

CASE STUDY: using and integrating two QSAR models for assessing bioconcentration factor

31/05/2023, LIFE CONCERT REACH Web-Seminars - (Q)SAR Models under REACH: Practical Examples

Claudia Ileana Cappelli knoell Germany GmbH qsar@knoell.com



Case study definition and access to the Gateway 01



TABLE OF CONTENTS



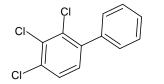
03 OCHEM/Gramatica & Papa (2005) predictions and assessment



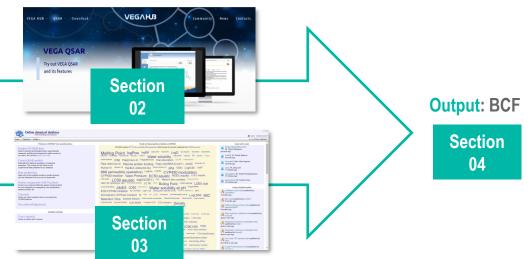
Case study definition



Aim: prediction of bioconcentration factor (BCF) in fish of 2,3,4-Trichlorobiphenyl using two QSAR models available in the LIFE CONCERT REACH Gateway, VEGA/CAESAR and OCHEM/Gramatica & Papa (2005)



Input: Clc1ccc(c2cccc2)c(Cl)c1Cl



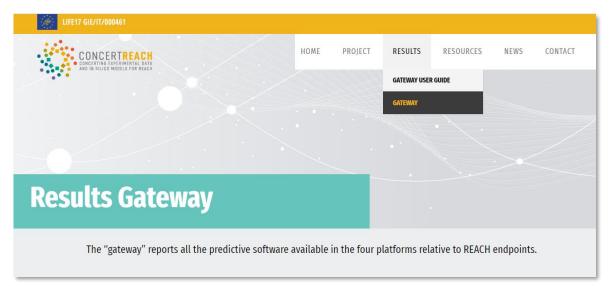
• Out of scope: documentation of results in IUCLID

Case studies: mutagenicity (R. Gonella-Diaza, 17/05); logKow (A. Szymoszek, 31/05)

The Gateway in LIFE CONCERT REACH



- Web-based system to guide users to **integrate data** from four platforms, i.e., **VEGAHUB, AMBIT, OCHEM and Danish (Q)SAR Database**, for evaluation of substance(s) under REACH
- Four main categories of endpoints: physicochemical, toxicological, ecotoxicological and endocrine disruption properties
- Freely accessible at https://www.life-concertreach.eu/results/results-gateway/



Model(s) selection

the endpoints list



9. ECOTOXICOLOGICAL INFORMATION

+ 9.1.1. Short-term toxicity testing on invertebrates (preferred species Daphnia)
+ 9.1.2. Growth inhibition study aquatic plants (algae preferred)
+ 9.1.3. Short-term toxicity testing on fish: the registrant may consider long-term toxicity testing instead of short-term.
+ 9.1.4. Activated sludge respiration inhibition testing
+ 9.1.5. Long-term toxicity testing on invertebrates (preferred species Daphnia)
+ 9.1.6. Long-term toxicity testing on fish
+ 9.2.1.1. Ready biodegradability
+ 9.2.2.1. Hydrolysis as a function of pH.
+ 9.3.1. Adsorption/desorption screening
+ 9.3.2. Bioaccumulation in aquatic species, preferably fish
+ 9.3.4. Further information on the environmental fate and behaviour of the substance and/or degradation products
Filter the models by

All VEGA AND To:		VEGA AND ToxRead DANISH QSAR DATABASE AMBIT OCHEM							
	End Point	Model	Туре	Dataset size	Training set size	Test set size	Cross- validation procedure	Platform	Remarks
	BCF	BCF (L/kg wet-wt) (EPI)	continuous					DanishQSARDatabase	—
	• For each endpoint, the gateway reports a list of								_
	• For e	each endpo	oint, th	ne ga	itewa	y re	eports	a list of	=
Who Biotra		each endpo els with ba of predic	sic inf	orma	ation:	mo	odel n	ame,	

• The user can look at the **whole list or** select models belonging to a **specific platform**

(QSAR Model Reporting Format, QMRF; papers)

• Currently **12 BCF-related models** from VEGA, Danish (Q)SAR Database and OCHEM



Model(s) selection and prediction(s)

9.3.2. Bioaccumulation in aquatic species, preferably fish

All	VEGA AND ToxRead	DANISH QSAR	DATABASE	AMBIT	OCHI	EM			
	End Point	Model	Туре	Dataset size	Training set size	Test set size	Cross- validation procedure	Platform	Remark
	onmental fate parameters 4. oncentration 2.4.a.BCF fish	BCF model (CAESAR)	continuous	473	378	95		VEGA	=
	ironmental fate parameters Bioconcentration . BCF fish	BCF model (Meylan)	continuous	662	516	146		VEGA	=
Bio	concentration: OECD 305	BCF model (Arnot-Gobas)	continuous	692	692	0		VEGA	=
ENV F	ATE 5.3.1. Bioaccumulation: aquatic	BCF model (KNN-Read- Across)	continuous	860	860	0		VEGA	=

