
Estimation of Melting Points of Brominated and Chlorinated Organic Pollutants using QSAR Techniques

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Persistent Organic Pollutants

- Do not undergo photolytic, biological, and chemical degradation.
 - low water solubility
 - high lipid solubility
 - semi-volatility
 - high molecular masses
- Found in pesticides
 - Transport by air and water
- Found at extremely low levels in 20% of US Food Supply
 - Continual exposure toxicity and effects are unknown

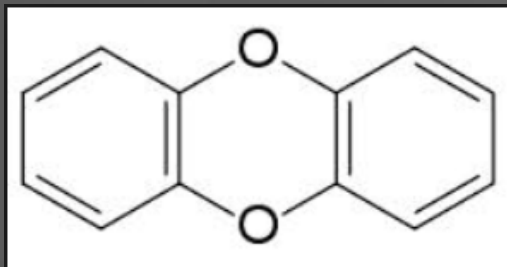
Previous Research

- Reserved to small sets (10 to 100 compounds) of hydrocarbons, substituted aromatics, aldehydes, amines, and ketones utilizing other descriptor types, regressions, and analysis techniques.
- Larger data sets were only of approximately 400 compounds.

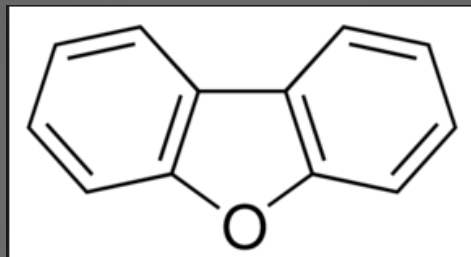
The Endeavor

- To estimate the melting points of 1436 chloro- and bromo-analogues of dibenzo-p-dioxins, dibenzofurans, biphenyls, naphthalenes, diphenyl ethers, and benzenes by utilizing quantitative structure—property relationship (QSAR) techniques.

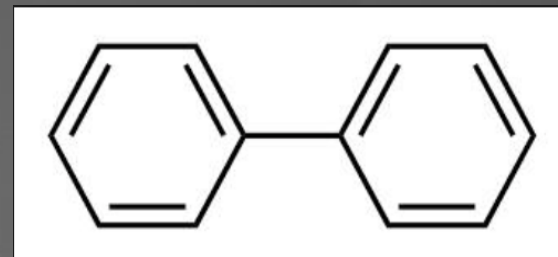
Chosen Structure Families



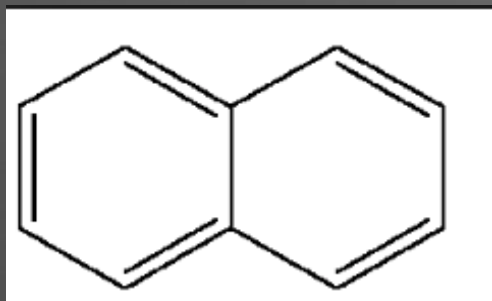
dibenzo-p-dioxins



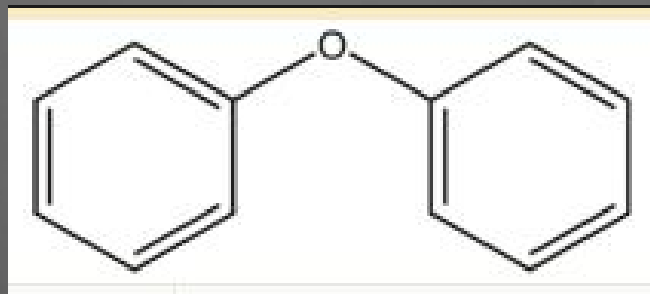
dibenzofurans



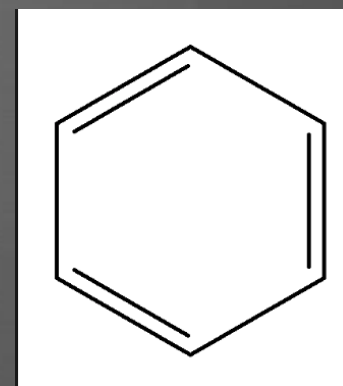
biphenyls



naphthalenes



diphenyl ethers



benzenes

Melting Point

- Specifies transition temperature
- Essential in biochemical and environmental studies
- Solubility estimation
- Difficult QSAR endpoint to predict

