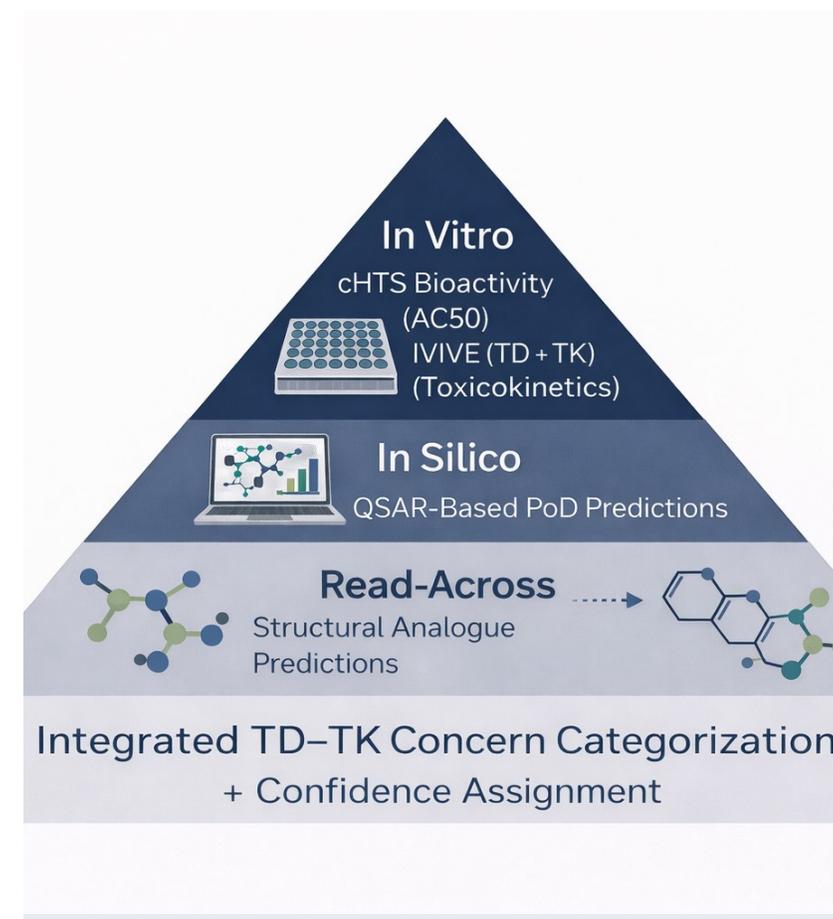


NAM-Based Assessment Framework for Classification of Chemicals for Systemic Toxicity Potential

ASCCT-ESTIV Award Winners Series
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Germany



The EPAA Designathon for Human Systemic Toxicity



The European Partnership
for Alternative Approaches to Animal Testing

Established in 2005, EPAA is a collaboration between the European Union (EU) and industry stakeholders from 9 sectors

Vision: Replacement, reduction and refinement (3Rs) of animal use for meeting regulatory requirements through better and more predictive science

