

Practical guide How to use and report (Q)SARs

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Version 1.0	First version	March 2010
Version 2.0	 Revision of the Practical guide addressing structure and content. The update includes the following: Correction of broken hyperlinks throughout the document. Changes in IUCLID screenshots with associated text improved for more clarity. Section 2 on 'How to get started with (Q)SARs' addressing approach to tools developed in research and development projects. 	December 2012
Version 3.0	 Revision of the Practical guide addressing structure and content. The update includes the following: Update of information in sections 2 and 3. Addition of a section (section 4) giving practical examples on how to assess the reliability of QSAR predictions with some QSAR programs. Addition of an appendix (Appendix 1) giving examples (a non-exhaustive list) of (Q)SAR programs available for each of the REACH required endpoints. 	March 2016
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Practical Guide - How to use and report (Q)SARs

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1. Introduction

REACH foresees in Annex XI that the standard testing regime can be adapted by the use of non-test methods, such as (quantitative) structure-activity relationships [(Q)SARs], if certain conditions are fulfilled.

This practical guide extends the chapter on (Q)SARs available in the practical guide on "How to use alternatives to animal testing to fulfil your information requirements for REACH registration".

It provides an overview of important aspects to consider when predicting properties of substances using (Q)SAR models as defined in the REACH Regulation – aspects which ECHA also takes into account to evaluate (Q)SAR results. This practical guide also gives useful examples for good prediction practices based on widely used and freely available (Q)SAR software programs.

Section 2 of this document gives general information about (Q)SARs and how to use them.

Section 3 explains the conditions that need to be fulfilled to use (Q)SAR predictions under REACH. Registrants are advised to explicitly include these considerations in their registration dossiers.

Section 4 gives practical examples based on freely available and commonly used (Q)SAR programs.

Appendix 1 gives examples (a non-exhaustive list) of (Q)SAR programs available for each of the REACH required endpoints.

Appendix 2 provides links to other guidance documents and tools that give further insights into the use of QSARs.

2. How to get started with (Q)SARs

2.1 Definitions

Structure-activity relationship (SAR) and quantitative structure-activity relationship (QSAR) models – collectively referred to as (Q)SARs – are theoretical models that can be used to quantitatively or qualitatively predict the physicochemical, biological (e.g. an (eco)toxicological endpoint) and environmental fate properties of compounds from the knowledge of their chemical structure.

A SAR is a qualitative relationship that relates a (sub)structure to the presence or absence of a property or activity of interest.

A QSAR is a mathematical model relating one or more quantitative parameters, which are derived from the chemical structure, to a quantitative measure of a property or activity.

In this document, the chemical for which an endpoint is being estimated by a (Q)SAR model is referred to as the target chemical. In other sources, this target chemical may be called a 'query compound' or 'input structure'.

2.2 Substance characterisation

The chemical structure needs to be well defined, following the <u>Guidance on identification and</u> <u>naming of substances under REACH</u>. All individual constituents of multi-constituent substances should be addressed. The composition of the well-defined substances also has to include known impurities (and additives, if any).

For substances of unknown or variable composition, complex reaction products or biological materials (UVCBs), expert judgement is needed to decide whether representative structures for the substance can be identified. Stable transformation products should also be identified. A suitable structural representation for the chemical (SMILES, mol file, etc.) is usually required.

2.3 Experimental results

In general, if reliable and adequate experimental (measured) results are available they should prevail over estimated values for the risk assessment and the classification and labelling of the substance.

Therefore, before using (Q)SAR models to predict a specific property of a substance, a critical first step is to assemble all of the available information on the substance. There are many information sources available for this purpose and which are further explained in the <u>Guidance</u> <u>on information requirements and chemical safety assessment - Chapter R.3: Information</u> <u>gathering</u>.

Among those sources, note that the OECD <u>QSAR Toolbox</u> includes one of the largest collections of publicly available data, and provides detailed information on how it should be used.

In addition, most of the (Q)SAR software programs will indicate if their training set¹ contains experimental results for the target chemical. In this case, the user should prioritise this existing experimental data over the predicted data, if there is sufficient indication that the experimental data is of good quality.

¹ Dataset used to construct the (Q)SAR model.

2.4 Conditions for using (Q)SAR results

Several (Q)SAR models have been integrated in software programs that are straightforward to use. However, experience and a thorough understanding of (Q)SARs is needed to verify their reliability and adequacy.

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

- (i) a (Q)SAR model where the scientific validity has been established should be used;
- (ii) the substance should fall within the applicability domain of the (Q)SAR model;
- (iii) the prediction should be fit for the regulatory purpose; and
- (iv) the information should be well documented.

An assessment of the first three points above is expected to be included in the registration dossier when substance properties are predicted using (Q)SAR models.

Section 3 of this practical guide provides detailed information on how to do this assessment.

2.5 Strategy for using (Q)SAR results

In general, it is recommended to use (Q)SAR results as part of a weight of evidence (WoE) approach or as supporting information. For instance, (Q)SAR predictions can support results from tests that have not been performed according to good laboratory practice (GLP) or according to accepted guidelines, if those predictions concur with the experimental results. A compilation of several predictions with unassignable quality cannot provide an adaptation on its own.

When using (Q)SARs, **it is recommended to run all (Q)SAR models** available to the registrant for the endpoint to be fulfilled, especially when models are independent from each other (e.g. the algorithms are based on different descriptors, structural alerts or training sets). Agreement among predictions generated from independent and scientifically-valid (Q)SAR models increases the confidence in relying on the predictions.

Predictions that fulfil only some conditions specified in REACH Annex XI (1.3) should be disregarded or the reason for providing these predictions should be explained if it is considered that there are some benefits to providing these predictions. If the remaining (valid and adequate) predictions show small quantitative differences, the most conservative result should be chosen for further consideration. If those remaining predictions show significant quantitative differences, the registrant must decide if these differences could affect the risk assessment (for demonstrating safe use) and/or classification and labelling.

If the (Q)SAR prediction outcome is a quantitative result, keep in mind that **the closer to a regulatory threshold the predicted result is, the more accurate the prediction needs to be**. For instance, if a (Q)SAR model predicts a LC_{50} (for fish at 96 hours) of 1.2 mg/L then this predicted value needs to be fully reliable to ensure that the actual LC_{50} of the substance is not below the CLP regulatory threshold of 1 mg/L. In contrast, if all (Q)SAR results (and even the worst case/over-conservative ones) do not go beyond the regulatory threshold of interest, this can support the waiving of the experimental study.

3. How to check a QSAR prediction

3.1 Is the (Q)SAR model valid?

As indicated in REACH Annex XI (1.3), the validity of the (Q)SAR model is the first condition to be fulfilled to use a (Q)SAR result. To check this, ECHA follows the OECD principles for validating (Q)SAR models². These are five principles dictating that a (Q)SAR model should be associated with:

