
	QMRF identifier (JRC Inventory): To be entered by JRC	
	QMRF Title: Insubria QSPR PaDEL-Descriptor model for Vapor Pressure prediction of PFC	
	Printing Date: Jan 20, 2014	

1. QSAR identifier

1.1. QSAR identifier (title):

Insubria QSPR PaDEL-Descriptor model for Vapor Pressure prediction of PFC

1.2. Other related models:

Bhhatrai B., Gramatica P., 2011, Prediction of Aqueous Solubility, Vapor Pressure and Critical Micelle Concentration for Aquatic Partitioning of Perfluorinated Chemicals, Environ. Sci. Technol., 2011, 45, 8120–8128 [8]

1.3. Software coding the model:

[1] PaDEL-Descriptor 2.18 A software to calculate molecular descriptors and fingerprints <http://padel.nus.edu.sg/software/padeldescriptor/index.html>

[2] QSARINS 1.2 Software for the development, analysis and validation of QSAR MLR models paola.gramatica@uninsubria.it www.qsar.it

2. General information

2.1. Date of QMRF:

05/12/2013

2.2. QMRF author(s) and contact details:

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2.3. Date of QMRF update(s):

2.4. QMRF update(s):

2.5. Model developer(s) and contact details:

[1] Paola Gramatica DiSTA, University of Insubria (Varese - Italy)
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[2] Stefano Cassani DiSTA, University of Insubria (Varese - Italy)
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2.6. Date of model development and/or publication:

July 2013

2.7. Reference(s) to main scientific papers and/or software package:

[1] Gramatica P., et al. QSARINS: A new software for the development, analysis and validation of QSAR MLR models, J. Comput. Chem. (Software News and Updates), 2013, 34 (24), 2121-2132 [1]

[2] QSARINS-Chem: Insubria Datasets and New QSAR/QSPR Models for Environmental Pollutants in QSARINS, submitted to J. Comput. Chem. (Software News and Updates)

2.8. Availability of information about the model:

The model is non-proprietary and published in a scientific peer-reviewed journal. All information in full details are available (e.g. training and prediction set, algorithm, ecc...).

2.9.Availability of another QMRF for exactly the same model:

No

3.Defining the endpoint - OECD Principle 1

3.1.Species:

No information available

3.2.Endpoint:

1.Physicochemical effects 1.4.Vapour pressure

3.3.Comment on endpoint:

Vapor Pressure (VP) is the pressure exerted by a vapor in equilibrium with the solid or liquid phase of the same substance.

3.4.Endpoint units:

mmHg

3.5.Dependent variable:

LogVP

3.6.Experimental protocol:

24 compounds from SRC PhysProp database[2] reported at 25 °C and 11 additional compounds from EU-FP6 PERFORCE report[3] at 20° or 25 °C for liquid or subcooled liquid were added (total 35 compounds). The data given in 20 °C temperature were extrapolated for 25 °C using the Wagner and Antoine equation

3.7.Endpoint data quality and variability:

No information available

4.Defining the algorithm - OECD Principle 2

4.1.Type of model:

QSAR - Multiple linear Regression Model (OLS - Ordinary least-squares)

4.2.Explicit algorithm:

LogVP (Full model)

OLS-MLR method. Model developed on a training set of 35 compounds

LogVP (Split by SOM modell)

OLS-MLR method. Model developed on a training set of 24 compounds

LogVP (Split by Ordered Response model)

OLS-MLR method. Model developed on a training set of 22 compounds

Full model equation: $\log VP = 4.47 - 0.46 \text{ nH} - 0.20 \text{ nsF} - 2.41 \text{ nHBint2}$

Split by SOM model equation: $\log VP = 4.47 - 2.71 \text{ nHBint2} - 0.19 \text{ nsF} - 0.43 \text{ nH}$

Split by Ordered Response model equation: $\log VP = 4.64 - 0.24 \text{ nsF} - 0.45 \text{ nH} - 2.07 \text{ nHBint2}$

4.3.Descriptors in the model:

[1]nH Number of hydrogen atoms

[2]nSF Count of atom-type E-State: -F

[3]nHBint2 Count of E-State descriptors of strength for potential Hydrogen Bonds of path length 2

4.4.Descriptor selection:

A total of 717 molecular descriptors of differing types (0D, 1D, 2D) were calculated in PaDEL-Descriptor 2.18. Constant and semi-constant values and descriptors found to be correlated pairwise were excluded in a pre-reduction step (one of any two descriptors with a correlation greater than 0.98 was removed to reduce redundant information), and a final set of 123 molecular descriptors were used as input variables for variable subset selection. The models were initially developed by the all-subset-procedure, and then GA was applied to obtain the final population of models (three variables). The optimized parameter used was Q2LOO (leave-one-out).

