## New Approach Methods (NAMs)-Based Next Generation Risk Assessment (NGRA) for Skin Allergy and Case Study

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### Assessing ingredient & product safety without animal testing

### Next Generation Risk Assessment (NGRA)



Is it safe to include x% of chemical y in product z?





### Principles of Next Generation Risk Assessment from ICCR Non-animal approaches in Cosmetic Risk Assessment



International Cooperation on Cosmetics Regulation

### Main overriding principles:

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

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

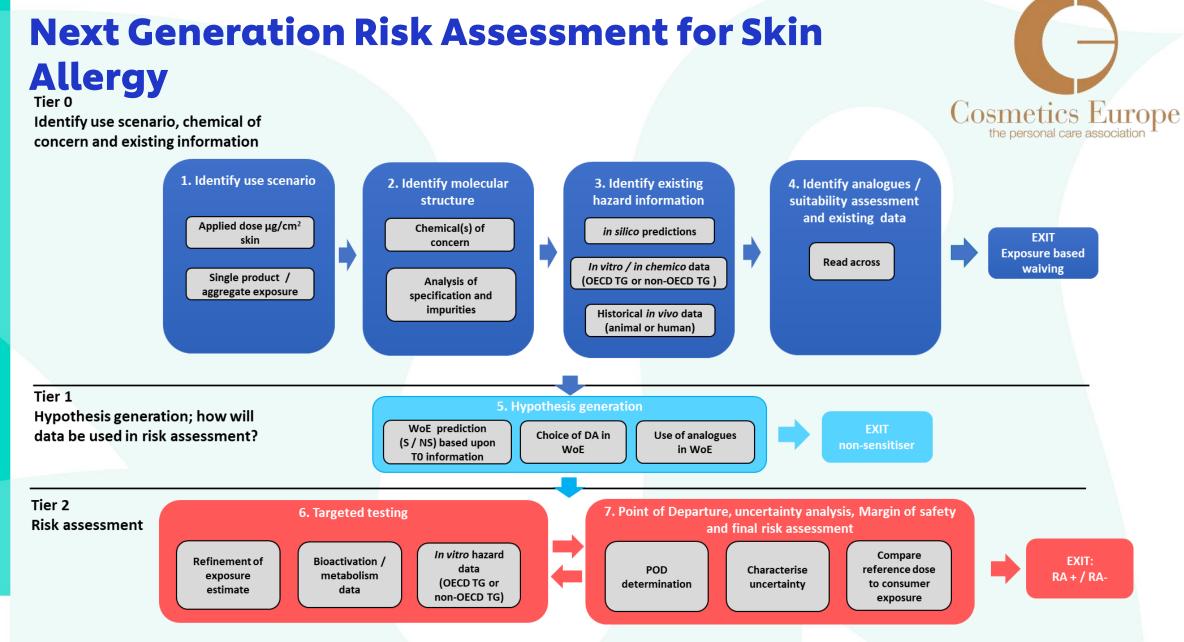
- » Following an appropriate appraisal of existing information
- » Using a tiered and iterative approach
- » Using robust and relevant methods and strategies

### Principles for documenting NGRA:

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



Dent *et al* (2018), Computational Toxicology, **7**, 20-26: <u>https://doi.org/10.1016/j.comtox.2018.06.001</u>

