# Case study application using non-animal approaches for local and systemic safety assessment of cosmetic ingredients

Renato Ivan de Ávila, PhD
Safety Scientist - Toxicology
Unilever Safety & Environmental Assurance Centre (SEAC)
Renato.de-Avila@unilever.com





# Can we use a new ingredient safely?

 Can we safely use x% of ingredient y in product z?





Risk = Hazard x Exposure



# Assuring consumer safety without animal testing: maximising use of existing information and animal-free approaches

- All available safety data (of suitable quality)
  - public domain, historical in-house data, supplier data etc
  - chemistry data, in vitro data, clinical data, epidemiological data, animal toxicology data etc
- Exposure-based waiving approaches (e.g. toxicological threshold of concern)
- History of safe use
- In silico predictions
- Read across
- Use of existing in vitro data and approaches
- Next Generation Risk Assessment (NGRA)

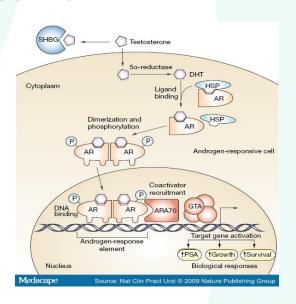


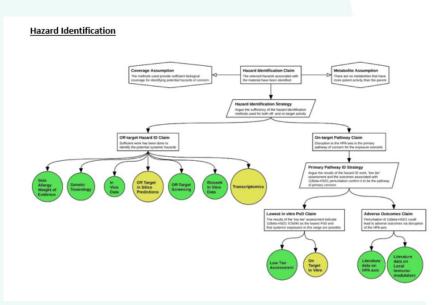
#### **Next Generation Risk assessment (NGRA)**

#### What is NGRA?

- Using new tools and approaches (NAMs New Approach Methods) to build a risk assessment to enable decisions to be made
- An exposure-led risk assessment solution to biological pathwayindicated hazard concerns









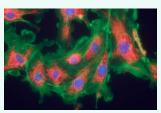
Mechanistic

**Hypothesis driven** 

### NGRA: Using relevant methods to test hypotheses

#### **New Approach Methods (NAMs)**







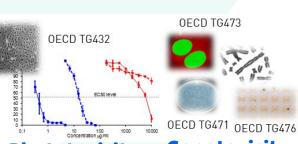
# OECD TG430/431 OECD TG437 OECD TG439

**Established Methods** 

#### **OECD test methods**

Skin and eye irritation





**Phototoxicity** Genotoxicity

#### **Cellular stress**

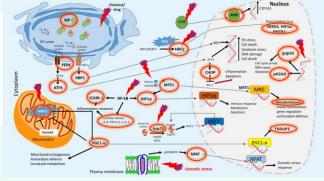
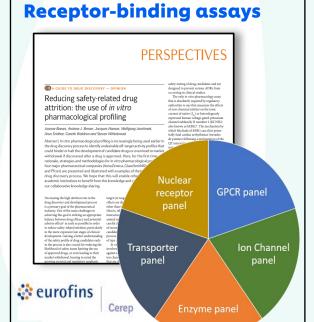


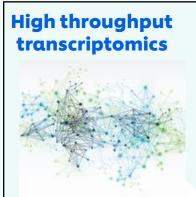
Image kindly provided by Paul Walker (Cyprotex)

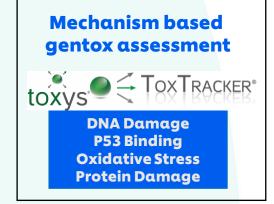
36 biomarkers identified that were representative of key stress pathways, mitochondrial toxicity and cell health.

Hatherell et al (2020), Toxicological Sciences, 176, 11-33





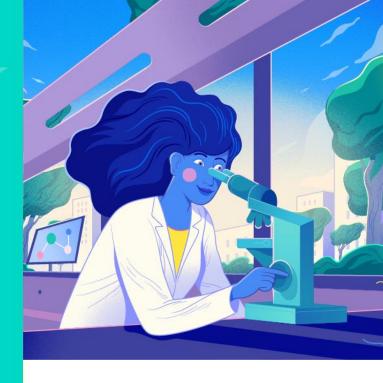






# Case study examples

- 1) Systemic effects
- 2) Local effect skin sensitisation





### Case study: hypothetical example for 0.1% coumarin in face cream

Baltazar *et al*. (2020) <u>A Next-Generation Risk Assessment Case Study for Coumarin in Cosmetic Products</u>. Toxicological Sciences, 176, 236-252



TOXICOLOGICAL SCIENCES, 176(1), 2020, 236-252

doi: 10.1093/toxsci/kfaa048 Advance Access Publication Date: April 10, 2020

#### A Next-Generation Risk Assessment Case Study for Coumarin in Cosmetic Products

Maria T. Baltazar, Sophie Cable, Paul L. Carmichael, Richard Cubberley, Tom Cull, Mona Delagrange, Matthew P. Dent, Sarah Hatherell, Jade Houghton, Predrag Kukic, Hequn Li, Mi-Young Lee, Sophie Malcomber, Alistair M. Middleton, Thomas E. Moxon , Alexis V. Nathanail, Beate Nicol, Ruth Pendlington, Georgia Reynolds, Joe Reynolds, Andrew White, and Carl Westmoreland

Unilever Safety and Environmental Assurance Centre, Colworth Science Park, Sharnbrook, Bedfordshire MK44 1LO. UK

 $^{1}$ To whom correspondence should be addressed. Fax: +44(0)1234264744. E-mail: maria.baltazar@unilever.com.

