

CASE STUDY: using and integrating two QSAR models for assessing *in vitro* mutagenicity in bacteria

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Aim: Prediction of in vitro gene mutation in bacteria with two (Q)SAR models and results documentation in IUCLID

Target molecule: 2,3,4-Trichlorobiphenyl

Knowledge-based model: VEGA – Mutagenicity (Ames test) model (ISS)

Statistical model: Danish (Q)SAR Database - Bacterial reverse mutation test (Ames test in S. typhimurium in vitro) – Battery model

Select input method: SMILES notation - c1ccc(cc1)c2ccc(c(c2Cl)Cl)Cl



Models for in vitro gene mutation in bacteria in the CONCERT REACH gateway





Models for in vitro gene mutation in bacteria

VEGA: four models for Ames test

- CAESAR Hybrid model (statistical + knowledge-based)
- KNN-Read-Across read-across model
- ISS knowledge-based structural alerts (Benigni-Bossa rulebase)
- SarPy-IRFMN statistical structural alerts



Danish (Q)SAR Database: 15 statistical models and 2 knowledge-based alert profilers for Ames test

• Bacterial reverse mutation test (Ames test in S. typhimurium in vitro)

To be considered only if the Bacterial reverse mutation test model gives a **positive in domain outcome**

- Direct acting Ames mutagens (without S9)
- Base pair Ames mutagens
- Frame shift Ames mutagens
- Potent Ames mutagens, reversions \geq 10 times controls

3 models for each

endpoint/mechanism +

Battery model

Combining the 3 outcomes

2 knowledge-based profilers from OECD QSAR Toolbox

DNA alerts for AMES by OASIS, alerts in parent only (OECD QSAR Toolbox v.4.2 Profiler)

In vitro mutagenicity (Ames test) alerts by ISS, alerts in parent only (OECD QSAR Toolbox v.4.2 Profiler)

Models for in vitro gene mutation in bacteria

How to select the appropriate model(s) for my substance?

- A priori selection is generally not possible
- However, **experience in using the models** might suggest which could give more reliable results for certain type of substances (e.g., industrial chemicals, active substances, etc.)
- Information on **compliance** of the target molecule **with the applicability domain of the model**
- Comparison with similar molecules with available experimental results
- It is generally required to use multiple and different models for evaluating the same endpoint *

Expert analysis of the results and supporting information is needed

Both VEGA and Danish (Q)SAR database provide information on applicability domain compliance and similar molecules can be extracted and analyzed

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VEGA: introduction





VEGA: Virtual models for Evaluating the properties of chemicals within a Global Architecture

- Developed mainly by Mario Negri Institute (Milan) and Kode s.r.l. (Pisa)
- Free platform developed based on contributions from EU projects
- Includes 110 statistical and knowledge-based (Q)SAR models for the prediction of (eco)toxicity, environmental fate and physico-chemical properties of chemicals.

VEGA: running predictions





GA in silico p	olatform - version 1.2.3	>
	Select models	VEG/
- Çç	Filter models: All available endpoints	•
SELECT	Mutagenicity (Ames test) Select all models Imagenicity (Ames test) model (CAESAR) - v. 2.1.14 Imagenicity (Ames test) model (ISS) - v. 1.0.3 Imagenicity (Ames test) model (ISS) - v. 1.0.3 Imagenicity (Ames test) model (ISS) - v. 1.0.4	.0.0
REDICT	Developmental toxicity Select all models	v

i



VEGA: analysis of the results

Knowledge-based model: Mutagenicity (Ames test) model (ISS) 1. Prediction Summary



Prediction for compound Molecule 0 -





- Colour code for prediction (e.g., red =toxic, green = non-toxic)
- Stars for reliability from 1 (low) to 3
- Summary of the evaluation

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VEGA: analysis of the results



Applicability Domain Index (ADI) ranges from 0 (not in AD) to 1 (in AD)

The ADI is calculated based on other indices, **each one taking into account a particular issue** of the applicability domain (AD)



Number of considered similar molecules, number and type of indexes and thresholds are **model dependent** Info in QMRF