4.5.Algorithm and descriptor generation:

Multiple linear regression (Ordinary Least Square method) was applied to generate the model.

Molecular descriptors were generated by PaDEL-Descriptor software. The input files for descriptor calculation contain information on atom and bond types, connectivity, partial charges and atomic spatial coordinates, relative to the minimum energy conformation of the molecule, and were firstly obtained by the semi empirical AM1 method using the package HYPERCHEM. Then, these files were converted by OpenBabel into MDL-MOL format and used as input for the calculation of descriptors in PaDEL-Descriptor.

4.6.Software name and version for descriptor generation:

PaDEL-Descriptor 2.18

A software to calculate molecular descriptors and fingerprints

Yap Chun Wei, Department of Pharmacy, National University of Singapore.

<http://padel.nus.edu.sg/software/padeldescriptor/index.html>

HYPERCHEM - ver. 7.03

Software for molecular drawing and conformational energy optimization

OpenBabel ver.2.3.2

Open Babel: The Open Source Chemistry Toolbox. Used for conversion between HYPERCHEM files (hin) and MDL-MOL files.

<http://openbabel.org>

4.7.Chemicals/Descriptors ratio:

Full model: 35 chemicals / 3 descriptors = 11.67

Split by SOM: 24 chemicals / 3 descriptors = 8

Split by Ordered response: 22 chemicals / 3 descriptors = 7.34

5. Defining the applicability domain - OECD Principle 3

5.1. Description of the applicability domain of the model:

The applicability domain of the model was verified by the leverage approach and fixed thresholds has been used to define both structural and response outliers (see section 5.4). The plot of leverages (hat diagonals) versus standardised residuals, i.e. the Williams plot, verified the presence of response outliers (i.e. compounds with cross-validated standardized residuals greater than 2.5 standard deviation units) and chemicals very structurally influential in determining model parameters (i.e. compounds with a leverage value (h) greater than $3p'/n$ (h^*), where p' is the number of model variables plus one, and n is the number of the objects used to calculate the model). For new compounds without experimental data, leverage can be used as a quantitative measure for evaluating the degree of extrapolation: for compounds with a high leverage value ($h > h^*$), that are structural outliers, predictions should be considered less reliable.

Response and descriptor space:

Range of experimental: logVP values: -6.37 / 4.39

Range of descriptor values: nH: 0 / 15; nHBint2: 0 / 2; nsF: 3 / 27

5.2. Method used to assess the applicability domain:

As it has been stated in section 5.1, the structural applicability domain of the model was assessed by the leverage approach, providing a cut-off hat value ($h^*=0.343$). HAT values are calculated as the diagonal elements of the HAT matrix:

$$H = X(X^T X)^{-1} X^T$$

The response applicability domain can be verified by the standardized residuals, calculated as: $r_i' = r_i / \sqrt{1-h_{ii}}$, where $r_i = Y_i - \hat{Y}_i$.

5.3. Software name and version for applicability domain assessment:

QSARINS 1.2

Software for the development, analysis and validation of QSAR MLR models

paola.gramatica@uninsubria.it

www.qsar.it

5.4. Limits of applicability:

Full model domain: outliers for structure, $hat > 0.343$ (h^*): 2-{ethyl[(heptadecafluorooctyl)sulfonyl]amino}ethyl acrylate (423-82-5), Sulfluramid [ISO] (4151-50-2); Outliers for response, standardised residuals > 2.5 standard deviation units: no

Split by SOM model domain: outliers for structure, $hat > 0.500$ (h^*): 2-{ethyl[(heptadecafluorooctyl)sulfonyl]amino}ethyl acrylate (423-82-5), Sulfluramid [ISO] (4151-50-2), N-ethyl-

6. Internal validation - OECD Principle 4

6.1.Availability of the training set:

Yes

6.2.Available information for the training set:

CAS RN:Yes

Chemical Name:Yes

Smiles:Yes

Formula:Yes

INChI:No

MOL file:Yes

6.3.Data for each descriptor variable for the training set:

All

6.4.Data for the dependent variable for the training set:

All

6.5.Other information about the training set:

The training set of the **Split by SOM Model** consists of 24 perfluorinated compounds with a range of logVP values from -6.37 to 4.39.

