



# Case studies assessing suitability of *in silico* modelling tools and read-across approaches for nanomaterial hazard assessment

SETAC Europe 34<sup>th</sup> annual meeting, Seville

9<sup>th</sup> May 2024

**Laurence Deydier Stephan**<sup>1</sup>, Marianne Matzke<sup>1</sup>, Andrea Richarz<sup>1</sup>, Dimitra-Danai Varsou<sup>2</sup>, Maria Antoniou<sup>2</sup> & Antreas Afantitis

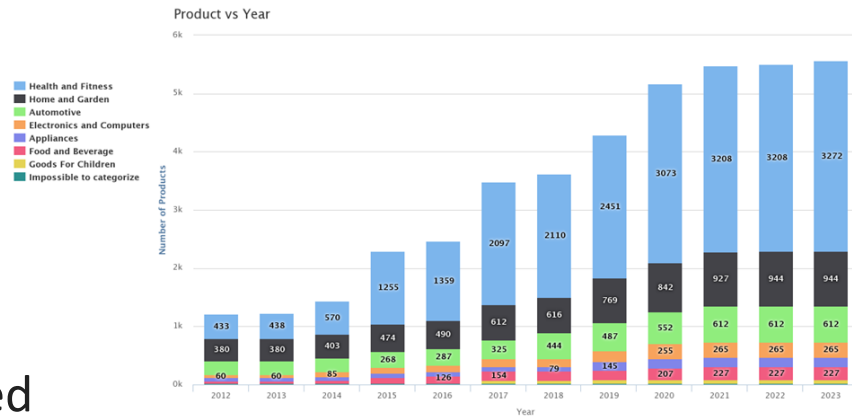
<sup>1</sup>European Chemicals Agency, Finland, <sup>2</sup>NovaMechanics MIKE, Piraeus Greece,

<sup>3</sup>NovaMechanics Ltd., Nicosia 1046, Cyprus

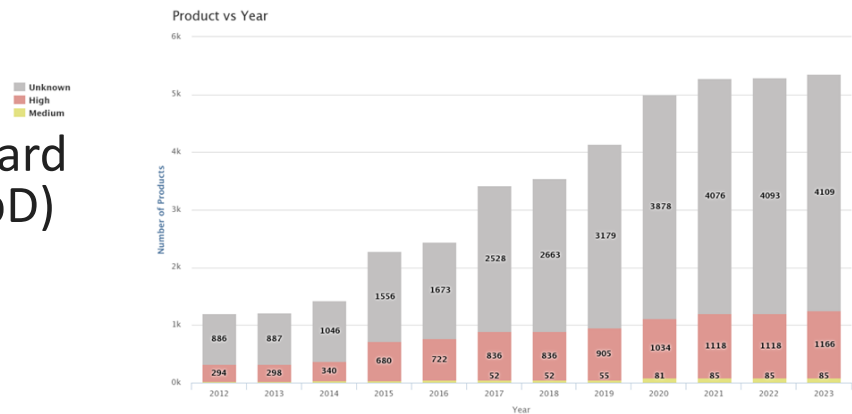


# Background

- Numerous novel nanomaterials (NMs) are used in a wide spectrum of commercial products.
- Many NMs are not yet thoroughly evaluated for potential harmful biological effects.
- Development and application of reliable, accurate computational models and tools.
- Contribution to the speeding up of the hazard & risk assessment and safety-by-design (SbD) of existing and novel NMs.



