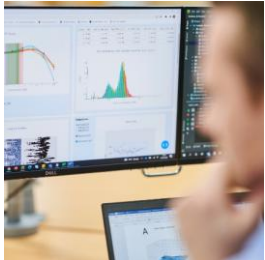


SARA-ICE A Model for Predicting a Human Relevant Point-of-Departure for Skin Sensitisation Risk Assessment

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SERS
Safety, Environmental
& Regulatory Science



A new tool for skin allergy risk assessment

Skin Allergy Risk Assessment – Integrated Chemical Environment (SARA-ICE) is a probabilistic model which has been developed into;

- a defined approach (DA) for point of departure (PoD) determination
- an extended model tool for flexible application in risk assessment and/or hazard classification

A quick tour:

- Development history
- Features and model structure of the SARA-ICE tools
- Evaluation of SARA-ICE at OECD
- Application using the publicly available user interface

Development Timeline of Skin Allergy Risk Assessment Models

2017-2019

A prototype Bayesian statistical model was developed at Unilever to estimate a no-effect-dose from HPPT data. This model was published in 2019.

2019-2021

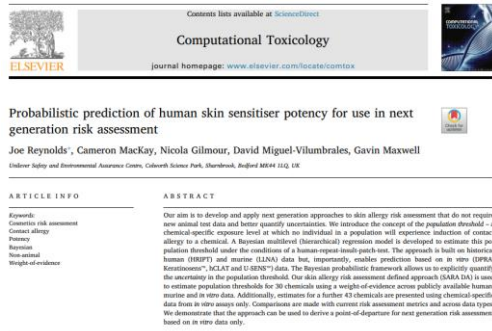
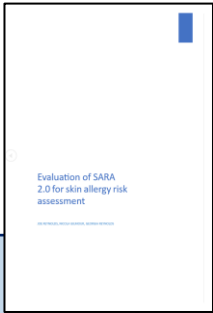
The model and underlying database are revised and expanded. Unilever performs an internal evaluation and applies the model for risk assessment.

2021-2022

SARA is published within a set of 3 papers describing the model and exploring its use in case study risk assessment scenarios.

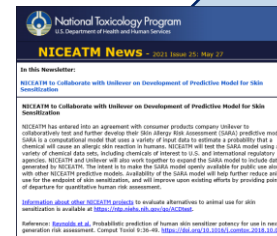
2023 - present

Unilever begin to develop SARA 2.0, starting from the SARA-ICE database and evaluate the model.



Evaluation of the Skin Allergy Risk Assessment (SARA) model for skin sensitisation risk assessment

Contents	
1	Overview of the SARA model..... 3
2	Definition of the SARA point of departure..... 3
3	SARA database..... 3
4	Modelling the relationships between in vivo and in vitro skin sensitisation hazard data..... 8
4.1	HPPT data..... 8
4.1.1	Model structure and priors..... 8
4.1.2	Validity of and sensitivity to the assumptions..... 10
4.2	LLNA data..... 11
4.2.1	Conversion of EC ₅₀ s to a SARA predictor..... 11
4.2.2	Model structure and priors..... 11
4.2.3	Validity of and sensitivity to the assumptions..... 13
4.3	In chemico and in vitro data..... 15



2021 - present

Unilever begin working with NICEATM to adapt SARA for regulatory use. The SARA database and the SARA-ICE model is developed and evaluated at the OECD.