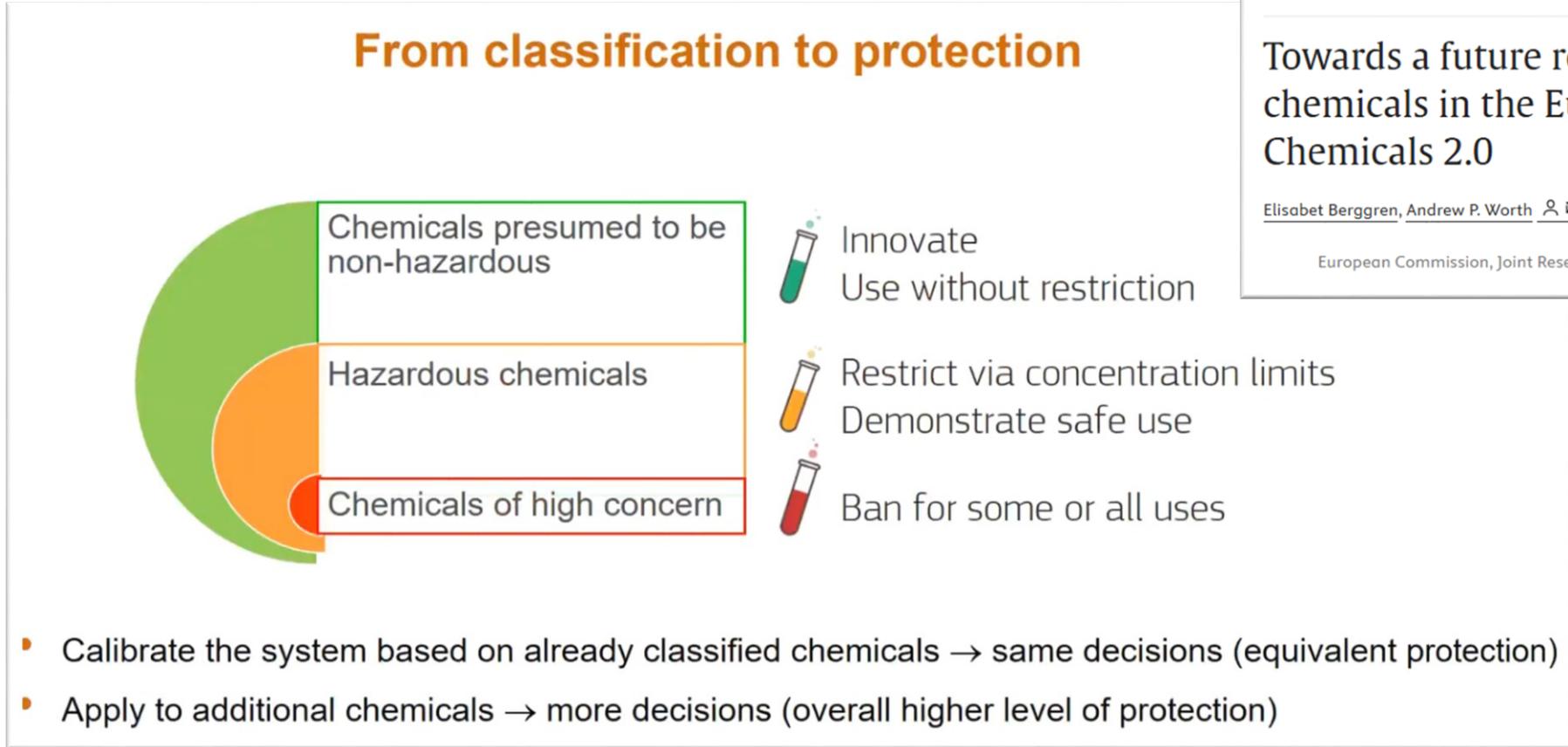


Goal: To assign a level of concern (high, medium and low) for systemic toxicity to each chemical based on “New Approach Methodology” (NAM) data related to their toxicodynamic and toxicokinetic properties

Proposal for a new classification matrix

		Activity (NAM-based toxicodynamics)		
		High	Medium	Low
Potential Systemic Availability (NAM-based toxicokinetics, based on ADME properties)	High	H	H	M
	Medium	H	M	L
	Low	M	L	L

From Classification to Protection



Towards a future regulatory framework for chemicals in the European Union – Chemicals 2.0

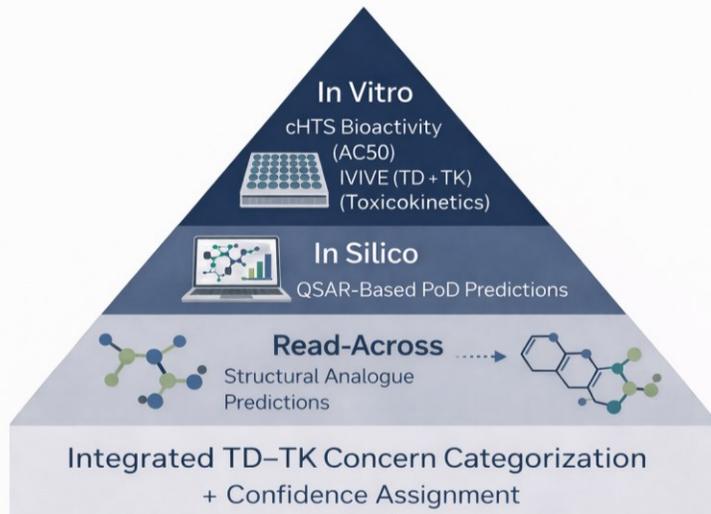
Elisabet Berggren, Andrew P. Worth

European Commission, Joint Research Centre (JRC), Ispra, Italy

NAM-based Assessment Workflow

Proposed Solution

- Simple and transparent stepwise workflow
- Open-access data and tools were used to classify the reference set of chemicals
- Integrating three lines of evidence
 - In vitro – cHTS Data
 - In silico – QSAR predictions
 - Data-gap Filling – Read-across predictions





German Federal Institute for Risk Assessment

NAM-based Assessment to Classify Chemicals for Systemic Toxicity Effects EPAA Designation on NAM-based solutions

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BACKGROUND

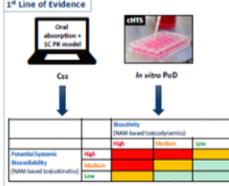
The NAM DESIGNATION is a research exercise aiming to explore a future hazard classification system for chemicals. The current goal is to assign a level of concern (high, medium and low) for systemic toxicity to each chemical based on "New Approach Methodology" (NAM) data related to their toxicodynamic and toxicokinetic properties. Here, we propose a simple, reproducible and transparent workflow using currently available open-access data and tools to classify the reference set of chemicals. The approach is a preliminary "screening-level" assessment which is, at the moment, neither sufficient to perform hazard classification of chemicals nor for any regulatory decision-making. However, the workflow can be used as a starting point for working towards a NAM-based solution in a future "Next Generation Risk Assessment" (NGRA) framework.

