



ParaSurf™ and ParaFit™

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- Cepos InSilico is a UK limited company located in Kempston, Beds
 - Company Secretary Paul Hillier
- Close links to the Universities of Erlangen and Portsmouth
- 100% owned German subsidiary (Cepos InSilico GmbH) located in Erlangen formed in December 2009
 - CEO Dr. Ute Seidel



News

From March 1st 2010, Cepos products will be distributed in North America by Cache Research LLC, Beaverton, Oregon.



<http://cacheresearch.com/home.html>

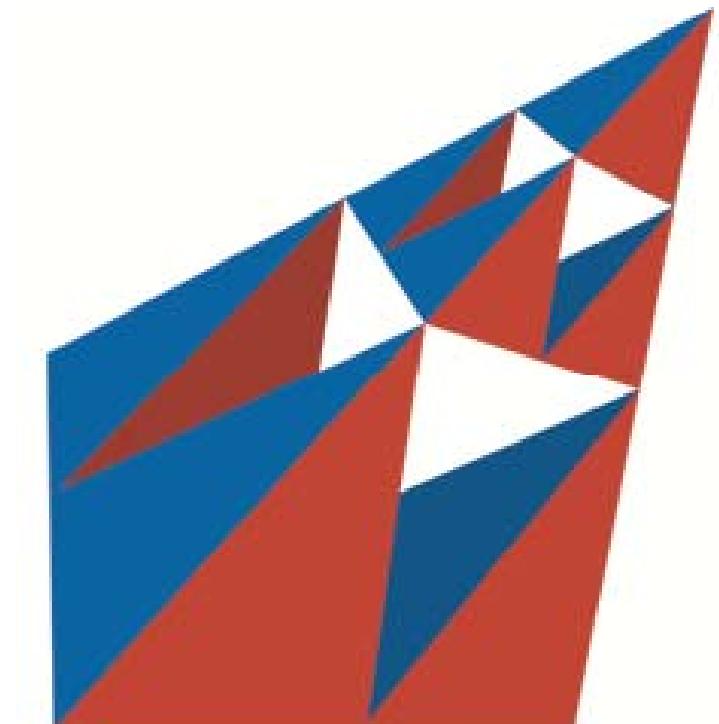
Cepos InSilico and InhibOx have agreed a joint distribution and technical cooperation effective immediately

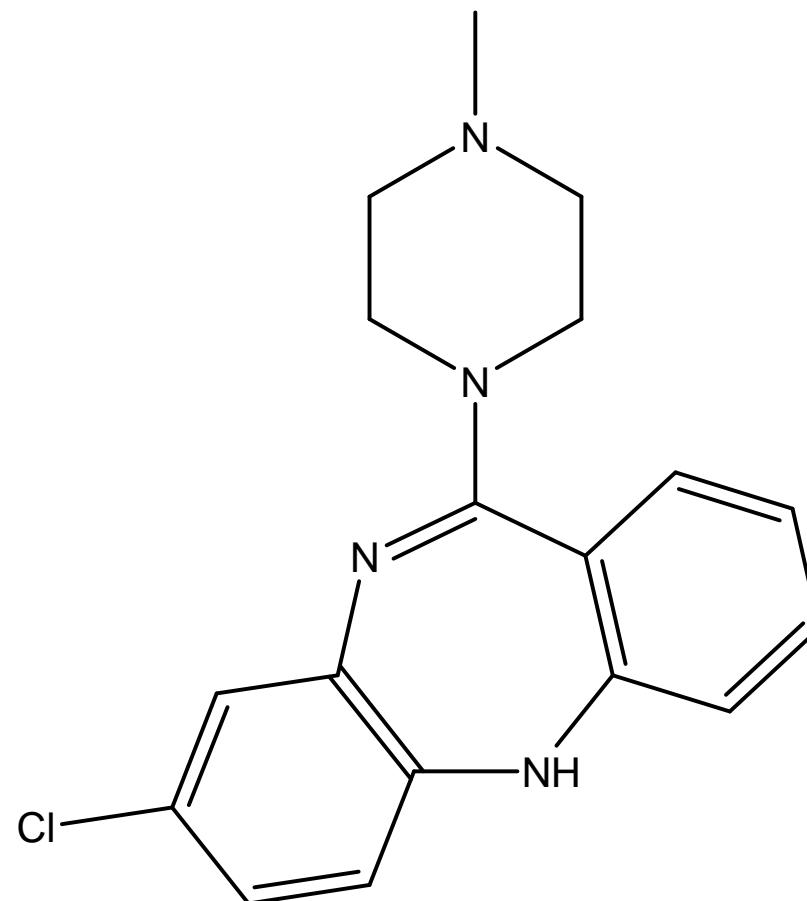


<http://www.inhibox.com/index.html>

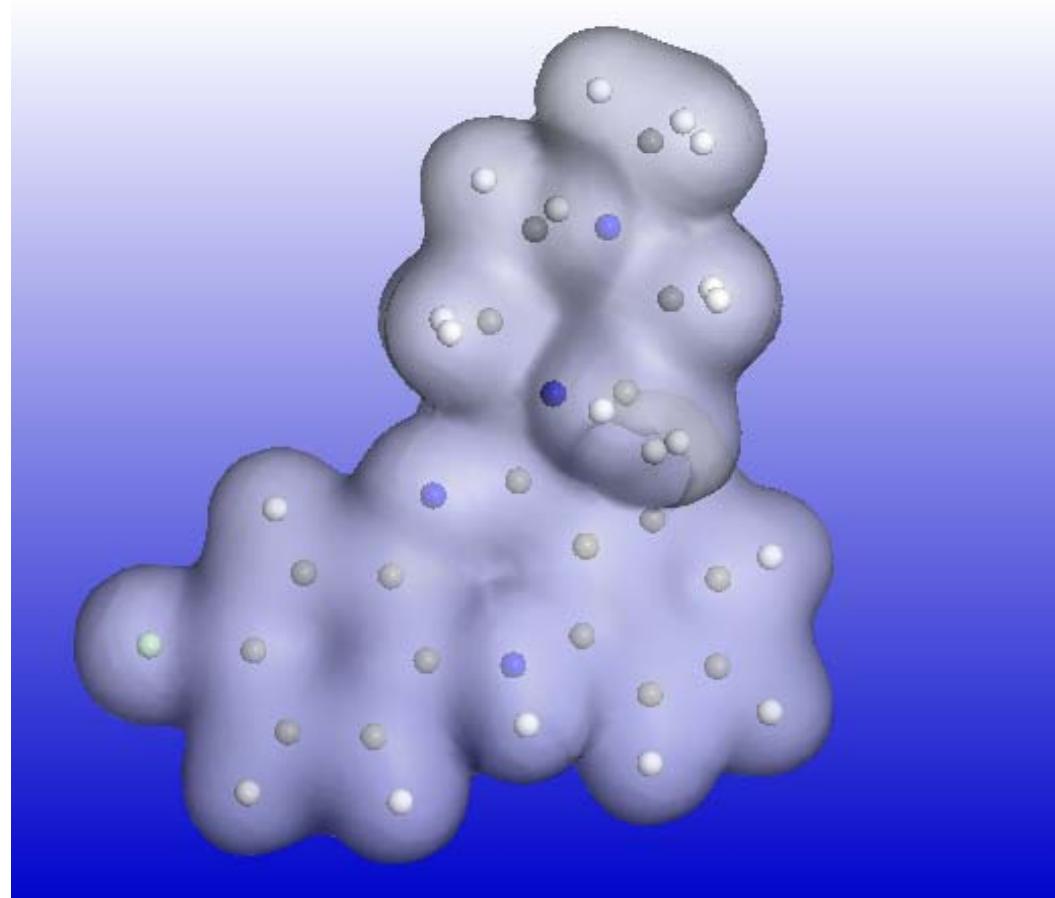
Cepos InSilico and Molcad
GmbH have agreed a joint
distribution and technical
cooperation effective
December 1st 2010.