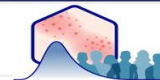
OECD Defined Approaches (DAs) for Skin Sensitisation (TG 497)

- In 2021, OECD Test Guideline 497 was adopted.
- TG 497 meets regulatory requirements for:
 - DAs that discriminate between sensitizers and non-sensitizers
 - DAs that discriminate strong from weak/moderate sensitizers (i.e., GHS potency categories)
- In 2021, the US and UK began a joint led feasibility study project under OECD for **evaluating a defined approach** that can provide a **point of departure** for quantitative risk assessment
- In 2024, the project began drafting an update to OECD TG 497 to incorporate DAs for PoD determination (i.e. SARA-ICE), expected to be released mid-2025.
- *In parallel, a self-contained version of the model and user interface have been developed, accessed via NICEATMs website.*



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SARA-ICE



SARA-ICE

The aim of the Unilever and NICEATM collaboration was to create a version of the SARA Model, SARA-ICE, which would be useful to wider industry, a model that could define points of departure (PoD) for use in risk assessment and have functionality for regulatory classification.

Database

The core dataset underpinning the model uses data in the ICE database.

434 chemicals

1,407 *in vivo* studies

2,575 *in vitro* studies



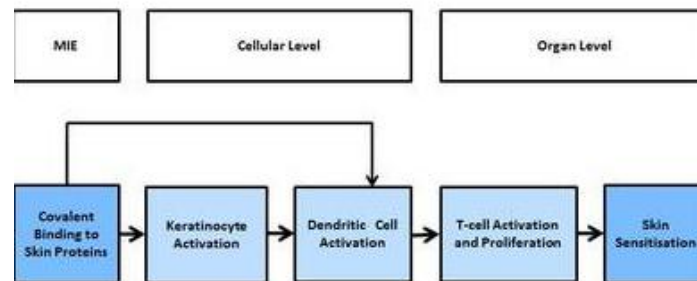
Integrated
Chemical
Environment

Input Assay Types

OECD TG NAM Assays aligned to key events in the skin allergy AOP.

- DPRA, kDPRA (KE1)
- KeratinoSens (KE2)
- U-Sens, hCLAT (KE3)

- Human (HMT/HRIPT) & LLNA studies may also be used.



Model Outputs

SARA-ICE, a Bayesian probabilistic model, gives a continuous measure of sensitiser potency: ED_{01} (1% sensitising dose in human patch test).

- A PoD (SARA-ICE DA)
- Or
- GHS Classification (SARA-ICE Extended)