All	VEGA AND ToxRead	DAI	VISH QSAR DA	TABASE	AMBIT	OCHEM			
	End Point	Model	Туре	Dataset size	Training set size	Test set size	Cross-validation procedure	Platform	Remark
	ioaccumulation in aquatic species, preferably fish	Linear	Continuous		179	59		OCHEM	



Section 02

6

After **model(s) selection**, click on the link in the **"Model" column(s)**: you will be redirected to the access page of the platform(s)

Online chemical database		112.64
Vern backering developed		S ling in construction
(c) Subject - State Control (C) State Control	Developmentalise-restlet development	Prevent Pr
Pediakat Jag	Middla Ce Cannol Biologi (2010). ¹⁶⁵ /RFL (addyos, Jigor 100. ¹⁶⁵ /RFL (2010). ¹⁶⁵ /RFL (2010).	US stand span ng:



Case study definition and access to the Gateway 01



TABLE OF CONTENTS

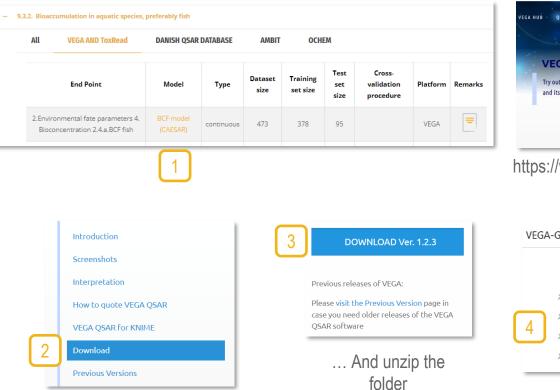


03 OCHEM/Gramatica & Papa (2005) predictions and assessment



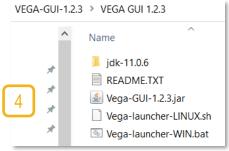
Access VEGA



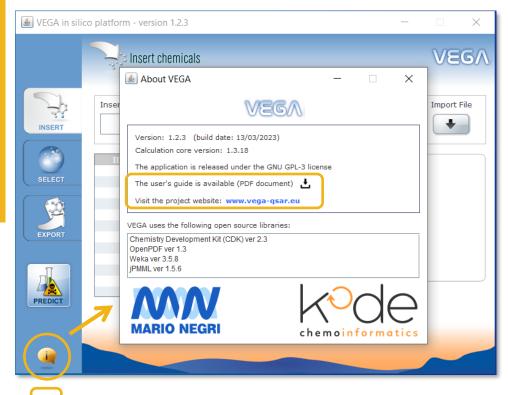




https://www.vegahub.eu/portfolio-item/vega-qsar/



Access VEGA



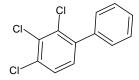


VEGA: Virtual models for Evaluating the properties of chemicals within a Global Architecture

- Developed mainly by Mario Negri Institute (Milan) and Kode s.r.l. (Pisa)
- Free platform developed based on contributions from EU projects
- Includes more than 100 statistical and knowledge-based (Q)SAR models for the prediction of (eco)toxicity, environmental fate and physicochemical properties of chemicals.

Insert the target chemical





Name: 2,3,4-Trichlorobiphenyl
SMILES: Clc1ccc(c2cccc2)c(Cl)c1Cl

Second Se	- 🗆 🗙	Second Se	- 🗆 X
Insert chemicals	VEGA	Insert chemicals	VEGA
Insert SMILES: Clc1ccc(c2cccc2)c(Cl)c1Cl	Import File	Insert SMILES:	Import File
ID SMILES 7		ID SMILES SELECT 0 clccc(ccl)c2ccc(c(c2Cl)Cl)Cl) State 8 CL	
EXPORT		EXPORT CI	
PREDICT		PREDICT	9