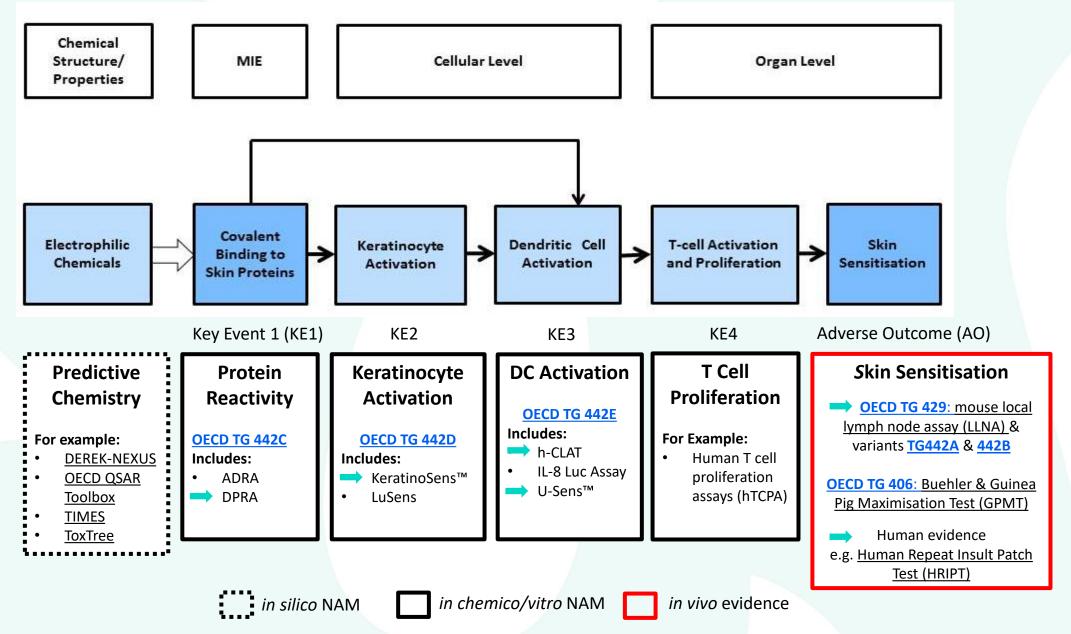


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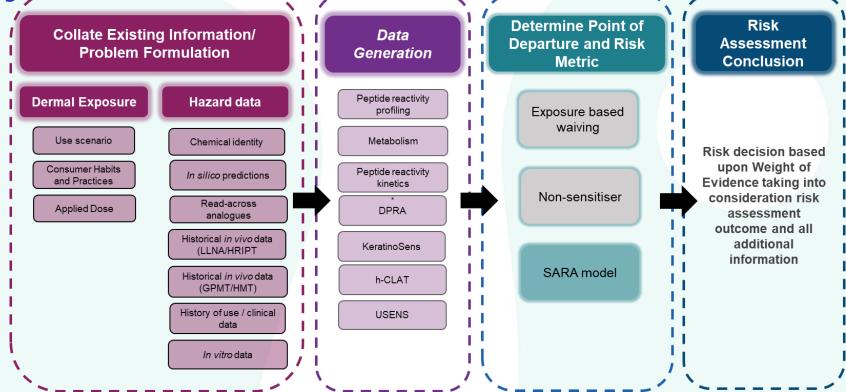
Gilmour et al, (2020), Regulatory Toxicology and Pharmacology, 116: https://doi.org/10.1016/j.yrtph.2020.104721

### **Skin Sensitisation AOP**

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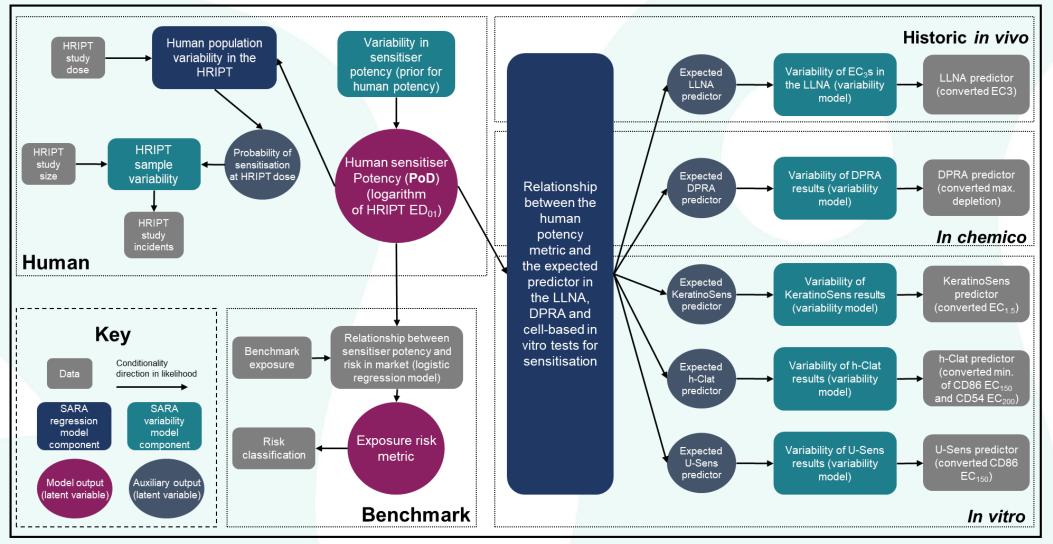
### Next Generation Risk Assessment (NGRA) Framework for Skin Allergy