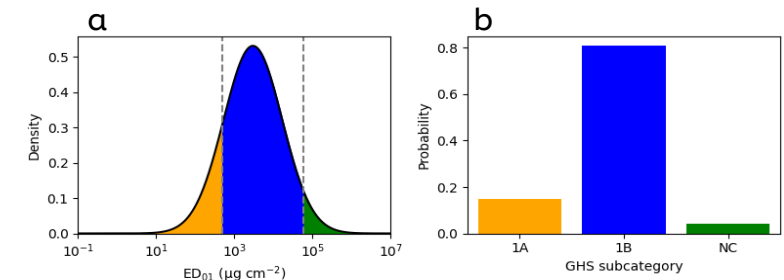
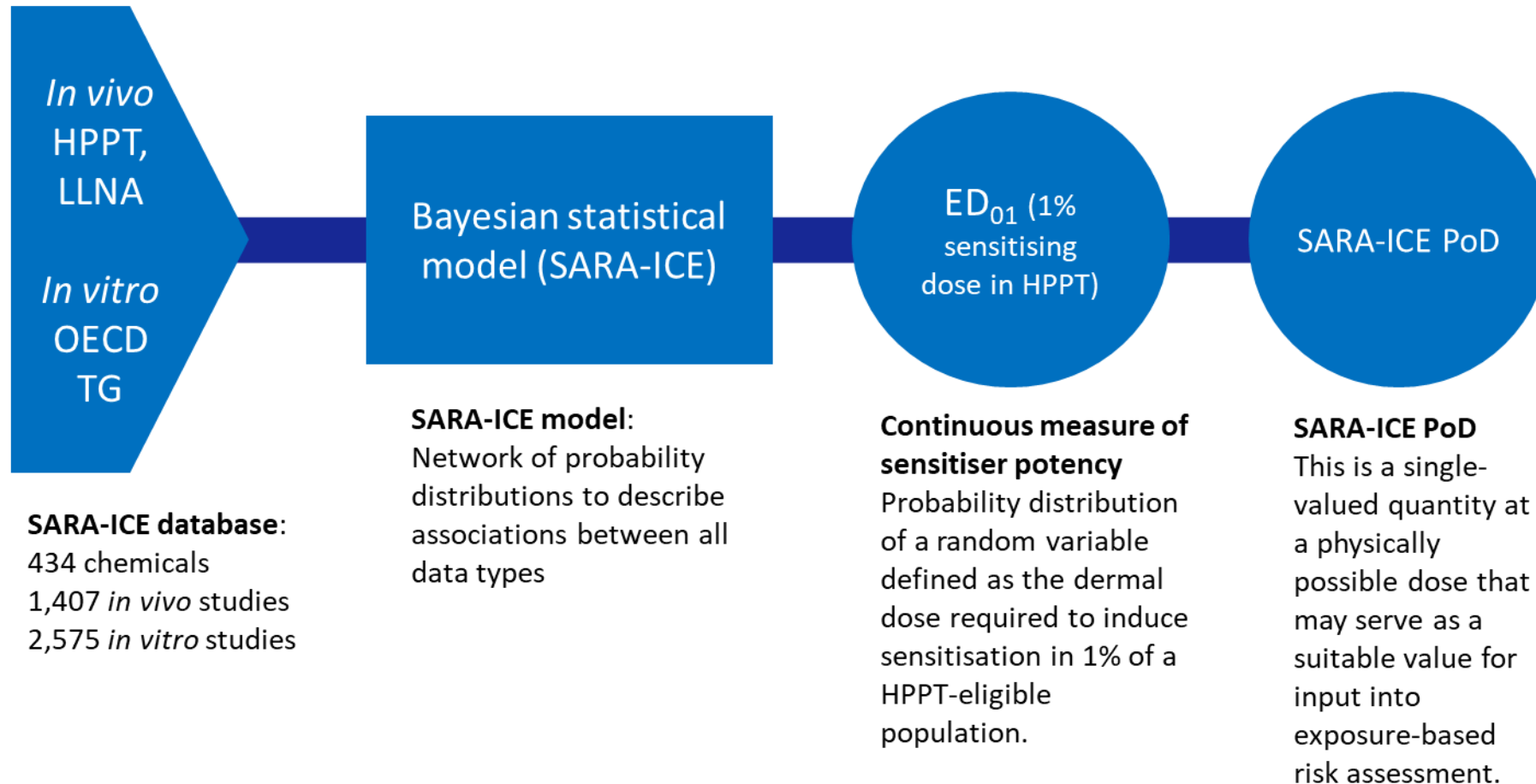


Figure (a) Example estimate of ED_{01} distribution with overlay of GHS subcategories 1A, 1B and NC defined thresholds, (b) probability of each GHS subcategory from ED_{01} distribution

SARA-ICE DA (Proposed OECD TG 497 Version)