- 1. A defined endpoint: the model must predict the same endpoint that would be measured to fulfil the requirements listed in REACH Annexes VII to X. For instance, predictions from a model generically predicting "mutagenicity" cannot be accepted as such. The model should predict the outcome of a specific test such as "positive", "negative" or "ambiguous" in a bacterial reverse mutation assay (i.e. Ames test required in REACH Annex VII, 8.4.1). Another example of an endpoint being too broad is a global prediction of a 'repeated dose toxicity lowest observed adverse effect level (LOAEL)' from a training set of LOAEL data based on a variety of mode of actions, target organs, species or test protocols. This principle links with the adequacy of the predictions described later in the document.
- 2. An unambiguous algorithm: the algorithm underlying the model must be available to ensure transparency and reproducibility of the calculation. Predictions from a model with an algorithm that is not available (to ECHA) to verify its functioning and to reproduce the predictions can hardly be accepted. In particular, special precautions are needed where non-transparent and difficult to reproduce methods have been used to build the (Q)SAR model (e.g. artificial neural networks using many structural descriptors).
- 3. A defined domain of applicability: the applicability domain (AD) and the limitations of the model have to be described to allow the assessment of the AD for the specific prediction (see Section 3.2 of this document). The most common methods for describing the AD are to consider the ranges of the individual descriptors and the presence of the structural fragments in the training set. Predictions from a model without information on the AD cannot be accepted.
- 4. **Appropriate measures of goodness-of-fit, robustness and predictivity:** this principle expresses the need for statistical validation of the model. Statistics on internal validation (goodness-of-fit and robustness) and external validation (predictivity) must be available. For instance, for regression models, the statistics of the regression model could be reported through the correlation coefficient (R^2), cross-validated (e.g. from leave-one-out procedure) correlation coefficient (Q^2) and the standard error of the model (*s*). It can be noted that an R^2 below 0.7, a Q^2 below 0.5 or an *s* above 0.3 should warn the (Q)SAR user of a potential low performance of the (Q)SAR model. The external validation should have been done by predicting compounds from an external set, i.e. not used for the model development. Statistics on the external validation are useful to estimate the uncertainty associated with the predictions.
- 5. A mechanistic interpretation, if possible: reasoning on the causal link between the descriptors used in the model and the predicted endpoint adds confidence in the reliability of the predictions e.g. a SAR model predicting skin sensitisation can be based on structural alerts. If reasoning is provided on how the structural alerts are associated

² <u>http://www.oecd.org/officialdocuments/publicdisplaydocumentpdf/?doclanguage=en&cote=env/jm/mono(2007)2</u>

with skin sensitisation (for example, they enclose electrophilic groups able to bind to proteins), then confidence in the prediction would increase due to the possibility to interpret the outcome of the model.

In general, a reference to a well-documented model or a (Q)SAR model reporting format (QMRF) attached to the registration dossier is recommended. See the *Guidance on information requirements and chemical safety assessment, Chapter R.6: (Q)SARs and grouping of chemicals (R.6.1.9.1)* for more details on the QMRF. The JRC's (Q)SAR Model Database (QMRF Inventory) is intended to provide information on (Q)SAR models submitted to the JRC for peer review.

There is no formal adoption process existing or foreseen for (Q)SAR models under REACH. The validity, applicability and adequacy of (Q)SAR models is assessed individually with the prediction generated for the target chemical.

NOTE: A valid (Q)SAR model does not necessarily produce a valid prediction. It is necessary to assess whether the substance falls within the applicability domain of the (Q)SAR model, that the results are adequate for the purposes of classification and labelling and/or risk assessment, and that adequate and reliable documentation of the applied method is provided.

3.2 Does the substance fall within the applicability domain of the (Q)SAR model?

It is important to verify that the target substance falls within the applicability domain (AD) of the model. The concept of AD was introduced to assess the probability of a chemical being covered by the (Q)SAR training set. Predictions outside the AD are normally not reliable and their use is hard to justify. A practical approach to check if a substance falls into the AD is to check the following elements:

- Descriptor domain: as indicated in the previous section, the AD of a model can be based on the range of the descriptors of the chemicals that are in the training sets (e.g. molecular weight, log Kow...). Therefore, if those ranges have been described, users of (Q)SAR models should check whether the target chemical falls into these ranges. It is expected that the descriptors follow normal distributions and if there are significant deviations, these should be explained. Clusters or other anomalies in the distribution of data (concerning one or both the dependent and the independent variables) may invalidate the model, and the prediction thereof.
- 2. **Structural fragment domain:** users of (Q)SAR models should check whether the (sub)structures (i.e. functional groups or meaningful fragments) of their target chemical are represented in the training set. In addition, the (Q)SAR users should also check whether their target chemical has more instances of a given fragment than the maximum for all training set compounds (e.g. if the target chemical has three hydroxyl groups and any compounds in the training do not have more than two hydroxyl groups, then the target chemical may be considered outside the AD).
- 3. Mechanistic and metabolic domains, if possible: to check these points is complex but still very valuable to support the reliability of the prediction. One approach could be to use the OECD QSAR Toolbox. Within this tool, the "profiling methods" can indicate whether the chemical shows relevant mechanisms of action that are not covered by the (Q)SAR model (i.e. not covered by its algorithm/mechanistic domain) while the "Metabolism/Transformations" – also available in the module "Profiling" of the QSAR Toolbox – can indicate metabolites/degradation products that should be considered. Significant potential differences in absorption, distribution, metabolism and elimination

between the target chemical and the chemicals in the training set might also invalidate the prediction from an otherwise valid model. These considerations are not explicitly addressed by the tools and might need to be considered separately from the model.

- 4. **Analogues in the training set:** having close structural analogues in the training set of the model increases the reliability of the prediction. Therefore, if the training set is available and the software does not do it automatically, the user should search for close analogues either manually (for small sets) or with the support of IT tools that provide structural search functionalities.
- 5. Accuracy of model predictions for analogues: if there are substances similar to the target substance, where the experimental results for the endpoint of interest are known (e.g. analogues in the training sets, in the validation sets or from any other source), the model can be run to predict these analogues and compare these predictions with the experimental results (to see how accurate the model is for these similar substances).
- 6. Considerations for specific substances: firstly, and as mentioned in Section 2.2, special considerations should be given to UVCBs, multi-constituents, additives, impurities, metabolites and degradation products. Secondly, most of the (Q)SAR models are developed for organic chemicals and do not address the specificity of some types of chemicals such as ionisable substances (e.g. salts, weak acids and bases), large molecular weight substances (e.g. polymers), potentially hydrolysable substances (e.g. esters, carbamates), surfactants (e.g. hydrocarbon chain with hydrophilic head) and isomers (e.g. stereoisomers, tautomers).

3.3 Is the prediction adequate for the purpose of C&L and/or risk assessment?

For a (Q)SAR prediction to be adequate, it should be not only reliable (i.e. derived from a valid (Q)SAR model and within its applicability domain), but also relevant for regulatory decision.

The adequacy of the prediction for the purposes of classification and labelling (C&L) and/or risk assessment is very much endpoint-dependent. Additional information may be needed to assess the adequacy of the prediction in the context of a regulatory decision.

Therefore, the validity (are the five OECD principles on scientific validity of a model fulfilled?), applicability (can reliable predictions be expected if the model is applied to the target substance?) and relevance (is the information which is needed for the risk assessment and/or C&L generated?) need to be assessed for each individual prediction.

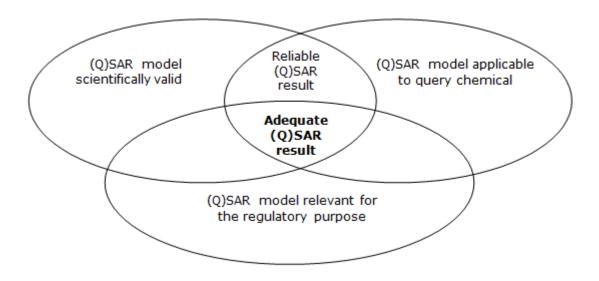


Figure 1: Criteria for identifying an adequate (Q)SAR model

C&L and risk assessment are based on well-defined requirements in terms of tests (and endpoints), thresholds and uncertainty analysis.

Therefore, results from (Q)SAR models should be equivalent to results obtained from the required experimental test.

Some examples of inadequacy are listed here:

- (Q)SAR models that are able to fully cover the complexity of higher-tier endpoints do not exist yet (e.g. repeated dose toxicity or reproductive toxicity). So far, the use of (Q)SARs as standalone information for these endpoints cannot be accepted. For instance, repeated dose toxicity tests provide many data points for effects in specific tissues (specific target organ toxicity) and it is not only the no observed adverse effect level (NOAEL) that matters. Indeed, effect results are needed to trigger other tests such as reproductive toxicity or for specific target organ toxicity single exposure/repeated exposure (STOT SE/RE) classification.
- If a quantitative outcome is needed (e.g. to derive a derived no-effect level (DNEL) or for classification) and the model only gives qualitative predictions (e.g. a negative or positive result), then the model is probably not adequate for the purpose.
- Uncertainty associated with predictions close to regulatory thresholds needs to be examined cautiously. For instance, if the predicted oral rat LD50 does not go beyond the threshold for classification but the standard error of the model and/or the error of the estimate is larger than this gap, then the prediction is probably not adequate.
- As required for an experimental bacterial reverse mutation assay (Ames test), the training set of the (Q)SAR model should include experimental results that cover the five bacterial strains in the presence and absence of metabolic activation (S9). This information has to be included in the documentation of the model and ideally also in the prediction report.
- (Q)SAR models for fish toxicity where the experimental results for the chemicals in the training set have been performed according to OECD test guideline 204 (14-day studies) cannot be used to predict long-term toxicity to fish because the test duration is too short.