#### ABSTRACT

Next-Generation Risk Assessment is defined as an exposure-led, hypothesis-driven risk assessment approach that integrates new approach methodologies (NAMs) to assure safety without the use of animal testing. These principles were applied to a hypothetical safety assessment of 0.1% coumarin in face cream and body lotion. For the purpose of evaluating the use of NAMs, existing animal and human data on coumarin were excluded. Internal concentrations (plasma C<sub>max</sub>) were estimated using a physiologically based kinetic model for dermally applied coumarin. Systemic toxicity was assessed using a battery of in vitro NAMs to identify points of departure (PoDs) for a variety of biological effects such as receptor-mediated and immunomodulatory effects (Eurofins SafetyScreen44 and BioMap Diversity 8 Panel, respectively), and general bioactivity (ToxCast data, an in vitro cell stress panel and high-throughput transcriptomics). In addition, in silico alerts for genotoxicity were followed up with the ToxTracker tool. The PoDs from the in vitro assays were plotted against the calculated in vivo exposure to calculate a margin of safety with associated uncertainty. The predicted C<sub>max</sub> values for face cream and body lotion were lower than all PoDs with margin of safety higher than 100. Furthermore, coumarin was not genotoxic, did not bind to any of the 44 receptors tested and did not show any immunomodulatory effects at consumer-

# 0.1% COUMARIN IN FACE CREAM (NEW FRAGRANCE)

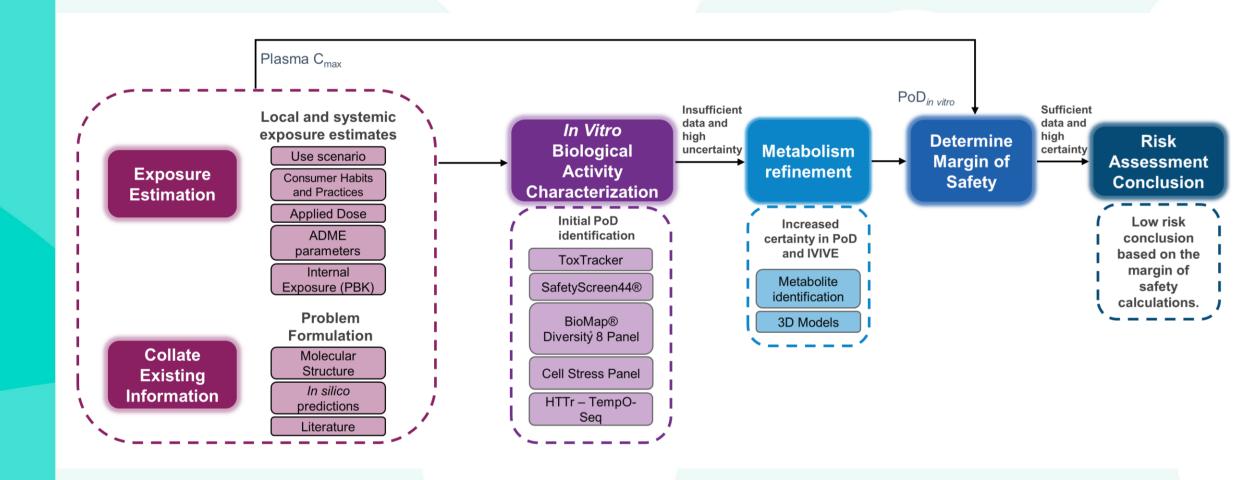


#### **Assumptions:**

- EU Market
- 100% purity
- no in vivo data was available such as animal data, History of Safe Use (HoSU) or Clinical data
- no use of animal data in Read Across
- In silico alerts known to be based on animal or in vivo data or on the structure of Coumarin itself were excluded



# Next-Generation Risk Assessment case study workflow for 0.1% coumarin in face cream





# **STEP ONE**

Exposure information and collation of existing information



### NGRA for 0.1% coumarin in face cream: exposure estimation



Table 2: Estimated daily exposure levels for different cosmetic product types according to Cosmetics Europe data (SCCNFP/0321/00; Hall et al., 2007, 2011).

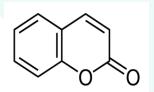
Product type	Estimated daily amount applied	Relative amount applied (mg/kg bw/d)	Retention factor <sup>1</sup>	Calculated daily exposure (g/d)	Calculated relative daily exposure (mg/kg bw/d)
Bathing, showerin	g				
Shower gel	18.67 g	279.20	0.01	0.19	2.79
Hand wash soap <sup>2</sup>	20.00 g	-	0.01	0.20 <sup>3</sup>	3.33
Hair care					
Shampoo	10.46 a		0.01	0.11	1.51
Hair condition-			201	0.04	0.60