QSAR Significance

- Inexpensive
- Time-efficient
- Safe
- Enhances the efficiency of drug design

Basic concept of QSAR modeling

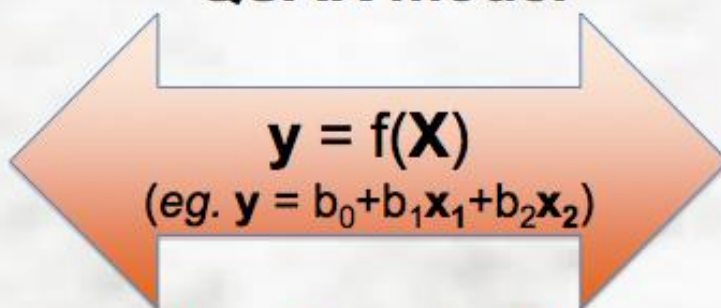
Endpoint
(experimentally measured)



y

- Activity (EC_{50})
- Phys/Chem property (K_{OW} , $t_{1/2}$)
- Retention parameters (t_R)
- Toxicity (LD_{50} , LC_{50})
- ...

QSAR model



- Linear Regression (LR)
- Multiple Linear Regression (MLR)
- Partial Least Squares (PLS)
- Artificial Neural Networks (ANN)
- ...

Structural descriptors



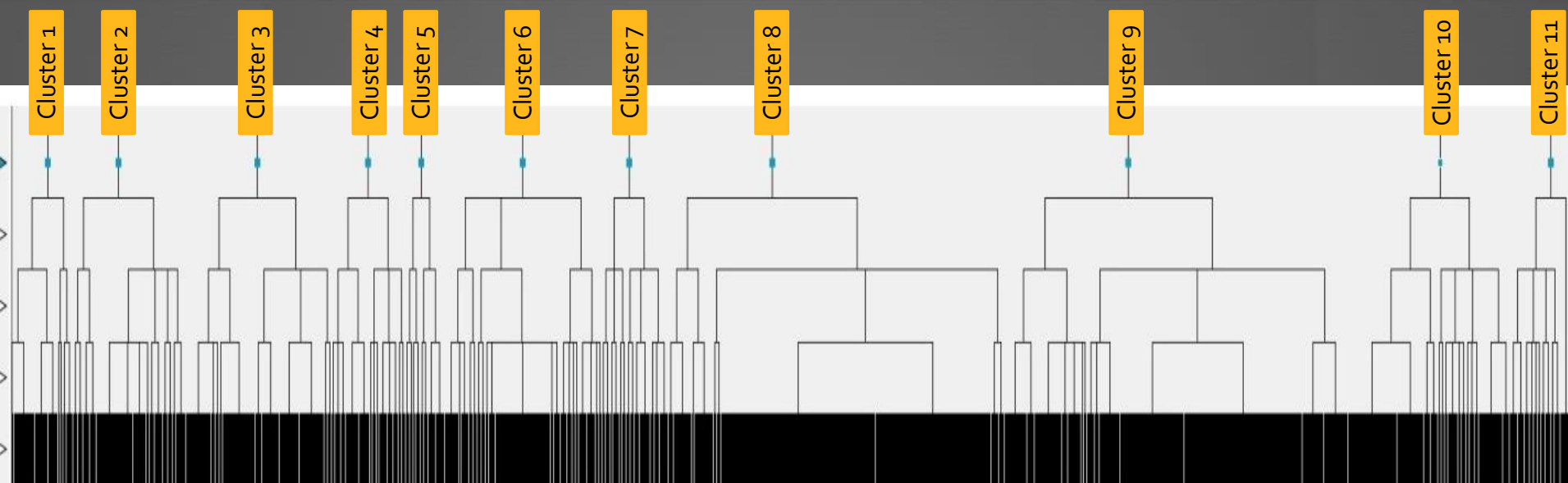
X

- Dipole moment
- Polarizability
- HOMO, LUMO
- Topological indexes
- Number of specific atoms/groups
- ...