DATA & METHODS

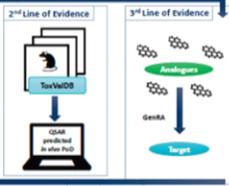
Tool/Model	Description	Estimated Value
Integrated Chemical Environment (ICE) Test (https://www.epa.gov/ice-test)	ICE is a user-friendly platform for accessing curated ToxCast/Tox21 (AC ₅₀) NAM data and computational tools. The Pharmacokinetics (PK) tool predicts tissue-level concentrations resulting from in vivo doses (C _{ss}).	Hazard/potency estimate (TD) The fifth percentile (P ₀₅) based on the distribution of all available AC ₅₀ values for assay endpoints relevant to systemic (repeated dose) toxicity. Systemic bioavailability estimate (TK) Modelled steady-state plasma concentrations (C _{ss}).
OECD QSAR Toolbox (https://www.oecd.org/chemicalsafety/nha-assessment/oecd-qsar-toolbox.htm)	The OECD toolbox incorporates information and tools from various sources into a logical workflow. Oral absorption profiler predicts oral absorption (high, medium and low).	Systemic bioavailability estimate (TK) Adjustment of C _{ss} (100%, 50% or 20%) based on estimated oral absorption (high, medium or low).
Generalised Read-across Framework (GenRA) (https://www.epa.gov/compton-toxic-generalised-read-across-genra)	The generalised read-across framework (GenRA) as an algorithmic approach to read-across to predict analogues for chemicals (targets) lacking in vitro data.	Hazard/potency estimate (TD) In vitro data from five analogues were used to get a read-across estimate of a PoD for the target chemical (toxicity or in vitro data-weighted average).
QSAR In Vivo Points of Departure (PoD) Model (https://doi.org/10.1016/j.comtox.2020.100139)	Quantitative structure-activity relationship (QSAR) model to predict quantitative points of departure (PoD) for in vivo repeated dose toxicity.	Hazard/potency estimate (TD) Prediction for in vivo PoDs expressed as external oral dose values in mg/kg/day.

Table 1. A brief description of the tools and models employed in workflow and the relevant predictions obtained from each of them.

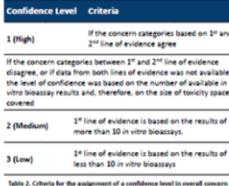
1st Line of Evidence



2nd Line of Evidence



3rd Line of Evidence



Confidence Level	Criteria
1 (High)	If the concern categories based on 1 st and 2 nd line of evidence agree If the concern categories between 1 st and 2 nd line of evidence disagree, or if data from both lines of evidence was not available, the level of confidence was based on the number of available in vitro bioassay results and, therefore, on the size of toxicity space covered
2 (Medium)	1 st line of evidence is based on the results of more than 10 in vitro bioassays.
3 (Low)	1 st line of evidence is based on the results of less than 10 in vitro bioassays.

Table 2. Criteria for the assignment of a confidence level in overall concern categories for each substance in the reference set of chemicals.

RESULTS

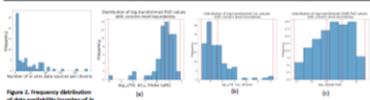


Figure 1. Frequency distribution of data availability for all reference chemicals. The x-axis represents the number of data points (0-100) and the y-axis represents the frequency (0-1000).

LIMITATIONS AND NEXT STEPS

Method	Problem	Possible Solution
In vitro data	Insufficient coverage of toxicological space related to human systemic toxicity. Uncertainty related to the individual assay results, with regard to human response.	Establishing an AOP network to identify critical toxic events or key biochemical and corresponding in vitro bioassays with high relevance to the individual assay results. Developing a standard test battery (MKNM) for systemic toxicity as a first tier testing strategy.
In silico	Insufficient coverage for metabolites. Concern categories do not provide information on the mechanistic effect/target organ.	Using metabolically competent primary human cell culture. Reporting mechanistic information in addition to the potency assessment to allow further assessments, e.g. in vitro.
GenRA	Limited toxicokinetics spectrum covered.	Including more substances to broaden the toxicity spectrum and refine the categorization.
QSAR in vivo PoD	Uncertainty in underlying animal data. Chemical predictions limited by the relevant chemical descriptors.	Developing new models on better curated data.
GenRA	Lack of relevant analogues. Limited data availability for relevant analogues.	Identifying new analogues. Applying more advanced PK models.
PK model	IC model may not be equally suitable for all chemicals.	Adding in vitro tests to the standard fit the testing strategy. Validating the applicability domain of the in vitro prediction.
In silico	Limited coverage of chemical space (chemical specific, e.g. 10, 10, 10) information not available for all reference substances.	Improving the prediction (adding more data). Including in silico in vitro bioassay models (SIT, etc.) mechanistic model in the testing strategy.
Overall	Low regulatory acceptance.	Validating TD and TK model components. Establishing human relevance. Conducting more case studies and uncertainty assessment.