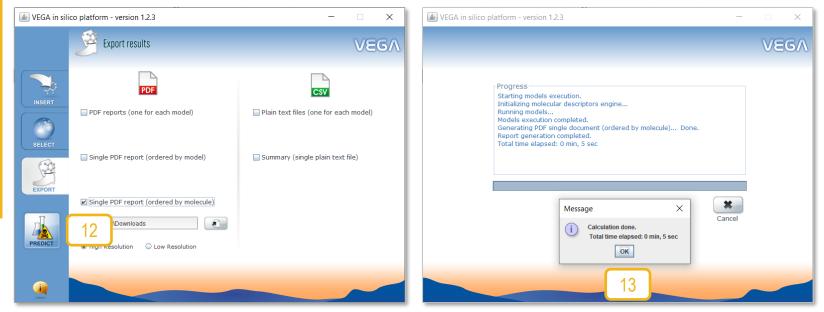


Select the model(s) of interest and preferred output(s) settings

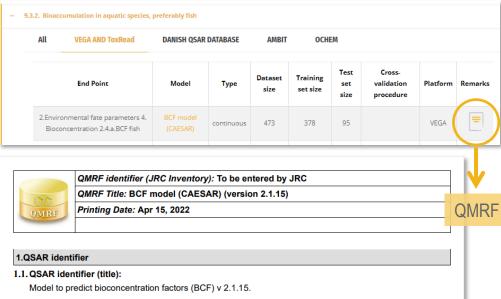
🛓 VEGA in silico platform - version 1.2.3 – 🗆 🗙	VEGA in silico platform - version 1.2.3 — 🗌 🗙
Select models VEGA	Export results VEGA
Filter models: EccToxicity Image: Select all models Image: Select all models Image: Select all models Image: Select all	Image: Single PDF report (ordered by model) Image: Summary (single plain text file)
Image: Second	Single PDF report (ordered by molecule) Single PDF report (ordered by molecule) Wigh Resolution High Resolution Low Resolution In Tick the layout(s) and choose the destination folder(s) for saving the report(s)
Model: BCF model (CAESAR) Version: 2.1.15 Available documentation: Training set (plain text with SMILES)	Full PDF reports: Simplified text reports:
Model information in QMRF, available at https://www.vegahub.eu/portfolio-item/vega-qsar-models-qrmf/	 prediction(s) results applicability domain experimental data of the target (if any) most similar substances other supporting info (if any)



Predict the target chemical



Assessment: scientific validity of the QSAR model (CAESAR)



1.2. Other related models:

Two models, model A and model B, have been used to build hybrid model, model C.

In the proposed approach, the outputs of the individual models (model A and B) were used as inputs of the hybrid model.

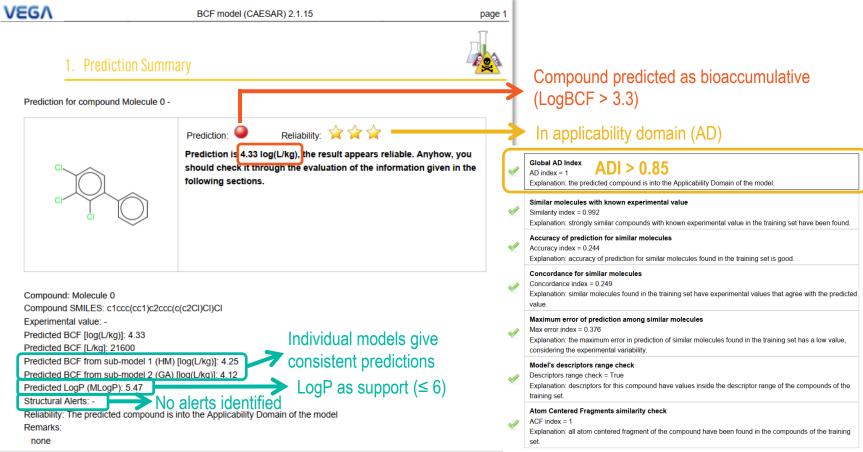
Model A was developed by Radial Basis Function Neural Networks (RBFNN) using a heuristic method to select the optimal descriptors; model B was developed by Radial Basis Function Neural Networks (RBFNN) using genetic algorithm for the descriptors selection.

https://www.vegahub.eu/portfolio-item/vega-qsar-models-qrmf/



- Quantitative prediction of BCF in fish (log of L/kg)
- **Hybrid model**, implementing the detection of **structural alerts** for reasoning (outliers, chemical classes with particular BCF behavior, polar groups)
- Applicability Domain Index (ADI) [0 not reliable; 1 fully reliable]
- Goodness-of-fit
 - n training set = 378; R2 = 0.81; RMSE = 0.58
- Robustness leave many out (20%) cross validation
 - R2cv= 0.79; SDEP = 0.66
- Predictivity external validation
 - n test set = 95; R2 = 0.78; RMSE = 0.62







VEGA

BCF model (CAESAR) 2.1.15

3.1 Applicability Domain: Similar Compounds, with Predicted and Experimental Values

Compound #1

CAS: 15862-07-4 Dataset id:185 (Training Set) SMILES: c1ccc(cc1)c2cc(c(cc2Cl)Cl)Cl Similarity: 0.998 Experimental value : 4.22 Predicted value : 4.332 Alerts (not found also in the target): Moiety (SMILES: Clc1cc(c2cccc2)c(Cl)cc1) (5 Compound #2 CAS: 16606-02-3 Dataset id:184 (Training Set) SMILES: c1cc(ccc1c2cc(ccc2Cl)Cl)Cl Similarity: 0.986 Experimental value : 3.95 Predicted value : 4.326 Alerts (not found also in the target): Moiety (SMILES: Clc1ccccc1c1ccc(Cl)cc1) (SI Moiety (SMILES: Clc1cc(c2ccccc2)c(Cl)cc1) (SR 09)

Compound #3

CAS: 7012-37-5 Dataset id:183 (Training Set) SMILES: c1cc(ccc1c2ccc(cc2Cl)Cl)Cl Similarity: 0.979 Experimental value : 4.33 Predicted value : 4.322

Alerts (not found also in the target): Moiety (SMILES: Clc1ccccc1c1ccc(Cl)cc1) (SF

Compound #4

CAS: 33284-53-6 Dataset id:194 (Training Set) SMILES: c1ccc(cc1)c2cc(c(c(c2Cl)Cl)Cl)Cl Similarity: 0.963 Experimental value : 4.39 Predicted value : 4,636

A similarity index (SI) of the target with respect to similar molecules with known experimental value is calculated. It takes into account how similar are the first two most similar compounds. Values near 1 mean that the predicted compound is well represented in the dataset used to build the model. otherwise the prediction could be an extrapolation.