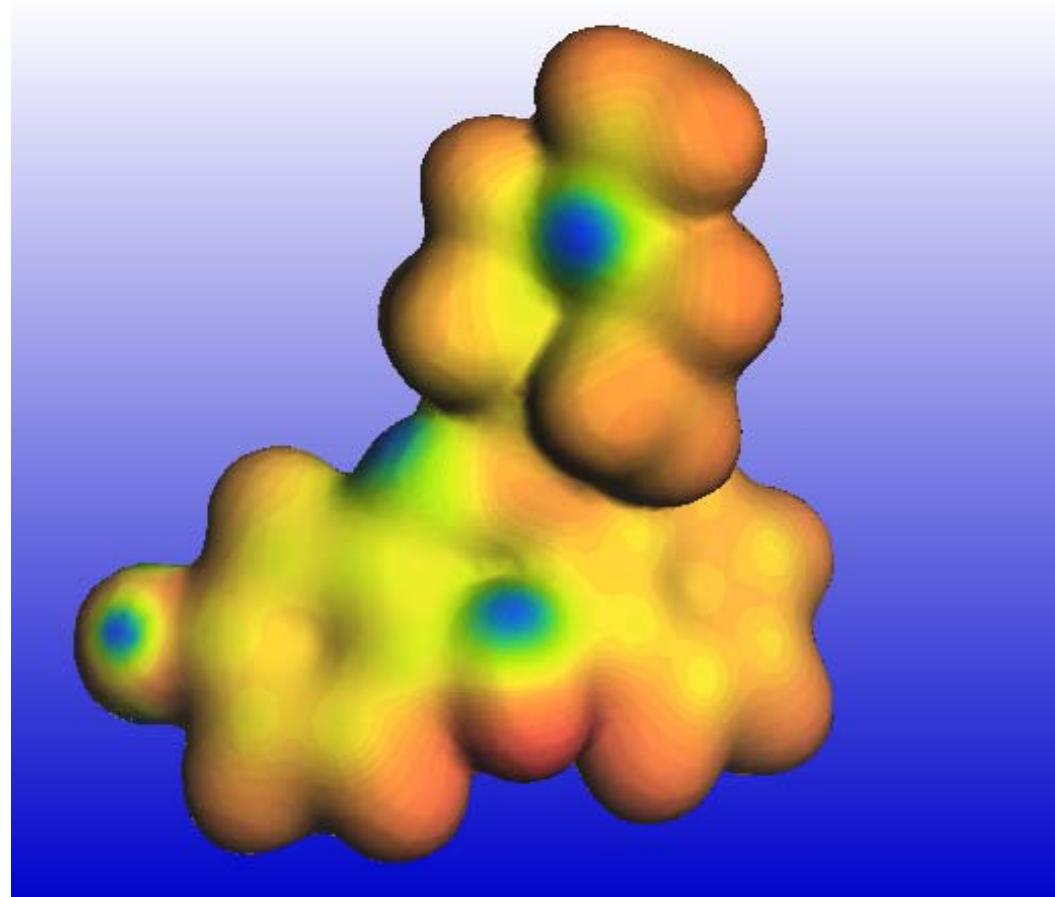
<http://www.molcad.de>











- ParaSurf™ describes intermolecular interactions completely, including
 - The shape (repulsions, electron density)
 - Intermolecular interactions
 - Coulomb
 - Donor/Acceptor
 - Acceptor/Donor
 - Dispersion

Local Properties

- To do this, it uses a series of local properties derived from semiempirical MO theory (usually AM1).
- These are:
 - The electron density
 - The Molecular Electrostatic Potential and the electrostatic field normal to the surface
 - The local ionization energy (Politzer et al.)
 - Local electron affinity
 - The local polarizability
 - Local hardness and local electronegativity (derived from the ionization energy and electron affinity)

Local Ionization Energy

$$IE_L(\mathbf{r}) = \frac{\sum_{i=1,HOMO} -\rho_i(\mathbf{r}) \varepsilon_i}{\sum_{i=1,HOMO} \rho_i(\mathbf{r})}$$

Sjoberg, P.; Murray, J. S.; Brinck, T.; Politzer, P. A., *Can. J. Chem.* 1990, **68**, 1440;
Murray, J. S.; Abu-Awwad, F.; Politzer, P., *THEOCHEM* 2000, **501-502**, 241;
Hussein, W.; Walker, C. G.; Peralta-Inga, Z.; Murray, J. S., *Int. J. Quant. Chem.* 2001, **82**, 160;
Politzer, P.; Murray, J. S.; Concha, M. C., *Int. J. Quant. Chem.* 2002, **88**, 19.

Other Local Properties

- Local Electron affinity:

$$EA_L(\mathbf{r}) = \frac{\sum_{i=LUMO,norbs} -\rho_i(\mathbf{r}) \varepsilon_i}{\sum_{i=LUMO,norbs} \rho_i(\mathbf{r})}$$

- Local Hardness:

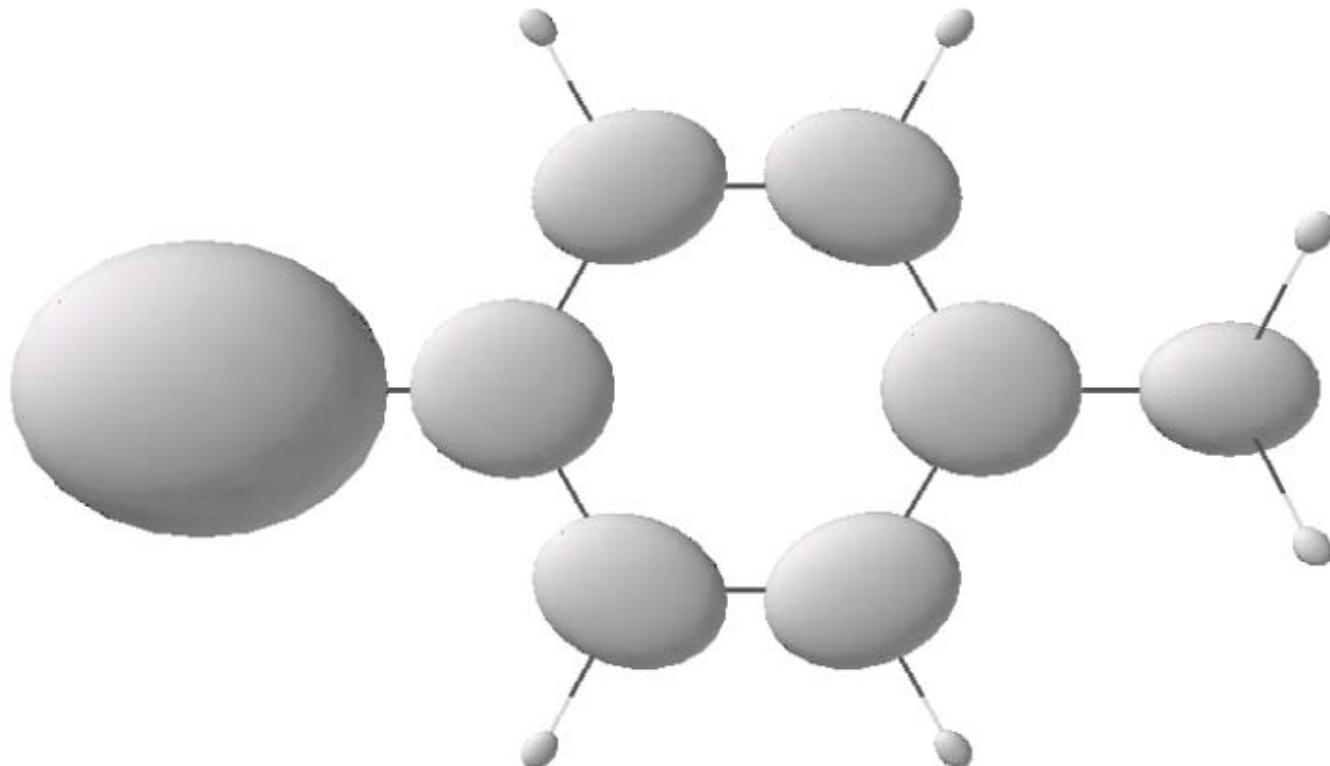
$$\eta_L(\mathbf{r}) = \frac{(IP_L(\mathbf{r}) - EA_L(\mathbf{r}))}{2}$$