B. Hall et al./Food and Chemical Toxicology 49 (2011) 408-422

Assessment is exposure-led and uses available habits and practices data



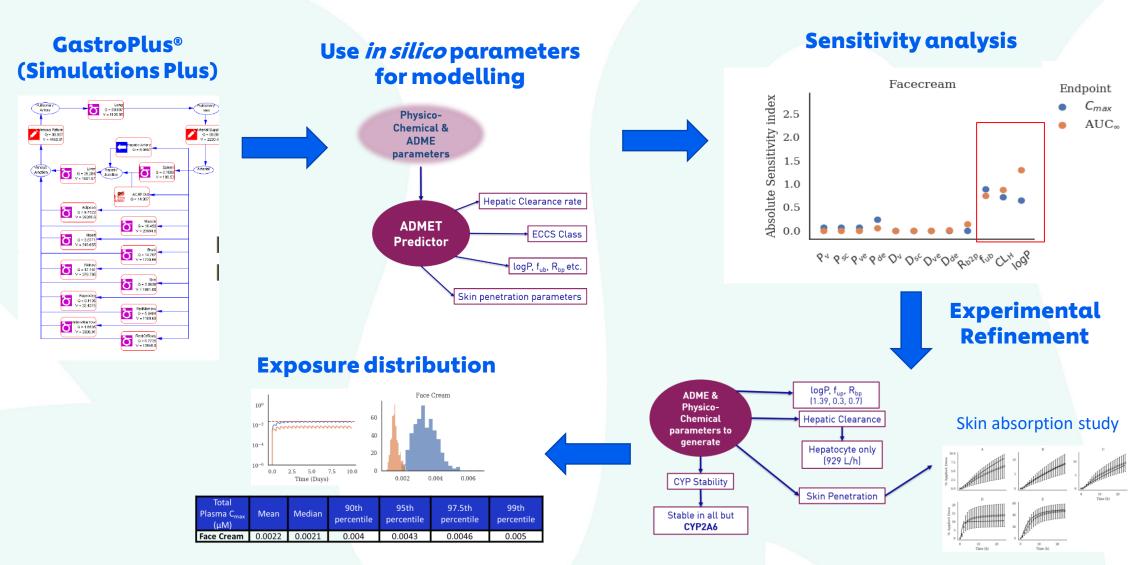




Parameter	Face cream
Amount of product used per day (g/day) using 90th percentile	1.54
Frequency of use	2 times/day
Amount of product in contact with skin per occasion (mg)	770
Ingredient inclusion level	0.1%
Skin surface area (cm2)	565
Exposure duration per occasion	12 hours
Amount of ingredient in contact with skin per occasion (mg)	0.77
Local dermal exposure per occasion (µg/cm2)	2.73
Systemic exposure per day (mg/kg)	0.02



# NGRA for 0.1% coumarin in face cream: exposure estimation- Internal concentration using PBK modelling - Model Inputs

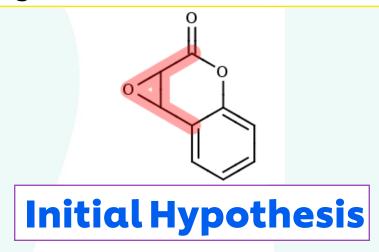




### NGRA for 0.1% coumarin in face cream: in silico predictions



Generation of hypothesis for potential Molecular Initiating events -ToxTree, MIE ATLAS\*, OECD toolbox



Next case study

- Coumarin might bind to proteins- MIE for induction of skin sensitisation
- DNA binding alert + epoxide formation MIE for genotoxicity
- Reactive metabolites might be formed with alerts for both genotoxicity and skin sensitisation
- No binding alerts for the 39 targets in MIE atlas



### NGRA for 0.1% coumarin in face cream: in vitro existing information

Identification of potential biological targets – PubChem and ToxCast



Only few active assays among multiple assays (≈ 5000)

Coumarin inhibited both <u>Monoamine oxidases</u> and <u>Carbonic anhydrases</u> at concentrations between 3 - 40 µM



The AC50 from dose-response curves was used a PoD for MoS calculation



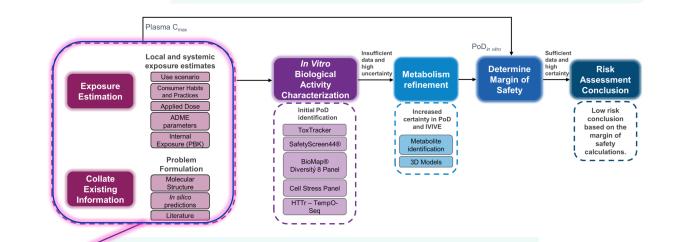
### NGRA for 0.1% coumarin in face cream: exposure estimation

### **Exposure Estimation**

- Total plasma Cmax values obtained from PBK model: 0.002 μM (mean), 0.005 μM (99<sup>th</sup> percentile)
- Stability assays indicated coumarin rapidly metabolized mainly via CYP2A6

#### Collate Existing Information

- Genotoxicity and skin sensitisation alerts for parent compound
- Hydroxylation predicted as main route of biotransformation
- Reactive metabolites (e.g. epoxides) predicted.
- Low bioactivity in ToxCast and Pubchem: binding to Carbonic Anhydrases and MAO-A/B reported
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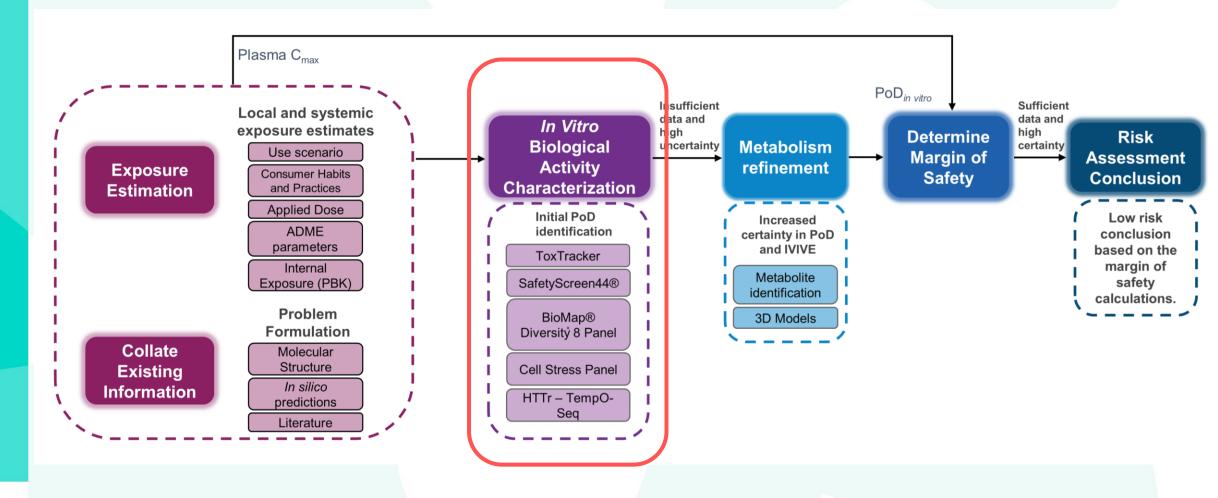


# STEP TWO

In vitro biological activity characterisation



# Next-Generation Risk Assessment case study workflow for 0.1% coumarin in face cream





# NGRA for 0.1% coumarin in face cream: In vitro biological activity characterisation: Genotoxicity assessment: ToxTracker

#### **Initial hypothesis:**

 DNA binding alerts for coumarin and metabolites





Standard ToxTracker assay +S9					
DNA damage		p53	Ox. s	stress	UPR
Bscl2	Rtkn	Btg2	Srxn1	Blvrb	Ddit3
	Standard ToxTracker assay -S9				
DNA da	amage	p53	Ox. s	stress	UPR
Bscl2	Rtkn	Btg2	Srxn1	Blvrb	Ddit3