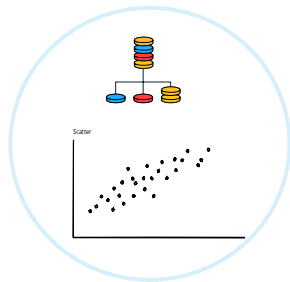
<https://nanodb.dk/>, 3/5/2023



<https://nanodb.dk/>, 3/5/2023

# Objective

- Conduct a study compiling **reliably and transparently** the information on the **existing computational approaches** for the **hazard/risk assessment and SbD of novel Nanomaterials and their (regulatory) relevance** (project supported by European Union Observatory for Nanomaterials)



nanoQSAR



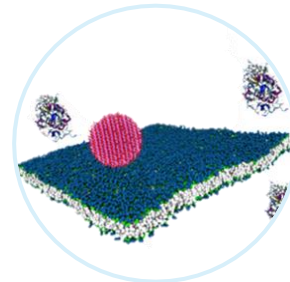
Grouping/read-across



AOPs

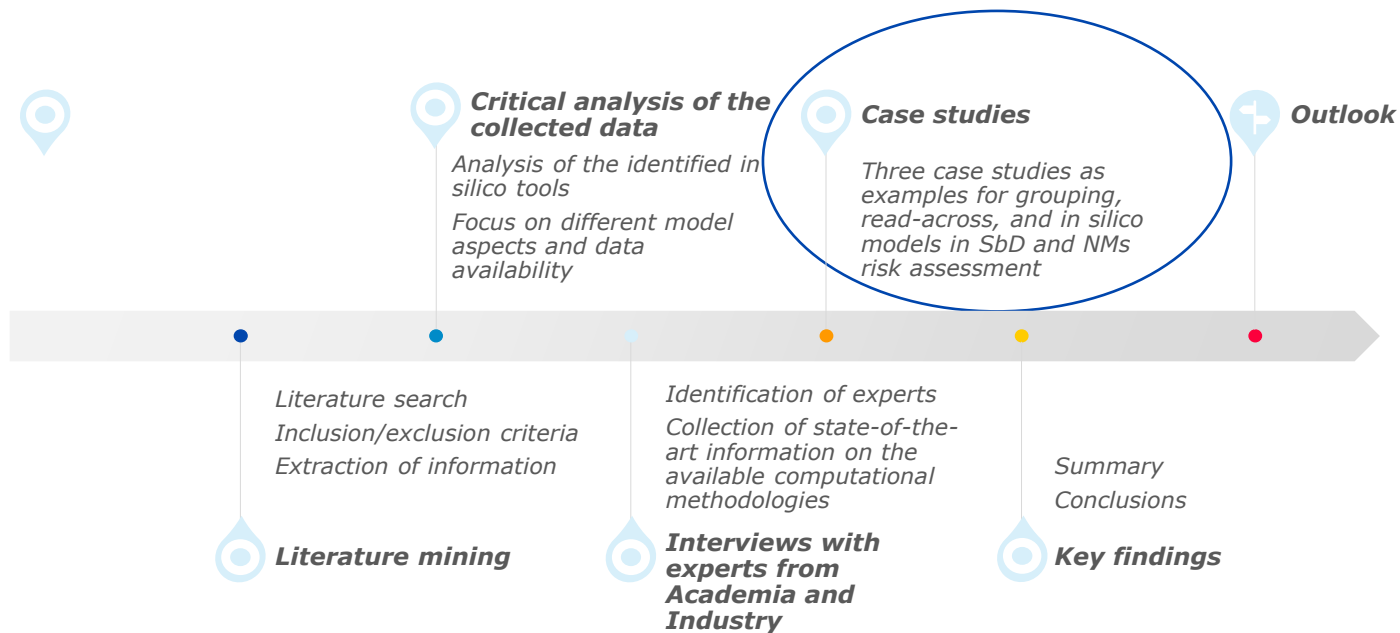


PBPK models



Simulations

# Project outline



# Case studies

**Read-across  
models for the  
assessment of the  
anti-microbial  
activity of various  
carbon-based NPs**

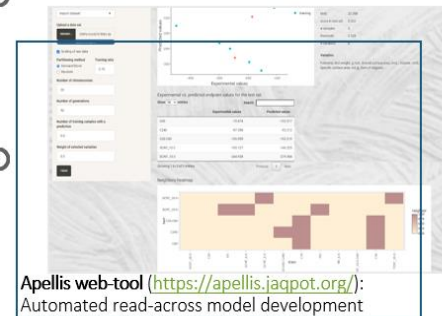
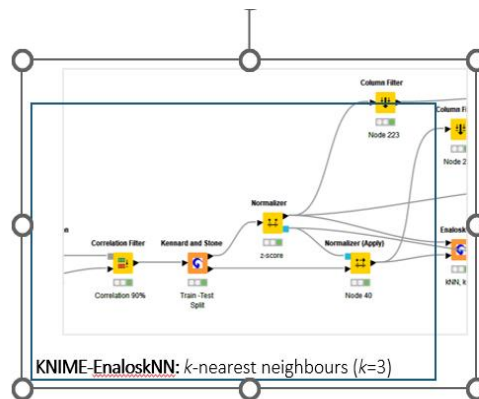
**Assessment of  
different tools for  
the prediction of  
TiO<sub>2</sub>-based NMs'  
endpoints**

**Combination of  
nanoinformatics  
models into  
SAPNets**

# Case study 1

## Development of read-across models for the assessment of the anti-microbial activity of various carbon-based NPs

- Dataset derived from Zhang *et al.*, 2021 (<https://doi.org/10.1016/j.ecoenv.2021.112357>) physicochemical descriptors of 17 CNPs
- Interactions between carbon-based NPs and a SARS-CoV-2 RNA fragment (potential of fragment stabilisation and application in anti-microbial coatings)
- Endpoint: Total Potential Energy ( $E_{int}$ )
- Modelling: Three read-across methods from the literature
- Consensus Model: Averages predictions of 2 individual models to avoid biases



# Case study 1

## Comparisons between models

- EnalokNN: user-friendly environment, data visualisation and flexibility, Enalos+ nodes require a license scheme.