The training set of the **Split by Ordered Response Model** consists of 22 perfluorinated compounds with a range of logVP values from -4.82 to 3.82.

6.6.Pre-processing of data before modelling:

The data was used as LogVP mmHg; The data given in 20 °C temperature were extrapolated for 25 °C using the Wagner and Antoine equation

6.7.Statistics for goodness-of-fit:

Split by SOM Model:

R^2 :0.95 ; CCContr[4]:0.97 ; RMSEtr: 0.65

Split by Ordered Response Model:

R^2 : 0.92 ; CCContr: 0.96 ; RMSEtr:0.71

6.8.Robustness - Statistics obtained by leave-one-out cross-validation:

Split by SOM Model:

Q^2_{loo} :0.92 ; CCCcv: 0.96 ; RMSEcv: 0.81

Split by Ordered Response Model:

Q^2_{loo} : 0.89 ; CCCcv: 0.94 ; RMSEcv: 0.83

6.9.Robustness - Statistics obtained by leave-many-out cross-validation:

Split by SOM Model: Q^2_{LMO} : 0.90

Split by Ordered Response Model: Q^2_{LMO} : 0.86

6.10.Robustness - Statistics obtained by Y-scrambling:

Split by SOM Model: R^2_{Yscr} :0.13

Split by Ordered Response Model: R^2_{Yscr} : 0.14

6.11.Robustness - Statistics obtained by bootstrap:

No information available (since we have calculated Q^2_{LMO})

6.12.Robustness - Statistics obtained by other methods:

No information available

7.External validation - OECD Principle 4

7.1.Availability of the external validation set:

Yes

7.2.Available information for the external validation set:

CAS RN:Yes

Chemical Name:Yes

Smiles:Yes

Formula:Yes

INChI:No

MOL file:Yes

7.3.Data for each descriptor variable for the external validation set:

All

7.4.Data for the dependent variable for the external validation set:

All

7.5.Other information about the external validation set:

To verify the predictive capability of the proposed models, the dataset (n=35) was split, before model development, into a training set used for model development and a prediction set used later for external validation. Two different splitting techniques were applied: **by Ordered Response** (n external validation set =13) and **by structural similarity (SOM)** (n external validation set =11).

7.6.Experimental design of test set:

In the case of split **by Ordered Response model**, chemicals were ordered according to their increasing activity, and one out of every three chemicals was put in the prediction set. The splitting **by SOM model** takes advantages of the clustering capabilities of Kohonen Artificial Neural Network (K-ANN), allowing the selection of a structurally meaningful training set and an equally representative prediction set.

7.7.Predictivity - Statistics obtained by external validation:

Split by SOM model: n prediction= 11 ; $R^2_{ext} = 0.98$; $Q^2_{ext} F1[5] = 0.88$; $Q^2_{ext} F2[6] = 0.88$; $Q^2_{ext} F3[7] = 0.89$; CCCex = 0.94; RMSEex = 0.94 ; MAEex = 0.81 .

Split by Ordered Response model: n prediction= 13 ; $R^2_{ext} = 0.94$; $Q^2_{ext} F1 = 0.92$; $Q^2_{ext} F2 = 0.92$; $Q^2_{ext} F3 = 0.89$; CCCex = 0.96; RMSEex = 0.86; MAEex = 0.71.

7.8.Predictivity - Assessment of the external validation set:

Range of response for prediction set (**SOM split**, n=11) compounds:

logVP (mmHg): -4.82 / 3.82 (range of corresponding training set: -6.37 / 4.39)

Range of modeling descriptors for prediction set (**SOM split**, n=11) compounds:

nH: 0 / 12 (range of corresponding training set: 0 / 10)

nHBint2: 0 / 1 (range of corresponding training set: 0 / 2)

nsF: 3 / 21 (range of corresponding training set: 4 / 27)

Range of response for prediction set (**Ordered Response split**, n=13) compounds:

logVP: -6.37 / 4.39 (range of corresponding training set: -4.82 / 3.82)

Range of modeling descriptors for prediction set (**Ordered Response split**, n=13) compounds: nH: 0 / 10 (range of corresponding training set: 0 / 12)

nHBint2: 0 / 2 (range of corresponding training set: 0 / 1) nsF: 3 / 27 (range of corresponding training set: 3 / 23) The distribution of response values of the chemicals in the two different training sets is comparable to the distribution of the response values of the two prediction set.