Positive (>2-fold induction)
Weak activation (1.5 to 2-fold induction)
Negative (<1.5-fold induction)



#### **Results:**

- ToxTracker negative
- Reactive coumarin metabolite(s) could induce DNA lesions secondary to oxidative stress



# NGRA for 0.1% coumarin in face cream: biological activity characterisation In vitro binding and enzymatic assays – Eurofins SafetyScreen44

To investigate possible interactions between coumarin and the 44 key targets involved in drug attrition

#### **PERSPECTIVES**



# Reducing safety-related drug attrition: the use of *in vitro* pharmacological profiling

Joanne Bowes, Andrew J. Brown, Jacques Hamon, Wolfgang Jarolimek, Arun Sridhar, Gareth Waldron and Steven Whitebread

Abstract | In vitro pharmacological profiling is increasingly being used earlier in the drug discovery process to identify undesirable off-target activity profiles that could hinder or halt the development of candidate drugs or even lead to market withdrawal if discovered after a drug is approved. Here, for the first time, the rationale, strategies and methodologies for in vitro pharmacological profiling at four major pharmaceutical companies (AstraZeneca, GlaxoSmithKline, Novartis and Pfizer) are presented and illustrated with examples of their impact on the drug discovery process. We hope that this will enable other companies and academic institutions to benefit from this knowledge and consider joining us in our collaborative knowledge sharing.

Decreasing the high attrition rate in the drug discovery and development process is a primary goal of the pharmaceutical industry. One of the main challenges in achieving this goal is striking an appropriate balance between drug efficacy and potential adverse effects' as early as possible in order to reduce safety-related attrition, particularly in the more expensive late stages of clinical development. Gaining a better understanding of the safety profile of drug candidates early in the process is also crucial for reducing the likelihood of safety issues limiting the use of approved drugs, or even leading to their market withdrawal, bearing in mind the

target (or targets), whereas secondary effects are due to interactions with targets other than the primary target (or targets) (that is, off-target interactions). Off-target interactions are often the cause of ADRs in animal models or clinical studies, and so careful characterization and identification of secondary pharmacology profiles of drug candidates early in the drug discovery process might help to reduce the incidence of type A ADRs.

of type A ADRS.

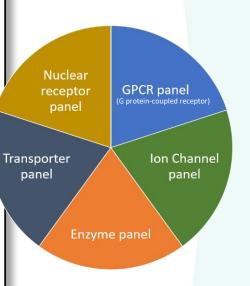
In vitro pharmacological profiling involves the screening of compounds against a broad range of targets (receptors, ion channels, enzymes and transporters) that are distinct from the intended.

safety testing of drug candidates and are designed to prevent serious ADRs from occurring in clinical studies.

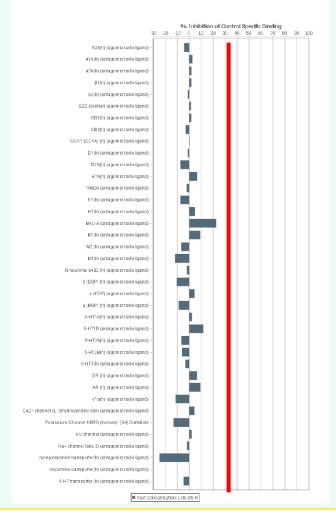
The only in vitro pharmacology assay that is absolutely required by regulatory authorities is one that measures the effects of new chemical entities on the ionic current of native  $(I_{\nu_e})$  or heterologously expressed human voltage-gated notassium channel subfamily H member 2 (KCNH2: also known as hERG)5. The mechanism by which blockade of hERG can elicit poten tially fatal cardiac arrhythmias (torsades de pointes) following a prolongation of the OT interval is well characterized7,8, and the seriousness of this ADR is one reason why this assay is a mandatory regulatory requi ment. Receptor binding studies are also recommended as the first-tier approach for the assessment of the dependence potential of novel chemical entities9

However, current regulatory guidance does not describe which targets should constitute an in vitro pharmacological profiling panel and does not indicate the stage of the discovery process at which in vitro pharmacological profiling should occur. Nevertheless, the general trend for most pharmaceutical companies is to perform this testing early in drug discovery to reduce attribute and to facilitate better prediction of ADRs in the later stages of drug discovery and development.

Here, for the first time, four major pharmaceutical companies (AstraZeneca, GlaxoSmithKime, Nowaris and Piñzer) share their knowledge and experiences of the innovative application of existing screening technologies to detect off-target interactions of compounds. The objective of this article is to describe the rationale and main advantages for the use of in vitro pharmacological profiling to discuss best practices and to.







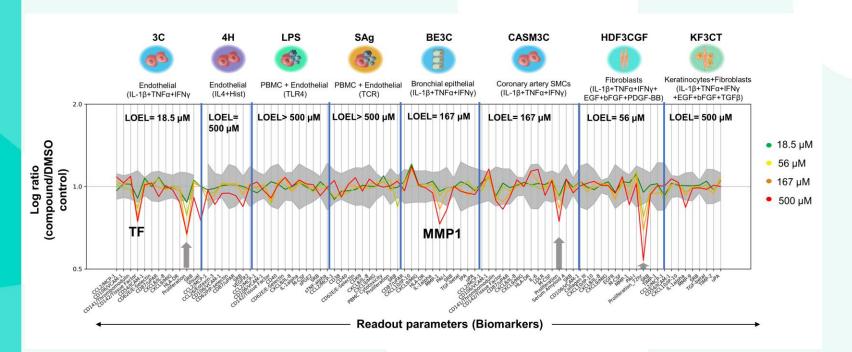
#### **Results:**

No significant activity was observed for all binding and enzymatic assay results



# NGRA for 0.1% coumarin in face cream: biological activity characterisation: Immunomodulatory screening assay - BioMap Diversity 8 Panel