- Local electronegativity

$$\chi_L(\mathbf{r}) = \frac{(IP_L(\mathbf{r}) + EA_L(\mathbf{r}))}{2}$$

B. Ehresmann, B. Martin, A. H. C. Horn and T. Clark, *J. Mol. Model.* 2003, 9, 342-347.

B. Ehresmann, M. J. de Groot, A. Alex and T. Clark, *J. Chem. Inf. Comp. Sci.* 2004, 44, 658-668.



G. Schürer, P. Gedeck, M. Gottschalk, T. Clark, *Int. J. Quant. Chem.*, 1999, 75, 17-31; B. Martin, P. Gedeck, T. Clark, *Int. J. Quant. Chem.*, 2000, 77, 473.

Local Polarizability

$$\alpha_L(\mathbf{r}) = \frac{\sum_{j=1}^{NAOs} \rho_j^1(\mathbf{r}) q_j \bar{\alpha}_j}{\sum_{j=1}^{NAOs} \rho_j^1(\mathbf{r}) q_j}$$

$\rho_j^1(\mathbf{r})$ Density due to a singly occupied atomic orbital j

q_j Coulson population of atomic orbital j

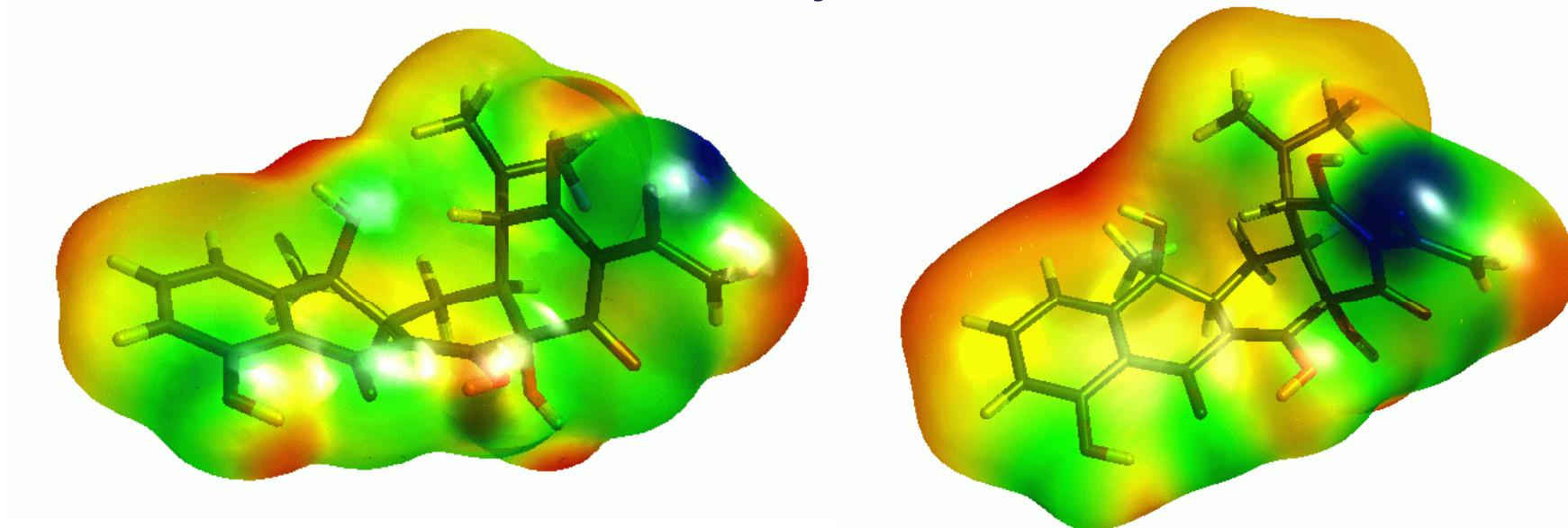
$\bar{\alpha}_j$ Mean polarizability calculated for atomic orbital j