Table 3. Summary of the total number of substances categorized using the workflow described above.

Table 4. A brief description of key general and workflow related limitations along with possible solutions.

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Data Sources and Computational Tools

Tool Model	Line of Evidence	Description	Estimated Quantity
Integrated Chemical Environment (ICE) Tool (https://ice.ntp.niehs.nih.gov/) A curated NTP/NICEATM resource integrating in vivo/in vitro data, chemical properties, predicted bioactivity, and computational tools (including ToxCast/Tox21) to support chemical safety assessment	In Vitro	IVIVE Module: In Vitro to In Vivo Extrapolation tool that translates in vitro assay activity concentrations into predicted in vivo equivalent doses using pharmacokinetic modeling	Hazard/potency estimate (TD) The fifth percentile (TD05) from available AC50 distribution for systemic (repeated-dose) endpoints Systemic bioavailability estimate (TK) Modelled steady-state plasma concentrations (C_{ss})
	Read-across	Chemical Quest Module: Similarity search tool using chemical structure fingerprints to identify similar chemicals with more data	TD and TK In vitro data from analogues were used to get a read-across TD and TK estimate for the target chemicals lacking in vitro data
QSAR In Vivo Points of Departure (PoD) Model (https://doi.org/10.1016/j.comtox.2020.100139) A structure-based QSAR model trained on in vivo repeat-dose toxicity data for ~3,500 chemicals (U.S. EPA's ToxValDB)	QSAR	QSAR model to predict quantitative PoDs (mg/kg/day) with confidence intervals to support hazard assessment in absence of experimental data	TD Prediction for in vivo PoDs expressed as external oral dose values in mg/kg/day.

Line of Evidence 1: In Vitro Bioactivity and IVIVE

Purpose

Support hazard estimation using bioactivity and IVIVE data

Toxicodynamics (TD)

Value: In vitro bioactivity data, AC_{50}

Source: Tox21 and ToxCast data

Focus: Assays relevant to systemic (repeated-dose) toxicity aggregated by mode of action

Data handling:

- Compiled all available AC_{50} values
- Calculated 5th percentile (TD05) of distribution as a conservative Point of Departure (PoD) for hazard estimation

Toxicokinetics (TK)

Value: Steady-state plasma concentration (C_{ss})

Assumptions:

- Model: 1-compartment (1C) population-based PK model from htk R package (implemented in ICE)
- Dose: 1 mg/kg/day, oral
- Species: Human
- Body Weight: 70.0 kg
- ADME Source: Default

The IVIVE tool uses pharmacokinetic models to predict the equivalent administered dose (EAD) from the activity concentration of selected assays. Integrated Chemical Environment

Run Reset

In Vitro Endpoint: AC_{50} , Species: human, Body Mass: 70.0, ADME Source: Default, Model: 1C, Exposure Route: NA

Exposure Route: NA

Exposure Interval, hours: [input field]

Exposure Length, hours: [input field]

Simulation Length, days: [input field]

Inhalation Dosing Method: Concentration

Inhalation Dosing Units: ppmv

httk: R Package for High-Throughput Toxicokinetics

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Affiliations — collapse

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² Division of the National Toxicology Program National Institute of Environmental Health Sciences 111 T.W. Alexander Dr., ML 62-17 Research Triangle Park, NC 27709, United States of America URL: <http://www.niehs.nih.gov/research/abrtiehs/labs/bmsb/>

The 1C model predicts the steady state plasma concentration from a single daily bolus dose without differentiation between exposure route. EADs calculated with the 1C model include an adjustment for fraction unbound. For details see User Guide [?].

Chemical Input

Select Chemicals

Quick List CASRNs

User Chemical Identifiers

23564-05-8
149961-52-4
139528-85-1
95-80-7
40722-80-3
17265-14-4
96-29-7
569-64-2
25637-99-4
33996-33-7
1303-86-2
693-23-2
77-90-7
101-21-3
63-25-2
13171-21-6
1309-84-4
25956-17-6
96.1P.4

Data Input

Select Assays

Assays	Description	Data Type
<input checked="" type="checkbox"/> DART - p53 Signaling Pathway	cHTS	in vitro
<input checked="" type="checkbox"/> DART - Cellular Response to Oxidative Stress	cHTS	in vitro
<input checked="" type="checkbox"/> DART - Cellular Response to Stress	cHTS	in vitro
<input checked="" type="checkbox"/> DART - Regulation of Hormone Levels	cHTS	in vitro

