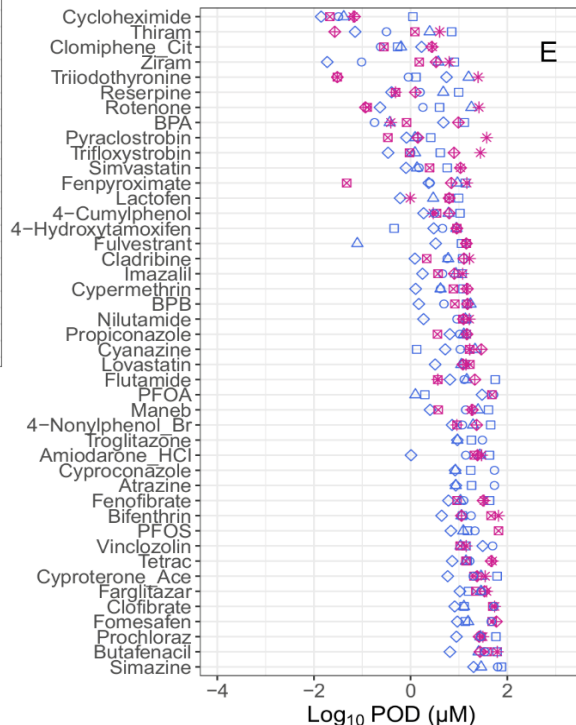
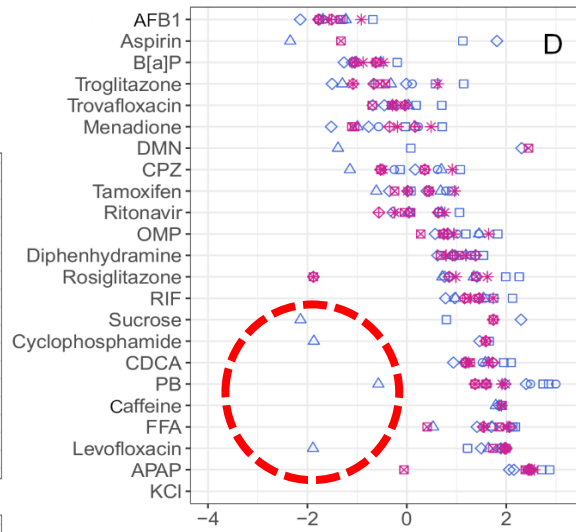
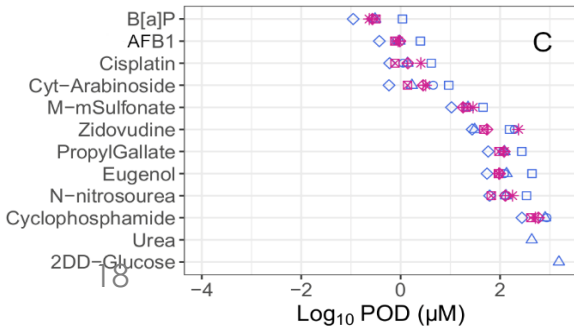
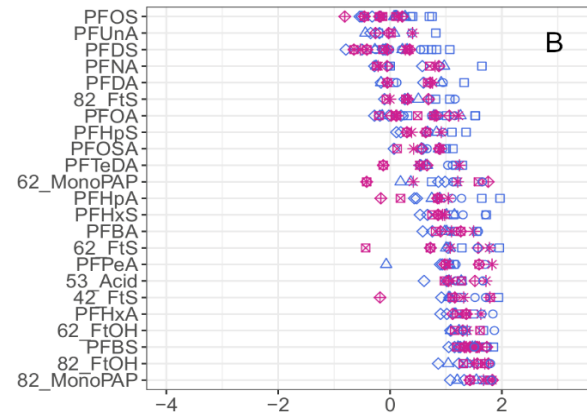
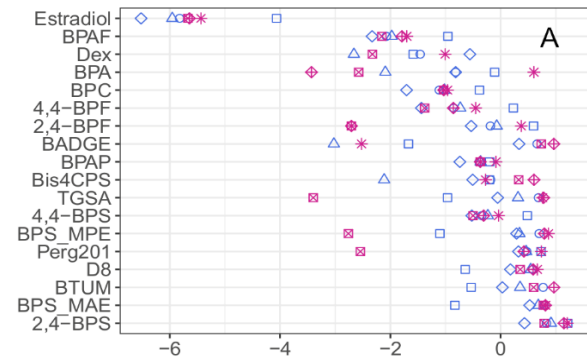
Calculate median gene set BMC