• (Q)SAR models predicting the biodegradation half-life of a compound cannot be used as a standalone replacement of a simulation test as they do not cover the need to identify the degradation products (REACH Annex IX, 9.2.3 requirements).

3.4 How to report a (Q)SAR prediction in IUCLID

As discussed previously in Sections 2.4, 3.1, 3.2 and 3.3, the following information has to be reported in IUCLID:

- Information on the validity of the (Q)SAR model;
- Verification that the substance falls within the applicability domain of the (Q)SAR model; and
- Assessment of the adequacy of the results for the purposes of classification and labelling and/or risk assessment.

These three pieces of information should be compiled according to the (Q)SAR model reporting format (QMRF) and the (Q)SAR prediction reporting format (QPRF).

These formats are respectively available in Chapters R.6.1.9.1 and R.6.1.10.1 of the *Guidance* on information requirements and chemical safety assessment.

The QMRF is normally provided by the model developer while the QPRF is prediction-specific and should be prepared by the registrant using the information in the software report and manual.

The endpoint study record for a (Q)SAR prediction has to be reported in IUCLID as follows.

Block "Administrative data"

- The field "Type of information" states "(Q)SAR".
- The field "Adequacy of study" states whether the estimate is used as a key study, as a supporting study or in a weight of evidence approach.
- The field "Reliability" states the reliability score, bearing in mind that for (Q)SAR predictions it should normally be a maximum of 2.

Administrative data \land			
®			
Endpoint			
partition coefficient	🔻	Remarks	
Type of information			
(Q)SAR	🔻	Other	Remarks
Adequacy of study			
weight of evidence	🔻		
Robust study summary			
Used for classification			
Used for SDS			
Study period			
Reliability			
2 (reliable with restrictions)	🔻	Other	

• The field "Justification for type of information" can be used to report information equivalent to the one listed in the QMRF and QPRF. A free text template to help fill this field is available by clicking on the A button and selecting "Type 'QSAR prediction'".

Justification for type of information	Free text templates	×
1	View / edit / insert freetext template as appropriate In case of several options, click the heading of the desired freetext template. Delete/add elements and edit text set in [] (if any) as appropriate	A,
	Type "Waiving of standard information'	
	Type 'Experimental study planned / Testing proposal on vertebrate animals'	
	Type 'QSAR prediction'	
	Type 'Read-across (analogue)'	

• Otherwise, the field "Attached justification" can be used to attach the QMRF and the QPRF.

Attached justification						
Attached justification	Reason / purpose					
QPRF_constituent1.docx	(Q)SAR prediction reporting (QPRF)					
QMRF_modelX.docx / 24.	(Q)SAR model reporting (QMRF)					

Block "Data source"

- The field "Title" is used to state the name and version of the program and/or the title of the publication. The field "Year" needs to include the year when the software program was released or when the (Q)SAR model was published. Additionally, the "Bibliographic source" has to give information on the (Q)SAR model.
- The field "Data access" provides information on the accessibility of the model.

Data source \land —										
Reference										
Title	Author	Reference type	Year	Bibliographic source	Testing labo	Report no.	Compan	Company s	Report d	Remarks
EPI Suite Version 4.11			2012	KOWWIN - Meylan and Howard (19						
(+) Add	🗙 Delete	e	🛧 Move up	🔶 Move down 🔰	Go to link targ	et				
Data access										
data published		`	Other			Remarks				

Block "Materials and methods"

Either the field "Guideline" (in table "Test guideline") or the field "Principles of method if other than guideline" should be filled in.

- In the field "Guideline", the user can select "other guideline" and provide text in the adjacent field. This text could, for instance, refer to the REACH Guidance on QSARs R.6 or to the test guidelines used to generate the data for the training set.
- Otherwise, in the field "Principles of method other than guideline", the user could provide further details/references on the (Q)SAR model.

Materials and m	nethods \land 👘					
Test guideline				1		
Qualifier		Guideline		/ersion / remarks	Deviations	
		other: REACH Guidance on	2SARs R.6			
		L				
(+) Add	J Edit	🗙 Delete 🔥	Move up	V Move down		
Principles of method if	other than guideline	· · · · · ·				
A ×						~
Meylan, W.M and P.H	Howard. 1995. Atom/f	ragment contribution method for	r estimating ocanol	water partition coefficients.	J. Pharm. Sci. 84: 83-92	

Block "Test materials"

- The field "Test material information" should include a link to the substance for which the prediction was made.
- The SMILES notation should be reported in the field "Specific details on test material used for the study".

Test material A		_	
Test material information			
4-Methyl-2-hexene	>	Ø	×
Specific details on test material used for the study			
			\sim
SMILES: CCC(C)C=CC			

NOTE: the registered substance may contain more than one constituent and/or impurities. In such cases, it may be useful to prepare an individual endpoint study record and a (Q)SAR prediction reporting format (QPRF) for each constituent/impurity to be able to address each chemical separately (recommended if constituents have different properties and therefore different models, assessments, etc. have to be applied).

Block "Results and discussion"

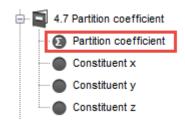
- The (Q)SAR-predicted result should be reported in the structured result fields. This
 allows the user to transfer information automatically from these result fields to the
 chemical safety report (CSR) when the IUCLID report generator is used. The list of
 fields to be filled in the "Results and discussions" block will vary depending on the
 endpoint.
- Therefore, we recommend consulting the manual on "How to prepare registration and PPORD dossiers" for instructions on how to fill in the results.

Results and discussion A									
Partition coefficient									
Key result	Туре	Partition	coefficient	Temp.	pН	Remarks on result			
	log Pow	3.49				other: / QSAR predicted value			
⊕ Add	🖋 Edit	🗙 Delete	↑ Move	up 🔶 Move dowr					
Details on results									
Any other info	mation on resu	Its incl. tabl	es 🔨 👘						
Normal	Default font	• •	A [▼] B I	⊻ ≣ ≣ ≡ 1=	📰 🗚 🖓				
KOWWIN pre	dicted that 4-Met	hyl-2-hexene	has a log Kow	= 3.49					

• If it is not possible to fill in all the structured result fields required to pass the technical completeness check, then the fields "Remarks" (at the right end of the table) or the field "Any other information on results incl. tables" could be used instead.

Endpoint study summary

In addition, it is advisable to create an endpoint study summary when more than one endpoint study record is available and to provide the overall assessment on the particular endpoint. This would enable the automatic transfer of this information to the CSR when the IUCLID report generator is used.



4. Practical examples

This section describes how to assess the reliability of QSAR predictions. The assessment depends on the software and on the target endpoint. The examples used in this section are based on computer programs that are widely used and freely available. The fact that these programs are used in these examples is not an endorsement from ECHA.

Usually, the use of QSARs is limited to experts. With these practical examples, the aim is to allow less experienced people to use and interpret QSARs for at least some endpoints (like in the following examples).

The programs used in the examples can predict several endpoints. However, only one endpoint per program (corresponding to one REACH requirement) has been used for each example. In most cases, predictions for different endpoints from the same program are reported (and can be assessed) in a similar way.