The 2 mostly similar compounds from the training set:

exhibit high similarity to the target

Similar molecules with known experimental value Similarity index = 0.992

Explanation: strongly similar compounds with known experimental value in the training set have been found.

SI > 0.9

have experimental values that agree with target prediction... .

> Concordance for similar molecules < 0.5

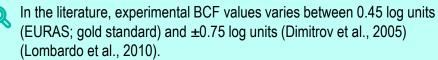
Concordance index = 0.249 Explanation: similar molecules found in the training set have experimental values that agree with the predicted value

0.5

...and their prediction accuracy is good.

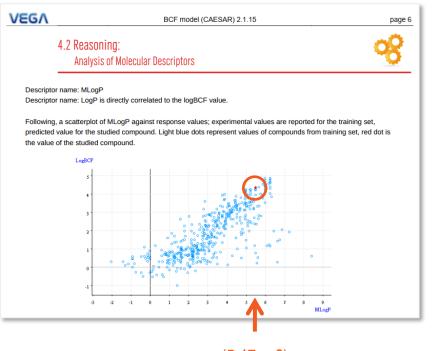
Accuracy of prediction for similar molecules Accuracy index = 0.244

Explanation: accuracy of prediction for similar molecules found in the training set is good.

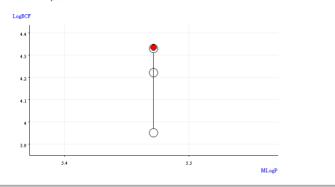


Alerts (not found also in the target): Moiety (SMILES: Clc1cc(c2cccc2)c(Cl)cc1) (Sn very





Following, a scatterplot of MLogP against response values only for 3 most similar compounds in the training set. Red dot is the value of the studied compound, black outlined circles represents experimental values of compounds from training set, black dots represents predicted value of the same compound; the size of the circle is proportional to the similarity to the studied compound.

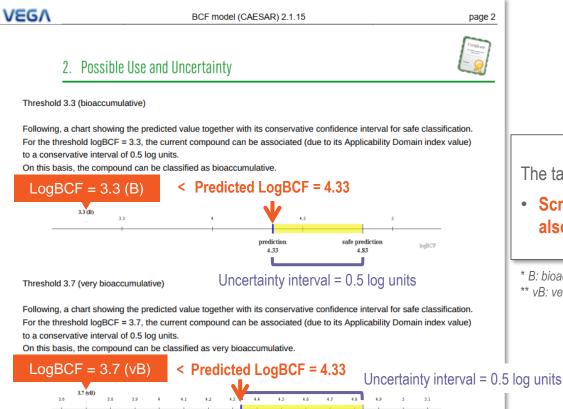


The most similar chemicals to the target also share the same logP

 $(5.47 \le 6)$ LogP as support



Assessment: relevance of the QSAR model to regulatory purpose



safe prediction

4.83

logBCF

prediction

4.33

The target **fulfils the criteria** for **B*** and **vB****

 Screening criterion (LogK_{ow} = 5.47 > 4.5) also fulfilled.

* B: bioaccumulative

** vB: very bioaccumulative

Assessment: adequacy of the QSAR result



Highly reliable → based on model validity, ADI and information to derive it

- Target well represented in the training set (i.e., strongly similar compounds, target descriptors within the range of the training set, all structural fragments found in the training set)
- Training set analogues experimental values: consistent with target prediction
- Training set analogues prediction accuracy: good
- The maximum error in prediction of training set analogues has a low value, considering the experimental variability.

Relevant → based on the purpose

• The target fulfils B and vB criteria.

Adequate

Klimisch 2 - results derived from a valid QSAR model and falling into its applicability domain, with adequate and reliable documentation/justification.