- Our NGRA framework for skin allergy is based upon the ICCR principles (Dent et al 2018) and the
  previously published NGRA frameworks for systemic tox {SEURAT-1} (Amaral et al 2018) and skin
  allergy {Cosmetic Europe} (Gilmour et al 2020).
- Designed to use a WoE based upon all available information, accommodate range of consumer product exposure scenarios and provide a quantitative point of departure and risk metric → SARA Model.



### **SARA Model**

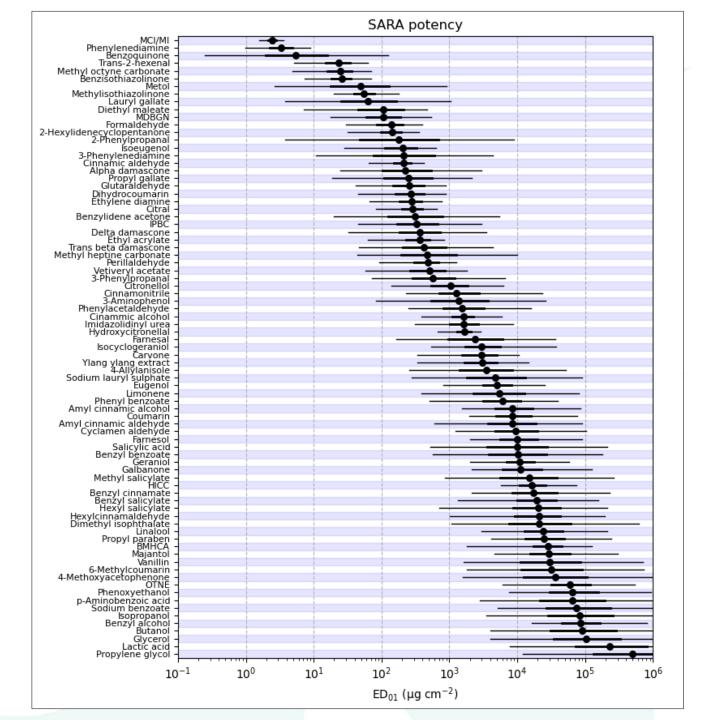


- The SARA model uses Bayesian statistics to infer a probability that a consumer exposure to some chemical can be considered low risk, to inform risk assessment decisions.
- The SARA Model uses a database of public NAM data covering AOP KEs 1-3, and historic LLNA and HRIPT data for the AOP AO.



### **SARA Defined Approach**

- The point of departure (PoD) metric is a dose with a 1% chance of human skin sensitisation (termed ED<sub>01</sub>).
- The SARA dataset contains 81 chemicals.
- The model accounts for variability in the DPRA, KeratinoSens™, h-CLAT and U-Sens™ and the *in vivo* data.
- The model has been expanded to incorporate benchmark exposure information.