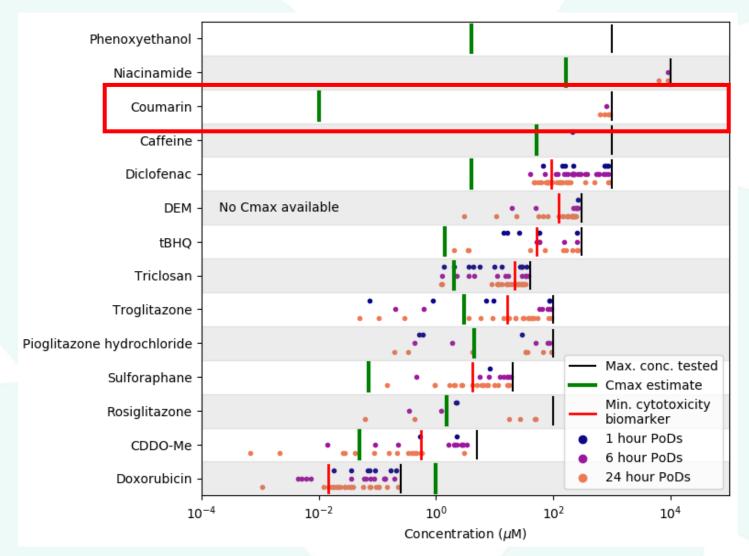
# To investigate possible effects on vascular inflammation, immune activation and tissue remodelling



Data suggested that
coumarin has no
immunomodulatory
effects at relevant
concentrations and is not
an anti-inflammatory
compound



### NGRA for 0.1% coumarin in face cream: biological activity characterisation In vitro cell stress panel



#### **Results:**

Coumarin not very active in comparison to known "high risk compounds" like doxorubicin

PoDs shown for HepG2 only

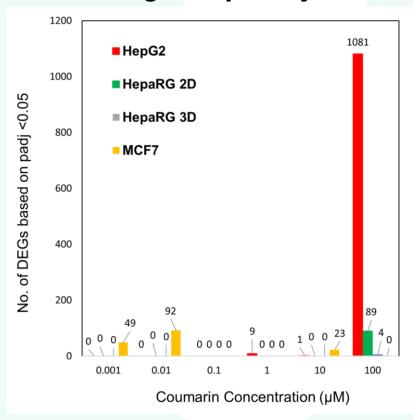


# NGRA for 0.1% coumarin in face cream: *In vitro* biological activity characterisation:

High-Throughput Transcriptomics (HTTr) using TempO-SEQ technology

Transcriptomics was applied as a broad non-targeted biological screen

# Differential expression analysis using DESeq2 analysis



#### **Results:**

Across the cell lines, treatment with coumarin resulted in limited gene-expression changes at concentrations below 100 µM, suggesting limited cellular effects at lower concentrations



### NGRA for 0.1% coumarin in face cream: Key results

### **Exposure Estimation**

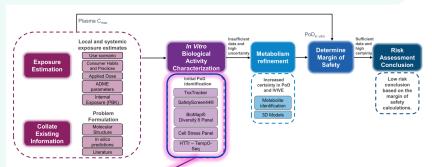
- Total plasma Cmax values obtained from PBK model: 0.002 μM (mean), 0.005 μM (99th percentile)
- Stability assays indicated coumarin rapidly metabolized mainly via CYP2A6

Collate Existing Information

- Genotoxicity and protein binding alerts for parent compound
- Hydroxylation predicted as main route of biotransformation
- Reactive metabolites (e.g. epoxides) predicted.
- Low bioactivity in ToxCast and Pubchem: binding to Carbonic Anhydrases and MAO-A/B reported
- Lowest PoD was 3 μM for carbonic anhydrase I (Figure 7)

In Vitro
Biological
Activity
Characterisation

- ToxTracker negative; weak activation of DNA damage reporters (only +S9)
- No immunomodulation potential
- Low bioactivity confirmed by binding/enzymatic assays, HTTr and cell stress panel.
- PoD range: 6-912 µM



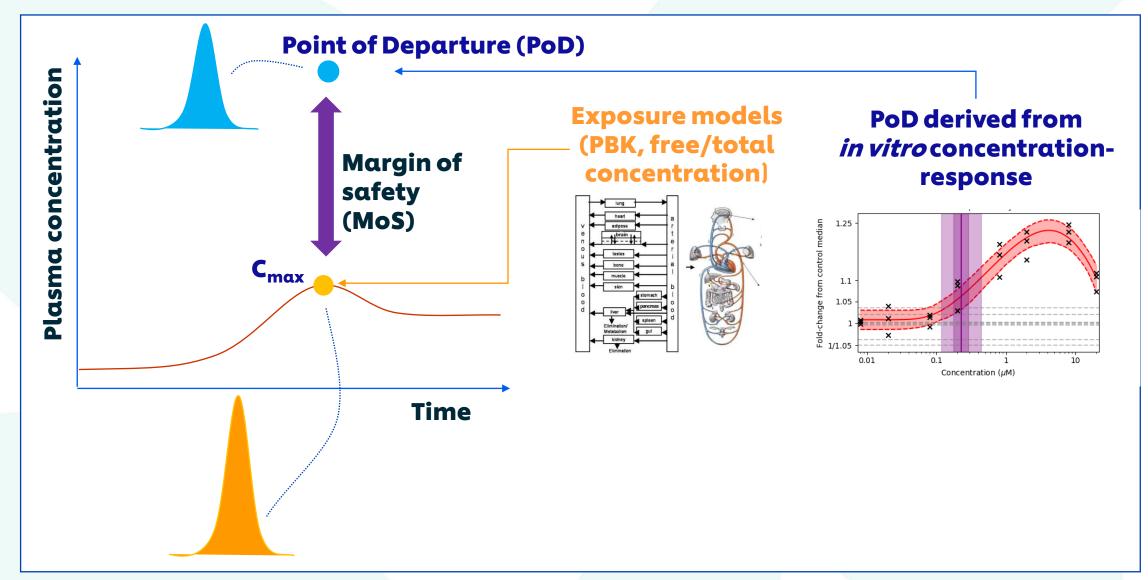


# STEP THREE

**Margin of Safety** 



### **Margin of Safety**





### NGRA for 0.1% coumarin in face cream: Preliminary Margin of Safety

Technology	Cell line/ Enzyme/Biomarker	Face cream Min. 5th percentile MoS	PoD provided as distribution?
Cell stress panel	HepG2 (ATP, 24h)	96738	Yes
Cell stress panel	NHEK (OCR 1h)	1330	Yes
HTTr	HepG2 (24h)	7223	No
HTTr	HepaRG (24h)	8864	No
Toxcast	MAO B (rat brain)	3711	No
PubChem	Carbonic Anhydrase Type I	706	No
PubChem	Carbonic Anhydrase Type II	2140	No
PubChem	Carbonic Anhydrase Type VI	14652	No