The four endpoints predicted in the following examples are log Kow, ready biodegradability, short-term toxicity to fish and acute mammalian toxicity. Those endpoints have been selected as representatives of REACH Annexes VII or VIII requirements for physicochemical properties, environmental fate, ecotoxicological and toxicological information.

4.1 Log Kow (EPI Suite)

a) Introduction

Partition coefficient n-octanol/water is a REACH requirement for all substances produced or imported above one tonne/year (REACH Annex VII). It is commonly expressed as a logarithmic value called log Kow or log P.

Many QSAR models are available to predict log Kow. KOWWIN – which is part of EPI Suite – is one of the most commonly used programs. KOWWIN uses a "fragment constant" method to predict Kow. Fragment constant methods divide the chemical structure into smaller structural fragments (atoms or larger functional groups). Each fragment is associated with a preassigned coefficient value called fragment constants. The predicted log Kow value is obtained by summing all the fragment constants appearing in the chemical structure.

At the time of writing this manual, the current version of EPI Suite[™] was version 4.11, which has been used to prepare this example.

Link to the (Q)SAR program: <u>http://www.epa.gov/tsca-screening-tools/epi-suitetm-estimation-program-interface</u>

Information on the QSAR model	Where to find this information	How to use this information
Description of the algorithm	Available in the KOWWIN Help ³ , in section "Methodology".	See Section 3.1 of this practical guide.
Statistics (goodness-of-fit and robustness)	Available in the KOWWIN Help, in section "Accuracy & Domain".	See Section 3.1 of this practical guide.
Information on	Where to find this	How to use this information
the applicability	information	
domain		
General applicability domain (chemical classes covered/not covered by the model)	Available on the front page of EPI Suite (<i>"The intended</i> <i>application domain is organic</i> <i>chemicals. Inorganic and</i> <i>organometallic chemicals are</i> <i>generally outside the</i> <i>domain."</i>) and in the KOWWIN Help, in sections "Ionisation" and "Zwitterion Considerations".	As indicated in Section 3.2 point 6 of this practical guide, special attention should be paid for some types of chemicals. KOWWIN includes some "corrections" for ionisable and zwitterionic substances to refine the low predictivity for these substances.
Descriptor domain	Available in the KOWWIN Help, in section "Accuracy & Domain".	The user should check that the target chemical is in the molecular weight range of the compounds in the training set (i.e. between 18 and 720).
Structural fragment domain	The KOWWIN results window lists the fragments (and their numbers) found in the target chemical. Appendix D of the KOWWIN Help gives the maximum number of fragments that occur in any individual compound of the training set.	The user should check whether the number of each fragment found in the target chemical (column "NUM" in the KOWWIN results window) does not exceed the maximum number of this fragment that occurs in any individual compound of the training set (column "Training set/Max" of Appendix D of the KOWWIN Help). Notes on specific substructures: - For some substructures, KOWWIN reports correction factors. In this case, the user should do the same verification described above for the numbers of fragments. - For some substructures, the coefficient has been estimated (if this is the case, it will be reported as a note in the KOWWIN results window). It should be kept in mind that this estimation brings additional uncertainty to the overall prediction.

b) How to check the reliability of the prediction

³ KOWWIN Help can be accessed by clicking on the tab "Help" at the top of the KOWWIN window.

Training set and validation set	Where to find this information	How to use this information
	The training and validation sets can be downloaded through links given at the bottom of the section "Accuracy & Domain" of the KOWWIN Help. The first link provides an Excel file with chemical names, experimental and estimated values of log Kow: http://esc.syrres.com/interkow /KowwinData.htm. The second link provides an SDF file with the same	The user should check whether there are close analogues in the training and validation sets (either manually in the Excel file or automatically with the SDF file). If there are close analogues, the user should check how well they have been predicted by KOWWIN.
	information as the Excel file plus structural information ⁴ : <u>http://esc.syrres.com/interkow</u>	
	/EpiSuiteData_ISIS_SDF.htm	

c) Additional information on EPI Suite

- The EPI Suite Help files do not have exactly the same table of contents between the various estimation programs (e.g. between KOWWIN, BIOWIN...). Therefore, the user should go through each specific Help file to identify the information needed to check the reliability of the prediction (as done for KOWWIN in the table above).
- Several programs of EPI Suite display the experimental result of the target chemical if it is part of the training or validation sets. For instance, KOWWIN contains a database of approximately 13 500 experimental log Kow and if an experimental log Kow value is available for the target chemical it will be displayed in the "Results" window (see green rectangle in the figure below). In this case, it is recommended to report this experimental data and the EPI Suite estimate in two different endpoint study records.

Kowwin Results	
Print Save Results Copy Remove Window Help	
	A
Log Kow(version 1.68 estim	nate): 2.82
Experimental Database Structure Match: Name : ATRAZINE	
CAS Num : 001912-24-9	
Exp Log P: 2.61	
Exp Ref : HANSCH,C ET AL. (1995)	
SMILES : n(c(nc(n1)NC(C)C)NCC)c1CL CHEM - 1 3 5-Triazing-2 A-diaming 6-chlor	

⁴ Many tools are available to read SDF files. These tools allow users to visualise the chemicals, to search for closest structural analogues or to search for specific substructures. Some of these tools are freely available (e.g. Knime or the OECD QSAR Toolbox).

4.2 Ready biodegradability (VEGA)

a) Introduction

Ready biodegradability is a REACH requirement for all substances produced or imported above one tonne/year (REACH Annex VII). The main outcome of a ready biodegradability test is the classification of the chemical either as "readily biodegradable" or as "not readily biodegradable".

The VEGA platform contains several QSAR models for various endpoints. One of these models predicts ready biodegradability (model developed by the Istituto di Ricerche Farmacologiche Mario Negri). This model is based on structural alerts.

Four sets of substructures (i.e. fragments) are included in this model and those sets are classified as "non-readily biodegradable", "possible non-readily biodegradable", "readily biodegradable" and "possible readily biodegradable". A target chemical is always considered non-biodegradable if at least one fragment related to "non-readily biodegradablity" is found.

At the time of writing this manual, the current version of VegaNIC is 1.1.0, which has been used to prepare this example.

Link to the (Q)SAR program: <u>http://www.vega-qsar.eu/</u>

Information on the (Q)SAR model	Where to find this information	How to use this information
Description of the algorithm	Available in the "Guide to Ready Biodegradability Model" ⁵ (in Sections 1.2, 1.4 and 1.5).	See Section 3.1 of this practical guide.
Statistics (goodness- of-fit and Robustness)	Available in the "Guide to Ready Biodegradability Model" (in Section 1.6).	See Section 3.1 of this practical guide.

b) How to check the reliability of the prediction

⁵ This guide can be downloaded from the VEGA program by clicking on the tab "SELECT", then on the tab "Environ", then on the question mark icon adjacent to the "Ready Biodegradability model (IRFMN)".

Information on the applicability	Where to find this information	How to use this information
domain General applicability domain (chemical classes covered/not covered by the model)	Partly available in the Vega prediction report.	If less than three "golden stars" are displayed in Section 1 of the Vega report, this indicates that at least one issue has been detected for the prediction and that, therefore, the prediction might not be reliable. In this case, the user should investigate the issues thoroughly. Note that the issues are further detailed in Section 3.2 of the VEGA report. In addition, as indicated in Section 3.2 point 6 of this practical guide, special attention should be paid to some types of chemicals.
Descriptor domain	Not applicable as the model is based only on fragments/structural alerts	The training set of the model is based on tests performed according to the OECD 301C guideline. Some substances led to unreliable results using this guideline (e.g. low water soluble, volatile or absorptive substances). Therefore, if the target substance has low water solubility, a high vapour pressure or a high log Koc, then the user should keep in mind that the prediction might be erroneous. In addition, the user should verify if there are close analogues in the training set (especially if the target substance has a large molecular weight).
Structural fragment domain	Available in the "Guide to Ready Biodegradability Model" and in the Vega prediction report.	If the target chemical does not contain any of the fragments listed in Sections 1.4 and 1.5 of the model's guide, then no prediction is given by the tool. In addition, in Section 3.2 of the Vega Prediction report, if the "Atom Centred Fragments similarity check" gives an ACF index <1, this would indicate that there is at least one atom centred fragment of the target chemical which has not been found in the compounds of the training set (or which is rarely present). In this case, the user should determine if these missing/rare fragments (listed in Section 4.1 of the report, if any) could have an impact on biodegradation.