Upload Custom Phys Chem Data

Select Assays

cHTS Mode of Action

Mode of Action

- Acute Lethality MOAs
- Endocrine MOAs
- Cancer MOAs
- Cardiotoxicity MOAs
- DART MOAs

File Name: [input field] MIME Type: [input field]

Line of Evidence 2: QSAR-based In Silico Predictions

Purpose

Support the assessment with an additional line of evidence

Model Basis

- Approach: In silico QSAR models for repeated-dose toxicity
- Training data: Large in vivo legacy dataset for ~3500 chemicals from US EPAs Toxicity Value database (ToxValDB)
- Output: Predicted in vivo PoDs (external oral dose, mg/kg/day) and a confidence interval

Conceptual Basis

- Predicted PoDs reflect both toxicodynamics (TD) and toxicokinetics (TK)
- Directly used for concern category assignment

ELSEVIER

Computational Toxicology
Volume 16, November 2020, 100139

Structure-based QSAR models to predict repeat dose toxicity points of departure

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Line of Evidence 3: Read-across for Data Gap Filling

Purpose

Address data gaps for chemicals lacking relevant in vitro AC50 data

Analog Identification

- ICE - Chemical Quest Tool
- Structural similarity search using Saagar fingerprints
- Similarity criterion: Tanimoto coefficient ≥ 0.7

Data Retrieval & Application

- Identified analogs routed to IVIVE functionality for in vitro data retrieval

Data-gap Filling

- Mean AC50 and C_{ss} values used for read-across to fill target chemical data gaps for TD and TK assessment

The Chemical Quest tool uses fingerprints to calculate structure similarity.

This tool uses fingerprints generated using Saagar features. Only 100 input chemical ids/structures allowed at a time.

Run Reset Search Custom Chemical List

Max hits per input: 10 Tanimoto Coefficient: 0.7 or greater

Chemical ID input (one per line):

- 1309-64-4
- 69-72-7
- 96-26-4
- 2797-51-5
- 135861-56-2
- 106-92-3
- 951659-40-8
- 5343-92-0
- 526-94-3
- 124-30-1
- 61-90-5

Smiles Structures for similarity search

Draw Enter

Chemical Quest Results

Send filtered results to: Select tool... Clear Filter

- Search
- Curve Surfer
- PBPK
- IVIVE
- Chem Characterization
- Copy CASRNs
- Copy DTXSIDs
- Copy SMILES
- Copy Qsar-Ready SMILES

Chemical Name: Pi
CASRN: 98-79-3
DTXSID: DTXSID604
Tanimoto: top 10 hi
Hit Count: 10
Passed Filter(s): 10
Selected Item(s): 0
View Results

Saagar—A New, Extensible Set of Molecular Substructures for QSAR/QSPR and Read-Across Predictions

Alexander Y Sedykh¹, Ruchir R Shah¹, Nicole C Kleinstreuer², Scott S Auerbach², Vijay K Gombur¹

Affiliations — collapse

Affiliations

- 1 Scioime LLC, Research Triangle Park, North Carolina 27709, United States.
- 2 National Institute of Environmental Health Sciences (NIEHS), National Toxicology Program (NTP), Research Triangle Park, North Carolina 27709, United States.

TD–TK Concern Level Categorization

Assumptions

- Log10-transformed TD, TK, and QSAR PoD values
- Equal number of reference chemicals per concern category

Proposal for a new classification matrix

		Activity (NAM-based toxicodynamics)		
		High	Medium	Low
Potential Systemic Availability (NAM-based toxicokinetics, based on ADME properties)	High	H	H	M
	Medium	H	M	L
	Low	M	L	L

Concern Categorization

• In Vitro

TD → AC50 5th percentile

TK → C_{ss} (95th percentile)

Log10 → tertiles → High/Medium/Low

• QSAR

Predicted in vivo PoD (mg/kg/day)