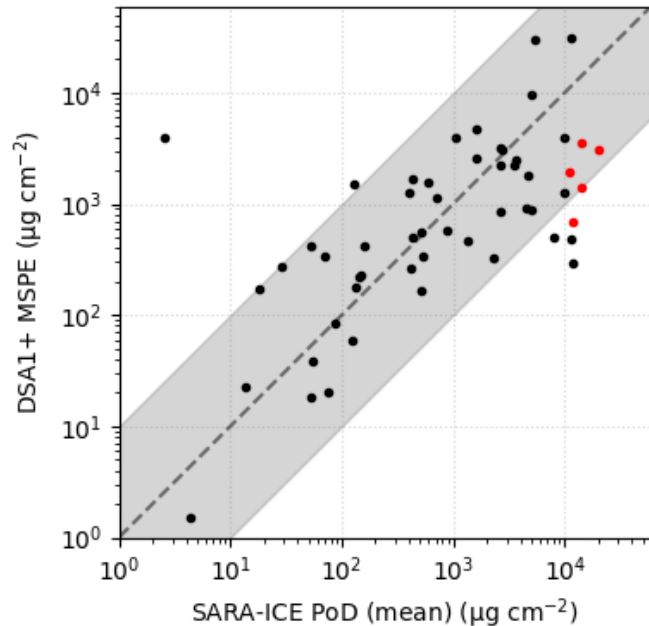
Input



Evaluation of the SARA-ICE PoD

SARA-ICE vs reference DSA1+

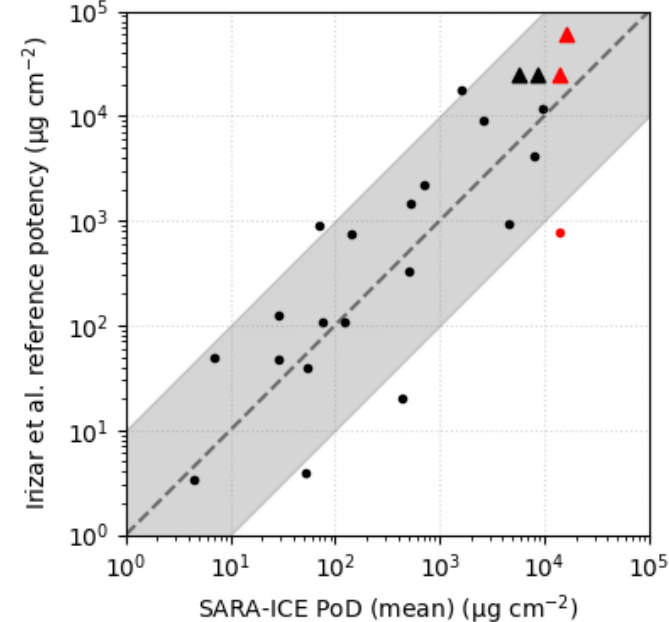
Pearson correlation: 0.64 (n=54)
 Geo. mean fold difference: 1.2
 Median fold difference: 1.3
 Geo. mean abs. fold difference: 3.7
 Median abs. fold difference: 2.9



SARA-ICE mean PoDs (from NAM data) relatively unbiased relative to reference DSA1+.
 PoDs on average around 3-fold away.

SARA-ICE PoDs vs Irizar et al. benchmarks

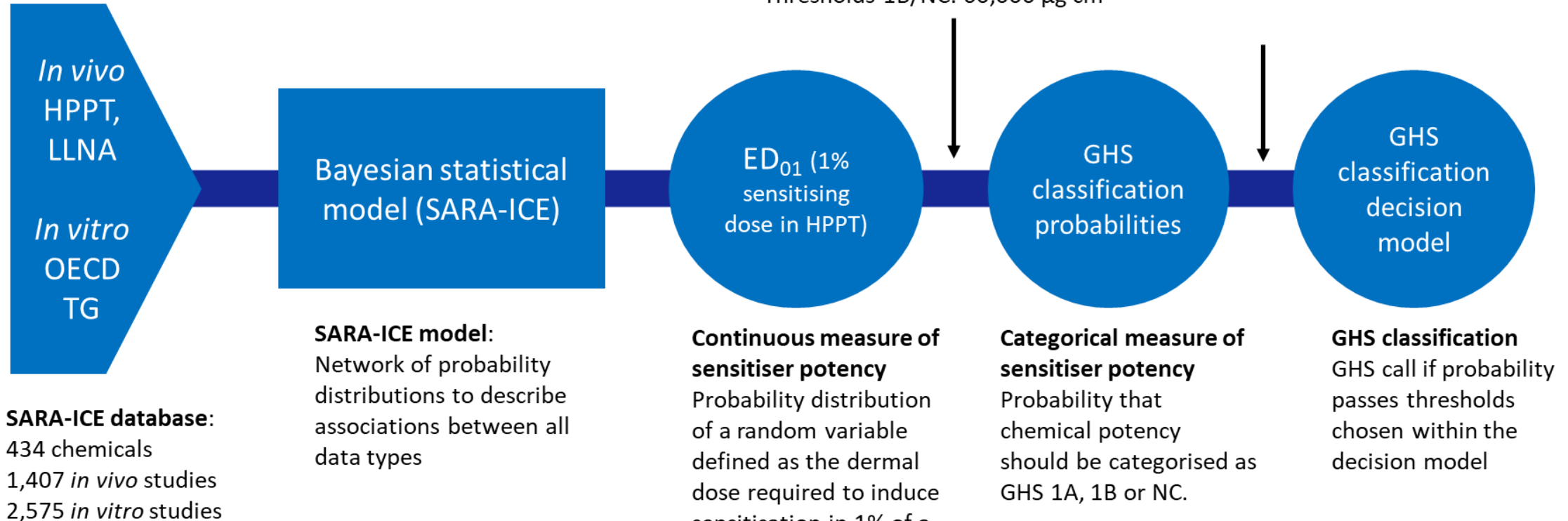
Pearson correlation: 0.84 (n=24)
 Geo. mean fold difference: 0.75
 Median fold difference: 0.58
 Geo. mean abs. fold difference: 3.6
 Median abs. fold difference: 3.3