# Use of consumer exposure information and clinical evidence to develop skin allergy risk benchmarks

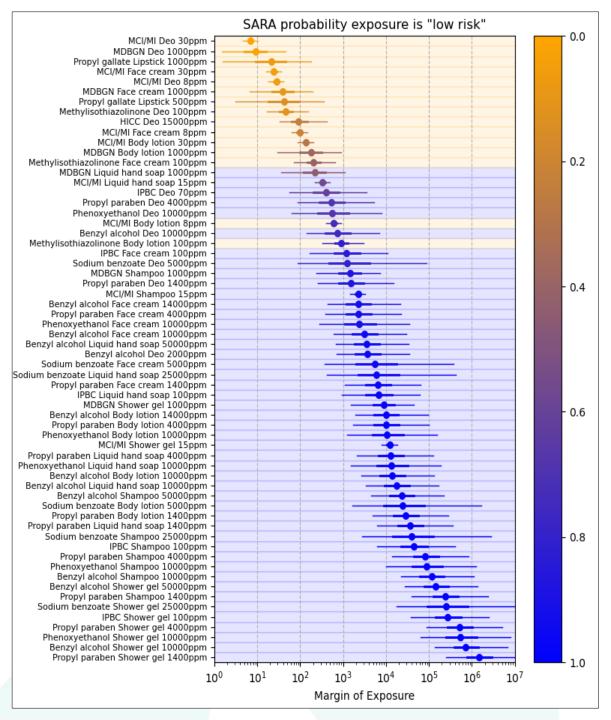
- Traditional risk assessment approaches for skin allergy use safety factors to rescale PoDs to market-equivalent safe doses for comparison against consumer exposure estimates.
- For NGRA, publicly available benchmark exposure information can be used to establish that an exposure is low risk and can be considered safe.
- To apply this concept, we established 62 low or high risk benchmark exposures using 10 human skin allergens (e.g. MCI/MI) with an established history of use in 7 cosmetic product types.

Material	Product type	Use level (ppm)	Consumer exposure to benchmark product (ng cm <sup>-2</sup> )	Induction risk
MCI/MI	Dee	30	350	HIGH
,	Deo	7.5	87.8	HIGH
	Face cream	30	100	HIGH
	race cream	7.5	25	HIGH
	Padulation	30	18	HIGH
	Body lotion	7.5	4	HIGH
	Liquid hand soap	15	7.3	LOW
	Shampoo	15	1.1	LOW
	Shower gel	15	0.2	LOW



#### Expansion of SARA model to use benchmark exposure information

- The SARA model was expanded to incorporate benchmark exposure information as an additional input alongside historic *in vivo* and NAM data.
- After fitting the model, and given some exposure scenario of interest, the model can calculate the SARA risk metric, defined as the probability that the exposure is low risk for human skin sensitisation induction.



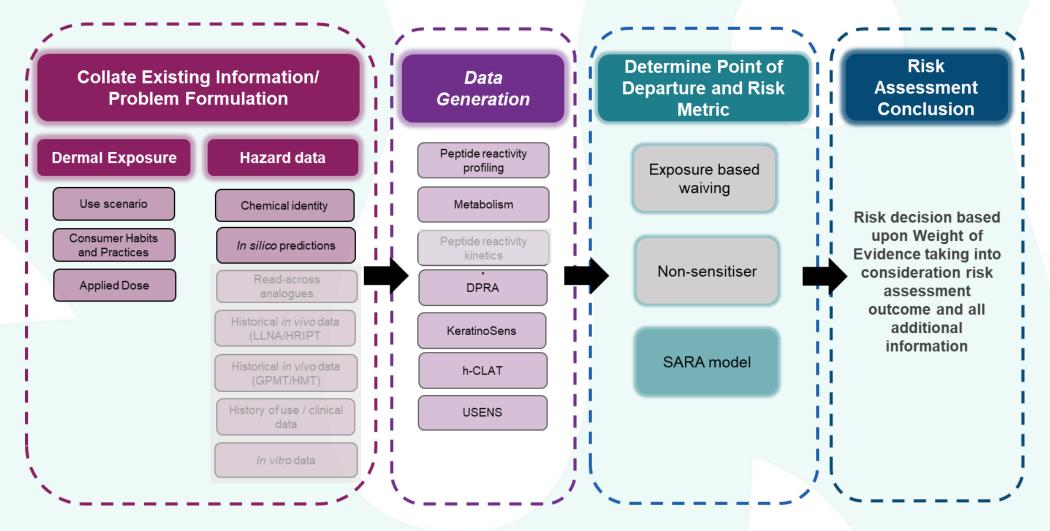


## **Case Study**



### **Application of NGRA framework for Skin Allergy**

This NGRA framework is applied to a hypothetical skin allergy assessment of a consumer product at two exposures - 0.1% coumarin in a face cream and 1% in a deodorant. For the purposes of the case study, *in vivo* data and read-across were not used, and the use of dermal sensitisation threshold (DST) was not appropriate.