- Apellis: user-friendly interface, automatic training (menus and buttons), convergence delays cause server timeouts.
- DTC-Lab: complex software and training material, program terminates easily, results are not explained.

CNP	Actual values [kJ/mol]	EnalokNN [kJ/mol]	DTC-Lab: Euclidean kernel [kJ/mol]	DTC-Lab: Gaussian kernel [kJ/mol]	DTC-Lab: Laplacian kernel [kJ/mol]	Consensus [kJ/mol]
C36	-109	-110	-109	-109	-113	-110
C60	-80	-101	-107	-100	-102	-102
C70	-96	-92	-87	-91	-101	-93
SCNT(6,6)	-153	-192	-158	-159	-163	-168
DCNT(6,6)	-262	-274	-249	-253	-286	-266
	MAE	15.4	10.8	8.2	12.9	9.1
	RMSE	20.4	14.0	10.5	15.6	12.2
	$Q_{\text{ext}}^2$	0.96	0.98	0.99	0.97	0.98

CNP	Actual values [kJ/mol]	Apellis predictions [kJ/mol]
C60	-80	-102
C240	-87	-92
C20@C60	-100	-102
SCNT(10,0)	-185	-146
DCNT(10,0)	-245	-280
	MAE [kJ/mol]	20.6
	RMSE [kJ/mol]	25.5
	$Q_{\text{ext}}^2$	0.93

# Case study 2

## Assessment of different tools for the prediction of TiO<sub>2</sub>-based NMs' endpoints

### QSAR for nano-mixtures

- Toxicological endpoints (EC<sub>50</sub>, immobilisation) of TiO<sub>2</sub>-based nano-mixtures.
  - Input data through menus and sliders.
  - Ecotoxicity risk level indication after results are generated
  - Complex manual, lacks explanation for “positive” and “negative” interactions.
  - Missing Domain of Applicability (DoA)
- > **Connections between phenotypic entities based on their effects on genes.**

### NanoMixHamster

- Cytotoxicity prediction of TiO<sub>2</sub> towards CHO-K1 cells using the SAPNet approach.
  - Step 1: Input of metallic NM composition, its additive electronegativity value is acquired.
  - Step 2: Toxicity prediction and graph visualisation of the DoA
  - Simple web application, tool is described in original publication in detail.
- > **Lung Exposure dose calculator**