B. Ehresmann, B. Martin, A. H. C. Horn and T. Clark, *J. Mol. Model.* 2003, 9, 342-347.

B. Ehresmann, M. J. de Groot, A. Alex and T. Clark, *J. Chem. Inf. Comp. Sci.* 2004, 44, 658-668.

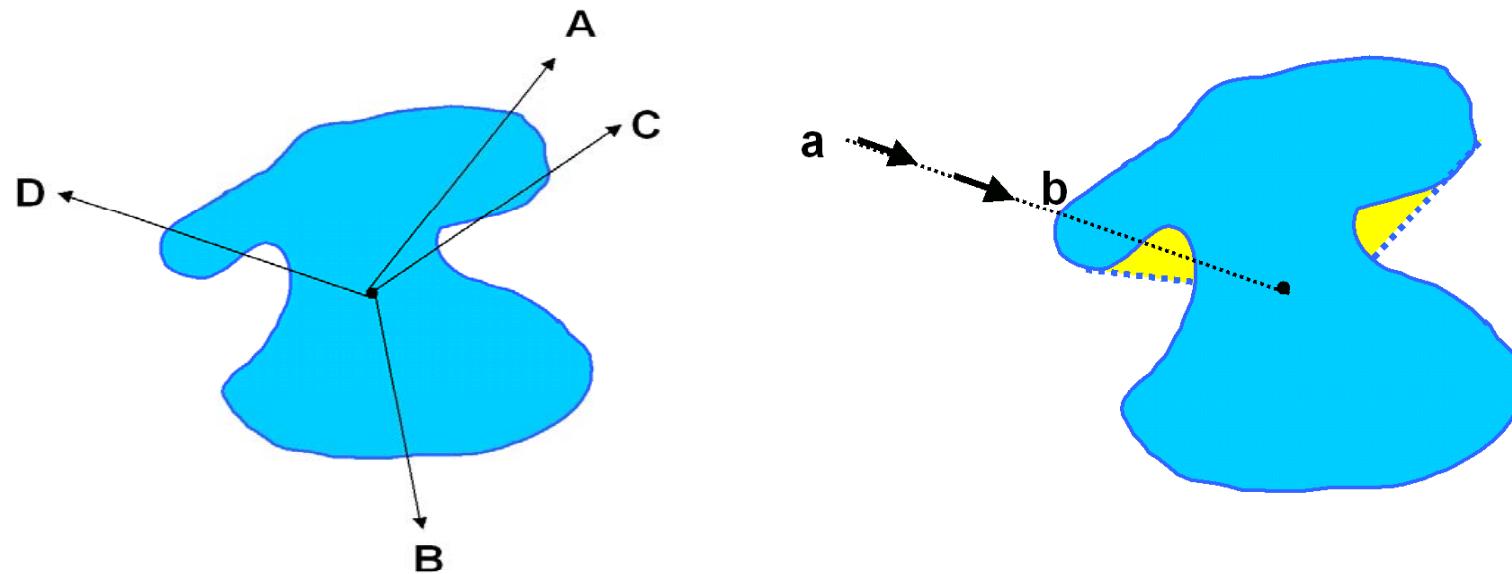
Alternative Surfaces

Tetracycline



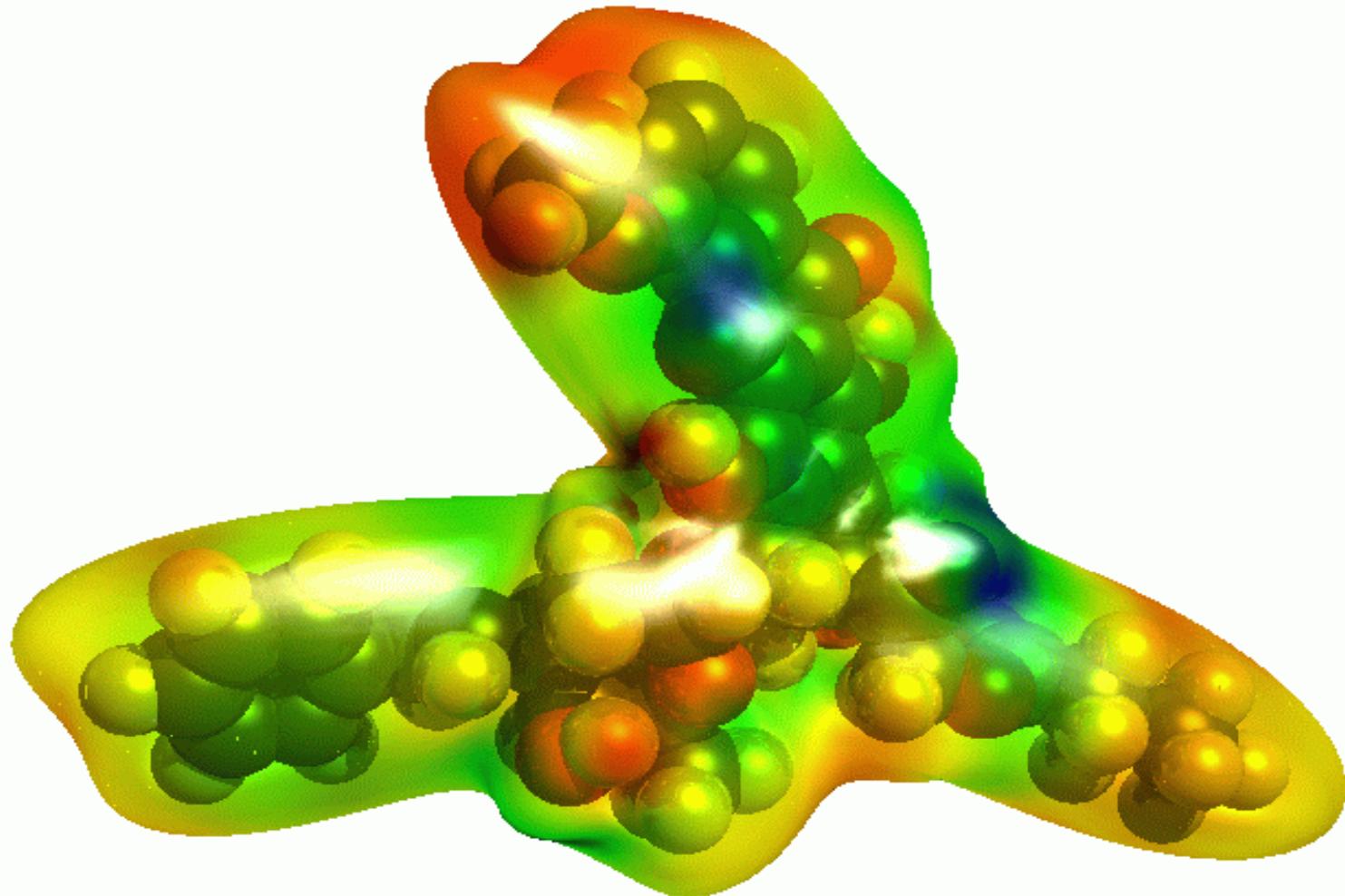
Marching-cube surface. Can be based on the electron density (isodensity surface) or on the solvent-excluded surface using van der Waals' radii

Alternative Surfaces



Shrink-wrap surface. Can also be based on the electron density (isodensity surface) or on the solvent-excluded surface using van der Waals' radii. These surfaces are single-valued along any given radial vector to the center, so may not "fit" as well as marching cube (yellow areas). However, they can be represented analytically.