Case study definition and access to the Gateway 01

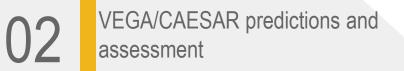


TABLE OF CONTENTS

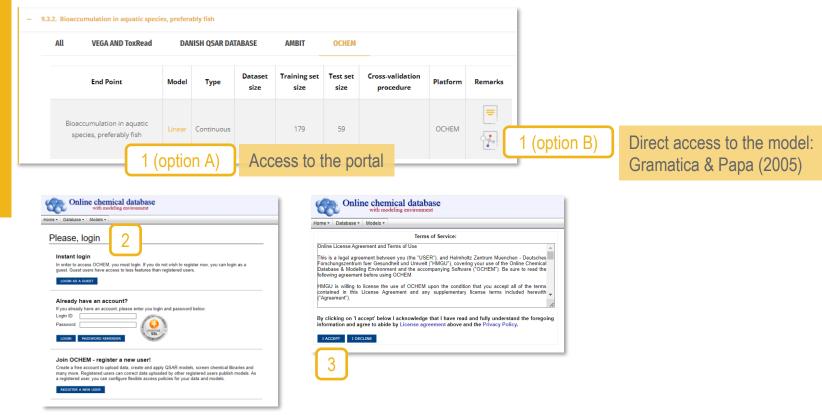


03 OCHEM/Gramatica & Papa (2005) predictions and assessment



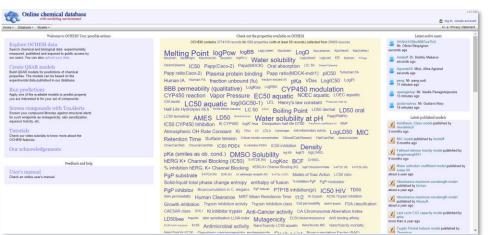


Access OCHEM



https://ochem.eu/login/show.do?render-mode=full

Access OCHEM





OCHEM: Online CHEmical database and Modeling environment

- Free web-based platform that provides tools for automation of steps to create a predictive QSAR/QSPR model
- It consists of a database of experimental measurements (> 1M chemical structures and 3M data points) integrated with a modeling framework, which supports all the steps to create a predictive model
- > 150 models are published on the web site, which can be used to predict new molecules.



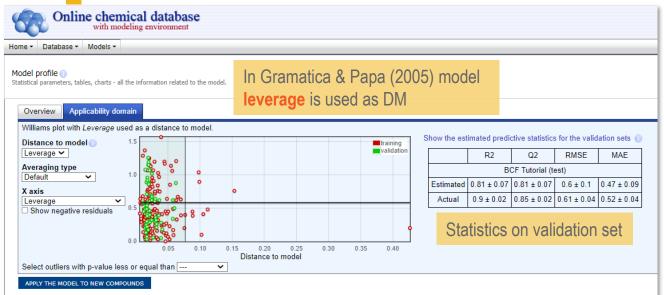


Models •	A+ a- Privacy staten
bles, charts - all the information related to the model.	
Equation and descrip	otors.
F tutorial , published in An Update of the BCF QSAR Model Based on Theoretical Molecular Descriptors Reference Descriptors refer to	[Dragon6 (blocks: 1-29)] Correl. limit: 0.0 Unique values: 0, Variance threshold: 0.0, Maximum value: 2147483647, ,
Predicted property, model type, molecular size, hydro	[IDDM, HIC, nHAcc, GATS1e, MATS1p] No validation
# K2 Y2 KW5L WAL	5 pre-filtered descriptors
BCF Tutorial (training) 179 records 0.8 ± 0.02 0.58 ± 0.03 0.48 ± 0.02 0.68 ± 0.02 0.68 ± 0.02 0.68 ± 0.02 0.68 ± 0.02 0.61 ± 0.04 0.52 ± 0.04 access to training and test sets, bonding capacity,	Y = -1.14 + 2.43*IDDM - 0.88*HIC - 0.481*nHAcc - 1.17*GATS1e - 1.95*MATS1p
statistics electronic properties (atomic polarizability electronegativity), molecular complexity	v and Calculated in 30 seconds Size: 11 Kb
Observed-vs-predicted chart	
10 20 30 40 50 Measured value	
ed records]	
Measured value	

Model page layout – "Applicability domain" tab



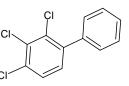
- The AD assessment in OCHEM is based on **distance to model (DM)**, i.e., any numerical measure of the prediction uncertainty for a given compound by the model
- DM assesses how "far" is the compound from the model: compounds with larger values of DM are expected to have lower prediction accuracy than compounds with smaller DM
- DMs estimate the **reliability** of predictions.



- High leverage values indicate that one starts extrapolating outside the training set range and it is no guaranteed that the model is valid and applicable
- Compounds with leverage exceeding a warning threshold h* are often outside the AD of the model.

Tetko, I.V. et al, J. Chem. Inf. Mod. 2008, 48(9), 1733-1746 Sushko I., Applicability Domain of QSAR models. Doctoral work. 2011. http://mediatum.ub.tum.de/node?id=1004002

Predict the target chemical



Name: 2,3,4-Trichlorobiphenyl SMILES: Clc1ccc(c2cccc2)c(Cl)c1Cl

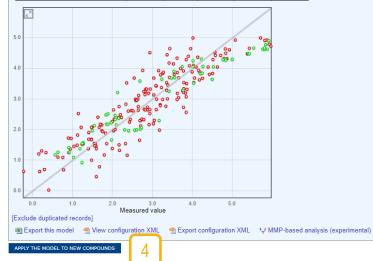
Next>>

Overview Applicability domain

Model name: BCF tutorial , published in An Update of the BCF QSAR Model Based on Theoretical Molecular Descriptors Public ID is 8