Training set and validation set	Where to find this information	How to use this information
	Available in the file called "Training set (plain text with SMILES)" ⁶ .	This file contains the SMILES of the training set compounds and of the test set compounds. The VEGA report displays the most similar compounds found in the training set and in the test set in Section 3.1. The user should check in this section whether these compounds are closely similar to the target chemical and if their experimental outcome agrees with the predicted one.

c) Additional information on VEGA

If there is an experimental result for the target chemical in the training set or in the test set, this data will be displayed in the VEGA report (in Section 1 of the report). In this case, the user should look for further information about this test – e.g. by searching for this experimental test within the QSAR Toolbox – and report the details of this test in an endpoint study record only dedicated to this experimental study.

⁶ This file can be downloaded from the VEGA program by clicking on the tab "SELECT", then on the tab "Environ", then on the question mark icon ¹/₂ adjacent to the "Ready Biodegradability model (IRFMN)".

4.3 Short-term toxicity to fish (ECOSAR)

a) Introduction

Short-term toxicity testing on fish is a REACH requirement for all substances produced or imported above 10 tonnes/year (REACH Annex VIII). The endpoint to be derived is the LC50, which is the concentration lethal to 50 % of the fish.

The ecological structure activity relationships (ECOSAR) class program is a collection of QSAR models estimating aquatic toxicity, including short-term toxicity to fish. Most of the ECOSAR models are based on the relationships between log Kow and toxicity (LC50 or EC50) and take into account different structural classes.

NOTE: Regarding EPI Suite and ECOSAR, ECOSAR is developed and maintained as a standalone program. Even if the latest version of ECOSAR (v.1.11) has been included in the latest version of EPI Suite (v.4.11), it is still recommended to use the ECOSAR standalone program because it will inform the user if an experimental value is available for the target chemical while the ECOSAR integrated in EPI Suite does not have this functionality.

Link to the (Q)SAR program: <u>http://www.epa.gov/tsca-screening-tools/ecological-structure-activity-relationships-ecosar-predictive-model</u>

Information on the (Q)SAR model	Where to find this information	How to use this information
Description of the algorithm	Available in the ECOSAR Help ⁷ , in section "Interpreting QSAR Class Ref Docs" and in each "QSAR Equation Document" of the various chemical classes.	See Section 3.1 of this practical guide.
Statistics (goodness- of-fit and robustness)	Available in the ECOSAR Help, in each "QSAR equation document" of the various chemical classes.	See Section 3.1 of this practical guide. For instance, the user should check that: - R^2 (i.e. correlation or determination coefficient) is not too low (e.g. not below 0.7), - The data points are not too few. This is especially important as several ECOSAR classes contain only one or few data points which can lead to unreliable predictions. - The linear regression is not built on scattered data points.

b) How to check the reliability of the prediction

⁷ ECOSAR Help can be accessed by clicking on the tab "Help" at the top of the ECOSAR window.

Information on the applicability domain	Where to find this information	How to use this information?
General applicability domain (chemical classes covered/not covered by the model)	Available in the ECOSAR Help, in sections "Chemicals that should not be profiled", "Surfactants" and "Special ECOSAR classes/dyes".	The user should check whether its chemical is one of those "that should not be profiled" with ECOSAR or if it should be predicted using a special ECOSAR class.
		In addition, it should be kept in mind that most of the ECOSAR models are based on the relationships between log Kow and toxicity (LC50 or EC50) addressing the uptake of chemicals through the aqueous phase.
		For very hydrophobic or very sorptive substances, the uptake from food can also be an important exposure pathway.
		Moreover, other properties of the substance can trigger specific modes of action (e.g. the substance is likely to be more reactive if it is protein binding).
		Finally, it should also be kept in mind that some chemical classes expressing excess toxicity (compared to the baseline toxicity model) have not yet been included in the ECOSAR program.
Descriptor domain	Available in the ECOSAR Help, in each "QSAR equation document" of the various chemical classes and in the ECOSAR results window.	The user should check that the molecular weight (MW) of the target chemical does not exceed 1 000. The MW of the target chemical is indicated in the ECOSAR results window.
		If the log Kow of the compound exceeds the maximum log Kow of the class-specific model (e.g. for Fish 96 hour LC50, the maximum log Kow varies between class models from 2.6 to 8.2, often being 5) or if the predicted effect (e.g. Fish 96 hour LC50) exceeds the water solubility of the compound, then the prediction might be unreliable and long-term toxicity data might be more appropriate. Detailed instructions on how to check these points is given in section c) below.

	"surfactants" special classes).	 provided (if not claimed confidential (CBI)). This requires manual work since the training sets are not in a structurally searchable format. Software programs⁸ are available to help the user derive the chemical structures if needed. In addition, the whole structural domain of the model cannot be determined if there are CBI compounds in the training set.
.	Where to find this information	How to use this information
	The training sets of all models (except the special classes for dyes and surfactants) are available from the "QSAR equation document" of each chemical class. However, some of the compounds in the training set have been kept confidential (CBI). Validation (test) sets do not exist in ECOSAR. ⁹	The user should check whether there are close analogues in the training set of the specific class model (based on the CAS numbers or on the chemical names provided). If there are close analogues, the user should compare their predicted and experimental values.

c) Additional information on ECOSAR

 In ECOSAR, all predictions (except for surfactants and dyes) are based on log Kow. By default, ECOSAR estimates Kow using KOWWIN. However, if users have a reliable measured log Kow they should insert it in the data entry screen (see screenshot below). This value will be taken into account by the model and will reduce the uncertainty of the prediction.

⁸ Several software programs allow users to derive the chemical structure from the CAS number or from the chemical name. Some of these programs are freely available (e.g. Chemspider or the OECD QSAR Toolbox).

⁹ However, there are several peer-reviewed publications assessing the external performance of ECOSAR.

Ecosar v1.11							
File Edit Function	ns BatchMode	ShowStructure	e Special_Cla	asses Help			
	Clear Screen	Previous	Get User	Save User	PhysProp	Calculate	
Enter SMILES:							
Enter NAME: NameLookup							
CAS Number:				User Entere	d Values:		
CAS Input		_		ubility (mg/L):			
Chemical ID 1: Melting Point (deg C): Log Kow:							

• If there is an experimental result for the target chemical in the training set then this data will be displayed in the ECOSAR results window (see "Available Measured Data from ECOSAR Training Set" in the figure below). In this case, it is recommended to report this experimental data and the ECOSAR estimate in two different endpoint study records.

Ecosar Result	5						X
Print Save Res	ults TableSave	Copy Remove	Window	Help			
Available Me	asured Data	from ECOSAR	Traini	ng Set			
				Measured			
CAS No	Organism	Duration	End Pt	mg/L (ppm)	Ecosar Class	Reference	
000050-00-0	Eich	======= 96-hr	LC50	 2h 1	Aldehydes (Mono)	DUL	
000020-00-0	11211	70-III	L020	24.1	niuengues (nono)		

- If the compound has been allocated to a specific class (e.g. "Aldehydes, mono" class), then the effect level of this class should be taken into account and not only the one from the "Neutral organics" class (i.e. baseline toxicity potential). In general, if the program identifies several classes, it is recommended to use the most conservative effect level from any of these classes (and to consider the potential synergistic toxicity effect of these various classes).
- Each of the "QSAR equation documents" is class-specific and some contain more information than others under the titles "APPLICATION" and "LIMITATIONS" (see examples below). Therefore, users should carefully read the "QSAR equation document" of the specific class/endpoint for which they want to perform a prediction.