Software Employed

- Structures were built and optimized using HyperChem 5.0
- Hierarchical clustering in ChemAxon's MCS Library Software.
- Descriptor Generation Software
 - DRAGON₅
- PLS Toolbox
 - Automated Descriptor Selection
 - Genetic Algorithm
 - Model Building

Hierarchical Cluster Results



Minimal MCS Size = 25

MCS Mode = Very Fast

Matching Parameters = Atom Type, Bond Type,
Charge

Clustering Options = Keep Rings

Required Cluster Count = 15

Maximum Level Count = 15

Cluster Analysis

Cluster #	Calibration Set	Validation Set	Prediction Set
1	R ² Cal: 0.89	R ² CV: 0.88	R ² Pred: 0.80
	Bias: 0	CV Bias: 0.004	Pred Bias: -1.07
	RMSEC: 8.04	RMSECV: 8.43	RMSEP: 13.24
3	R ² Cal: 0.76	R ² CV: 0.72	R ² Pred: 0.66
	Bias: -4.26e-14	CV Bias: -0.20	Pred Bias: 5.46
	RMSEC: 21.66	RMSECV: 23.64	RMSEP: 23.72
4	R ² Cal: 0.89	R ² CV: 0.86	R ² Pred: 0.91
	Bias: -2.84e-14	CV Bias: -0.13	Pred Bias: -5.29
	RMSEC: 17.32	RMSECV: 19.30	RMSEP: 15.71
7	R ² Cal: 0.94	R ² CV: 0.80	R ² Pred: 0.63
	Bias: -2.84e-14	CV Bias: -5.24	Pred Bias: -3.82
	RMSEC: 18.56	RMSECV: 34.79	RMSEP: 26.26

Cluster Analysis

Cluster #	Calibration Set	Validation Set	Prediction Set
8	R ² Cal: 0.75	R ² CV: 0.56	R ² Pred: 0.81
	Bias: 4.26e-14	CV Bias: -1.19035	Pred Bias: 13.38
	RMSEC: 21.12	RMSECV: 28.5243	RMSEP: 34.0649
9	R ² Cal: 0.84	R ² CV: 0.54	R ² Pred: 0.83
	Bias: 0	CV Bias: 2.41	Pred Bias: -13.43
	RMSEC: 35.51	RMSECV: 67.81	RMSEP: 44.05
10	R ² Cal: 0.85	R ² CV: 0.791392	R ² Pred: 0.709519
	Bias: 5.68e-14	CV Bias: 1.66	Pred Bias: 6.91
	RMSEC: 22.39	RMSECV: 27.13	RMSEP: 56.34
11	R ² Cal: 0.77	R ² CV: 0.73	R ² Pred: 0.80
	Bias: 0	CV Bias: -0.14	Pred Bias: -0.059
	RMSEC: 26.61	RMSECV: 29.19	RMSEP: 25.52