Predicted property: BCF Training method: MLRA

Data Set	#	R2	q2	RMSE	MAE
• Training set: BCF Tutorial (training)	179 records	0.8 ± 0.02	0.8 ± 0.02	0.58 ± 0.03	0.48 ± 0.02
• Test set: BCF Tutorial (test) [x]	59 records	0.9 ± 0.02	0.85 ± 0.02	0.61 ± 0.04	0.52 ± 0.04



Model profile X Apply a model X Model Applier Provide the compound(s) to predict Please provide compounds for which you want to predict the target property Several options are available: $_{\bigcirc}$ Upload compounds from a file Choose File No file chosen SDF. MOL2. SMILES or an Excel sheet Draw Molecule click on depiction to the right to draw [molecule profile] Name/CASRN/SMILES: Clc1ccc(c2cccc2)c(Cl)c1Cl load structure e.g., "CC=CCC" or "Aspirine" O Choose a previously prepared set: [...] 5 O Select molecules by a tag: [...] Additional options Disable prediction cache 8





Assessment: scientific validity of the QSAR model (Gramatica & Papa, 2005)

All	VEGA AND ToxRead	DAI	VISH QSAR DA	TABASE	AMBIT	OCHEM			
	End Point	Model	Туре	Dataset size	Training set size	Test set size	Cross-validation procedure	Platform	Remark
	ccumulation in aquatic ecies, preferably fish	Linear	Continuous		179	59		OCHEM	

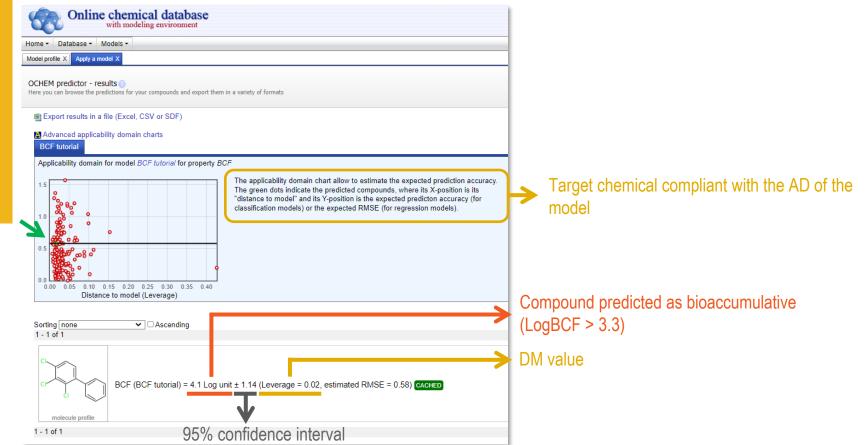
- Quantitative prediction of BCF in fish (log of L/kg)
- Multiple linear regression (Ordinary Least Square regression method)
- The AD was checked by leverage approach (high leverage compounds: hat value > 0.1). Outliers for the response were identified by analysis of the standardized residuals (values > 2.5 standard deviation units)
- Goodness-of-fit
 - n training set = 179; R2 = 80.7%; RMSE = 0.56
- Robustness leave many out cross validation
 - Q2_{LMO(25%)}=79%; Q2_{LMO(50%)}=78.2%
- Predictivity external validation
 - n test set = 59; Q2ext = 86.4%; R2ext = 90.5%; RMSE = 0.57

	QSAR & Combinatorial Science
An Update of the BCF QSAR Mod Molecular Descriptors	lel Based on Theoretical
Paola Gramatica* and Ester Papa	
QSAR and Environmental Chemistry Research Unit, Department of St Dunant 3, 21100 Varese (Italy); http://www.qsar.it; E-mail: paola.gramati	
Keywords: BCF; Molecular descriptors; Genetic Algorithm; Validation; Molecular size	Hydrogen bonding;
Received: February 22, 2005; Accepted: April 18, 2005	
DOI: 10.1002/gsar.200530123	

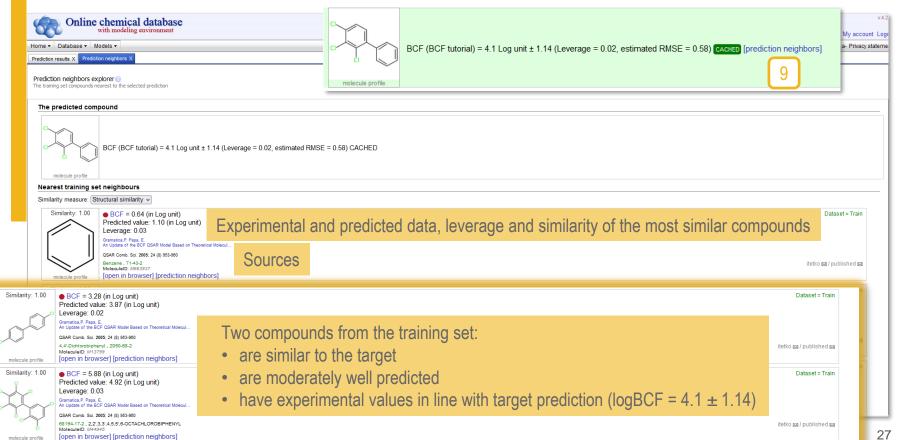
QMRF identifier (JRC Inventory):Q13-24a-001 QMRF Title:QSAR for bioconcentration fac Printing Date:Dec 11, 2019	
QMR	QMRF available
1.QSAR identifier	at QsarDB
1.1.QSAR identifier (title): QSAR for bioconcentration factor in fish	
1.2.Other related models: Gramatica P & Papa E (2003). QSAR Modeling of Bioco	
theoretical molecular descriptors. QSAR & Combinatoria 374-385.	I Science, 22,