SARA-ICE mean PoDs (from NAM data) relatively unbiased relative to Irizar et al. reference potency.
 PoDs on average around 3.5-fold away.

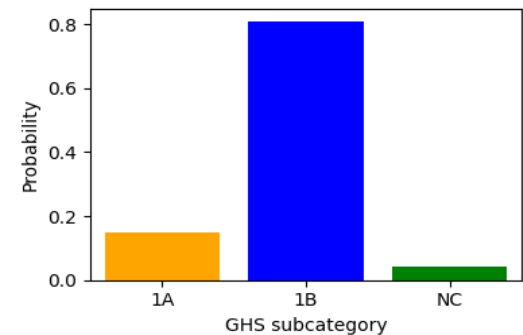
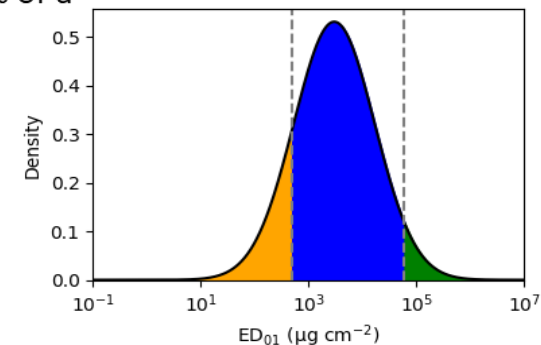
SARA-ICE DA (Extended Version)