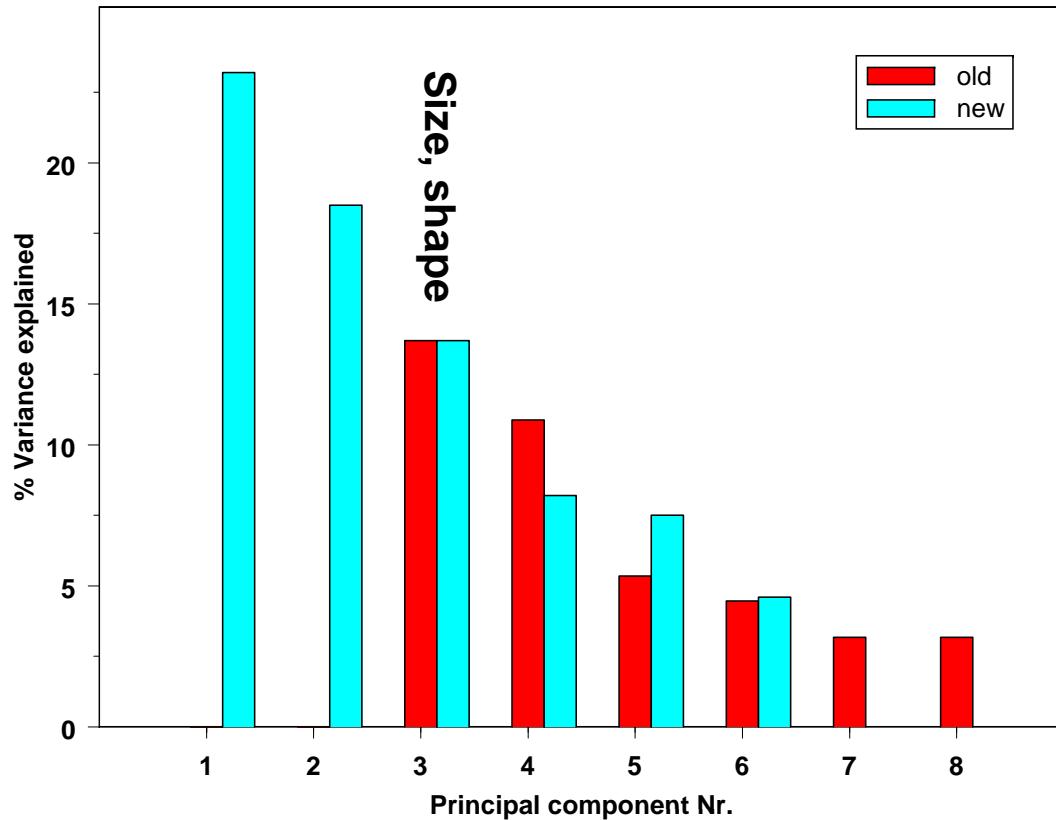
A Shrink-Wrap Surface



Descriptors

- How can we use the local properties?
 - Statistical descriptors based on the distribution of the properties on the surface:
 - Mean, maximum, minimum, range
 - Variance, skew, kurtosis
 - Balance parameter,
 - The descriptors are calculated for each of the local properties to give a comprehensive set of high information content for QSAR

Variance Explained



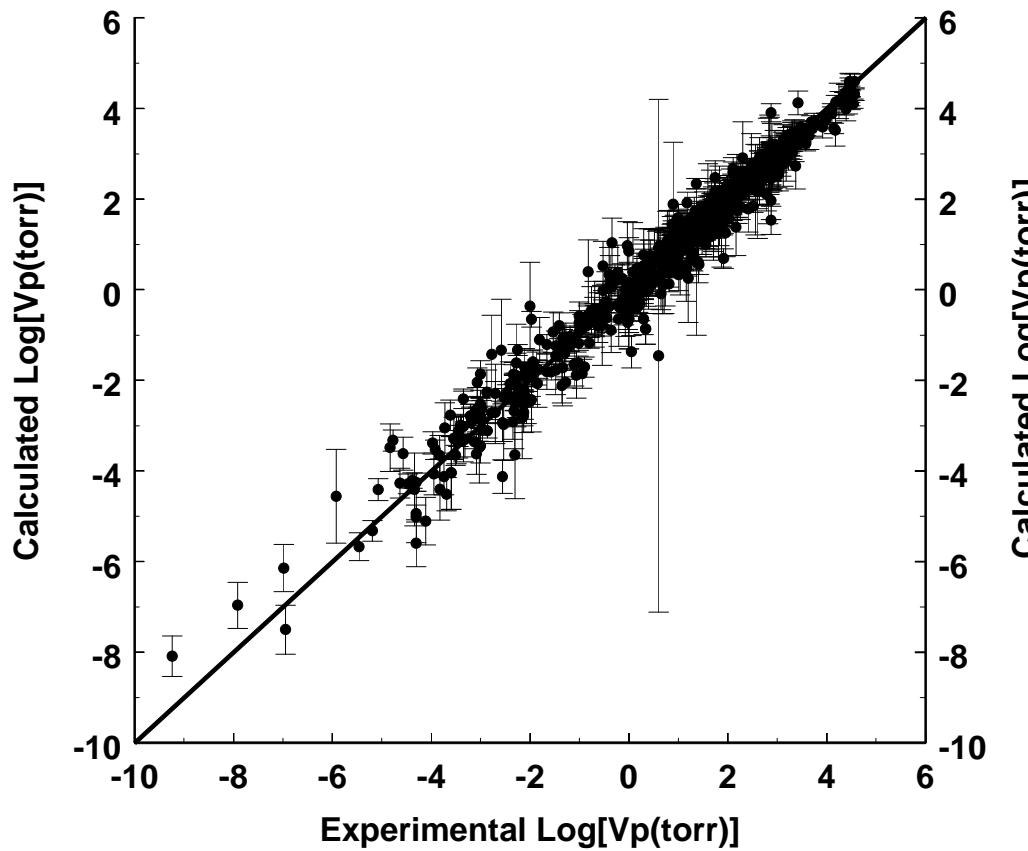
B. Ehresmann, M. J. de Groot, A. Alex and T. Clark, *J. Chem. Inf. Comp. Sci.* 2004 , 44 , 658-668.

- Maybridge database
- Principal components
- “Old” descriptors contain 2D-components
- “New” descriptors purely surface-based or global (MWt, volume etc.)
- Size, shape descriptor is almost identical
- Two most significant principal components are mostly local ionization energy and electron affinity

A QSPR Model for Vapor Pressure

- *8,542* measured vapor pressures at temperatures between 76 and 800 K for *2,349* compounds extracted from the Beilstein database
- Therefore include the temperature as a descriptor
- *A Temperature-dependent Quantum Mechanical/Neural Net Model for Vapor Pressure*, A. J. Chalk, B. Beck and T. Clark, *J. Chem. Inf. Comput. Sci.* 2001, *41*, 1053.

