

Oral LD₅₀ Toxicity Modeling of Per- and Polyfluorinated Chemicals on Rat and Mouse



Barun Bhatarai, Paola Gramatica
 Department of Structural and Functional Biology, University of Insubria, QSAR Research Unit in Environmental Chemistry and Ecotoxicology,
 Via Dunant 3, 21100 Varese, paola.gramatica@uninsubria.it



INTRODUCTION

Quantitative structure-activity relationship (QSAR) and chemometric methods were applied to **Perfluorinated Chemicals (PFCs)** – fluorinated carbon chain (C₄ to C₁₆) containing linear or cyclic chemicals, which are considered as ‘emerging pollutants’. They are found widely distributed in the environment, released due to their widespread use in different household and industrial products as cleansers, fire-fighting foams, micelles, oil and water repellants for leather, paper, and textiles etc. Continuous exposure of these chemicals is found to be the source of bio-accumulation in body parts of human, wildlife and is ultimately becoming the cause of toxic reactions. However, there are more than 650 PFCs, linear and cyclic, that are found in ECHA (European Chemical Agency) pre-registration list of compounds and these chemicals need to be identified, if, they belong to Substances of Very High Concern (SVHC). Experimental data for majority of these compounds are unavailable or are proprietary and a need to use the existing available data to predict the activity of these compounds is necessary. Thus, a dataset of **oral lethal dose 50% (LD₅₀)** was compiled for short and long chain PFCs on two species of rodents – **Rat (Rattus)** and **Mouse (Mus)**. The oral exposure analysis was chosen as it is an important indicator of food and accidental domestic poisonings, and/or occupational poisonings. QSAR was then applied to model the available data and predict the oral LD₅₀ toxicity of other chemicals including those listed in ECHA for which toxicity data is not available. The set of descriptors which best describes the structure-toxicity relationship, the similarities, and the differences observed related to two species are discussed. Principal Component Analysis (PCA) was used to select most toxic compounds from those within the structural applicability domain (AD) of both the models. QSAR study on LC₅₀ inhalation data of PFCs on rodents has been published earlier and combining with the result of current LD₅₀ oral study, a comparative toxicity analysis of two different end-points on rodents and consensus prediction and prioritization of hazardous PFCs is performed. The **prioritized chemicals** will be further subjected to experimental test under the EU-FP7 funded CADASTER project.

MATERIALS AND METHODS

Data Set: 58 Mouse and 50 Rat LD₅₀ oral data were used. Training and prediction sets were prepared *a priori* from available experimental datasets in terms of structure (SOM) and random by response approach and these sets were used to derive statistically robust and predictive (both internally and externally) models. 26% to 37% splitting were used. Structural applicability domain (AD) of the models were verified on 376 per- and polyfluorinated chemicals including those in REACH pre-registration list
Molecular Descriptors: More than 600 molecular descriptors (0D–3D) were calculated by the software DRAGON [1] from the XYZ coordinates in Hyperchem using AM1 [2].

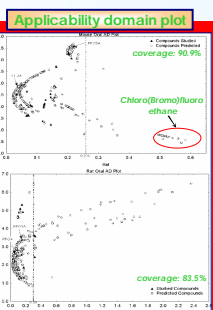
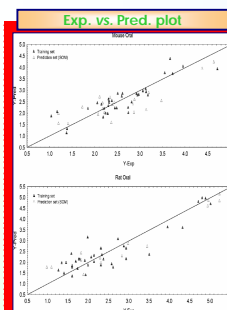
Multiple Linear Regression (MLR) and **Genetic Algorithm-Variable Selection (GA-VSS)** were performed by the software MOBY DIGS [3] using the Ordinary Least Square regression (OLS) method.

Validation: The robustness of the models and their internal predictive ability were evaluated by both Q² based on leave-one-out cross-validation and bootstrap. The proposed models were also checked for reliability and robustness by permutation testing [4]: new models were recalculated for randomly reordered response (R_{Y-scrambling}). The external validation was performed by developing the model on the training set and then using those models to predict the test set [5].

RESULTS AND DISCUSSION

$$\text{Mus log } 1/\text{LD}_{50} = 4.543 - 2.450 (\pm 0.312) \text{HATS2u} + 1.362 (\pm 0.203) \text{B09[C-O]} - 0.142 (\pm 0.032) \text{F01[C-O]} - 0.486 (\pm 0.174) \text{B04[C-F]}$$

n=58, r²=75.93, q²=71.89, s=0.41, F=41.8
 Equation 1



$$\text{Rat log } 1/\text{LD}_{50} = -2.277 + 0.041 (\pm 0.003) \text{D/Dr09} + 2.943 (\pm 0.580) \text{MATS1e} + 8.838 (\pm 1.712) \text{E1u} + 1.166 (\pm 0.211) \text{H8m}$$

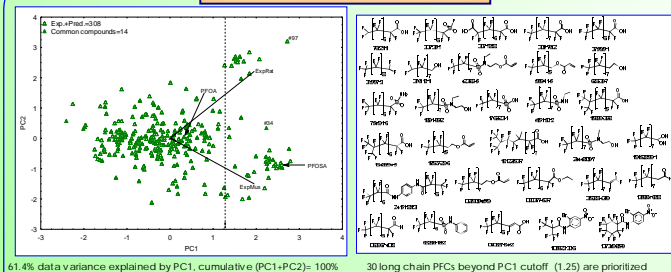
n=50, r²=88.28, q²=85.50, s=0.44, F=84.78
 Equation 2