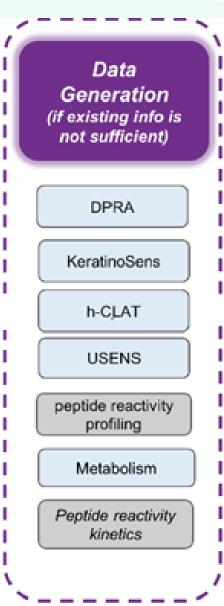
### Local exposure + Collate Existing Information/ Problem

Local Exposure Estimation	Collate Existing Information/ Problem Formulation In vitro data Historical in vivo data (LLNA/HRIPT) Read-Across analogues Molecular Structure In silico predictions (PCA) History of use / clinical data	Product type	Face cream	Deodorant
Use scenario		Product used per day (90 <sup>th</sup> percentile) (g/day)	1.54	1.5
Consumer Habits and Practices Applied Dose		Ingredient inclusion level (%)	0.1	1
		Skin surface area (face / axilla) (cm²)	565	200
		Leave-on or Rinse-off	Leave-on	Leave-on
	Historical in vivo data (e.g. GMPT/HMT)	Local dermal exposure (µg/cm²)	2.7	75

- In silico chemistry predictions for the sensitiser potential of coumarin: TIMES-SS predicts coumarin and metabolites non-sensitisers; Derek Nexus, ToxTree and OECD QSAR Toolbox all predict sensitiser potential. ToxTree and OECD QSAR Toolbox predicted a Michael Acceptor mechanism. Both direct and indirect (pro-hapten) mechanisms were indicated.
- Meteor Nexus identified hydroxylation as the main route of biotransformation. Most metabolites were predicted to bind to protein, a flag for skin sensitization. 7-OH coumarin was identified as one of the main metabolites in an investigation in human hepatocytes.



#### **Data Generation**

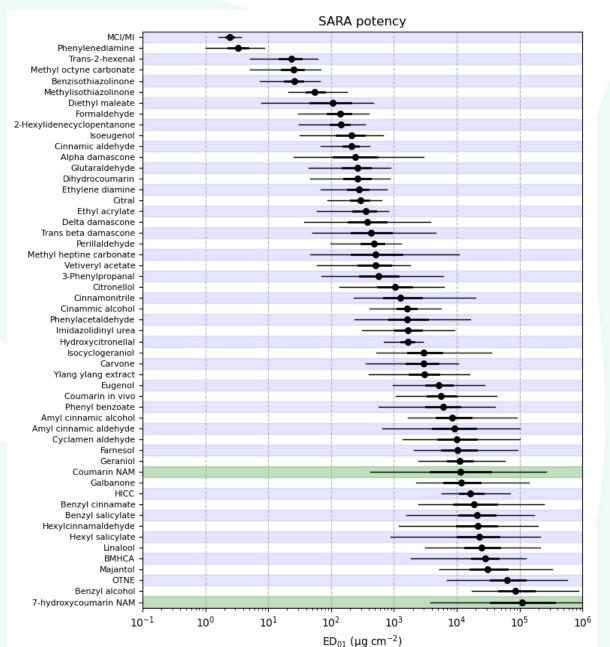


	DPRA (TG442C)		KeratinoSens ™ (TG 442D)	h-CLAT (TG 442E)		U-SENS™ (TG 442E)
	%cys depl.	%lys depl.	EC1.5 (µM)	CD86 (EC200 µg/mL)	CD54 (EC150 µg/mL)	CD86 (EC150 µg/mL)
Coumarin	1.3	0	187.5	<178	>637	95.5
7-OH Coumarin	0*	0	>2000	>566	>566	182

- Coumarin was positive in all tests, except for DPRA where peptide depletion was too low to meet positive threshold.
- 7-OH coumarin was negative in KeratinoSens<sup>™</sup> & h-CLAT, positive in USENS<sup>™</sup>, inconclusive in DPRA. \*Peptide profiling was completed which identified cysteine depletion to be caused by dimerization and therefore the DPRA value was adjusted.