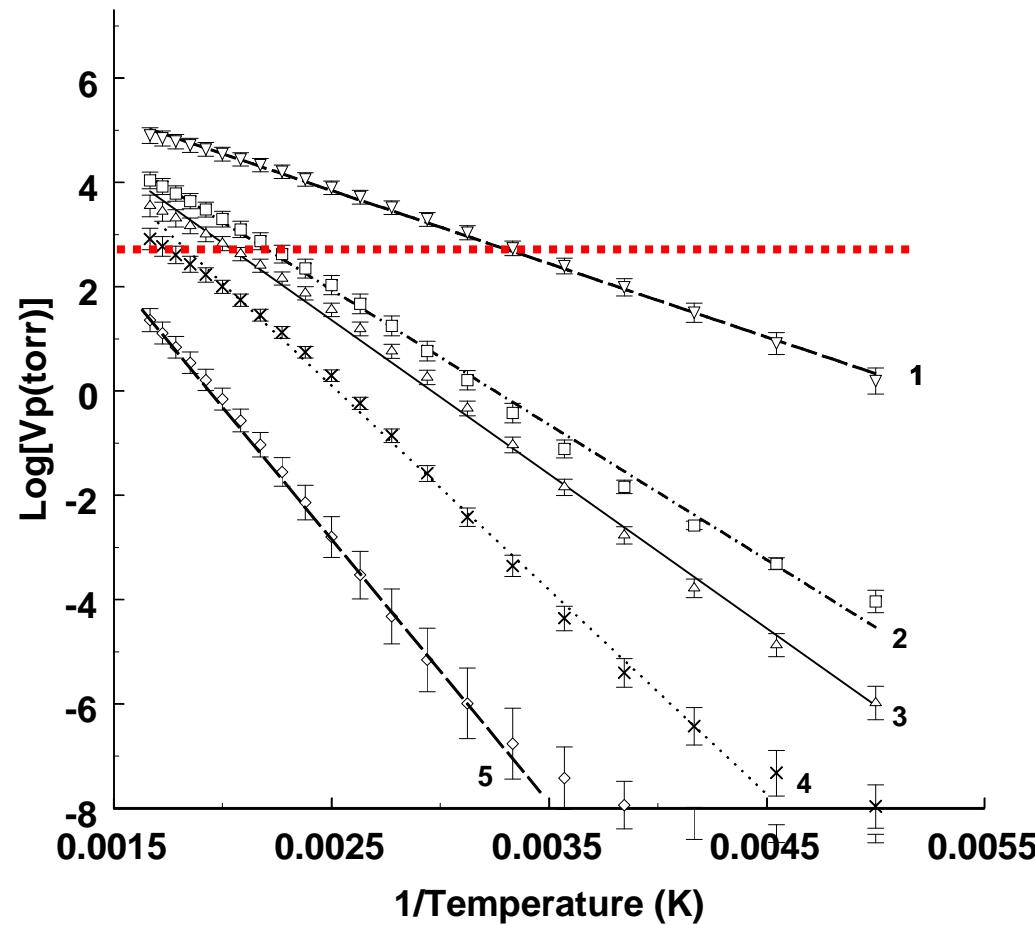
Validation Set



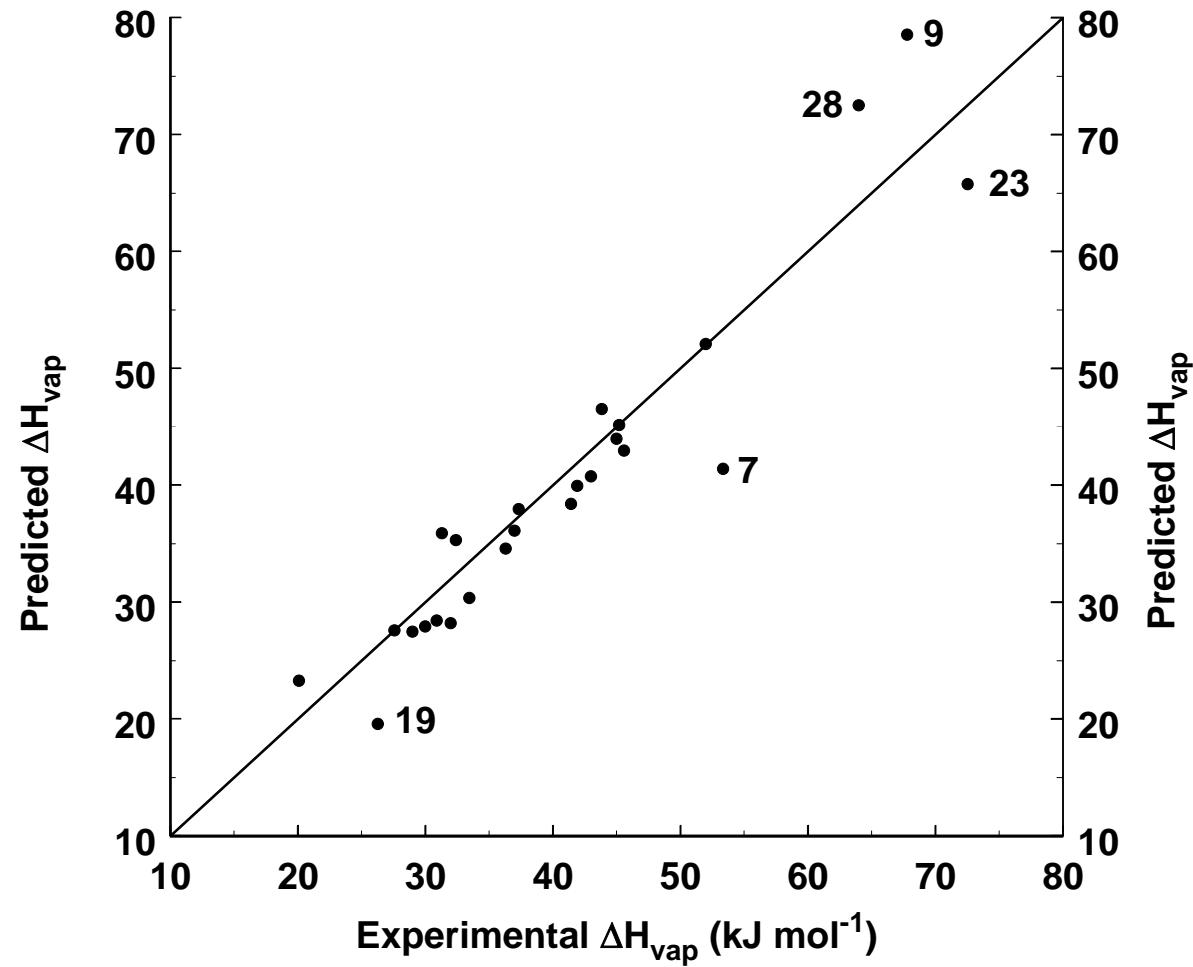
$N(\text{train}) = 7,681$
 $N(\text{validate}) = 861$

	Mean	Cross-validation	Validation set
R^2	0.976	0.952	0.976
σ	0.322	0.459	0.326
MUE	0.211	0.291	0.213
LE	2.02	4.15	2.06