VEGA: Example of critical evaluation of the automated AD / CONCERTING EXPERIMENTAL DATA AND IN SILICO MODELS FOR REACH reliability evaluation IEE47 CIE /IT /000464 3.1 Applicability Domain: Similar Compounds, with Predicted and Experimental Values Compound #1 Prediction: Reliability: 🍃 🍃 😭 CAS: N.A. Prediction is NON-Mutagenic, but the result shows some critical aspects. Dataset id:441 (Training Set) SMILES: c1cc(c(cc1c2ccc(c(c2Cl)Cl)Cl)Cl)Cl)Cl which require to be checked Similarity: 0.925 some similar molecules found in the training set have experimental values Experimental value : NON-Mutagenic < that disagree with the predicted value Predicted value : NON-Mutagenic Compound #2 CAS: 91-94-1 Two most similar Dataset id:458 (Training Set) SMILES: Nc1ccc(cc1Cl)c2ccc(N)c(c2)Cl Similarity: 0.86 Compound: Molecule 0 molecules considered Experimental value : Mutagenic Compound SMILES: c1ccc(cc1)c2ccc(c(c2CI)CI)CI Predicted value : Mutagenic Experimental value: Predicted Mutagen activity: NON-Mutagenic Alerts (not found also in the target): SA28 Primary aromatic amine, hydroxyl amine and its Structural Alerts: derived esters (with restrictions) Reliability: The predicted compound could be out of the Applicability Domain of the model Remarks none CAS: 72-54-8 Dataset id:473 (Training Set) SMILES: c1cc(ccc1C(c2ccc(cc2)Cl)C(Cl)Cl)Cl Global AD Index Similarity: 0.828 Experimental value : NON-Mutagenic AD index = 0.801Predicted value : Mutagenic Explanation: The predicted compound could be out of the Applicability Domain of the model. Alerts (not found also in the target): SA8 Aliphatic halogens Affected by one Compound #4 Similar molecules with known experimental value Similarity index = 0.889 CAS: 72-55-9 Dataset id:176 (Training Set) molecule, with a Explanation: Strongly similar compounds with known experimental value in the training set have been ... SMILES: c1cc(ccc1C(c2ccc(cc2)Cl)=C(Cl)Cl)Cl Similarity: 0.815 Experimental value : NON-Mutagenic Accuracy of prediction for similar molecules different alerts profile Predicted value : NON-Mutagenic Accuracy index = 1Explanation: Accuracy hoon ai te or similar molecules round in the Compound #5 CAS: 50-29-3 concordance for similar molecules Dataset id:751 (Training Set) Concordance index = 0.521 SMILES: c1cc(ccc1C(c2ccc(cc2)Cl)C(Cl)(Cl)Cl)Cl Similarity: 0.813 Explanation: some similar molecules found in the training set have experimental values that disagree with the Experimental value : NON-Mutagenic Predicted value : NON-Mutagenic predicted value. Atom Centered Fra ilarity check A higher reliability could be assigned to the negative prediction, ACF index = 1 Explanation: all atom centered fragment of the compound have been found in the compounds of the training also considering that all other similar molecules (mostly with the set. 13 same "no alerts" profile) are experimentally negative

VEGA: take home message

CONCERTREACH CONCERTING EXPERIMENTAL DATA AND IN SILICO MODELS FOR REACH

- Full documentation of all models is available, in QMRF format
- Supporting information (AD compliance, similar molecules) is provided, allowing expert evaluation
- AD compliance is affected by identified similar molecules
- Current **similarity evaluation does not consider important parameters** (e.g. different alerts, leading to different mechanism)
- Automated AD compliance check is not perfect and requires user critical check
 - > This affects other tools as well, including commercial ones

Relevant for REACH dossier preparation in IUCLID

A novel tool called VERA has been developed, aiming also at improving similarity evaluation and AD compliance check (Next presentation, today)



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Danish (Q)SAR Database: introduction





- Developed by the Danish Technical University
- Freely accessible online database of (Q)SAR predictions for a little over 650.000 chemical organic mono-constituent structures
- For each single contained substance, the database generates a report upon user request, which contains more than 200 (Q)SAR results
- Offers advanced searching tools for identifying potential similar molecules for supporting (Q)SAR predictions or candidate analogues for read-across