APPLICATION:

This SAR may be used to estimate the toxicity of aldehydes (mono) with log Kow values of less than 5.0 and molecular weights less than 1000. <u>Acrolein is about 1400 times more toxic than predicted by this SAR.</u>

LIMITATIONS:

Aliphatic polyamines with greater than 3 aliphatic amines and/or an amine-nitrogen composition of $\ge 25\%$ exhibit excess toxicity based on available CBI data. Insufficient data were available to construct a QSAR, but fish toxicity test data of compounds with 27% amine-nitrogen resulted in LC₅₀ values that were ~100x more toxic than estimations predicted from the aliphatic amine class.

• As indicated in the table from section b) (see descriptor domain), if the log Kow of the compound exceeds the maximum log Kow of the class-specific model or if the predicted effect value exceeds the water solubility of the compound, then the prediction might be unreliable (see ECOSAR results window below).

Ecosar Results					
Print Save Results TableSave Copy Remove	Window Help				
Values used to Generate ECOSAR Prof	 ile 				
Log Kow: 5.252 (EPISuite Kowwi Wat Sol: 0.052 (mg/L, PhysProp	n v1.68 Estim DB exp value	ate))			
ECOSAR v1.1 Class-specific Estimati	 ons 				
Neutral Organics				Duedieted	
ECOSAR Class Organi	5M =======	Duration		······	
Neutral Organics : Fish Neutral Organics : Fish		96-hr	LC50 ChV	0.140 0.020	
Note: * = asterisk designates: Chemical may not be soluble enough to measure this predicted effect. If the effect level exceeds the water solubility by 10%, typically no effects at saturation (NES) are reported.					
Class Specific LogKow Cut-Offs					
If the log Kow of the chemical is greater than the endpoint specific cut-offs presented below, then no effects at saturation are expected for those endpoints.					
Neutral Organics:					
Maximum LogKow: 5.0 (Fish 96-hr LC5	0; Daphnid LC	50, Mysid	LC50)		

- Ecotoxicity chronic values (ChV) can be predicted with ECOSAR. However, users should pay attention to the following points:
 - The (Q)SAR models available for predicting these chronic values are often built on small to very small training sets.
 - The chronic value (ChV) is defined as the geometric mean of the no observed effect concentration (NOEC) and the lowest observed effect concentration (LOEC). However, under REACH, NOECs are the effect concentrations used to assess long-term toxicity data to aquatic organisms. A proxy to derive the NOEC is to divide the ChV by √2.
 - Acute-to-chronic ratios (ACRs) are used by ECOSAR when measured data are lacking within a class. Such predictions are flagged with an exclamation mark (!) in the ECOSAR results window (see screenshot below) and should be considered with caution.

Print Save Results TableSave Copy Remove Window Help						
Phenol Amines : Fish ChV 0.565 👖 🔺						
NOTE: ! = exclamation designates: The toxicity value was estimated through						
application of acute-to-chronic ratios per methods outlined in						
the ECOSAR Methodology Document provided in the ECOSAR Help Menu.						

4.4 Acute toxicity to rats (T.E.S.T.)

a) Introduction

Acute toxicity by oral route is a REACH requirement for all substances produced or imported above one tonne/year (REACH Annex VII). The preferred test species according to the OECD test guidelines is the rat, and the endpoint to be derived is the LD50 (generally expressed in mg/kg body weight).

Most of the software programs that predict acute oral toxicity are commercial except T.E.S.T. (Toxicity Estimation Software Tool), which is made freely available by the US EPA. This tool offers four different methods to predict acute oral toxicity to rats: hierarchical, FDA, nearest neighbour and consensus methods.

The consensus method predicts the toxicity simply by taking an average of the predicted toxicities from the three other methods (hierarchical, FDA and nearest neighbour methods).

This consensus method should be the preferred one to use as it achieved the best results for prediction accuracy and leverage compared to the three other methods. To check the reliability of the predictions coming from this consensus method, the user should assess the reliability of the three other methods. Therefore, the information given in the following table addresses all of these four methods.

At the time of writing this manual, the current version of T.E.S.T. is 4.1, which has been used to prepare this example.

Link to the (Q)SAR program: <u>http://www2.epa.gov/chemical-research/toxicity-estimation-software-tool-test</u>

Information on the (Q)SAR model	Where to find this information	How to use this information
Description of the algorithm	Available in the T.E.S.T. User's Guide ¹⁰ , in Sections 1.2 and 2.2 called "QSAR methodologies".	

b) How to check the reliability of the prediction

¹⁰ This User's Guide can be accessed by clicking on the tab "Help" at the top right of the T.E.S.T. window.

Statistics (goodness-of-fit and robustness)	Available in the T.E.S.T. User's Guide, in Sections 2.2.1, 2.2.2, 2.3.1 and 4.4.1.	In Section 4.4.1 of the User's Guide, it is written that R^2 is less than 0.6 for the hierarchical, FDA and nearest neighbour methods and that $\frac{R^2-R_0^2}{R^2}$ is more than 0.1 for all of the four methods. Therefore, those methods do not satisfy the conditions for an acceptable predictive power as indicated in Section 2.3.1 of the T.E.S.T. User's Guide. In addition, it is stated in Section 4.4.1 of the User's Guide that <i>"The prediction statistics for this endpoint were not as good as those for the other endpoints.</i> <i>This is not surprising since this endpoint has a higher degree of experimental uncertainty and has been shown to be more difficult to model than other endpoints"</i> .
Information on the applicability domain	Where to find this information	How to use this information
General applicability domain (chemical classes covered/not covered by the model)	Available in the T.E.S.T. User's Guide, in Section 3.4.	For instance, in Section 3.4 of the User's Guide, it says that "salts, undefined isomeric mixtures, polymers, or mixtures were removed [from the training set]". Therefore, those types of substances should not be predicted with T.E.S.T.
Descriptor domain	Information can be found in the T.E.S.T. User's Guide (Sections 2.1 and 2.2.1, 2.2.2).	For the hierarchical and FDA methods, 797 descriptors can be used in the model equation depending on the target chemical.
		These methods give a prediction only if the target chemical is within the ranges of descriptor values for the chemicals in the cluster (for the descriptors appearing in the cluster model).
		Concerning the nearest neighbour method, the descriptor differences between the target chemical and its structural analogues are not accounted for by T.E.S.T.
		To increase the confidence in the predictions given by these three methods, the user could potentially check that the target chemical is within the ranges of log Kow and molecular weight of the compounds in the cluster (or of the three nearest neighbours).

Structural fragment domain	Information related to that can be found in the T.E.S.T. User's Guide (Sections 2.2.1 and 2.2.2).	The hierarchical and FDA methods give a prediction only if the compounds in the cluster have at least one example of each of the fragments contained in the target chemical. Concerning the nearest neighbour method, the structural differences between the target chemical and its structural analogues are not accounted for by T E S T
		for by T.E.S.T. Therefore, users should check this visually from the T.E.S.T. report.
Training set and	Where to find this	How to use this information
validation set	information The results report of each of the four methods displays the predictions for the most similar chemicals in the validation set (i.e. prediction set or external test set) followed by the predictions for the most similar chemicals in the training set. In addition, the training and validation sets can be downloaded as SDF files through the link given at the bottom of this web page: http://www2.epa.gov/chemical -research/toxicity-estimation- software-tool-test	 In the results report of each method, for the predictions performed on the analogues in the validation and training sets, the user should check: whether these predicted values are close to the experimental values; and whether the mean absolute error (MAE) for these analogues is not more than the MAE of the entire set (this would be displayed as a red cell instead of a green cell).

c) Additional information on T.E.S.T.

• If there is an experimental result for the target chemical in the training set or in the validation set (i.e. prediction set or external test set), then this data will be displayed in the T.E.S.T. results report (see "^aNote" in the screenshot below). The user can check this experimental value by clicking on the name of the source (in the example below by clicking on "ChemidPlus"). In this case, it is recommended to report this experimental data and the T.E.S.T. estimate in two different endpoint study records.