Response (Compounds)	Descriptors Input	Selected	Splitting criteria	Compounds	R ²	Q ² _{LOO}	Q ² _{BOO}	Q ² _{ext}	Q ² _{int}	RMSE _{ext}	RMSE _{int}	R _Y
Mouse Oral LD ₅₀ (58)	690	HATS2u; B09[C-O]; F01[C-O]; B04[C-F]	SOM 37.9% Random by Activity 32.7% Full model	Train: 36, Predict: 22 Train: 39, Predict: 19 75.9	82.8	75.7	74.2	65.6	55.6	0.32	0.51	11.5
Rat Oral LD ₅₀ (50)	635	D/Dr09; MATS1e; E1u; H8m	Random by Activity 26.0% Full model	Train: 36, Predict: 14 Train: 37, Predict: 13 88.3	85.5	80.3	70.9	91.1	81.4	0.41	0.46	12.2
					90.7	87.5	85.6	80.7	75.1	0.36	0.59	11.3
					88.3	85.5	82.2	-	-	0.42	0.47	8.0

Mouse Oral descriptors:
 HATS2u (-0.589), 3D GETAWAY
 → leverage-weighted autocorrelation of lag 2/unweighted.
 B09[C-O] (0.478), 2D binary finger-print
 → the presence of [C-O] at n=9
 F01[C-O] (-0.303), 2D frequency finger-print
 → the frequency of [C-O] at n=1
 B04[C-F] (-0.198), 2D binary finger-print
 → the presence of atom pair [C-F] at n=4

Rat Oral descriptors:
 D/Dr09 (0.765), 2D topological
 → distance/debouring index of order 9
 MATS1e (0.331), 2D Moran autocorrelation
 → electronegativity
 E1u (0.316), 3D WHIM
 → 1st component accessibility directional index/unweighted
 H8m (0.294), 3D GETAWAY
 → H autocorrelation of lag 8 (weighted by atomic mass)

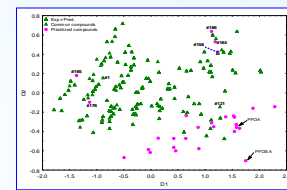
Toxicity Trend – LD₅₀ Oral



Overview – LC₅₀ Inhalation [6]

Chem. Res. Toxicol. 2010, 23, 528–539
Per- and Polyfluoro Toxicity (LC₅₀ Inhalation) Study in Rat and Mouse Using QSAR Modeling
 Barun Bhatarai and Paola Gramatica*
 QSAR Research Unit in Environmental Chemistry and Ecotoxicology, Department of Structural and Functional Biology (DSFB), University of Insubria, via Oltramerio 3, Varese 21100, Italy
 Received July 24, 2009
 Fully or partially fluorinated compounds, known as per- and polyfluorinated chemicals are widely distributed in the environment and released because of their use in different household and industrial products. Few of these long chain per- and polyfluorinated chemicals are classified as emerging pollutants.

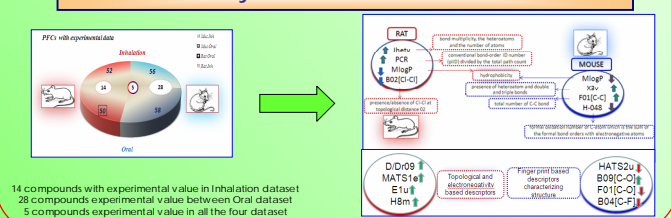
Endpoint	Descriptors	N _{test}	R ²	Q ² _{LOO}	Q ² _{ext}	RMSE _{ext}	AD% _{100 PFCs}
Mouse Inhalation	X3v, H-048, MlogP, F01[C-O]	56	79.8	76.31	71.62-85.11	0.74	75.6%
Rat Inhalation	Jhetv, PCR, MlogP, B02[C-F]	52	78.1	73.85	66.70-75.47	0.86	76.8%



MDS plot of 7 molecular descriptors highlighting the structural diversity of prioritized compounds among 180 PFCs within the AD Rodents' LC₅₀ models

- Two QSAR models each on Mouse and Rat LC₅₀ inhalation data were published [6]
- Prioritization study on LC₅₀ data was performed → 28 long chain PFCs predicted

Combined toxicity trend: LC₅₀ Inhalation and LD₅₀ Oral



CONCLUSIONS

- Toxicity of PFCs on rodents was studied by developing reliable, robust and predictive QSAR models
- Models for Rodents Oral LD₅₀ shows combination of electronic (MATS1e, HATS2u) and fingerprint based descriptors (F01[C-O], B04[C-O])
- Models on Rodents Oral and Inhalation data shows importance of following main descriptors for overall toxicity
 - negative hydrophobicity (MlogP),
 - positive electronegativity (Jhetv, X3v and MATS1e)
 - descriptors representing the position and the frequency of atom pairs like C-C, C-F and C-O that counts for the main functional groups of long chain PFCs
- Prioritized most toxic long chain PFCs will be suggested to CADASTER partners for the experimental design

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Financial support by European Union through the project CADASTER FP7-ENV-2007-1-212668