https://qsardb.org/repository/handle/10967/110









Assessment: adequacy of the QSAR result



Highly reliable → based on model validity, AD compliance and information to derive it

- Strongly similar compounds to the target
- Training set analogues experimental values: in line with target prediction
- Training set analogues prediction accuracy: moderately good.

Relevant → based on the purpose

• The target fulfils B and vB criteria.

Adequate Klimisch 2 - results derived from a valid QSAR model and falling into its applicability domain, with adequate and reliable documentation/justification.





Case study definition and access to the Gateway 01

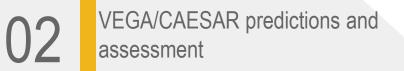


TABLE OF CONTENTS



03 OCHEM/Gramatica & Papa (2005) predictions and assessment



Conclusion



- CAESAR and Gramatica & Papa (2005) models are documented in **QMRF** and provide supporting information (e.g., **AD compliance**, **similar molecules**), allowing **expert evaluation**
- In the present case study, CAESAR and Gramatica & Papa (2005) models provide consistent, reliable and adequate predictions and the target molecule 2,3,4-Trichlorobiphenyl can be assessed as B and vB (logBCF = 4.33 as a worst case).

The result may be used in the context of REACH:

- To cover the endpoint fully
- Together with other information (e.g., experimental data) as supporting

data or part of WoE

Acknowledgements:

- knoell Academy team
- Igor V. Tetko (BigChem GmbH) and all partners of the LIFE CONCERT REACH project
- Antje Gerloff-Elias and the QSAR team at knoell
- The speakers of today and of 17/05





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

Thanks for your attention

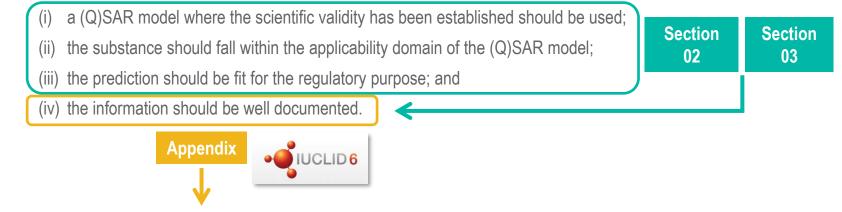
worldwide registration

knoell

Adequacy and reliability of documentation: IUCLID



• Results of (Q)SARs may be used instead of testing when the conditions set in REACH Annex XI (1.3) are met:

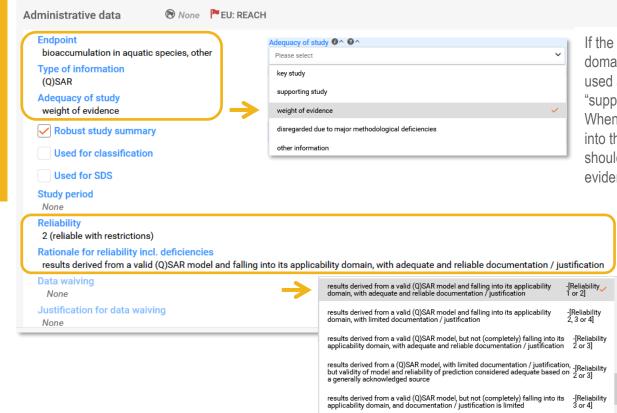


Compile one Robust Study Summary (RSS) for each QSAR result (at least two RSS of the predicted endpoint of interest for the target molecule, which are related to the two most reliable predictions derived from different models).

->> ECHA Practical guide "How to use and report (Q)SARs" Version 3.1 – July 2016

Adequacy and reliability of documentation: administrative data





results derived from a (0)SAR model with limited documentation /

-IReliability

If the molecule is compliant with the applicability domain of the model, the QSAR result can be used as "key study", "weight of evidence" or "supporting study", depending on your case. When the molecule does not (completely) fit into the applicability domain, the QSAR result should be used only within a "weight of evidence" approach or as "supporting study".

> Reliability and its justification is case by case decision. As a general rule, all (Q)SAR entries should be of RL2, because RL1 is reserved for high quality experimental studies and RL3 or RL4 can be considered only in exceptional cases.