### **Determine Point of Departure (PoD) using SARA Model**

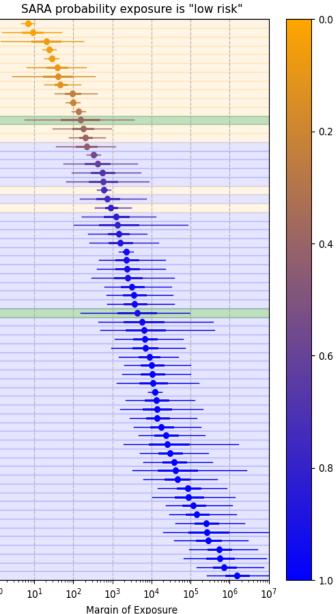


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- The generated DPRA, KeratinoSens<sup>™</sup>, hCLAT and USens<sup>™</sup> data were used as inputs into the SARA Model to define a human relevant PoD (ED<sub>01</sub> i.e the 1% sensitising dose for a HRIPT population).
- For coumarin, the expected SARA Model derived  $ED_{01}$  is 11,000µgcm<sup>-2</sup>, whilst for 7-OH coumarin the expected  $ED_{01}$  is 110,000µgcm<sup>-2</sup> i.e. 7-OH coumarin is estimated to be 10-fold less potent than coumarin).
- Therefore, a risk assessment based on coumarin potency data only would be conservative.

#### **Determine Margin of Exposure (MoE)**

MCI/MI Deo 30ppm MDBGN Deo 1000ppm Propyl gallate Lipstick 1000ppm MCI/MI Face cream 30ppm · MCI/MI Deo 8ppm · MDBGN Face cream 1000ppm Propyl gallate Lipstick 500ppm Methylisothiazolinone Deo 100ppm HICC Deo 15000ppm MCI/MI Face cream 8ppm MCI/MI Body lotion 30ppm Coumarin NAM Deodorant 10000ppm MDBGN Body lotion 1000ppm Methylisothiazolinone Face cream 100ppm MDBGN Liquid hand soap 1000ppm MCI/MI Liquid hand soap 15ppm IPBC Deo 70ppm Propyl paraben Deo 4000ppm Phenoxyethanol Deo 10000ppm MCI/MI Body lotion 8ppm Benzyl alcohol Deo 10000ppm Methylisothiazolinone Body lotion 100ppm IPBC Face cream 100ppm Sodium benzoate Deo 5000ppm MDBGN Shampoo 1000ppm Propyl paraben Deo 1400ppm MCI/MI Shampoo 15ppm Benzyl alcohol Face cream 14000ppm -Propyl paraben Face cream 4000ppm Phenoxyethanol Face cream 10000ppm Benzyl alcohol Face cream 10000ppm Benzyl alcohol Liquid hand soap 50000ppm -Benzyl alcohol Deo 2000ppm Coumarin NAM Face cream 1000ppm Sodium benzoate Face cream 5000ppm Sodium benzoate Liquid hand soap 25000ppm Propyl paraben Face cream 1400ppm IPBC Liquid hand soap 100ppm MDBGN Shower gel 1000ppm · Benzyl alcohol Body lotion 14000ppm -Propyl paraben Body lotion 4000ppm -Phenoxyethanol Body lotion 10000ppm MCI/MI Shower gel 15ppm Propyl paraben Liquid hand soap 4000ppm Phenoxyethanol Liquid hand soap 10000ppm Benzyl alcohol Body lotion 10000ppm Benzyl alcohol Liquid hand soap 10000ppm Benzyl alcohol Shampoo 50000ppm Sodium benzoate Body lotion 5000ppm Propyl paraben Body lotion 1400ppm Propyl paraben Liquid hand soap 1400ppm · Sodium benzoate Shampoo 25000ppm IPBC Shampoo 100ppm Propyl paraben Shampoo 4000ppm Phenoxyethanol Shampoo 10000ppm Benzyl alcohol Shampoo 10000ppm Benzyl alcohol Shower gel 50000ppm Propyl paraben Shampoo 1400ppm Sodium benzoate Shower gel 25000ppm IPBC Shower gel 100ppm Propyl paraben Shower gel 4000ppm Phenoxyethanol Shower gel 10000ppm -Benzyl alcohol Shower gel 10000ppm Propyl paraben Shower gel 1400ppm 10<sup>0</sup>