### NanoToxRadar

- Cytotoxicity to A549 cell line and zeta potential predictions.
  - Required input data: NM core composition, doping parameters and coating materials.
  - Risk indication according to cytotoxicity prediction.
  - Simulation of TiO<sub>2</sub> internal biodistribution
  - Missing manual, warning limits explanation and results interpretation.
- > **IATA**



# Case study 2

## Assessment of different tools for the prediction of TiO<sub>2</sub>-based NMs' endpoints

- Fast sensitivity analyses or virtually screen newly synthesised NMs.
- Suitable for use in an SbD framework → explore the correlations between different properties and the nanotoxicity → define the limits within which NMs are safe.

- Missing domain of applicability limits (even if it exists in the relevant publication).
- Missing tutorials (or too technical manuals, missing results interpretation).
- Inconsistencies with original publications.
- Slow execution.
- Need for frequent update.

NanoMix Hamster

Application for predicting cytotoxicity of TiO<sub>2</sub>-based multicomponent nanomaterials.

Overview

Network

Publication

Workflow diagram: Step 1 (Input of data) → Step 2 (Prediction of additive electronegativity) → Summary (Decision diamond) → Dataset Generator (Text box).

Text box: Here is the first step of SAPNet dedicated to predicting the toxicity of nanoparticles towards the CHO-K1 cell line. Please describe your nanomaterial. It should be TiO<sub>2</sub>-based nanomaterial, which is modified with Au, Ag, Pt, or Pd nanodots (Au<sub>n</sub>, TiO<sub>2</sub>).

Please fill in the form below to predict additive electronegativity of the metallic system,  $\chi_{mix}$

Metal 1 name	Metal 2 name
Au	Pt
Metal concentration (%)	Metal concentration (%)
0.5	0.5

Predicted value of additive electronegativity  $\chi_{mix} = 1.4$

NanoTox Reader

Nanotoxicity Prediction Program

3D model of a TiO<sub>2</sub> nanoparticle with a core of Cd (3%).

Parameters:

- Shape:
- Core:
- Doping:
- Coating:
- Diameter:

R2MIE nano

Network browser

Center node: TiO<sub>2</sub>

Center size: 21

Center color: Red

Network: 7

Nodes: 7

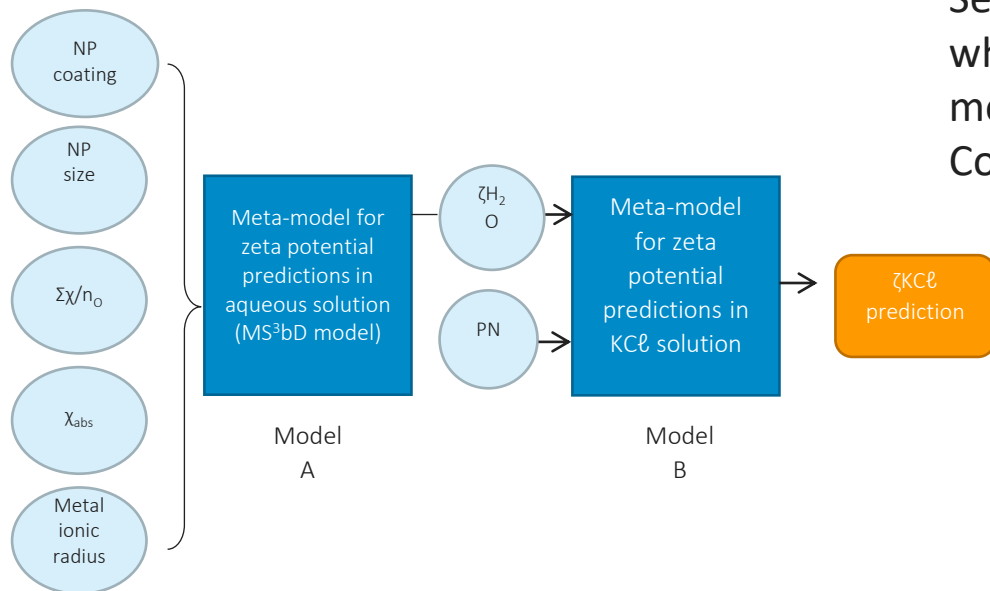
Edges: 1

Network description: Network description: This network contains nodes for TiO<sub>2</sub>, Ag, Au, Pt, Pd, and Au-Ag, Au-Pt, Au-Pd, and Au-Ag-Pt. The network is a complete graph where every node is connected to every other node.