Based on total concentrations for both  $C_{max}$  and PoDs

- The lowest MoS across all assays was derived using the PoD (represented by Ki) for the inhibition of carbonic anhydrase I
- All PoD are higher than predicted exposure



### NGRA for 0.1% coumarin in face cream: Key results

# **Exposure Estimation**

- Total plasma Cmax values obtained from PBK model: 0.002 μM (mean), 0.005 μM (99th percentile)
- Stability assays indicated coumarin rapidly metabolized mainly via CYP2A6

#### Collate Existing Information

- Genotoxicity and protein binding alerts for parent compound
- Hydroxylation predicted as main route of biotransformation
- Reactive metabolites (e.g. epoxides) predicted.
- 90-100% coumarin predicted to be freely available in vitro
- Low bioactivity in ToxCast and Pubchem: binding to Carbonic Anhydrases and MAO-A/B reported
- Lowest PoD was 3 µM for carbonic anhydrase I (Figure 7)

# *In Vitro* Biological Activity Characterisation

- ToxTracker negative; weak activation of DNA damage reporters (only +S9)
- No immunomodulation potential
- Low bioactivity confirmed by binding/enzymatic assays, HTTr and cell stress panel.
- PoD range: 6-912 µM

Potential metabolitedriven bioactivity not addressed Determine Margin of Safety

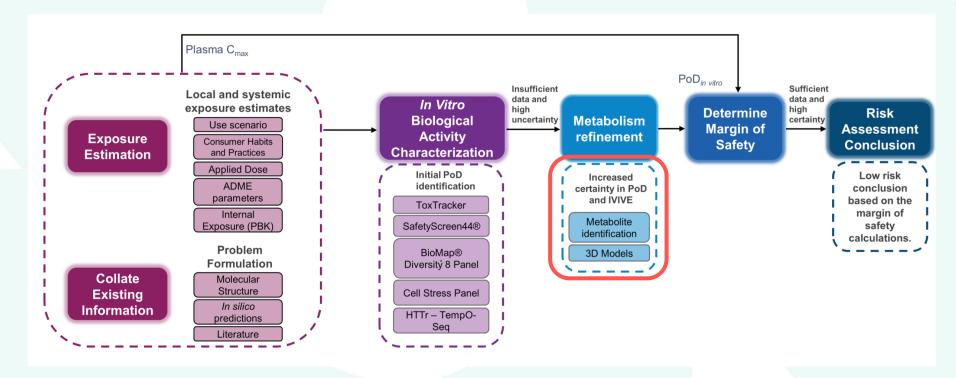
**Preliminary MoS** 

706 - 96738



### NGRA for 0.1% coumarin in face cream: Next steps for refinement

- Coumarin metabolism in primary human hepatocytes investigation of metabolites formed in human in vitro liver models
- 2. Short and long-term exposure in 3D tissues longer exposure durations in a 3D HepaRG model with potentially higher metabolic capacity and in vivo-like physiology than HepG2 cells



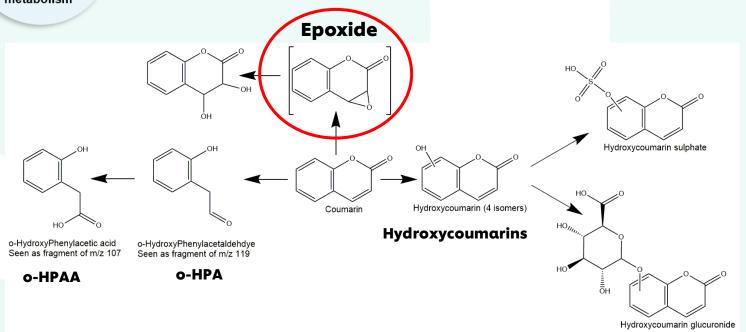


# NGRA for 0.1% coumarin in face cream: Coumarin metabolism in primary human hepatocytes



# Metabolism study to investigate if reactive metabolites are likely to be formed at consumer relevant concentrations

Human *In vitro* metabolism



Coumarin's proposed metabolic pathway based on the in vitro experiments.

#### **Results:**

- Coumarin is preferentially detoxified to hydroxycoumarins and respective glucuronides
- Reactive metabolites such as the epoxide, o-HPAA and o-HPA were only detected at the highest concentration (1mM)
- Not expected to be formed in vivo for our consumer exposure scenario



# NGRA for 0.1% coumarin in face cream: Short and long-term exposure in 3D tissues

#### To increase our confidence in the initial PoDs from the 2D cell models

Technology	Cell line/ Enzyme/Biomarker	Face cream Min. 5th percentile MoS	PoD provided as distribution?
Cell stress panel	HepG2 (ATP, 24h)	96738	Yes
Cell stress panel	NHEK (OCR 1h)	1330	Yes
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PubChem	Carbonic Anhydrase Type I	706	No
PubChem	Carbonic Anhydrase Type II	2140	No
PubChem	Carbonic Anhydrase Type VI	14652	No
Cell stress panel	HepaRG_3D (cell mem perm 168h)	9601	Yes
HTTr	HepaRG_3D_24h	9538	No