Input

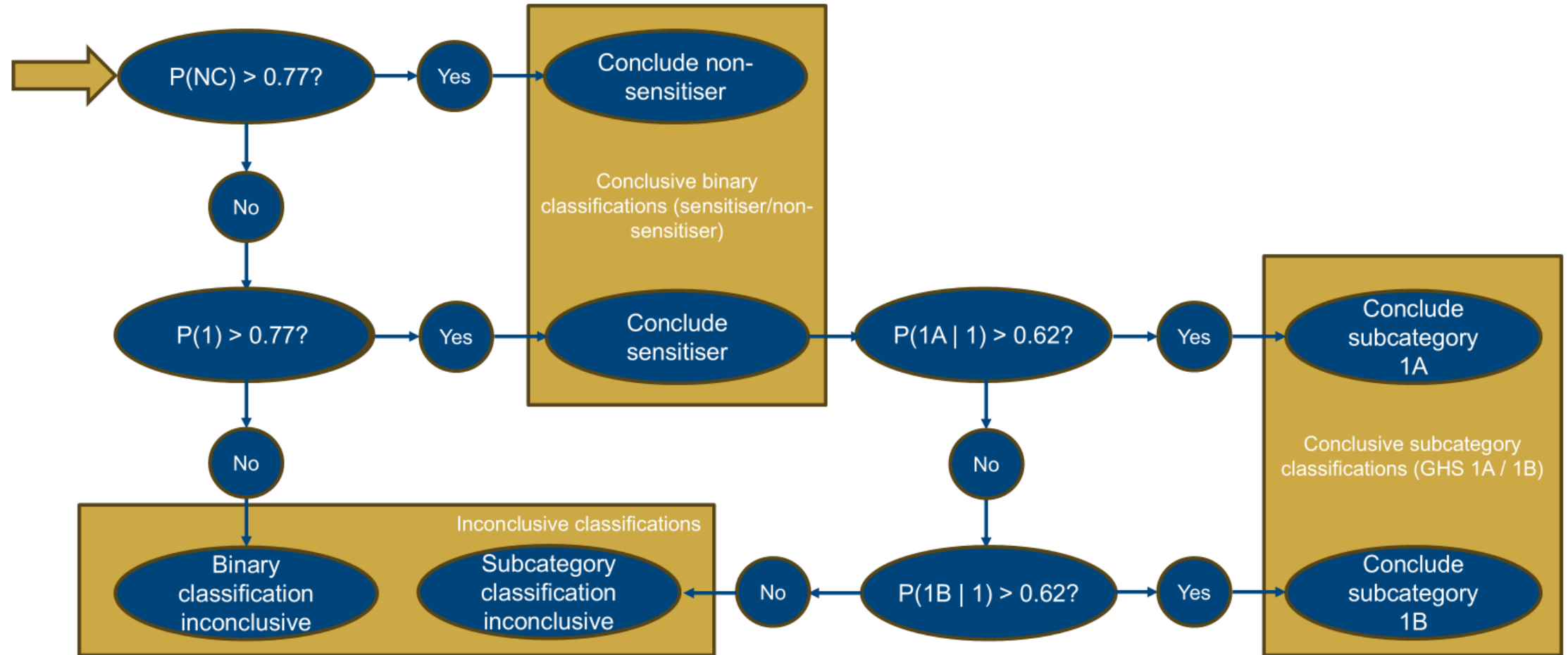


θ_{bin} = selected probability threshold for making a binary classification (1/NC)

θ_{sub} = selected threshold for making a sub-classification of 1A of 1B, contingent on class 1 being true



GHS Classification Decision Model (SARA-ICE Extended)



SARA-ICE NAM vs OECD DASS benchmarks

The SARA-ICE decision model has been evaluated against OECD benchmark classifications.

Binary classifications

Human, $\Theta_{bin} = 0.77$	SARA-ICE 1	SARA-ICE NC	Inconclusive	Total
Reference 1	37	5	13	55
Reference NC	0	5	6	11
Total	37	10	19	66
Sensitivity: 88%				
Specificity: 100%				
Balanced accuracy: 94%				
Inconclusive rate on reference class 1: 24%				
Inconclusive rate on reference class NC: 55%				
LLNA, $\Theta_{bin} = 0.77$	SARA-ICE 1	SARA-ICE NC	Inconclusive	Total
Reference 1	89	9	37	135
Reference NC	2	19	12	33
Total	91	28	49	168
Sensitivity: 91%				
Specificity: 90%				
Balanced accuracy: 91%				
Inconclusive rate on reference class 1: 27%				
Inconclusive rate on reference class NC: 36%				