7.9. Comments on the external validation of the model:

no other information available

8. Providing a mechanistic interpretation - OECD Principle 5

8.1. Mechanistic basis of the model:

The model was developed by statistical approach. No mechanistic basis was defined a priori.

8.2. A priori or a posteriori mechanistic interpretation:

The DRAGON model published in Bhatarai B. and Gramatica P [8] is:

$$\log VP = 7.97 - 0.16 F03[C-F] - 3.16 AAC - 0.64 nDB$$

where F03[C-F]: 2D frequency fingerprint descriptor, meaning the frequency of C-F at topological distance 03 (these values are higher for branched and cyclic compounds)

AAC: 2D information indices, particularly mean information index on atomic composition (increases with higher atomic weight atoms or larger molecule)

nDB: 0D constitutional descriptor is the number of double bonds

All these descriptors being inversely related to VP have an influence in decreasing the vapor pressure.

The equation of the new PaDEL-descriptor model included in QSARINS is:

$$\log VP = 4.47 - 0.46 nH - 0.20 nSF - 2.41 nHBint2$$

where nH= Number of hydrogen atoms

nSF= Count of atom-type E-State: -F nHBint2= Count of E-State descriptors of strength for potential Hydrogen Bonds of path length 2

In the two models there is a high correlation between F03[C-F] and nSF (0.96), which encode for the same structural information related to fluorine atoms, and acceptable correlation also among AAC and nH (0.73).

8.3. Other information about the mechanistic interpretation:

no other information available

9. Miscellaneous information

9.1. Comments:

To predict VP for new PFC chemicals without experimental data, it is suggested to apply the equation of the Full Model, developed on all the available chemicals (N=35), thus ensuring a wider applicability domain.

The equation (reported also in section 4.2) and the statistical parameters of the full model are:

Full model equation: $\log VP = 4.47 - 0.46 \text{ nH} - 0.20 \text{ nSF} - 2.41 \text{ nHBint2}$

$N = 35$; $R^2 = 0.93$; $Q^2 = 0.91$; $Q^2_{LMO} = 0.90$; $CCC = 0.96$; $CCC_{cv} = 0.95$; $RMSE = 0.72$; $RMSE_{cv} = 0.82$.

9.2. Bibliography:

[1] Gramatica P., et al. QSARINS: A new software for the development, analysis and validation of QSAR MLR models, J. Comput. Chem. (Software News and Updates), 2013, 34 (24), 2121-2132.

[2] SRC PhysProp database. <http://www.syrres.com>

[3] Krop, H.; de Voogt, P. EU-FP6 PERFORCE (PERFluorinated ORganic Chemicals in the European environment) 2, IBED-ESPM, 2008

[4] Chirico N. and Gramatica P., Real External Predictivity of QSAR Models. Part 2. New Intercomparable Thresholds for Different Validation Criteria and the Need for Scatter Plot Inspection, J. Chem. Inf. Model. 2012, 52, pp 2044– 2058

[5] Shi L.M. et al. QSAR Models Using a Large Diverse Set of Estrogens, J. Chem. Inf. Comput. Sci. 41 (2001) 186–195.

[6] Schuurman G. et al. External Validation and Prediction Employing the Predictive Squared Correlation Coefficient - Test Set Activity Mean vs Training Set Activity Mean, J. Chem. Inf. Model. 48 (2008) 2140-2145.

[7] Consonni V. et al. Comments on the Definition of the Q2 Parameter for QSAR Validation, J. Chem. Inf. Model. 49 (2009) 1669-1678

[8] Bhattacharai B., Gramatica P., Prediction of Aqueous Solubility, Vapor Pressure and Critical Micelle Concentration for Aquatic Partitioning of Perfluorinated Chemicals, Environ. Sci. Technol., 45, 8120–8128

9.3. Supporting information:

Training set(s) Test set(s) Supporting information

10. Summary (JRC Inventory)

10.1. QMRF number:

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10.2.Publication date:

To be entered by JRC

10.3.Keywords:

To be entered by JRC

10.4.Comments:

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