COMP-238 Model Applicability Domains: When Can I Use my Model? 235<sup>th</sup> National Meeting of The American Chemical Society New Orleans, LA April 9, 2008

# Combining global and local approaches to model domain applicability

<u>Rajarshi Guha</u>, School of Informatics, Indiana University



David T. Stanton, Modeling & Simulations Department, Corporate Research, Procter & Gamble



### The Big Picture

A molecule that is not similar to the TSET is probably going to be poorly predicted

> We shouldn't rely on such a prediction

- If a molecule is similar to the TSET, it may or may not be well predicted
  - > Can we rely on such a prediction?
  - Why should a molecule similar to the TSET be poorly predicted?



### **Traditional Approaches**

- Most approaches to model applicability consider the similarity to TSET
- "Extra" features in a molecule may be invisible to the model descriptors
  - The extra features cause the observed property to differ from that of similar molecules in the TSET
  - > We see the effect of this as a large residual

How do we decide that a molecule, similar to the training set, is **actually** similar to the training set



### Simple Example – Boiling Point Model

- C<sub>4</sub> C<sub>12</sub> Non-aromatic, containing at least 1 oxygen
  - Model development: 80 esters, ethers, ketones
    - ➢ 64 in Training set
    - 16 in Validation set
  - New External Prediction Set: 32 alcohols
- ➢ 6-variable model
  - ▶ R<sup>2</sup> = 0.970
  - > Good fit for Validation set (r = 0.973)
- > Applied model to predict BP for alcohols





#### Simple Boiling Point Model - Results Model Development Results **External Prediction Results** 300 TSET R<sup>2</sup> = 0.970 250 þ þ 250 200 200 150 150 100 100 2 2 50 100 150 200 250 150 250 350 450 50 **Observed Boiling Point (°C) Observed Boiling Point (°C)** Training Set Training Set (esters, ketones, ethers) $\Delta$ **External Prediction Set** Test Set (alcohols)

### The model under predicts most of the new observations



### A Hierarchical Strategy



### Model-Space Characterization

- Evaluate mean and std dev of the distances of the TSET points to the centroid of the TSET
  - A PSET point is *in model-space* if its distance to the centroid is within 2SD of the mean TSET distances
- Crude classification
- But we can safely ignore points that lie outside model space







### Neighborhood Characterization

- Evaluate the average pairwise distance between 5-NN of each TSET point
- Summarize the whole TSET in terms of mean and standard deviation
  - For a PSET point, determine the average 5-NN distance in the TSET
  - Apply a similar classification rule as before
- Isolated TSET points can skew the mean and std dev







### Neighborhood Characterization

- The R-NN curve method numerically measures the density of space around a given point
  - The measure is called R<sub>max(S)</sub> and higher values indicate a more sparse location
  - Makes no assumptions about the spatial distribution of points





### **Global Features - Structural Fingerprints**





### **Global Features - Structural Fingerprints**



- Summarize the TSET in terms of bit frequencies
- Identify bits that have low frequencies
- Identify bits that have intermediate frequencies

Indiana University School of

**or**matics



**Infrequent Features** 



### Augmenting Descriptor Spaces

- Is a PSET molecule truly within the TSET?
- Augment the model space with another dimension
  - If the distance and neighborhood classifications don't hold, maybe it really isn't in model space







### Augmenting Descriptor Spaces

- We don't know what this extra dimension should be
- > We might need more than one extra dimension
- But we can provide a few constraints
  - The topology of the TSET should be the same in both spaces
  - The extra dimensions should not be highly correlated to the model descriptors
  - The distributions in the extra dimensions should not be the same for the TSET and PSET





### Searching for Suitable Spaces



### **Using Augmented Descriptor Spaces**



Augmented Descriptor Space



If the nearest neighbors stay the same in the augmented space, the PSET molecule is most likely similar to the TSET



### Simple BP Model - Local Assessments

	Local Model IN	Local Model OUT
Error IN	6	1
Error OUT	23	2

- Most of the PSET is considered to be within the TSET
- The model descriptors are not capturing molecular features that differentiate the PSET molecules from the TSET





### Simple BP Model - Descriptor Augmentation

Augmented with 2 descriptors

- Used a brute force method to identify suitable sets
- Most sets cause relatively low loss of order
- Chose two H-bonding descriptors
  - > RPCS, RNCS
  - Very low correlation to the model descriptors
  - Non-zero values in the TSET and PSET





### Simple BP Model - Global Assessments

	Local Model IN	Local Model OUT
Error IN	6	1
Error OUT	23	2

#### 25% Correct Assessment

True Positives		
	Global Model IN	Global Model OUT
Error IN	0	6
Error OUT	0	0

False Positives		
	Global Model IN	Global Model OUT
Error IN	0	0
Error OUT	0	23



Beilstein Data Set BP Models
 Published boiling point models\*
 Four very diverse training sets

 
 Furans & Tetrahydrofurans (N=209)
 Thiophenes (N=134)
 Pyrans (N=146)

 Combo-1
 Combo-1





<sup>\*</sup> J. Chem. Inf. Comput. Sci., **1991**, 31, 301-310 J. Chem. Inf. Comput. Sci., **1992**, 32, 306-316

### **Beilstein BP Models - Local Assessments**

#### **Furan / Tetrahydrofuran Model**

	Local Model IN	Local Model OUT
Error IN	24	1
Error OUT	35	3

42.8% Correct Assessment

#### **Combo-2 Model**

	Local Model IN	Local Model OUT
Error IN	63	0
Error OUT	0	0

**100% Correct Assessment** 

Indiana University School of

### Beilstein BP Models - Global Assessments

#### **Furan / Tetrahydrofuran Model**

	Local Model IN	Local Model OUT
Error IN	24	1
Error OUT	35	3

42.8% Correct Assessment

True Positives		
	Global Model IN	Global Model OUT
Error IN	20	4
Error OUT	0	0

False Positives		
	Global Model IN	Global Model OUT
Error IN	0	0
Error OUT	30	5



### Beilstein BP Models - Global Assessments

#### **Combo-2 Model**

	Local Model IN	Local Model OUT
Error IN	63	0
Error OUT	0	0

**100% Correct Assessment** 

Global Model OUT
21
0



### More Sophistication?

- Model applicability is equivalent to novelty detection
- Augmented spaces are conceptually similar to the theory of SVM's
  - Go to a high-D space, to get better separation



The original problem remains what is a set of suitable descriptors?



### Summary

It's not enough to just consider the model space

- So what is a good global space?
  - > Augmented space approach is intuitive
  - Molecular diversity
  - Probably more important than methodology
- > Fix the false positives, leave the true positives
- Becomes much harder when applied to diverse training sets







## Simple BP Model - Fingerprint Classification

Molecules that are similar to the TSET by distance have very few "extra" features





- This does not correspond well with residual classes
- Molecules with no "extra" features may still have a high residual



### A Hierarchical Strategy - Local Information



### A Hierarchical Strategy - Global Information



### Simple BP Model - Descriptor Augmentation

From a pool of 160 descriptors, we identified 3 descriptors

Maintained the topology of the TSET perfectly

- > They were all H-bonding descriptors
  - Values for the TSET were 0, non-zero for the PSET molecules
- > We consider SSAH as the augmenting descriptor



### Simple BP Model - Descriptor Augmentation

### In terms of the distance to centroid class

- All PSET molecules are outside of model space in the augmented space
- In terms of the number of common nearest neighbors
  - All PSET molecules have the same neighborhood in both spaces