Danish (Q)SAR Database



Danish (Q)SAR Database: gathering (Q)SAR results



CONCERTING EXPERIMENTAL DATA

Danish (Q)SAR Database: results for *in vitro* gene mutation in bacteria



In vitro Genotoxicity - 1	Bacterial I	Reverse M	utation Test (<u>Ames</u> test)				3 models + ba endpoints to	attery (consensus) be considered on) for Ames test and ly if the outcome fo	or Ames is Positive		
		Exp	Battery	CASE Ultra	Leadscope	SciQSAR			and in domain (POS_IN)				
Ames test in S. typhimurium	(in vitro)		NEG_IN	NEG_IN	NEG_IN	NEG_IN							
*Direct Acting Mutagens (wi	ting Mutagens (without S9) N/A NEG_IN NEG_IN NEG_IN NEG_IN						The target molecule was evaluated as compliant with AD of all						
*Base-Pair Ames Mutagens		N/A	NEG_IN	NEG_IN	NEG_IN	NEG_OUT		Ames mode	s. which generate	ed consistent ne d	ative predictions.		
*Frameshift Ames Mutagens	3	N/A	NEG_IN	NEG_IN	NEG_IN	NEG_OUT		The	other four model	s should not be co	unsidered		
*Potent Ames Mutagens, Re 10 Times Controls	eversions ≥	N/A	POS_IN	POS_OUT	POS_IN	POS_IN	The other four models should not be considered.						
DTU-developed models										ACH regulto from	the four VECA models		
* The four models (Direct Ac Potent Ames Mutagens) sho indication of mechanism or p	cting mutager ould not be us potency for ca	is (without S ed to determ ases where th	 Base-Pair Ame ine if substances ne main Ames mo 	es Mutagens, Fra are <u>Ames</u> mutag odel (Ames test in	ameshift Ames I gens, but can be n S. typhimuriur	Mutagens, e used for n (<i>in vitro</i>)) is		VVIUIIII LII	and the Consens	us model have bee	en integrated		
								VEGA					
								ISS	CAESAR	SarPy	KNN		
	VEGA		Mut. / Non	-mut. scores	Used model	s		NEG_Mod	NEG_Low	NEG_Low	POS_Good		
Mutagenicity consensus	NEG		0.23 / 0.25		4			Four individual models in mutagenicity consensus model version 1.0.2 contained in VEGA version 1.1.4 with calculation core version 1.2.4					
Mutagenicity (Ames) conser 1.2.4	nsus model ve	ersion 1.0.2 d	contained in VEG	A version 1.1.4 w	ith calculation o	core version		Prediction: POS = N Non-mutagenic, Exp	/utagenic, NEG = Non-mutag o = experimental value, Good	enic, SUSP.POS = Suspected = Good reliability, Mod = Mode	l mutagenic, POSS.NEG = Possible lerate reliability, Low = Low reliability.		
Prediction: POS = Mutageni	c, NEG = <u>No</u>	<u>n-mutagenic</u> .											
							(DNA alerts for AME	S by OASIS, alerts in:				
								- parent only No alert found					
								In vitro mutagenicity	(Ames test) alerts by ISS, ale	erts in:			
								- parent only		No alert found			
Structural a	lerts ide	ntified b	y two end	point-spec	itic profile	rs present	In	OECD QSAR Toolb	ox v.4.2 profilers				
	the OECD QSAR Toolbox							Profiler predictions are supporting information to be used together with the relevant QSAR predictions					



- Stepwise approach *
- Danish (Q)SAR Database can be searched for molecules, based on available experimental data, (Q)SAR predictions, structural alerts, etc., for the endpoint of interest

Our case:

- Search for experimentally **positive and negative** molecules for Ames;
- Target molecule has no Alerts for DNA binding or in vitro gene mutation in bacteria, similar molecules can be selected with the same "no alerts" profile.
- For each query, a list of molecules is retrieved
- The lists can be merged, using logical operator such as AND or OR

* Example of stepwise approach provided in appendix of this presentation

Human health

AND Intersect results

OR

NOT

Unite results

Complement results MAX/MIN More combinations 1:



New search	Searches	< >	Re	esults	Substances
	1.	Exact match:	>	1	
ID	2.	POS Bacterial Reverse Mutation Test (Ames test	>	2361	
Structure and name	3.	NEG Bacterial Reverse Mutation Test (Ames test	>	3971	
	4.	NO alert in P: DNA binding by OASIS, OECD QSAR Toolbox v.4	>	476842	
	5.	NO alert in P: DNA binding by OECD, OECD QSAR Toolbox v.4	>	275408	
	6.	NO alert in P: DNA alerts for AMES by OASIS, OECD QSAR Too	>	594272	
PhysChem	7.	NO alert in P: In vitro mutagenicity (Ames test) alerts by	>	456044	
Environment	8.	1. OR 2. OR 3.	>	6333	
	9.	4. AND 5. AND 6. AND 7. AND 8.	>	2091	
ADME					