Subcategory classifications

Human, $\Theta_{bin} = 0.77, \Theta_{sub} = 0.62$	SARA 1A	SARA 1B	SARA NC	Inconclusive	Total
Reference 1A	14	2	0	5	21
Reference 1B	3	7	5	16	31
Reference NC	0	0	5	6	11
Total	17	9	10	27	63
Sensitivity 1A: 88%, Specificity 1A: 85%, Balanced accuracy 1A: 86%					
Sensitivity 1B: 47%, Specificity 1B: 90%, Balanced accuracy 1B: 69%					
Sensitivity NC: 100%, Specificity NC: 84%, Balanced accuracy NC: 92%					
Average balanced accuracy: 82%					
Inconclusive rate on reference class 1A: 24%					
Inconclusive rate on reference class 1B: 52%					
Inconclusive rate on reference class NC: 55%					
LLNA, $\Theta_{bin} = 0.77, \Theta_{sub} = 0.62$	SARA 1A	SARA 1B	SARA NC	Inconclusive	Total
Reference 1A	27	3	0	8	38
Reference 1B	12	22	8	43	85
Reference NC	0	1	19	13	33
Total	39	26	27	64	156
Sensitivity 1A: 90%, Specificity 1A: 81%, Balanced accuracy 1A: 85%					
Sensitivity 1B: 52%, Specificity 1B: 92%, Balanced accuracy 1B: 72%					
Sensitivity NC: 95%, Specificity NC: 89%, Balanced accuracy NC: 92%					
Average balanced accuracy: 83%					
Inconclusive rate on reference class 1A: 21%					
Inconclusive rate on reference class 1B: 51%					
Inconclusive rate on reference class NC: 39%					

Application of the SARA-ICE Models

Example Case Study: Geraniol

- Using NAM data only, generate a PoD (SARA-ICE DA) and GHS Classification (SARA-ICE Extended)

SARA-ICE Input Data:

Substance Name	CASRN	MW (g/mol)
Geraniol	8007-13-4	154.25

DPRA	
Depletion Cys (%)	Depletion Lys (%)
12.3	2.6

kDPRA
log Kmax (M ⁻¹ s ⁻¹)
-3.4

KeratinoSens
EC1.5 (uM)
209.8

hCLAT	
CD54, EC200 (ug/ml)	CD86, EC150 (ug/ml)
>168	123

USENS
CD86, EC150 (ug/ml)
53.6

SARA-ICE DA (Proposed OECD TG 497 Version)

An official website of the United States government [Here's how you know](#)



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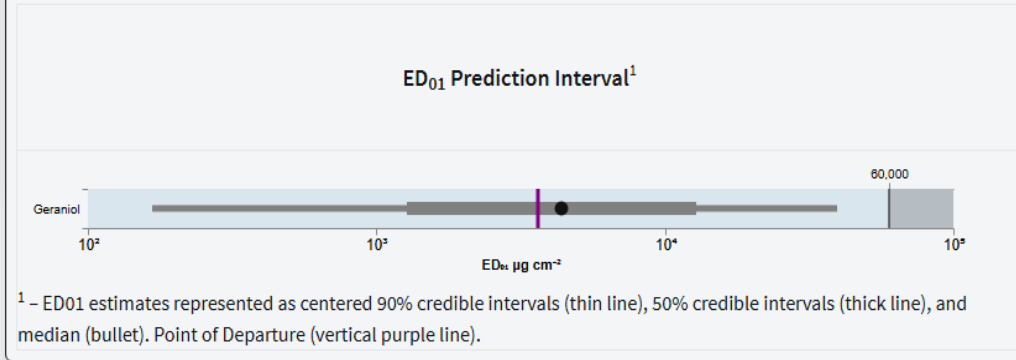
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SARA-ICE DA OECD TG497 (v1.0) Results

Substance	CASRN	POD*	ED ₀₁ Percentiles ($\mu\text{g cm}^{-2}$)		
			5th	50th	95th
Geraniol	8007-13-4	3,600	170	4,400	39,000

POD* - Point of departure - geometric mean of the ED₀₁, predicated on being a sensitizer