# NGRA for 0.1% coumarin in face cream: Key results

# **Exposure Estimation**

- Plasma Cmax obtained (range 0.002- 0.02 µM) from PBK models (Table 2)
- Stability assays indicated coumarin rapidly metabolized mainly via CYP2A6

#### Collate Existing Information

- Genotoxicity and protein binding alerts for parent compound
- Hydroxylation predicted as main route of biotransformation
- Reactive metabolites (e.g. epoxides) predicted.
- 90-100% coumarin predicted to be freely available in vitro
- Low bioactivity in ToxCast and Pubchem: binding to Carbonic Anhydrases and MAO-A/B reported
- Lowest PoD was 3 µM for carbonic anhydrase I (Figure 7)

# In Vitro Biological Activity Characterisation

- ToxTracker negative; weak activation of DNA damage reporters (only +S9)
- No immunomodulation potential
- Low bioactivity confirmed by binding/enzymatic assays, HTTr and cell stress panel.
- PoD range: 6-912 µM
- Potential metabolitedriven bioactivity not addressed

# Metabolism refinement

- Hydroxylation confirmed as main route of biotransformation at 10 µM
- Reactive metabolites not formed at consumer relevant exposures
- Low bioactivity also found in a metabolic competent cell model (HepaRG 3D)
- PoDs range: 41-871
   µM (Table 4 and 5).

Determine Margin of Safety

**Updated MoS** 

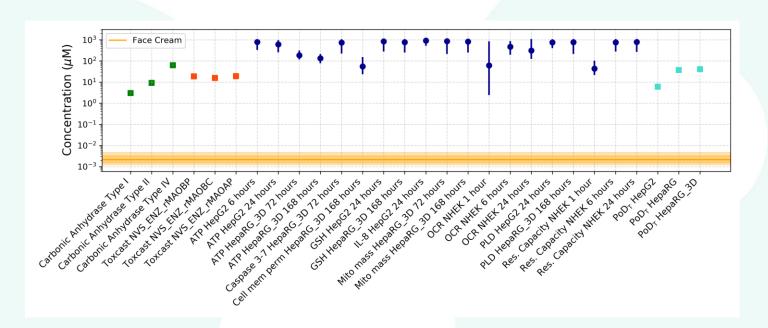
9538-9601

Preliminary MoS

706 - 96738



#### NGRA for 0.1% coumarin in face cream: Risk assessment conclusion

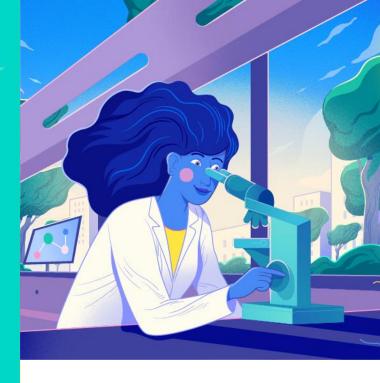


- The predicted  $C_{max}$  values for face cream were lower than all PoDs with a MoS (the  $5^{th}$  percentile) higher than 100
- Coumarin is not genotoxic, does not bind to any of the 44 targets and does not show any immunomodulatory effects at consumer relevant exposures
- Weight of evidence suggests that the inclusion of 0.1% coumarin in face cream is low risk for the consumer



# Case study examples

- 1) Systemic effects
- 2) Local effect skin sensitisation





# Case study: hypothetical example for 0.1% coumarin in face cream

Reynolds et al. (2021). A hypothetical skin sensitisation next generation risk assessment for coumarin in cosmetic products. Reg. Tox. Pharm., 127, 2021.

Regulatory Toxicology and Pharmacology 127 (2021) 105075



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journal homepage: www.elsevier.com/locate/yrtph





#### A hypothetical skin sensitisation next generation risk assessment for coumarin in cosmetic products

G. Reynolds, J. Reynolds, N. Gilmour, R. Cubberley, S. Spriggs, A. Aptula, K. Przybylak, S. Windebank, G. Maxwell, M.T. Baltazar

Unilever Safety and Environmental Assurance Centre, Colworth Science Park, Sharnbrook, Bedfordshire, MK44 1LQ, UK

#### ARTICLE INFO

Handling Editor: Dr. Lesa Aylward

Keywords and highlights: Skin sensitisation Allergic contact dermatitis Next generation risk assessment Non-animal alternatives New approach methodologies Consumer exposure Uncertainty analysis Decision making Metabolism

#### ABSTRACT

Next generation Risk Assessment (NGRA) is an exposure-led, hypothesis-driven approach which integrates new approach methodologies (NAMs) to assure safety without generating animal data. This hypothetical skin allergy risk assessment of two consumer products - face cream containing 0.1% coumarin and deodorant containing 1% coumarin - demonstrates the application of our skin allergy NGRA framework which incorporates our Skin Allergy Risk Assessment (SARA) Model. SARA uses Bayesian statistics to provide a human relevant point of departure and risk metric for a given chemical exposure based upon input data that can include both NAMs and historical in vivo studies. Regardless of whether NAM or in vivo inputs were used, the model predicted that the face cream and deodorant exposures were low and high risk respectively. Using only NAM data resulted in a minor underestimation of risk relative to in vivo. Coumarin is a predicted pro-hapten and consequently, when applying this mechanistic understanding to the selection of NAMs the discordance in relative risk could be minimized. This case study demonstrates how integrating a computational model and generating bespoke NAM data in a weight of evidence framework can build confidence in safety decision making.