- The target molecule (2,3,4-Trichlorobiphenyl)
- 2: All molecules from the database, **experimentally positive** for Ames test
- 3: All molecules from the database, **experimentally negative** for Ames test
- 4 to 7: All molecules with no alerts for each of the relevant profilers (DNA binding-related, Ames testrelated)
- 8: Target + experimentally positive + experimentally negative (1, 2 and 3 combined with OR)
- 9: Subset of 8, including molecules with no alerts for the four relevant profilers (8, 4, 5, 6 and 7, combined with AND)

Query for 2 to 7 is performed in the "Human health" section)



7. AND 9. AND 10. AND 11. AND 14.: Page 1 Previous Next 1 2 3 210 Structures 1-10 of 2091 Click to calculate similarity with the target molecule Bacterial Reve... Bacterial Reve.. Structure Id Similarity Experimental Battery and automatically sorting by highest similarity Use "+" to add columns to the overview xxxxx-xx-x 1.0 NEG_IN Columns can be added to show: Experimental data for the target endpoint Predictions for the target endpoint XXXX-XX-X Other information useful for evaluating similar molecules NEG IN 0.791 NEG (e.g. structural alerts from OECD QSAR Toolbox profilers) XXXXX-XX-X 0.791 NEG NEG_IN XX-XX-X 0.791 NEG NEG IN The generated overview **can be used** to prepare a statement for supporting the reliability of the (Q)SAR prediction for the XXXXX-XX-X 0.784 NEG NEG IN target molecule.

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Danish (Q)SAR Database: take home message



- Full documentation of the models is available, in QMRF format
- Supporting information (AD compliance, similar molecules) is either provided or can be retrieved, allowing expert evaluation
- Identification of similar molecules is not automated; however, more information can be taken into account compared to other tools (e.g. alerts profile)
- The generated reports do not include full applicability domain compliance evaluation; however, the QMRF includes the definition of the applicability domain.



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VEGA outcome reported according to ECHA Practical guide "How to use and report (Q)SARs" Version 3.1 – July 2016

Administrative data 🛛 🕲 None 🕲 N	lone
Endpoint @ ^ @ ^ in vitro gene mutation study in bacteria	
Type of information (Q)SAR	
Adequacy of study None Weight	t of evidence OR supporting study
Robust study summary	
Used for classification	
Used for SDS	
Study period None	According to ECHA Practical guide "it should normally be a maximum of 2"
Reliability None	IUCLID includes several possibilities for explaining the assigned reliability.
None	Appropriate rationale should be chosen considering both VEGA AD and reliability evaluation and expert assessment

Justification for type of information 1. SOFTWARE

- 2. MODEL (incl. version number)
- 3. SMILES OR OTHER IDENTIFIERS USED AS INPUT FOR THE MODEL

4. SCIENTIFIC VALIDITY OF THE (Q)SAR MODEL

[[Explain how the model fulfils the OECD principles for (Q)SAR model validation. Consider attaching the QMRF and/or QPRF or providing a link]

- Defined endpoint:
- Unambiguous algorithm:
- Defined domain of applicability:
- Appropriate measures of goodness-of-fit and robustness and predictivity:
 Mechanistic interpretation:

5. APPLICABILITY DOMAIN

[Explain how the substance falls within the applicability domain of the model]

- Descriptor domain:
- Structural domain:
- Mechanistic domain:
- Similarity with analogues in the training set:
- Other considerations (as appropriate):

6. ADEQUACY OF THE RESULT

[Explain how the prediction fits the purpose of classification and labelling and/or risk assessment]

VEGA v1.2.3

Mutagenicity ISS Model (version 1.0.3)

c1ccc(cc1)c2ccc(c(c2Cl)Cl)Cl

(QMRF) can be attached (next slide) and referenced here

VEGA report can be attached and used as reference. However, if expert assessment is performed, it can be described here.

procedure used to identify similar molecules in Danish QSAR database con be explained here