Log10 → tertiles → High/Medium/Low

• Read-across

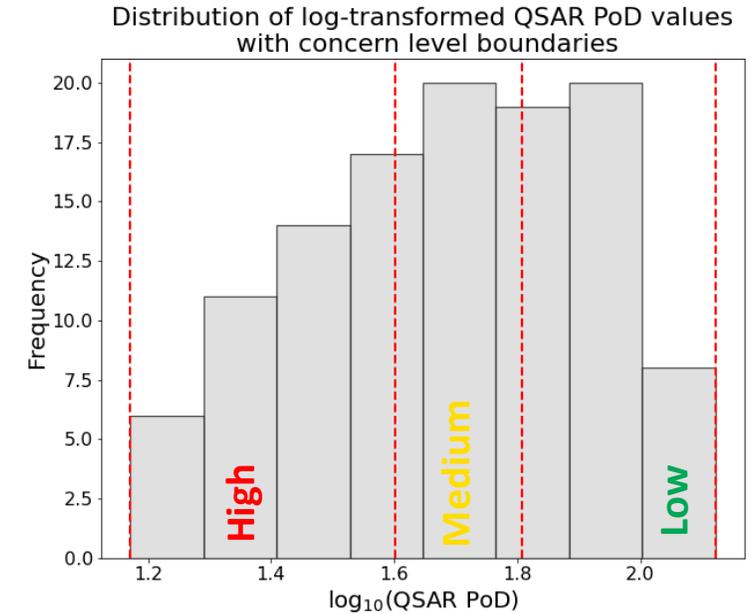
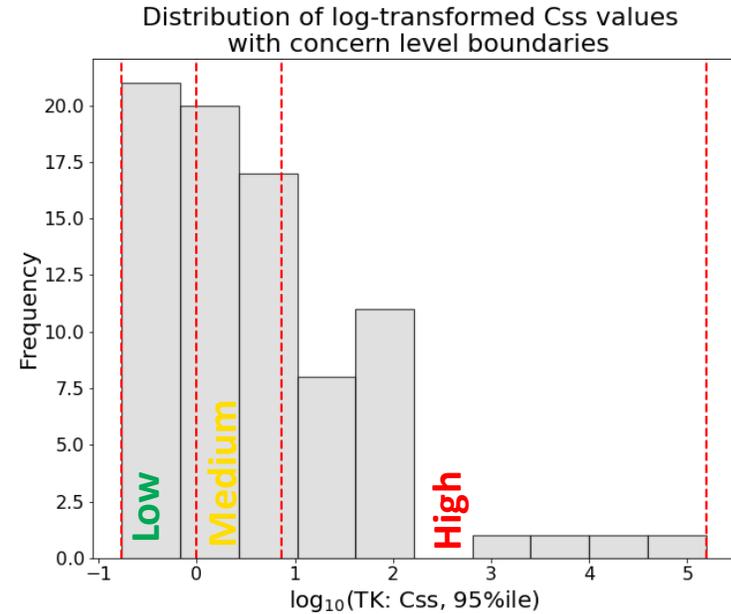
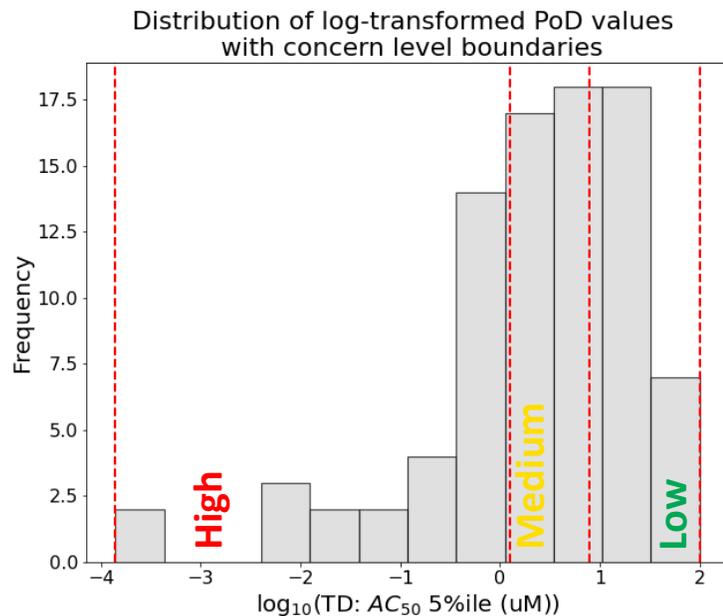
TD → Analog-based TD05

TK → Analog-based C_{ss} (95th percentile)