Identify lowest gene set (min. 5% of pathway, 3 genes)

tPOD

A General Comparison of Approaches to Derive tPODs

2. Derivation



- Overall good agreement between data points (robust transcriptomic data)
- Commonly employed approaches may not be reliable
 - Using percentiles (e.g., 5th percentile)
- Other tPODs provide sound alternatives
 - Distribution of single genes (e.g., *the 25th Ranked Gene*)
 - Using gene sets from pathway databases (*lowest and most sensitive gene set*)

- BMC Distribution
- Gene Set (Pathway)
- 1st Mode ○ 25th Gene △ 5th Percentile ◇ LCRD
- ⊠ GO * KEGG ⊠ Reactome

Ratio of In Vitro Derived AEDs to Apical PODs

3. Conversion

- Convert the transcriptomic point of departure to an administered equivalent dose (httk R-package)

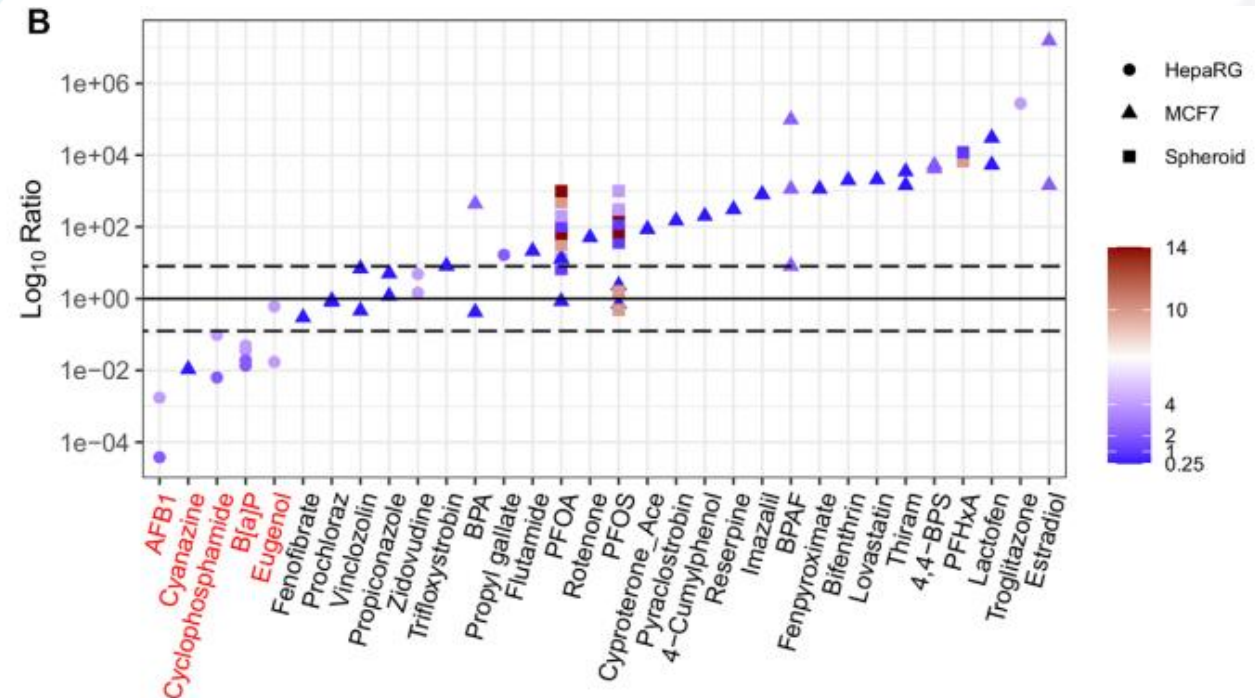
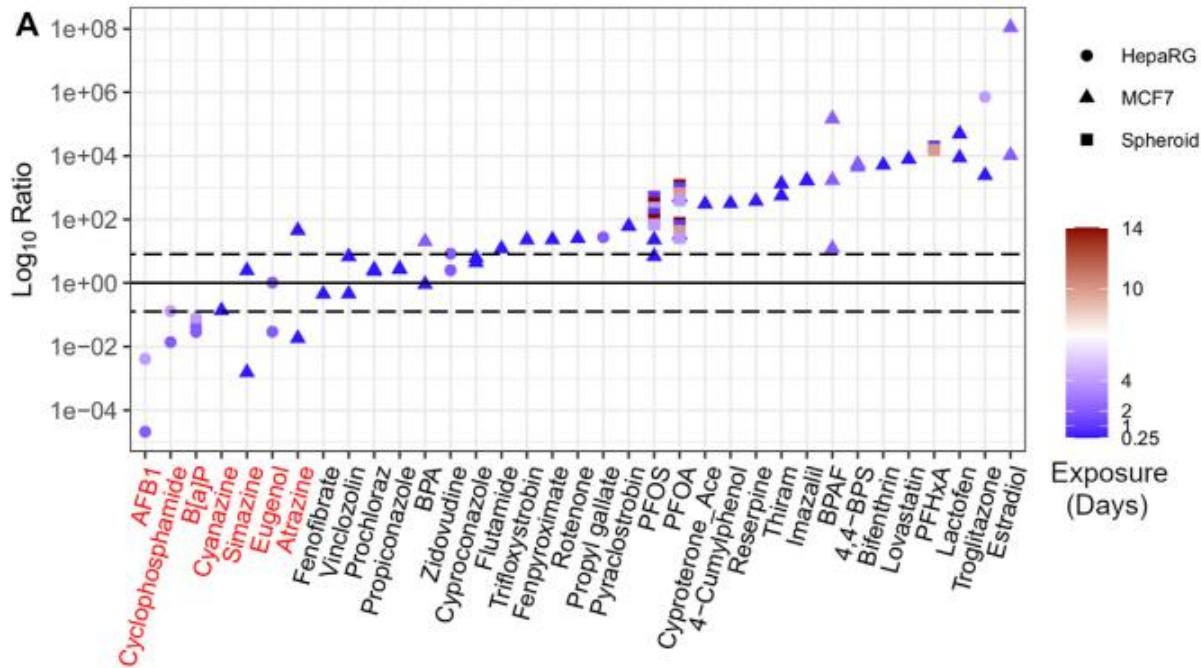
4. Comparison

- Comparison to apical PODs from in vivo studies, from available databases: EPA ToxVal, REACH dossier, OECD Toolbox, and the Health Canada dossier

$$\text{Log}_{10}\text{Ratio} = \text{Log}_{10}\text{POD}_{\text{Traditional}} - \text{Log}_{10}\text{AED}_{\text{NAM}}$$