Cluster 11 Results/Discussion

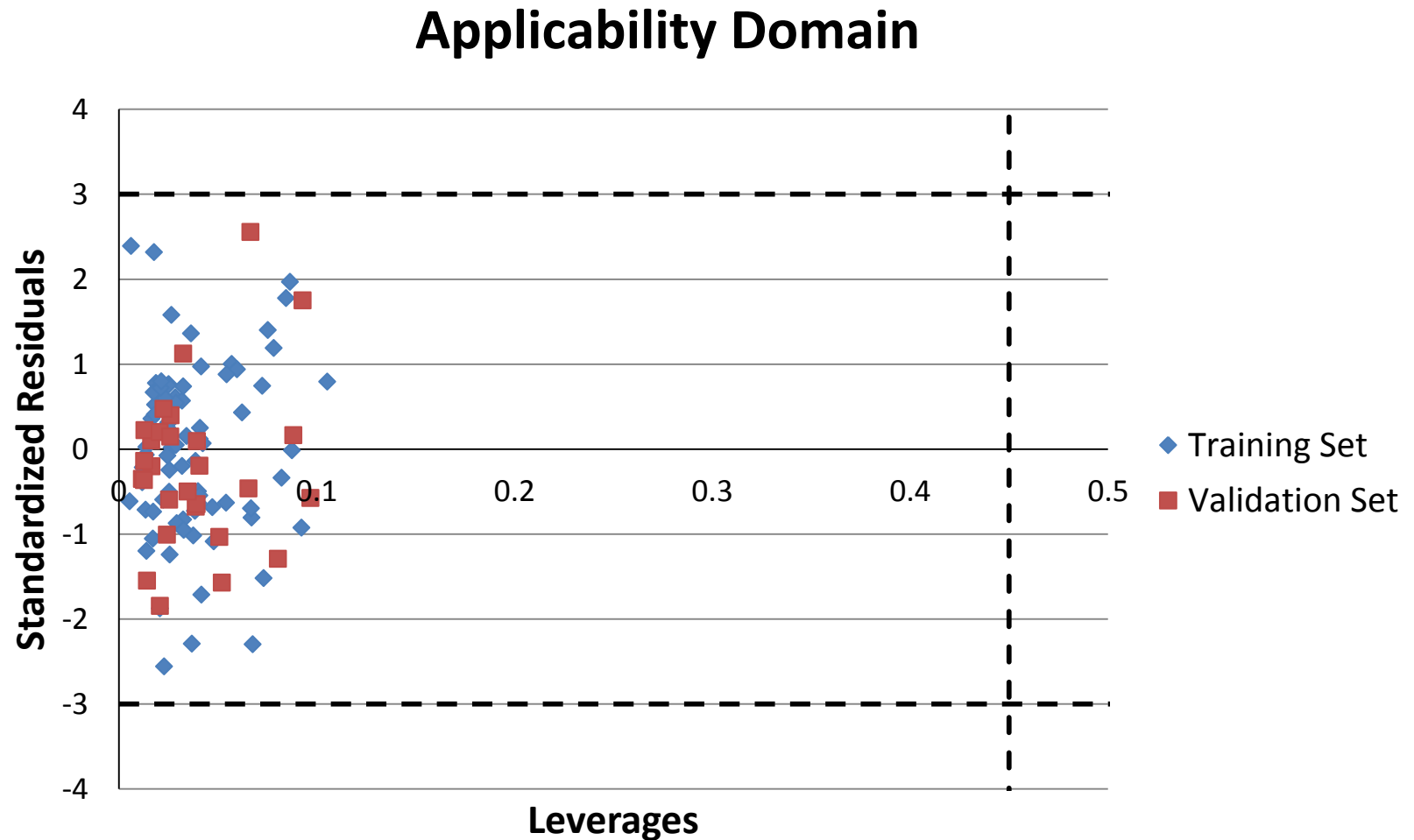
- Algorithm

- $MP = 0.213208(D/D) + 0.011536(Wap) + 101.442(J) + 2.9733(S1K) + 0.093732(TPC) + 0.000737(piID) + -0.238566(PCR) + 0.063346(PCD) + 1.99058(MATS8m) + 161.158(MATS1v) + 46.4475(MATS2v) + 67.106(MATS3v) + -194.013$