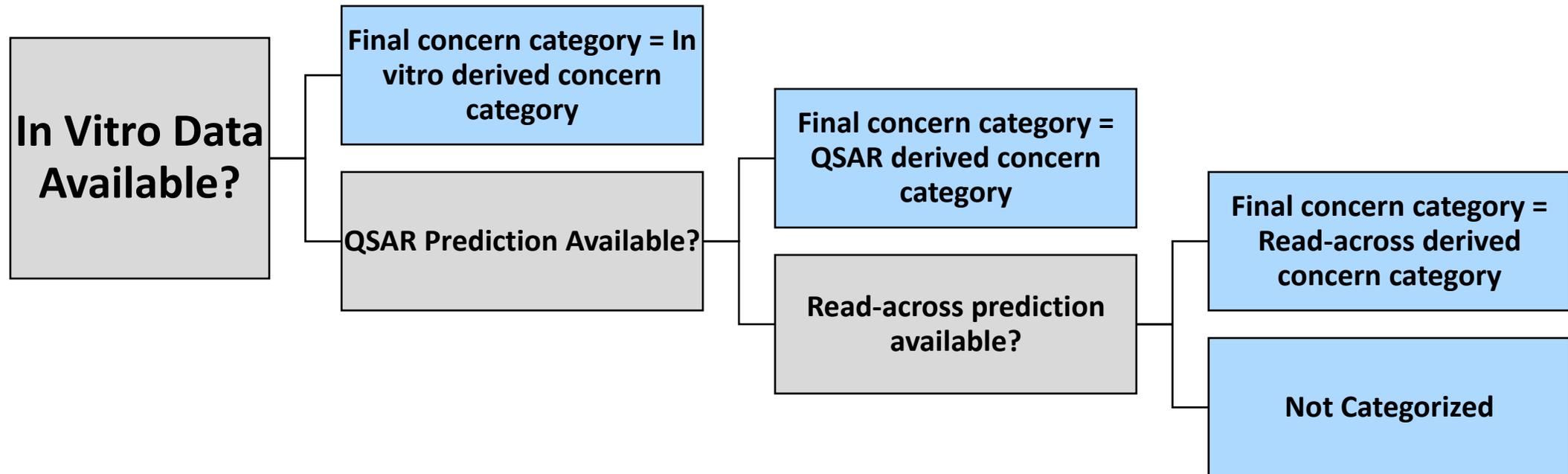
Log10 → tertiles → High/Medium/Low

Concern Category Assignment Framework

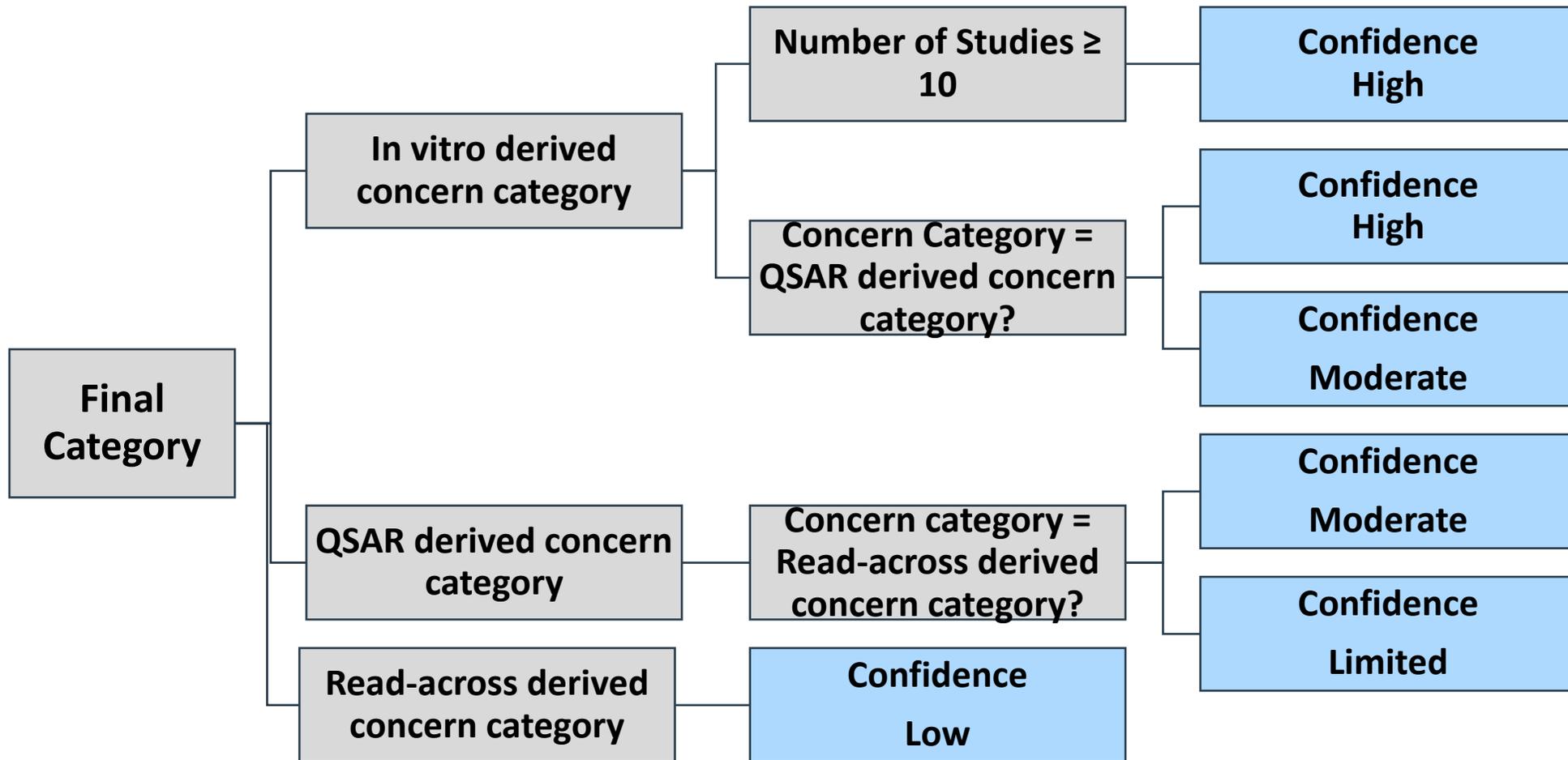
Assuming an equal number of reference chemicals in each concern category, the distribution of the calculated hazard values was used to set the range for potency (cut-off values: 33rd percentile, 66th percentile) for each level