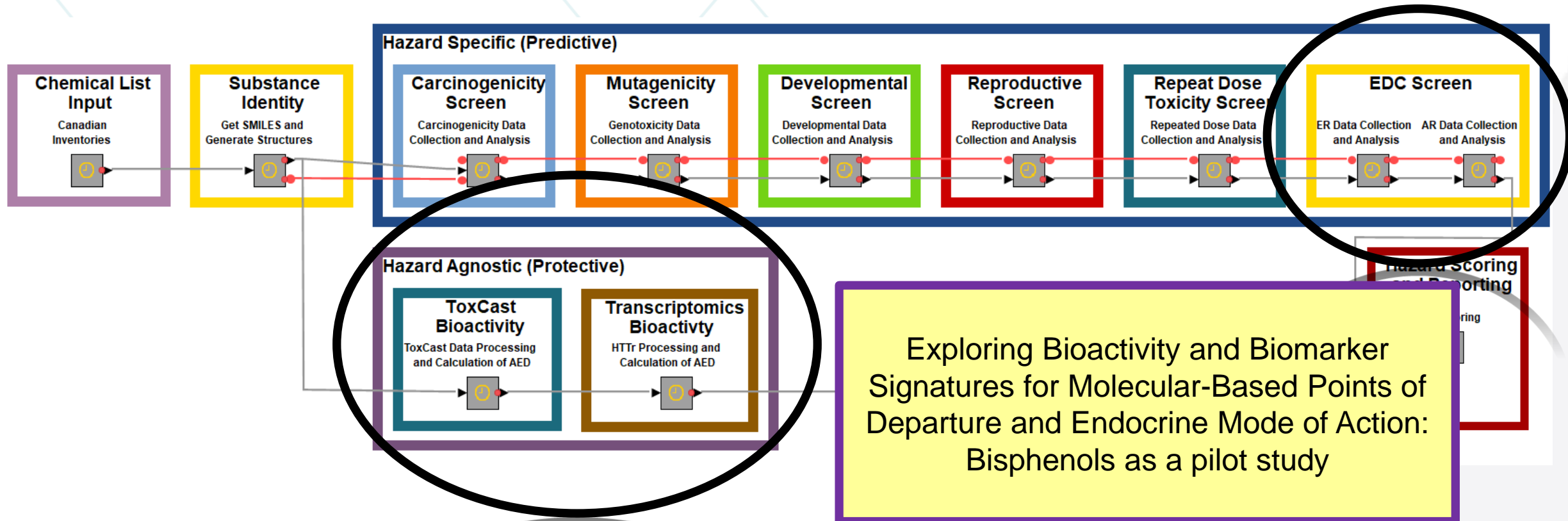
Identifying outliers using ratios:

- Compared the lowest derived AED from BMC Distribution approaches (**A**) or Gene Set approaches (**B**) to apical PODs
- Select chemicals (**highlighted in red**) had apical PODs that were lower than in vitro estimates were flagged



Integrating Data Sources to Support Screening and Assessment

- Integrated Approaches to Testing and Assessment (IATA) represent a flexible framework
 - can include a range of different methods and sources of information;
 - can be assembled in different ways;
 - can be used in different regulatory decision-making contexts depending on assessment questions and protection goals.



Unclassified

English - Or. English

15 December 2022

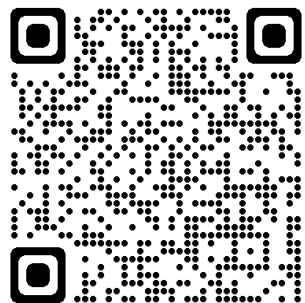
ENVIRONMENT DIRECTORATE
CHEMICALS AND BIOTECHNOLOGY COMMITTEE



Case Study on the use of Integrated Approaches to Testing and Assessment for potential Systemic Toxicity and Estrogen Receptor Activation of a Group of Bisphenols and Select Alternatives