SARA-ICE DA OECD TG497 (v1.0) Prediction Intervals



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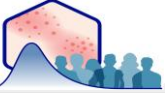
SARA-ICE DA (Extended Version)



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SARA-ICE

Overview > Load Input File > Select Model > **View Analysis** > Download Analysis

GHS Thresholds

Users can select Probability Thresholds for GHS Call for their own individual use requirements. Sliders are restricted to minimum allowable thresholds for predicting hazard or sub-category as defined in the evaluation found in Reinke et al., 2025. Evaluation of categorization performance was carried out using thresholds of 0.77 (hazard) and 0.62 (sub-category), as described in Supplementary Data 2: Performance of SARA-ICE against OECD benchmark GHS dataset.

GHS Hazard Probability Threshold
Category 1 vs Not Categorized (NC)
Minimum: 0.67

GHS Sub-Cat Probability Threshold
1A vs 1B (assuming Category 1)
Minimum: 0.5

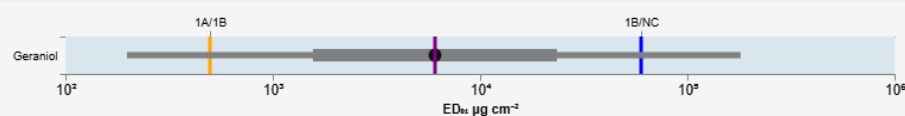
SARA-ICE Extended Model (v1.0) Results

Substance	CASRN	Mean ¹	SPUR ²	ED ₀₁ Percentiles (µg cm ⁻²)			SARA-ICE Probability GHS Subcategory				GHS Call
				5th	50th	95th	1A	1B	1	NC	
Geraniol	8007-13-4	6,100	30	200	6,000	>60,000	0.11	0.76	0.87	0.13	1B

¹ Mean¹ – Geometric Mean of the ED₀₁ distribution
² SPUR² – (SARA-ICE Prediction Uncertainty Ratio) fold-difference between the median (50th percentile) and the 5th percentile

SARA-ICE Extended Model (v1.0) Prediction Intervals

ED₀₁ Prediction Interval¹



¹ – ED₀₁ estimates represented as centered 90% credible intervals (thin line), 50% credible intervals (thick line), and median (bullet). Geometric Mean (vertical purple line).

NAM PoD Margin of Safety (MoS) in Risk Assessment

Acceptable MoS = a value above which a risk assessor may usually conclude low risk for their safety assessment

Traditionally, values of 100 or above have been used as acceptable MoS.

To convert **acceptable MoS for a human derived NESIL** → **acceptable MoS for NAM PoDs**, statistically analyse

- differences between NESILs vs benchmark exposures
- differences between NESILs and NAM PoD.

$$\log_{10} (\text{Acceptable NAM MoE}) = \beta + \beta_{\text{NAM}} + (\log_{10} (\text{Acceptable NESIL MoE}) - \beta) \sqrt{1 + \frac{\sigma_{\text{NAM}}^2}{\sigma^2}}$$

Acceptable MoE NESIL	Acceptable MoE SARA-ICE PoD NAM inputs (geometric mean)
100	100
300	360
1000	1,500

Summary

- SARA-ICE DA fulfils a gap in the current OECD TG 497 on defining a PoD for risk assessment
- SARA-ICE Extended enables a more flexible use of the model, and allows for GHS classifications to be made
- SARA-ICE allows flexible use of a range of OECD TG NAMs (as well as historical LLNA/Human data)
- SARA-ICE WebApp is nearly ready for public release and will be available on the NICEATM website (<https://ntp.niehs.nih.gov/whatwestudy/niceatm>)
- A margin of safety can be calculated and applied for NAM PoDs to provide equivalent protectiveness against human benchmarks as a traditional NESIL for skin sensitisation risk assessment (Reynolds et al., manuscript in preparation)

Thank You



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The NICEATM Group