Prediction results				
Endpoint	Experimental value (CAS= 28785-06-0) Source: <u>ChemidPlus</u>	Predicted value ^a		
Oral rat LD ₅₀ -Log10(mol/kg)	1.97	1.92		
Oral rat LD ₅₀ mg/kg	1599.21	1762.18		
^a Note: the test chemical was present in the external test set.				

• Some methods (hierarchical, FDA and group contribution methods) give the "Prediction interval" (i.e. 90 % confidence interval) in the results report. When using these, the

user should check that this interval does not overlap with a regulatory threshold (e.g. CLP, persistent, bioaccumulative and toxic (PBT) or risk assessment thresholds).

Appendix 1. QSAR models related to REACH endpoints

The QSAR computer programs listed in this appendix are widely known and are given to inform REACH registrants on the QSAR models availability for each of the REACH endpoints. However, it constitutes neither an exhaustive list of available programs nor a list of regulatory validated QSAR models. So far, most of the toxicological and ecotoxicological information required under REACH can rarely be fulfilled with QSAR predictions alone.

a. Physicochemical endpoints

Endpoint	Software tool	Models/Modules	Free or
			Commercial
Melting/freezing	EPI Suite (US EPA)	MPBPVP	Free
point			
Boiling point	EPI Suite (US EPA)	MPBPVP	Free
	T.E.S.T. (US EPA)	Normal boiling point	Free
	ACD/Percepta	Boiling Point/Vapor Pressure	Commercial
	(ACD/Labs)	Module	
Relative density	T.E.S.T. (US EPA)	Density	Free
Vapour pressure	EPI Suite (US EPA)	MPBPVP	Free
	T.E.S.T. (US EPA)	Vapor pressure at 25°C	Free
	ACD/Percepta	Boiling Point/Vapor Pressure	Commercial
	(ACD/Labs)	Module	
Surface tension	T.E.S.T. (US EPA)	Surface tension at 25°C	Free
Water solubility	EPI Suite (US EPA)	WSKOW and WATERNT	Free
	T.E.S.T. (US EPA)	Water solubility at 25°C	Free
	ACD/Percepta	Aqueous Solubility Module	Commercial
	(ACD/Labs)		
	ADMET Predictor	Physicochemical and	Commercial
	(Simulations Plus)	Biopharmaceutical Module	
	Discovery Studio	ADMET Descriptors	Commercial
	(Accelrys)		
Partition coefficient	EPI Suite (US EPA)	KOWWIN	Free
(log Kow)	VEGA (IRFMN)	LogP Models	Free
	ACD/Percepta	LogP Module	Commercial
	(ACD/Labs)		
	ADMET Predictor	Physicochemical and	Commercial
	(Simulations Plus)	Biopharmaceutical Module	
	JChem (ChemAxon)	LogP/logD predictor	Commercial
Flash point	T.E.S.T. (US EPA)	Flash point	Free
Dissociation constant	Danish QSAR Database	pKa from ACD/Labs	Free
	(DTU)		
	ACD/Percepta	pKa Module	Commercial
	(ACD/Labs)		
	ADMET Predictor	Physicochemical and	Commercial
	(Simulations Plus)	Biopharmaceutical Module	
	JChem (ChemAxon)	pKa predictor	Commercial
Viscosity	T.E.S.T. (US EPA)	Viscosity at 25°C	Free

Endpoint ¹¹	Software tool	Models/Modules	Free or Commercial
Hydrolysis	EPI Suite (US EPA)	HYDROWIN	Free
Ready biodegradability	Danish QSAR Database (DTU)	Not Ready Biodegradability model from DTU	Free
	EPI Suite (US EPA)	BIOWIN and BioHCwin	Free
	VEGA (IRFMN)	IRFMN model	Free
	CATALOGIC (LMC)	Several OECD 301 models	Commercial
	Discovery Studio (Accelrys)	Aerobic Biodegradability model	Commercial
	Meta-PC (MultiCASE)	Aerobic Microbial Biodegradation expert rules	Commercial
Bioaccumulation in	EPI Suite (US EPA)	BCFBAF	Free
aquatic species	T.E.S.T. (US EPA)	Bioaccumulation factor	Free
	VEGA (IRFMN)	CAESAR, Meylan and KNN/Read-Across models	Free
	CASE Ultra (MultiCASE)	EcoTox model bundle	Commercial
	CATALOGIC (LMC)	Two BCF base-line models	Commercial
Adsorption/desorption screening	EPI Suite (US EPA)	KOCWIN	Free

b. Environmental fate and pathways endpoints

¹¹ The REACH requirement "Simulation testing in water, soil or sediment and identification of degradation products" is not listed in this table because – to our knowledge – there are no QSAR tools/models available for this endpoint.

c. Ecotoxicological endpoints

Endpoint ¹²	Software tool	Models/Modules	Free or Commercial
Short-term toxicity to fish	Danish QSAR Database (DTU)	Fathead minnow 96h LC50 from DTU	Free
	ECOSAR (US EPA)	Fish, 96-hr, LC50	Free
	T.E.S.T. (US EPA)	Fathead minnow LC50 (96 hr)	Free
	VEGA (IRFMN)	SarPy/IRFMN classification and KNN/Read-Across model	Free
	ADMET Predictor (Simulations Plus)	Toxicity module	Commercial
	CASE UItra (MultiCASE)	EcoTox model bundle	Commercial
	Discovery Studio (Accelrys)	Fathead Minnow LC50	Commercial
Long-term toxicity to fish	ECOSAR (US EPA)	Fish, ChV ¹³	Free
Short-term toxicity to aquatic invertebrates	Danish QSAR Database (DTU)	Daphnia magna 48h EC50 from DTU	Free
(daphnia)	ECOSAR (US EPA)	Daphnid, 48-hr, LC50	Free
	T.E.S.T. (US EPA)	Daphnia magna LC50 (48 hr)	Free
	VEGA (DEMETRA)	Daphnia Magna LC50 (48 h)	Free
	ADMET Predictor (Simulations Plus)	Toxicity module	Commercial
	Discovery Studio (Accelrys)	Daphnia EC50	Commercial
Long-term toxicity to aquatic invertebrates (daphnia)	ECOSAR (US EPA)	Daphnid, ChV ¹¹	Free
Toxicity to aquatic plants (algae)	Danish QSAR Database (DTU)	Pseudokirchneriella s. 72h EC50 from DTU	Free
	ECOSAR (US EPA)	Green Algae, 96-hr, EC50	Free
Short-term toxicity to terrestrial invertebrates	ECOSAR (US EPA)	Earthworm, 14-day, LC50	Free

¹² The following REACH requirements are not listed in this table because – to our knowledge – there are no QSAR tools/models available for these endpoints: Toxicity to aquatic micro-organisms (activated sludge respiration inhibition testing),

Long-term toxicity to sediment organisms,

Long-term toxicity to terrestrial invertebrates,

Short-term toxicity to terrestrial plants,

Long-term toxicity to terrestrial plants,

Toxicity to terrestrial micro-organisms, and

Long-term toxicity to birds.