Series on Testing and Assessment
No. 373



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

Toxicological Sciences, 2023, **191(2)**, 266–275

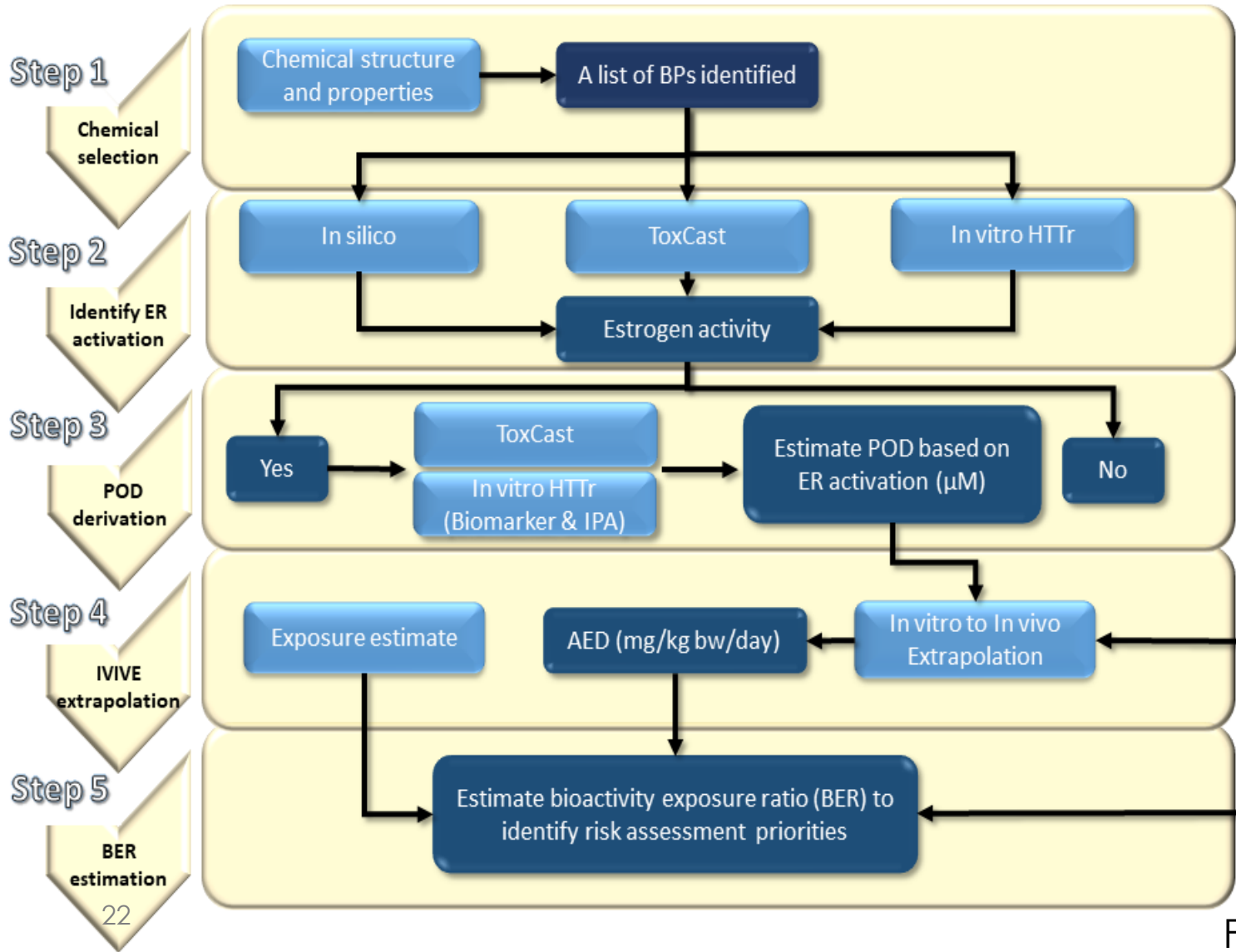
<https://doi.org/10.1093/toxsci/kfac127>
Advance Access Publication Date: 19 December 2022