Expert assessment is needed





Attached	d justification	🕂 New item 🔹	Import file 💙			LIFE17 GIE/IT/000461
#	Attached justi	ication	Reason / purpose		Actions	QPRF can also be attached, if
1	QMRF_MUTA_I	SS.pdf	(Q)SAR model report	ting (QMRF)		prepared by the user
Data sou	rce					
Refere NE Data ad data p	nce GA v1.2.3 2023 ccess ublished					Basic information about the software and model are sufficient
Material	s and methods					
Test g	uideline -	New item 🛛 🖄 Import	file 💙			
#	Qualifier	Guideline	Version / remarks	Deviations	Actions	Otherwise, the test guidelines
1	according to g	uideline other: REACH Guidance on (R.6	H QSARs None	None		used to generate the data for the training set

Principles of method if other than guideline

The Benigni / Bossa rulebase for mutagenicity and carcinogenicity - a module of Toxtree, (2008) by R. Benigni, C. Bossa, N. Jeliazkova, T. Netzeva, and A. Worth. European Commission report EUR 23241 EN R. Benigni, C. Bossa, T. Netzeva, A. Rodomonte, and I. Tsakovska (2007) Mechanistic QSAR of aromatic amines: new models for discriminating between mutagens and nonmutagens, and validation of models for carcinogens. Environ mol mutag 48:754-771 Benfenati E, Manganaro A, Gini G. VEGA-QSAR: Al inside a platform for predictive toxicology Proceedings of the workshop "Popularize Artificial Intelligence 2013", December 5th 2013, Turin, Italy Published on CEUR Workshop Proceedings Vol-1107 Information from QMRF section 2.7 - Reference(s) to main scientific papers and/or software package

Test material	
Test material information 2,3,4-Trichlorobiphenyl 2,3,4-trichlorobiphenyl 1,2,3-trichloro-4-phenylbenzene 55702-46-0 Additional test material information None	Test ma
Specific details on test material used for the study SMILES: c1ccc(cc1)c2ccc(c(c2Cl)Cl)Cl Specific details on test material used for the study (confidential)	lf multipl substance,



Test material must reflect the evaluated structure

If multiple constituents are assessed for one substance, the Practical Guide suggest to prepare separate entries

Results and discussion 0 ^

Test res	ults 🕂 N	lew item 🛛 💧 Import fi	le 💙							
#	Key result	Species / strain	Metabolic act	Genotoxicity	Cytotoxicity /	Vehicle contr	Untreated ne	True negative	Positive contr	Actions
1		None	None	None	None	None	None	None	None	
Addition None Remarks no muta Any othe	al information on r s on result sgenic potential (bas rr information on r	esults ed on QSAR/QSPR predi results incl. tables	ction)		In the resu incluc	Its section, t les (Q)SAR-	he remarks p specific items	vicklist S		
None										
verall rer	marks, attachme	nts								
Overall r	emarks									

None

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- Katarzyna Bucior and the QSAR team at knoell
- The speakers of today and of 31 May





LIFE CONCERT REACH – Web-seminars on practical examples on using (Q)SAR for REACH

Thanks for your attention

worldwide registration

knoell



Appendix A

Importance of using multiple models for addressing the same endpoint







Statistical and KB models: second integration layer





- Focus on **specific substructures** (alerts);
- Might not be able to evaluate potential influence of the structure on the toxicity of the alert;
- Some models encode "exclusion rules", still based on expert knowledge:
 - Structural modifications that decrease or remove the toxicity;
 - Constraints on physico-chemical parameters.

E.g. Benigni-Bossa rulebase (ISS and CAESAR)



Ar—

Ar = Any aromatic/heteroaromatic ring

- Chemicals with ortho-disubstitution, or with an ortho carboxylic acid substituent are excluded.
- Chemicals with a sulfonic acid group (-SO3H) on the same ring of the nitro group are excluded.

Statistical and KB models: second integration layer

SA2

SA4 0 11 5=0

сі—⟨∖







E.g. SarPy/IRFMN model: structural alert for non-mutagenicity



- Might be able to **consider the whole structure** for the toxicity prediction:
 - User evaluation **limited by "Black box"** effect;
- Might be able to identify structural features that decrease the toxicity of structural alerts:
 - Critical expert evaluation is needed.





Appendix B

Danish QSAR Database: similar molecules identification





7. To account for mechanistic similarity, molecules with the same alerts profile can be extracted

In this case, according to the report generated for the target, no alerts for any of the profilers