¹³ See section 4.3 c) of this practical guide for further information on these chronic values.

d. Toxicological endpoints

Endpoint	Software tool	Models/Modules	Free or
	Danish QSAR Database	Modele for equite toxicity in	Commercial
Acute toxicity	(DTU)	Models for acute toxicity in rodents from ACD/Labs	Free
	T.E.S.T. (US EPA)	Oral rat LD50	Free
	ACD/Percepta	Acute Toxicity Module	Commercial
	(ACD/Labs)		Commercial
	ADMET Predictor	Toxicity module	Commercial
	(Simulations Plus)		
	CASE Ultra (MultiCASE)	AcuteTox model bundle	Commercial
	Discovery Studio	Rat oral LD50 and rat inhalation	Commercial
	(Accelrys)	toxicity LC50	
Skin irritation	Danish QSAR Database	Skin irritation model	Free
or skin	(DTU)		
corrosion	OECD QSAR Toolbox	Skin irritation/corrosion Inclusion (and Exclusion) rules by BfR	Free
	ToxTree (JRC)	Skin irritation / skin corrosion	Free
	ACD/Percepta	Irritation Module	Commercial
	(ACD/Labs)		
	CASE Ultra (MultiCASE)	SkinEye Toxicity model bundle	Commercial
	Derek (Lhasa)	Irritation (of the skin) alerts	Commercial
	Discovery Studio	Skin irritancy	Commercial
	(Accelrys)		
Eye irritation	OECD QSAR Toolbox	Eye irritation/corrosion Inclusion	Free
		(and Exclusion) rules by BfR	_
	ToxTree (JRC)	Eye irritation and corrosion	Free
	ACD/Percepta	Irritation Module	Commercial
	(ACD/Labs) CASE Ultra (MultiCASE)	SkipEvo Tovicity model bundle	Commercial
	Derek (Lhasa)	SkinEye Toxicity model bundle Irritation (of the eye) alerts	Commercial
	Discovery Studio	Ocular irritancy	Commercial
	(Accelrys)		commercial
Skin	Danish QSAR Database	Allergic Contact Dermatitis model	Free
sensitisation	(DTU)		
	OECD QSAR Toolbox	Protein binding alerts for skin	Free
		sensitisation by OASIS	
	ToxTree (JRC)	Skin sensitisation reactivity	Free
		domains	
	VEGA (IRFMN)	CAESAR model	Free
	ACD/Percepta	Irritation Module	Commercial
	(ACD/Labs)		O
	CASE Ultra (MultiCASE)	SkinEye Toxicity model bundle Skin sensitisation	Commercial
	Derek (Lhasa)	Skin sensitization	Commercial
	Discovery Studio (Accelrys)		Commercial
	TIMES (LMC)	Skin sensitization with	Commercial
		autoxidation	
Repeated dose	ADMET Predictor	Toxicity module	Commercial
toxicity	(Simulations Plus)		
	CASE Ultra (MultiCASE)	Several model bundles associated	Commercial
	, , ,	with repeated dose toxicity	
	Derek (Lhasa)	Several endpoints associated	Commercial
		with repeated dose toxicity	

Endpoint	Software tool	Models/Modules	Free or
			Commercial
	Discovery Studio (Accelrys)	Rat Chronic (Oral) LOAEL	Commercial
	Leadscope	Several models associated with repeated dose toxicity	Commercial
<i>In vitro</i> gene mutation in	Danish QSAR Database (DTU)	Models for Ames test	Free
bacteria (Ames test)	OECD QSAR Toolbox	Several profilers (alerts) associated with this endpoint	Free
,	T.E.S.T. (US EPA)	Mutagenicity	Free
	ToxTree (JRC)	<i>In vitro</i> mutagenicity (Ames test) alerts by ISS	Free
	VEGA (IRFMN)	CAESAR, SarPy/IRFMN, ISS and KNN/Read-Across models	Free
	ACD/Percepta (ACD/Labs)	Genotoxicity Module	Commercial
	CASE Ultra (MultiCASE)	Bacterial mutagenicity model bundle	Commercial
	Derek and Sarah (Lhasa)	Mutagenicity in vitro	Commercial
	Discovery Studio (Accelrys)	Ames Mutagenicity	Commercial
	Leadscope	Genetox Expert Alerts Suite and Non-human Genetic Toxicity Suite	Commercial
	TIMES (LMC)	Ames mutagenicity	Commercial
Mutagenicity (other	Danish QSAR Database (DTU)	Models for genotoxicity endpoints	Free
endpoints than in vitro gene	OECD QSAR Toolbox	Several profilers (alerts) associated with mutagenicity	Free
mutation in bacteria)	ToxTree (JRC)	Several decision trees associated with mutagenicity	Free
	CASE Ultra (MultiCASE)	EcoTox model bundle	Commercial
	Derek (Lhasa)	Chromosome damage in vitro	Commercial
	Leadscope	Non-human Genetic Toxicity Suite	Commercial
	TIMES (LMC)	Several models associated with mutagenicity	Commercial
Reproductive toxicity	Danish QSAR Database (DTU)	Models for Endocrine endpoints and model for Teratogenic Potential in Humans	Free
	VEGA (IRFMN)	CAESAR and PG models	Free
	ADMET Predictor (Simulations Plus)	Toxicity module	Commercial
	CASE Ultra (MultiCASE)	Several model bundles associated with reproductive and developmental toxicity	Commercial
	Derek (Lhasa)	Several endpoints associated with reproductive toxicity	Commercial
	Discovery Studio (Accelrys)	Developmental Toxicity Potential	Commercial
	Leadscope	Several models associated with reproductive and developmental toxicity	Commercial
	TIMES (LMC)	Androgen, AHR and Estrogen (receptor) binding affinity models	Commercial

e. Information on the Danish (Q)SAR Database

A new version of the Danish (Q)SAR database has been released in November 2015 and is publicly available at the following link: <u>http://qsar.food.dtu.dk/</u>.

This database contains (Q)SAR predictions for physicochemical properties, ecotoxicity, environmental fate, ADME and toxicity of more than 600 000 chemical structures.

When possible, models from the Technical University of Denmark and some commercial models have been modelled in the three software systems Leadscope, CASE Ultra (MultiCASE) and SciQSAR. Some model predictions from ACD/Labs and the US EPA (EPI Suite and ECOSAR) have also been integrated in the database.

However, it should be noted that the database does not provide the possibility to refine the predictions as some of the source software programs do (e.g. ECOSAR). In addition, the database does not provide as much details on the results as the source software programs and it is not updated on a regular basis. Therefore, whenever possible, the predictions given by the database should be compared to the results obtained from the source software programs themselves.

A REACH registrant who would like to report in its IUCLID registration dossier a prediction coming from the Danish (Q)SAR database should also check that the (Q)SAR model is valid – by comparing the points given in Section 3.1 of this practical guide with the information given in the model's QMRF – and should attach a QPRF (for each prediction) to the IUCLID endpoint study record.

Appendix 2. Further guidance documents and links

a. Guidance documents providing information about (Q)SARs

Various chapters of the *Guidance on information requirements and chemical safety assessment* provide useful information:

Chapter R.4: Evaluation of available information, with a summary on how to use non-testing data obtained by applying (Q)SARs (chapter R.4.3.2.1 (Q)SAR data)

Chapter R.6: (Q)SARs and grouping of chemicals contains a dedicated part on computational methodologies (chapter R.6.1 Guidance on (Q)SARs)

Chapter R.7: Endpoint specific Guidance offers relevant tools and approaches for the endpoints of interest, and specific for each endpoint.

Chapter R.8: Characterisation of dose [concentration]-response for human health: identifies for human health, the available (Q)SARs which may be suitable mostly for hazard identification, in particular in a weight of evidence approach.

Chapter R.10: Characterisation of dose [concentration]-response for environment presents several (Q)SAR tools which may be used to determine the Predicted No-Effect Concentrations (PNECs) (chapter R.10.2.2.2 (Q)SAR and grouping approaches).

Chapter R.11: PBT Assessment refers to information on the use of non-testing degradation and bioaccumulation data for persistent, bioaccumulative and toxic (PBT) chemicals.

b. Other useful links

OECD (Q)SAR Toolbox: <u>http://www.qsartoolbox.org/</u>

OECD Global Portal (eChemPortal):

http://www.echemportal.org/echemportal/index?pageID=0&request_locale=en

Manual 'How to prepare registration and PPORD dossiers': <u>http://echa.europa.eu/manuals</u>

'Practical guide on how to use alternatives to animal testing to fulfil your information requirements for REACH registration': <u>http://echa.europa.eu/practical-guides</u>

'Practical guide for SME managers and REACH coordinators on how How to fulfil your information requirements at tonnages 1-10 tpa and 10-100 tpa': <u>http://echa.europa.eu/practical-guides</u>

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