- The MoE was calculated from the ED<sub>01</sub> for coumarin and the dermal exposures for each product type using the SARA Model.
- The median MoE for face cream exposure ranks with the low-risk benchmarks whilst the median MoE for the deodorant exposure ranks with the high-risk benchmarks.
- The SARA DA probability that the exposure is low risk is calculated to be 0.90 for the face cream dermal exposure and 0.39 for the deodorant dermal exposure.
- Coumarin exposure at 0.1% in a face cream is low risk for skin sensitisation whereas coumarin exposure at 1% in a deodorant is high risk



### **NICEATM-Unilever CRADA**



National Toxicology Program U.S. Department of Health and Human Services

#### 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

NICEATM has entered into an agreement with consumer products company Unilever to collaboratively test and further develop their Skin Allergy Risk Assessment (SARA) predictive model. SARA is a computational model that uses a variety of input data to estimate a probability that a chemical will cause an allergic skin reaction in humans. NICEATM will test the SARA model using a variety of chemical data sets, including chemicals of interest to U.S. and international regulatory agencies. NICEATM and Unilever will also work together to expand the SARA model to include data generated by NICEATM. The intent is to make the SARA model openly available for public use along with other NICEATM predictive models. Availability of the SARA model will help further reduce animal use for the endpoint of skin sensitization, and will improve upon existing efforts by providing points of departure for quantitative human risk assessment.

<u>Information about other NICEATM projects</u> to evaluate alternatives to animal use for skin sensitization is available at <u>https://ntp.niehs.nih.gov/go/ACDtest</u>.



Reference: <u>Reynolds et al.</u> Probabilistic prediction of human skin sensitizer potency for use in next generation risk assessment. Comput Toxiol 9:36-49. <u>https://doi.org/10.1016/j.comtox.2018.10.004</u>

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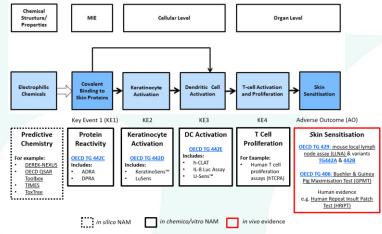
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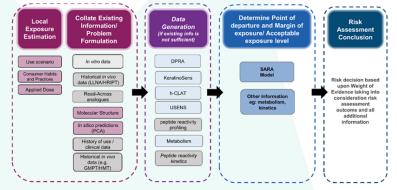
### Conclusions

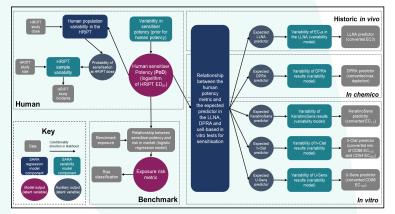
- Significant progress has been made in the last decade to apply non-animal experimental data using Defined Approaches & tiered frameworks.
- Bayesian DAs enable experimental data variability to be modelled and uncertainty in PoDs & risk metrics to be factored into decision-making.

### **Next Steps**

- SARA DA & Skin Allergy Risk Benchmarks manuscripts submitted for publication.
- NICEATM collaboration established to test SARA, expand the approach and make it publicly available.
- In-house work ongoing to explore new SARA inputs & expand the database, including risk benchmarks.









### Scientific partnership 科学合作





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# GLOBAL PRODUCT SAFETY COLLABORATION 全球产品安全合作 替代方面的合作



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## **Thank you!**

# 谢谢!