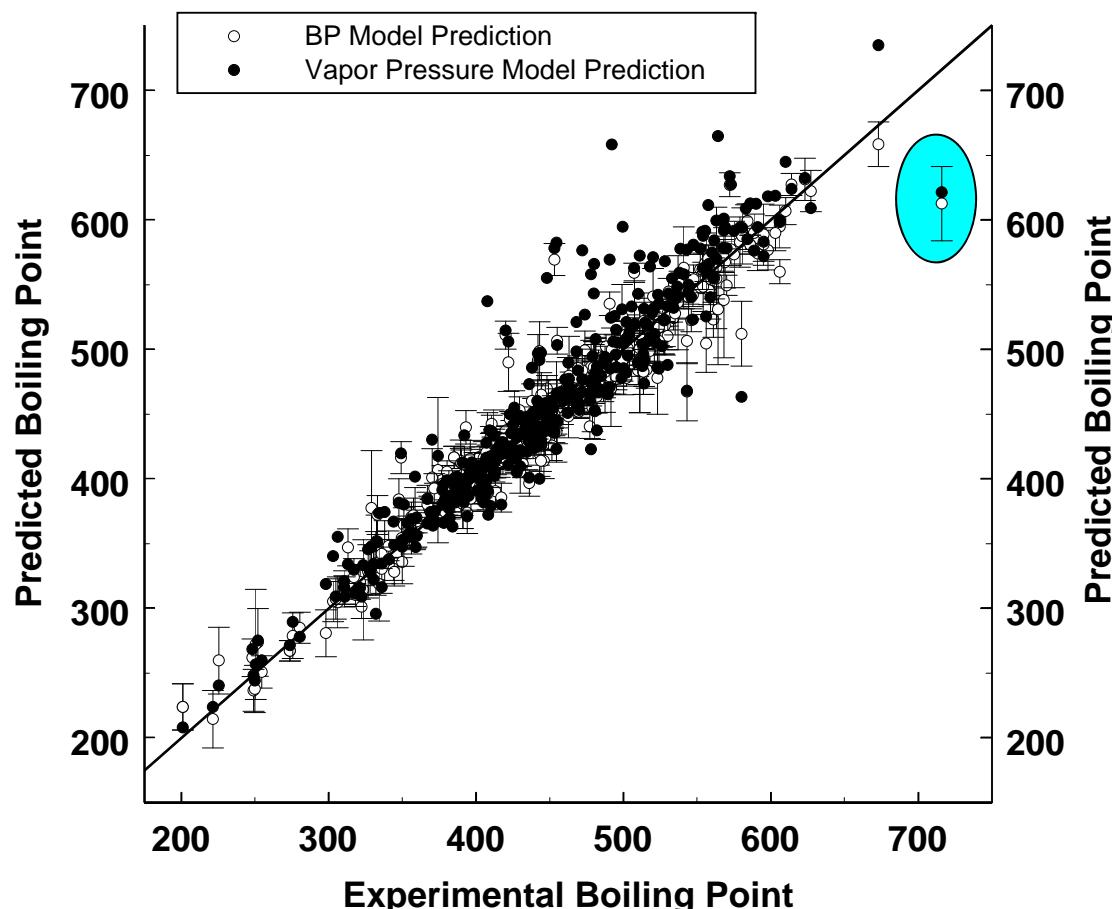
Temperature-Dependence



Heats of Vaporization



Boiling Points



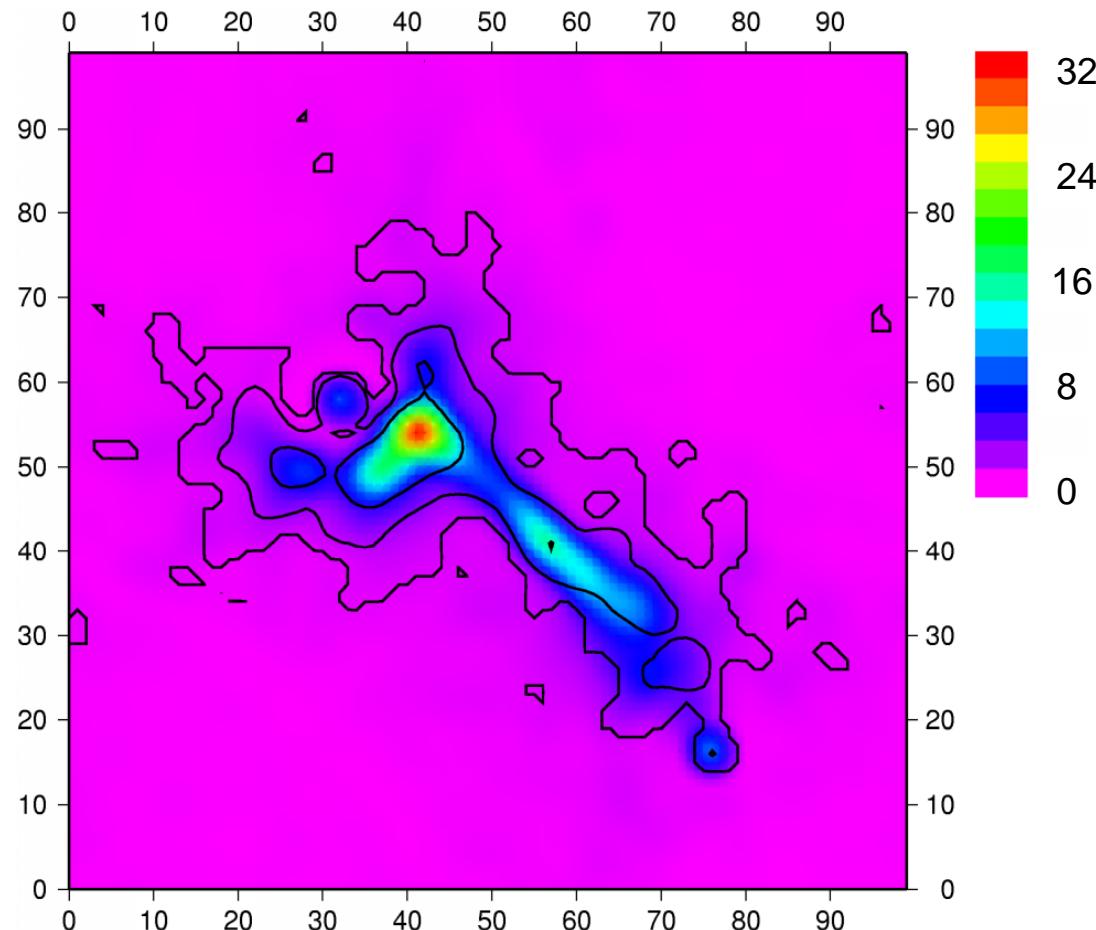
	Vapor pressure model	Boiling point model ^a
R ²	0.889	0.928
σ	28.6	21.9
MUE	18.7	13.5
LE	166.1	140.0

^a *A Quantum Mechanical/Neural Net Model for Boiling Points with Error Estimation*, A. J. Chalk, B. Beck, and T. Clark, *J. Chem. Inf. Comput. Sci.* 2001, 41, 457.

- Maybridge used as the non-drug dataset
- 2,105 compounds selected from the World Drug Index as real drugs used as the drug dataset
- Select descriptors for a Kohonen map by using recursive partitioning to separate the two datasets
- *Descriptors, Physical Properties and Drug-Likeness*, M. Brüstle, B. Beck, T. Schindler, W. King, T. Mitchell and T. Clark, *J. Med. Chem.* 2002 , 45, 3345.

World Drug Index

(42,131 compounds with 150 atoms or less)



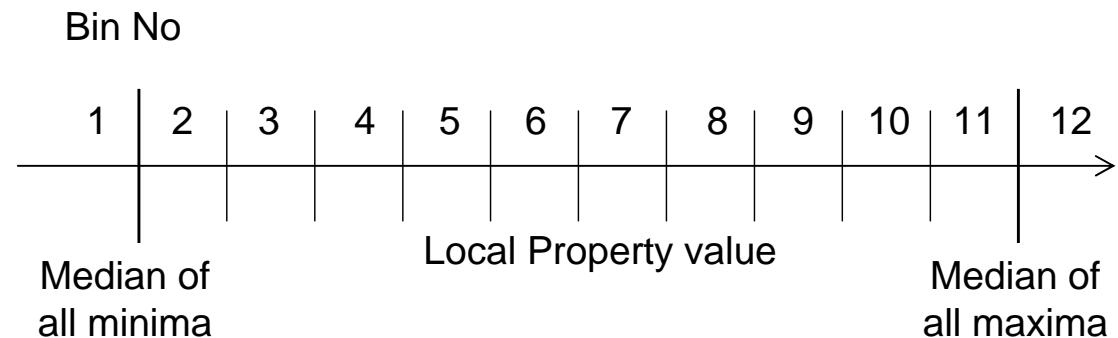
Surface-integral models

$$P = \sum_{i=1}^{ntri} f\left(V^i(\mathbf{r}_i), IE_L^i(\mathbf{r}_i), EA_L^i(\mathbf{r}_i), \alpha_L^i(\mathbf{r}_i), \eta_L^i(\mathbf{r}_i)\right) \cdot A^i$$

- P = target property
- A_i = area of triangle i
- $ntri$ = number of triangles
- \mathbf{r}_i = center of triangle i

Surface-Integral QSPR Models: Local Energy Properties, B. Ehresmann, M. De Groot and T. Clark, *J. Chem. Inf. Model.*, 2005, 45, 1053-1060.