- Data Characteristics

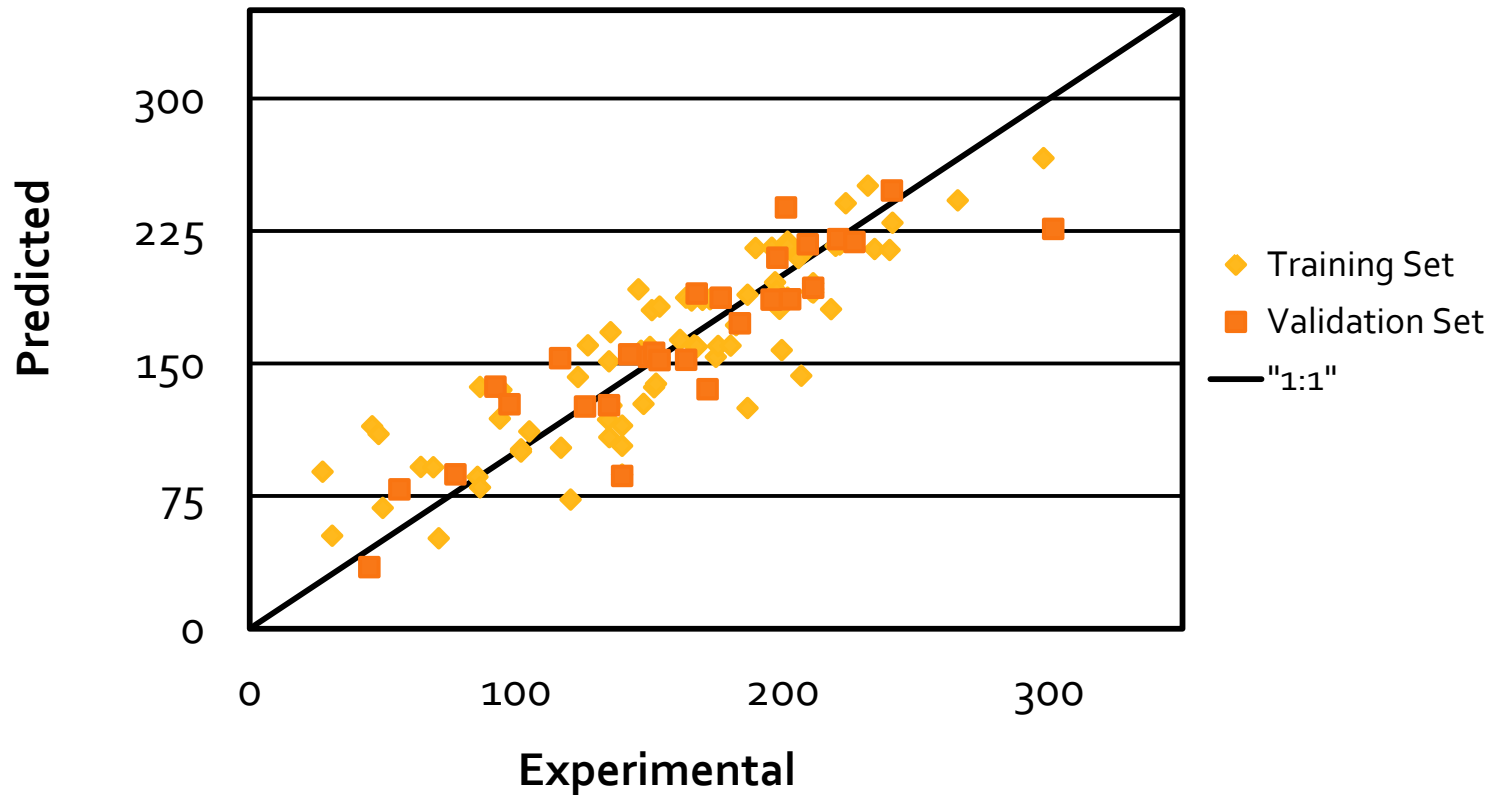
- $N_{trn} = 80$ $N_{val} = 28$ $N_{pred} = 386$

Cluster 11 Validation



Cluster 11 Validation

Melting Point Experimental vs Predicted



Missing Clusters

- Cluster 2: still being validated and analyzed
- Cluster 5: has been validated and results are currently being compiled
- Cluster 6: shown to have a poor correlation in multiple aspects
 - Possibly due to data set
 - Endpoint more complex than originally believed

Conclusions

- The melting points of the series of POPs have been modeled using a set of calculated descriptors
 - Suggests that existing cheminformatics descriptors fail to adequately describe interactions in the crystalline solid phase
 - May be a significant cause of error in melting point prediction
- Original data set contains too many similar compounds to produce a valid model
 - Hierarchical clustering method employed via MCS Library
 - Improved predictions significantly

Future Endeavors

- Explore other possible methodology and descriptors
 - Nonlinear methods
 - Other descriptors - for example, quantum-chemical descriptors
- Model other endpoints (biological activity, solubility, etc.) with these data sets

Acknowledgements

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Questions???