Concern Category Integration Framework



Confidence Assignment Framework



Concern Category Distribution and Concordance Analysis

Assessments	Count (Percentage)
Total number of chemicals assessed	123/150 (82%)
No. of chemicals assessed in the 1 st line of evidence (In vitro)	83/150 (55%)
No. of chemicals assessed in the 2 nd line of evidence (QSAR)	115/150 (77%)
No. of chemicals assessed in the 3 rd line of evidence (Read-across)	13/67 (19%)
No. of chemicals assessed in both, 1 st and 2 nd lines of evidence	75/150 (50%)
Concern categories agree between the 1 st and the 2 nd line of evidence	28/76 (37%)
No. of chemicals categorized as high concern	46/123 (37%)
No. of chemicals categorized as medium concern	40/123 (33%)
No. of chemicals categorized as low concern	37/123 (30%)
Concern category matches with current C&L classification	39/123 (32%)
Concern category less or more protective	Less: 37/123 (30%) More: 43/123 (35%)

Summary

- The proposed framework provides a structured, NAM-based screening/categorization approach that enables systematic prioritization of chemicals
- It integrates in vitro bioactivity, IVIVE-derived data, QSAR predictions, and read-across within a transparent decision logic and confidence framework

However,

- Not all reference chemicals could be categorized due to limitations in availability of relevant data
- While there are technical limitations (regarding tools) that can be addressed in the future, the relevance of the data needs to be demonstrated
- Improvements are needed for various aspects both on the TD and TK data/tools

Limitations for NAM-Based Regulatory Screening

Method	Problem	Possible Solution
Toxicodynamic properties		
<i>In vitro</i> bioactivity data	<ul style="list-style-type: none"> Insufficient coverage of toxicological space related to human systemic toxicity Uncertainty related to the individual assay results with respect to human relevance 	<ul style="list-style-type: none"> Establishing an AOP network to identify central key events or key characteristics and corresponding <i>in vitro</i> bioactivity tests Developing a standard test battery (IATA/DA) for systemic toxicity as a first-tier testing strategy
	Insufficient coverage for metabolites	<ul style="list-style-type: none"> Using metabolically competent primary human cell culture
	Concern categories do not provide information on the MoA/toxicological effect/target organ	<ul style="list-style-type: none"> Reporting mechanistic information in addition to the potency assessment to allow further assessments, e.g. mixtures
	Limited toxicity/potency spectrum covered	<ul style="list-style-type: none"> Including more substances to broaden the toxicity spectrum and refine the categorization
QSAR <i>in vivo</i> PoD prediction	<ul style="list-style-type: none"> Variability in underlying animal data Chemical predictions limited by the relevant chemical descriptors 	<ul style="list-style-type: none"> Developing new models on better-curated data
Read-across	<ul style="list-style-type: none"> Lack of relevant analogues Limited data availability for relevant analogues 	

Limitations for NAM-Based Regulatory Screening

Method	Problem	Possible Solution
Toxicokinetic properties		
PK model	<ul style="list-style-type: none"> 1C model may not be equally suitable for all chemicals 	<ul style="list-style-type: none"> Applying more advanced PBK models
	<ul style="list-style-type: none"> Limited coverage of chemical space (chemical-specific (e.g. fu, CLint) information not available for all reference substances) 	<ul style="list-style-type: none"> Adding <i>in vitro</i> TK tests to the standard first-tier testing strategy Extending the applicability domain of the <i>in silico</i> prediction
Oral absorption profiler	<ul style="list-style-type: none"> Limited coverage of chemical space (chemical properties are outside of the threshold values for parametric boundaries; lack of data) 	<ul style="list-style-type: none"> Improving the prediction (adding more data?)
	<ul style="list-style-type: none"> Not all routes of exposure covered 	<ul style="list-style-type: none"> Including <i>in vitro/in silico</i> barrier models (GIT, skin, respiratory tract) in the testing strategy
Overall	<ul style="list-style-type: none"> Low regulatory acceptance 	<ul style="list-style-type: none"> Validating TD and TK method components Establishing human relevance Conducting more case studies and uncertainty assessment

THANK YOU

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