Network diagram showing connections between nodes: Ag, Au, Pt, Pd, Au-Ag, Au-Pt, Au-Pd, Au-Ag-Pt, and TiO<sub>2</sub>.

# Case study 3

## Combination of nanoinformatics models into SAPNets



Series of interconnected predictive models, where descriptors are predicted by other “meta-models”, as proposed by Rybińska-Fryca *et al.*  
Combination of two models:

Model A: MS<sup>3</sup>bD model by Papadiamantis *et al.*, a fully validated *k*NN predictive model for zeta potential in water (pH=7).

Model B: MLR model by Wyrzykowska *et al.* that predicts zeta potential in KCl solution.

$$\zeta_{KCl} = 3.98 + 21.68 \cdot \zeta_{H_2O} + 7.88 \cdot PN$$

# Case study 3

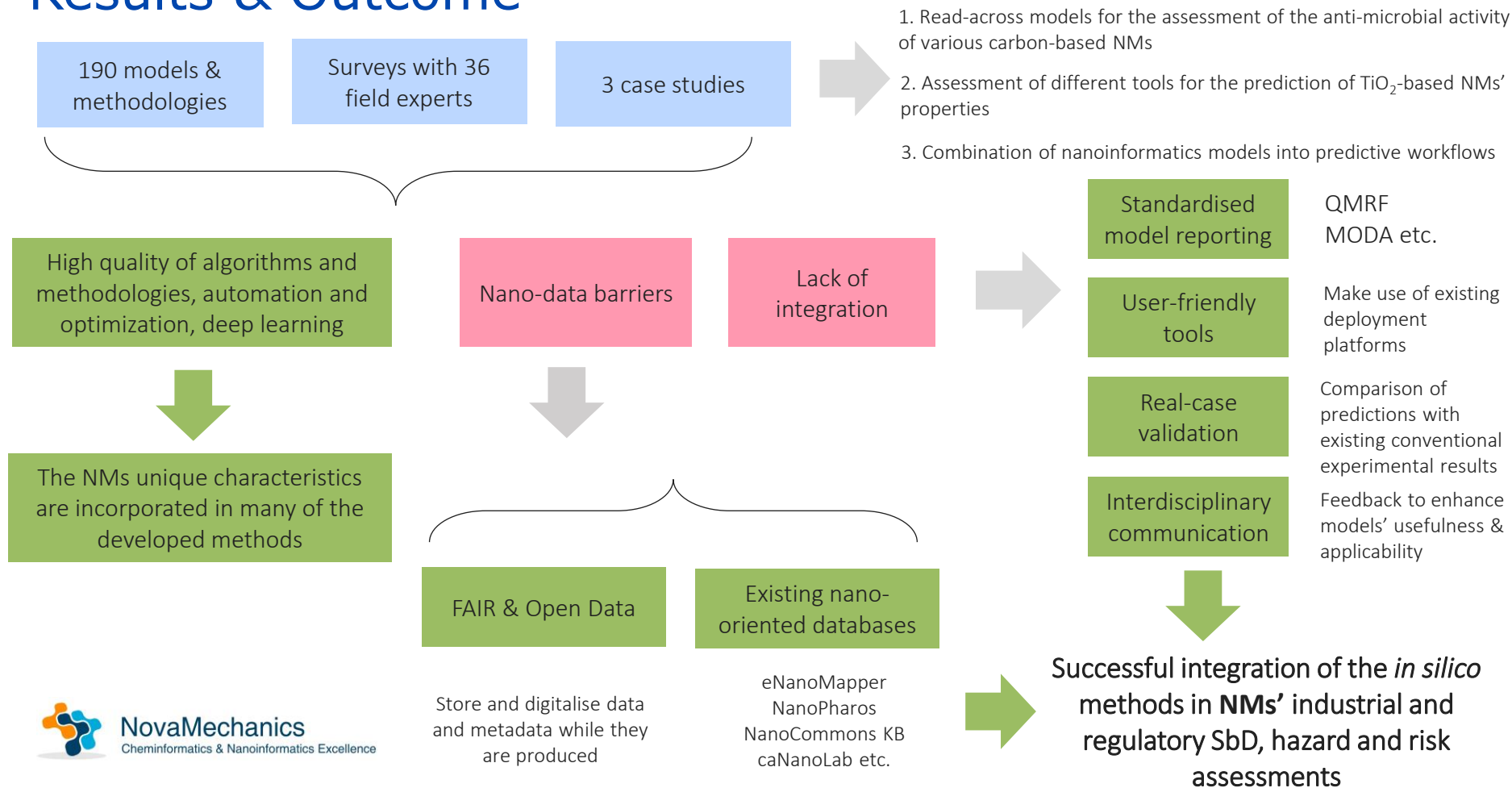
- Solely  $\zeta\text{H}_2\text{O}$  predictions characterised as “reliable” in model A were fed into model B.
  - Domain of Applicability definition and model requirements should be clearly stated.
- Unknown experimental conditions that regulate NM behaviour (pH, temperature etc.), impossible to determine the models’ compatibility.
- Large errors propagation.
  - Experimental and computational data and meta-data sharing is essential.
  - SAPNets-models integration can be applied to the filling of nano-data gap while no experimental data generation is needed.