Research article

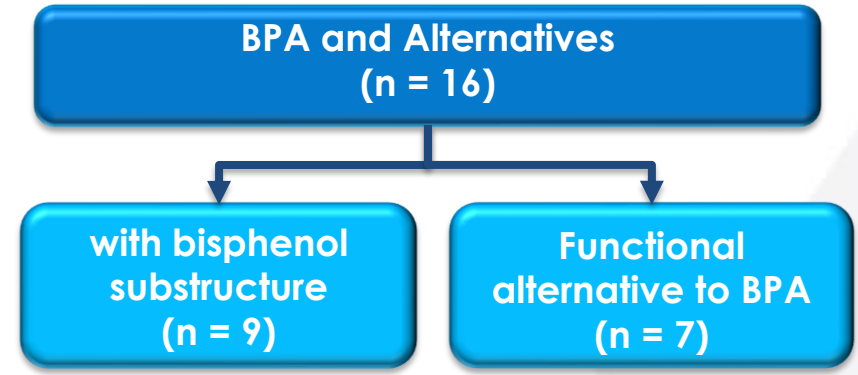
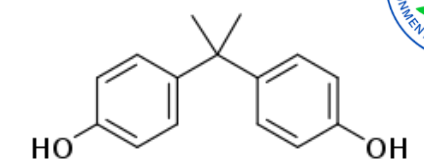
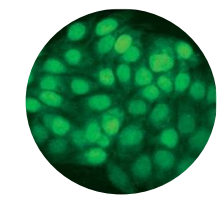
In vitro transcriptomic analyses reveal pathway perturbations, estrogenic activities, and potencies of data-poor BPA alternative chemicals

Geronimo Matteo ^{1,2} Karen Leingartner,^{1,2} Andrea Rowan-Carroll,^{1,2} Matthew Meier,^{1,2} Andrew Williams,^{1,2} Marc A. Beal,^{3,4} Matthew Gagné,⁴ Reza Farmahin,⁴ Shamika Wickramasuriya,⁴ Anthony J. F. Reardon,⁴ Tara Barton-Maclaren,⁴ J. Christopher Corton ⁵ Carole L. Yauk ^{2,*} Ella Atlas ^{1,6,*}

Pilot Example: An Integrated Approach to Testing and Assessment to Evaluate BPA and Select Alternatives



Human MCF-7 Cells



Retrieve tPOD values for all BPs based on non-specific toxicity from Matteo et al. (2023) and ToxCast data

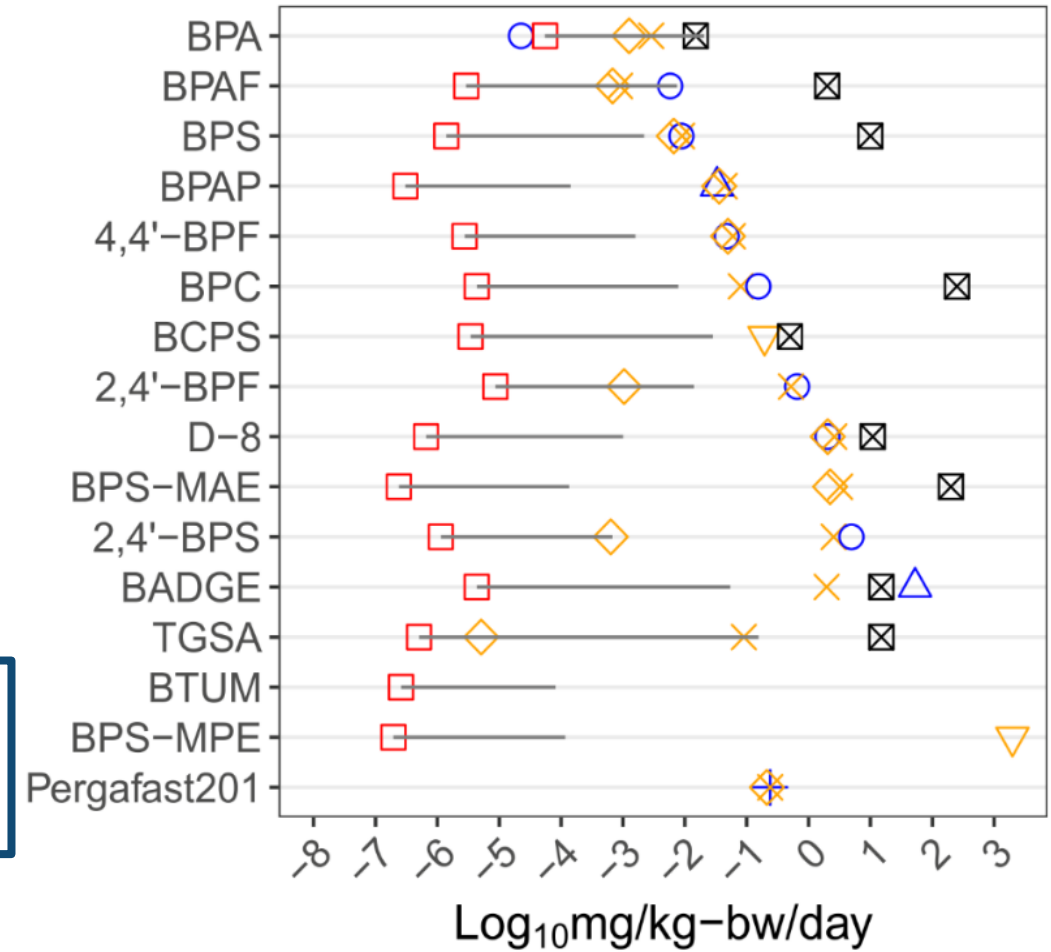
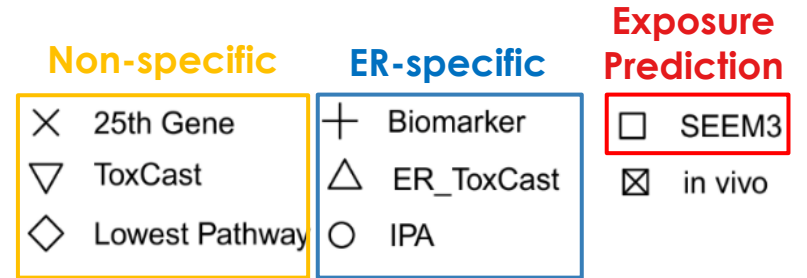
Lowest available in vivo PODs