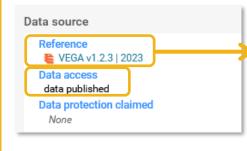


Adequacy and reliability of documentation: administrative data

Justification for type of information 1. SOFTWARE VEGA v1.2.3	
2. MODEL (incl. version number) CAESAR v2.1.15	
3. SMILES OR OTHER IDENTIFIERS USED AS INPUT FOR THE MODEL Cle1ccc(c2ccccc2)c(Cl)c1Cl	
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: - 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] Attached justification + New item C Import file C Im	
# Attached justification	Reason / purpose
# 1 QMRF_BCF_CAESAR.pdf	(Q)SAR model reporting (QMRF)
# 2 report_BCF_CAESAR_trichlorobiphenyl.pdf	(Q)SAR: supporting information

Adequacy and reliability of documentation: data source





General information	
Reference Type None	
Title* VEGA v1.2.3	
Author None	
Year 2023	
Bibliographic source CAESAR BCF Model v2.1.15	



Adequacy and reliability of documentation: materials and methods

Materials and methods				
Test guideline + New item 🔞 Import file 🗸				
# Qualifier Guideline	Ver	sion / remarks	Deviations	Actions
1 according to guideline other: REACH Guidance R.6 on QSARs and grouping	of chemicals Nor	ле	None	
Principles of method if other than guideline - Software tool(s) used including version: Vega v1.2.3 - Model(s) used: CAESAR BCF Model version 2.1.15 Full reference and details of the used formulas can be found in: <!--</td--><td></td><td>nods'</td><td></td><td></td>		nods'		
no Test material Test material information [2] 2,3,4-trichlorobiphenyl_QSAR 2,3,4-trichlorobiphenyl 1,2,3-trichloro-4-phenylbenzene 55702-46-0	Edit 2,3,4-trichlorobiphe This IUCLID information associated data.	nyL_QSAR In is a re-usable data element. Note that any	é Go to source→ 🗙 rmodification will impact all 🛛 🗙	
	Name* 2,3,4-trichlorobiphenyl_QS/ Composition	AR		
	Composition +	New item 🔹 Import file 🗸		
	# Туре	Reference sub Concentration	Remarks Actions	
	1 Constituent	Phi P.3.4- trichlorobiphenyl 11,2,3-trichloro- None 4-phenylbenzene 155702-46-0	None	
	Composition / purity: othe not applicable for in silico Other characteristics Test material form None Details on test material None Confidential details on test - SMILES: clocc(cc1)c2ccc	study st material 🔺		



Adequacy and reliability of documentation: materials and methods

Test organisms	
Test organisms (species) other: Fish	
Details on test organisms None	
Study design	
Route of exposure aqueous	
Justification for method minimised test method used to support BCF estimates based on QSAR	
Test type other: calculation	
Water / sediment media type natural water: freshwater	

Details on estimation of bioconcentration BASIS FOR CALCULATION OF BCF

- Estimation software: Vega version 1.2.3, CAESAR v 2.1.15

- Result based on measured OR calculated log Pow of: <value>

Any other information on materials and methods incl. tables

Supporting data for "Justification of type of information", e.g. tables.



Adequacy and reliability of documentation: results and discussion

Re	esults and	discussion										
	Lipid cont	ent + New it	em 🝓 Import file 🗸									
	#	Lipid content			Time point			Remarks on result		Act	ions	
(Bioaccum	ulation factor	🕇 New item 💧 Imp	oort file 🗸								
	#	Key result	Conc. / dose	Temp.	рН	Туре	Value	Basis	Time of plateau	Calculation basis	Remarks on result	Actions
	₿1		None	None	None	other: Log BCF	4.33 dimensionless	whole body w.w.	None	None	None	
	∦ 2		None	None	None	BCF	21600 L/kg	whole body w.w.	None	None	None	

The performance of the mo	del on similar substances as given in the attached report is summarized in the table:							
Name and/or CAS No.	SMILES	Index of similarity to the test	Experimental result	Predicted result	Quality of prediction*			
		compound	(log BCF)	(log BCF)				
15862-07-4	c1ccc(cc1)c2cc(cc2Cl)Cl)Cl	0.998	4.22	4.33	Good			
16606-02-3	c1cc(ccc1c2cc(ccc2Cl)Cl)Cl	0.986	3.95	4.33	Good			
7012-37-5	c1cc(ccc1c2ccc(cc2Cl)Cl)Cl	0.979	4.33	4.32	Good			

*Absolute difference (experimental-predicted): < 0.5 good, 0.5-1.0 moderate, >1.0 poor The performance of the model on similar molecules is... (characterise: good, moderate, etc)

Overall remarks, attachments

Overall remarks

E.g. conclusion about adequacy of the result for the regulatory purpose under REACh regulation (EC) No 1907/2006.

This section includes information on the most similar substances to the target, as provided by the VEGA model (experimental and predicted data, similarity index). Quality of prediction is assigned by the user, as indicated. If the most similar substances are not provided automatically by the software, related information can be searched by the user, e.g., using the OECD QSAR Toolbox.

In this section the user has to conclude on applicability domain compliance of the target and validity of the prediction.