Rybińska-Fryca, A., *et al.* Structure-activity prediction networks (SAPNets): A step beyond Nano-QSAR for effective implementation of the safe-by-design concept. *Nanoscale* **12**, 20669–20676 (2020).

Papadiamantis, A. G. *et al.* Computational enrichment of physicochemical data for the development of a  $\zeta$ -potential read-across predictive model with Isalos Analytics Platform. *NanoImpact* **22**, 100308 (2021).

Wyrzykowska, E., *et al.* Development of a novel *in silico* model of zeta potential for metal oxide nanoparticles: A nano-QSPR approach. *Nanotechnology* **27**, 1–8 (2016).

# Results & Outcome



# Key findings

## Is it sensible ?

- Reliable NMs models with satisfactory quality levels
- Including various unique NM characteristics
- Ongoing work to integrate optimisation functions and user-friendly tools

## Is it accessible ?

Public repositories, platforms and databases applying FAIR principles increase their visibility and use

## Can it be used/integrated in the research & regulatory context ?

- through fully available, consistent and standardised data and meta data
- through transparent & clear communication on models' development/validation/DoA/Manuals
- already used as NAMs and considered within NAMS for NM framework program (ECHA/EFSA)

[https://euon.echa.europa.eu/documents/2435000/3268573/ECHA\\_2022\\_61\\_study\\_report.pdf/739900b3-bd9c-a4f0-d3bc-88f4aa801f68?t=1694691997584](https://euon.echa.europa.eu/documents/2435000/3268573/ECHA_2022_61_study_report.pdf/739900b3-bd9c-a4f0-d3bc-88f4aa801f68?t=1694691997584)

# Questions?



## Thank you

[info@novamechanics.com](mailto:info@novamechanics.com)

[laurence.deydier@echa.europa.eu](mailto:laurence.deydier@echa.europa.eu)

[echa.europa.eu/subscribe](https://echa.europa.eu/subscribe)



Connect with us



[echa.europa.eu/podcasts](https://echa.europa.eu/podcasts)



European Chemicals Agency



[@one\\_healthenv\\_eu](https://www.instagram.com/one_healthenv_eu)



[@EU\\_ECHA](https://twitter.com/EU_ECHA)



[@EUECHA](https://www.facebook.com/EUECHA)



[EUchemicals](https://www.youtube.com/EUchemicals)