#### 1. Introduction

Skin sensitisation, which can ultimately result in development of allergic contact dermatitis (ACD), is a common human health problem (Alinaghi et al., 2019). To ensure consumer safety, particularly for proliferation of antigen-specific T-cells (OECD 2014). The developed NAMs adopted by the OECD align to the first three key events and can be used in various combinations incorporated into defined approaches (DA) within a framework for a weight of evidence (WoE) chemical safety assessment (Gilmour et al., 2020; Kleinstreuer et al., 2018; OECD 2017,

### 0.1% COUMARIN IN FACE CREAM (NEW FRAGRANCE)

#### **Assumptions:**

- **EU Market**
- 100% purity
- no in vivo data was available such as animal data, History of Safe Use (HoSU) or Clinical data
- no use of animal data in Read Across
- In silico alerts known to be based on animal or in vivo data or on the structure of Coumarin itself were excluded

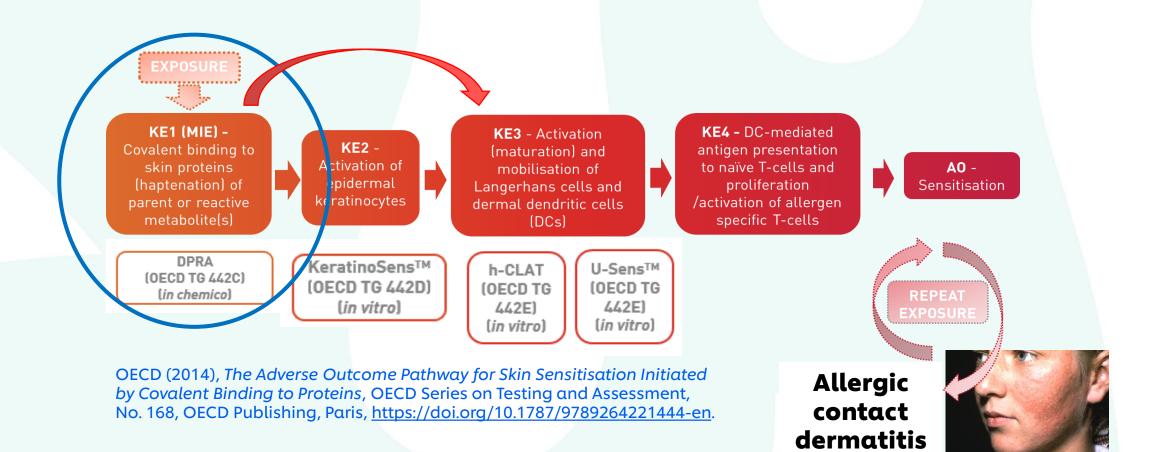






#### **Initial hypothesis:**

Protein binding alerts for coumarin and metabolites





#### Step 1: Generation of *in vitro* results for Coumarin

	DPRA (TG442C)		KeratinoSens (TG 442D)	h-CLAT (TG 442E)		U-SENS (TG 442E)
Call	-ve		+ve	+\	re	+ve
Model Input	<b>%cys</b> depletion	<b>%lys</b> depletion	<b>EC1.5</b> (μM)	<b>CD54</b> (EC200 μg/mL)	<b>CD86</b> (EC150 μg/mL)	<b>CD86</b> (EC150 μg/mL)
Results	1.3	0	187.5	<178	>637	95.5

#### **Initial results:**

- Coumarin is a skin sensitiser
- to metabolites (-ve DPRA)



# Step 2. Generation of PoD for risk assessment- Skin allergy risk assessment (SARA) Defined approach (DA)

• The SARA DA is a Bayesian probabilistic model, which estimates the human sensitiser potency via a prediction of a HRIPT 1% sensitising dose ( $ED_{01}$ ) (i.e PoD) for a selected chemical.

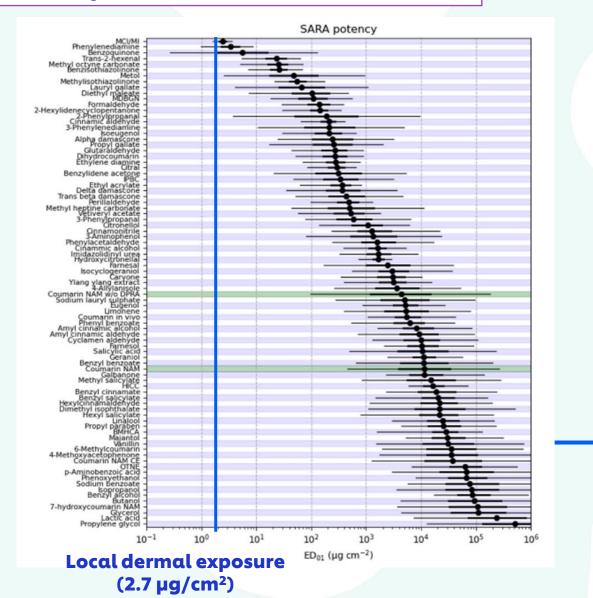
- The SARA model uses Bayesian statistics to infer a probability that a consumer exposure to some chemical can be considered low risk, to inform risk assessment decisions.
- The SARA Model uses a database of public NAM data covering AOP KEs 1-3, and historic LLNA and HRIPT data for the AOP AO.

#### **SARA Model Inputs**

- Historical Local lymph node assay (LLNA)
- Historical Human repeated insult patch test (HRIPT)
- In vitro data: DPRA (TG442C), KeratinoSens (TG 442D), h-CLAT (TG 442E), U-SENS (TG 442E)
- ❖ First publication dataset of 30 chemicals expanded to 53 core + 49 in vitro only



#### Step 2: PoD for risk assessment



The ED01 for coumarin has a ranging from 420 - 260,000 µg/cm²



#### **Results:**

- lower than the predicted ED<sub>01</sub>
- Low risk conclusion



### **Concluding remarks**

- NGRA is a framework of non-standard, bespoke data-generation, driven by the risk assessment questions
  - Exposure led
  - Human relevant
  - · in silico
  - · in vitro
  - weight of evidence
- Margin of safety is determined by the ratio of human exposure to the point of departure for the most sensitive assay
- NGRA tools are available now and research into more approaches continues



### Recordings of past webinars are available:

# https://www.afsacollaboration.org/event/





#### **Invitation:**

Bianca Marigliani, an AFSA partner, will present the AFSA E&T Platform, 27<sup>th</sup> May, 14:40 (sala D):

"Animal-Free Safety Assessment of Cosmetics: a global education and training program"



# Acknowledgements

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Congratulations to RENAMA on your 10th anniversary!