Creating Practical Estimates for Risk Assessment with Bioactivity Data



- In vitro bioactivity data demonstrated to be robust, with good agreement between approaches
 - 25th ranked gene well aligned
 - lowest gene set overly conservative
- Transcriptomics AEDs for non-specific toxicity produced similar values to ER-specific estimates (when pathway specific values are available)
- Bioactivity based-AEDs typically lower than apical PODs from animal data (RDT, Repro, Dev)

The integration of bioactivity estimates provides practical information on potency and mode of action for hazard and risk characterization



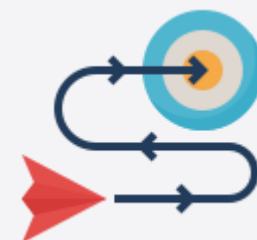
Learnings & Insights

- Data collection, interpretation and integration workflows facilitate the use of new and increasingly complex information
 - Maintain flexibility to update analysis methods and use best available science
 - Transparent, reproducible & efficient evidence-based decisions through integrated knowledge
- The IV-MBM refined the IVIVE approach for genotoxicants, providing more predictive AEDs from NAM data
 - Greater impact for compounds with higher LogKOW / LogKAW
 - Requires assay specific information often difficult to obtain; reduces throughput
- Technical and practical challenges remain
 - Access to curated datasets -> many transcriptomic studies in literature but lack common data formats and repository
 - Experimental design -> cell lines, organoids, MPS, metabolism, testing strategies
 - Standardization -> 'acceptable' practices - consistent, transparent use in risk assessment (e.g. for tPOD – R-ODAF, OORF)
 - Characterizing uncertainty -> including consideration of uncertainty factors
- Much work to be done to address toxicity endpoints of regulatory interest using predictive approaches



Final Thoughts

- Various NAM techniques are currently available and are being used for supporting hazard assessment and prioritization; context of use is important
- The combination of in vitro bioactivity data with IVIVE provides the opportunity to apply NAM-based AEDs in approaches that are protective of human health
 - Currently being targeted as an early tier assessment
 - Ongoing refinement of interpretation approaches and consideration of mechanistic/pathway analysis
- Research-Regulatory collaborations are imperative
- Continue to build a common vision and commitment to advance alternative methods and maintain excellence in science-based decision making



Questions?



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