Standardized binning scheme:



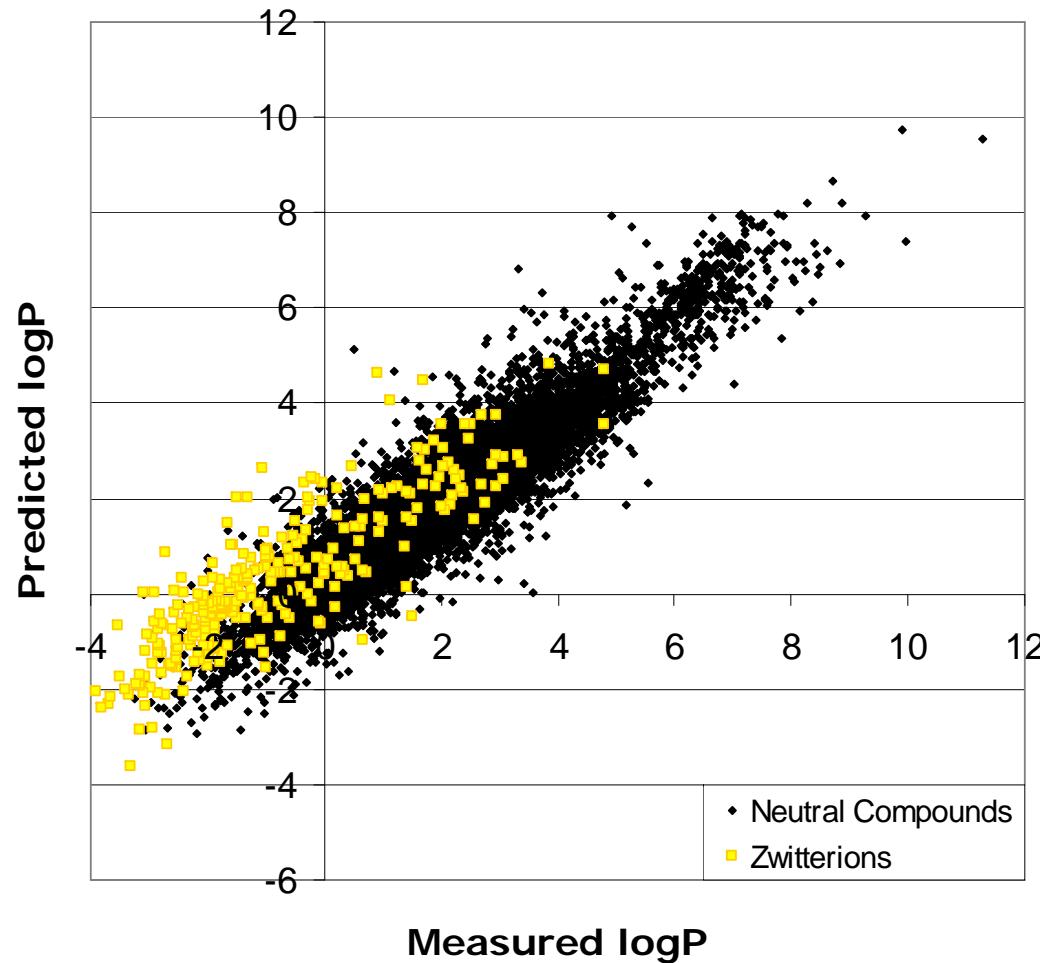
Areas as descriptors:

$$D^N = \sum_{i=1}^{Ntri} \delta_i^N A_i$$

D^N = descriptor for bin N
 δ_i^N = 1 if the triangle is within the bin, 0 otherwise.

A Surface-Integral Model for logP_{OW}, C. Kramer, B. Beck and T. Clark, J. Chem. Inf. Model., 2010, 50, ASAP (DOI: 10.1021/ci900431f).

Results (Test Set)



logP Model

Model	Public Validation Set	
	RMSE	
ACDlabs logP	0.45	
ClogP (BioByte)	0.52	
SIM-logP	0.64	
SlogP (MOE)	0.68	
logP_o/w (MOE)	0.77	
AlogP (TSAR)	0.86	

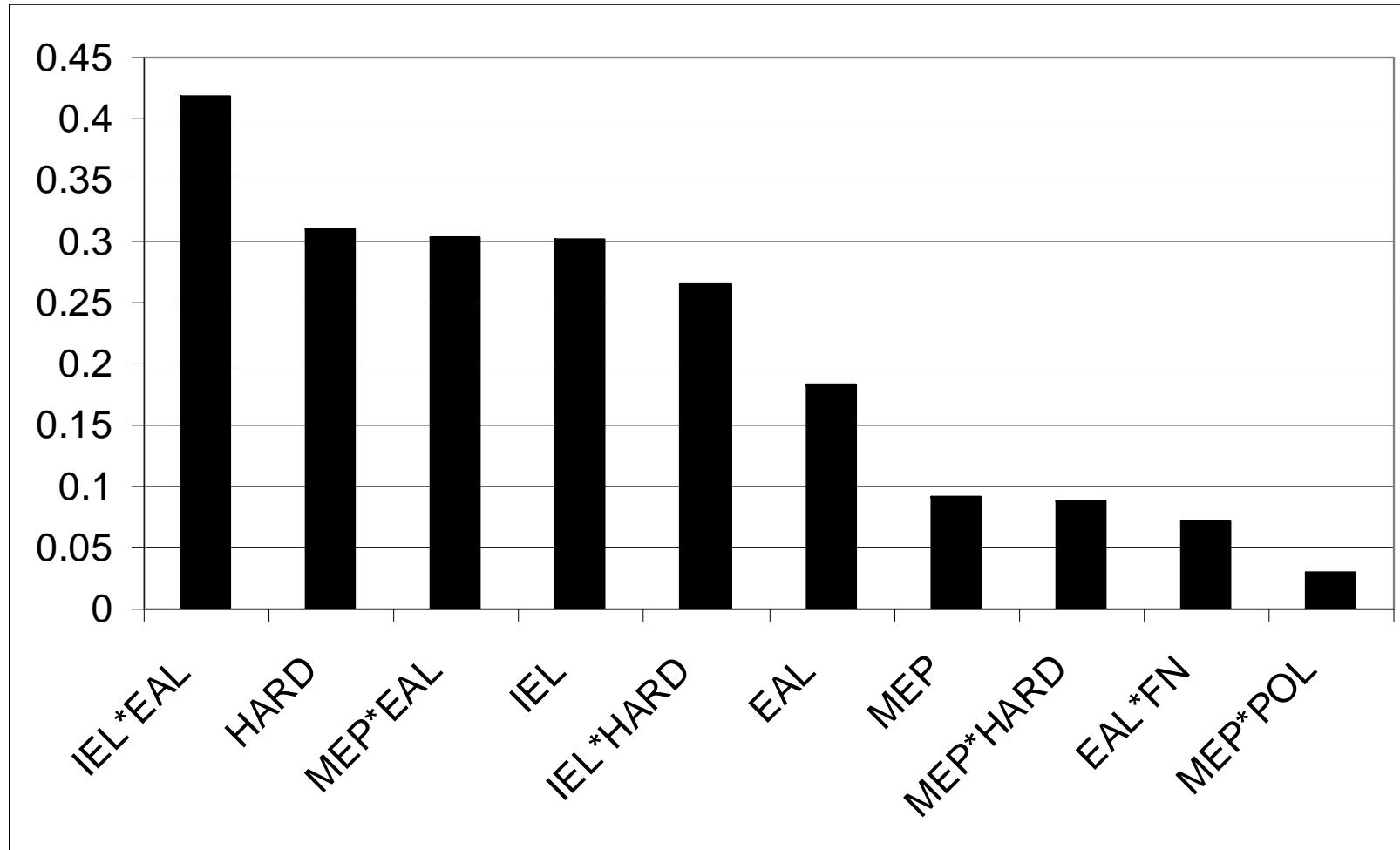
A Surface-Integral Model for logP_{OW}, C. Kramer, B. Beck and T. Clark, J. Chem. Inf. Model., 2010, 50, ASAP (DOI: 10.1021/ci900431f).

logP Model

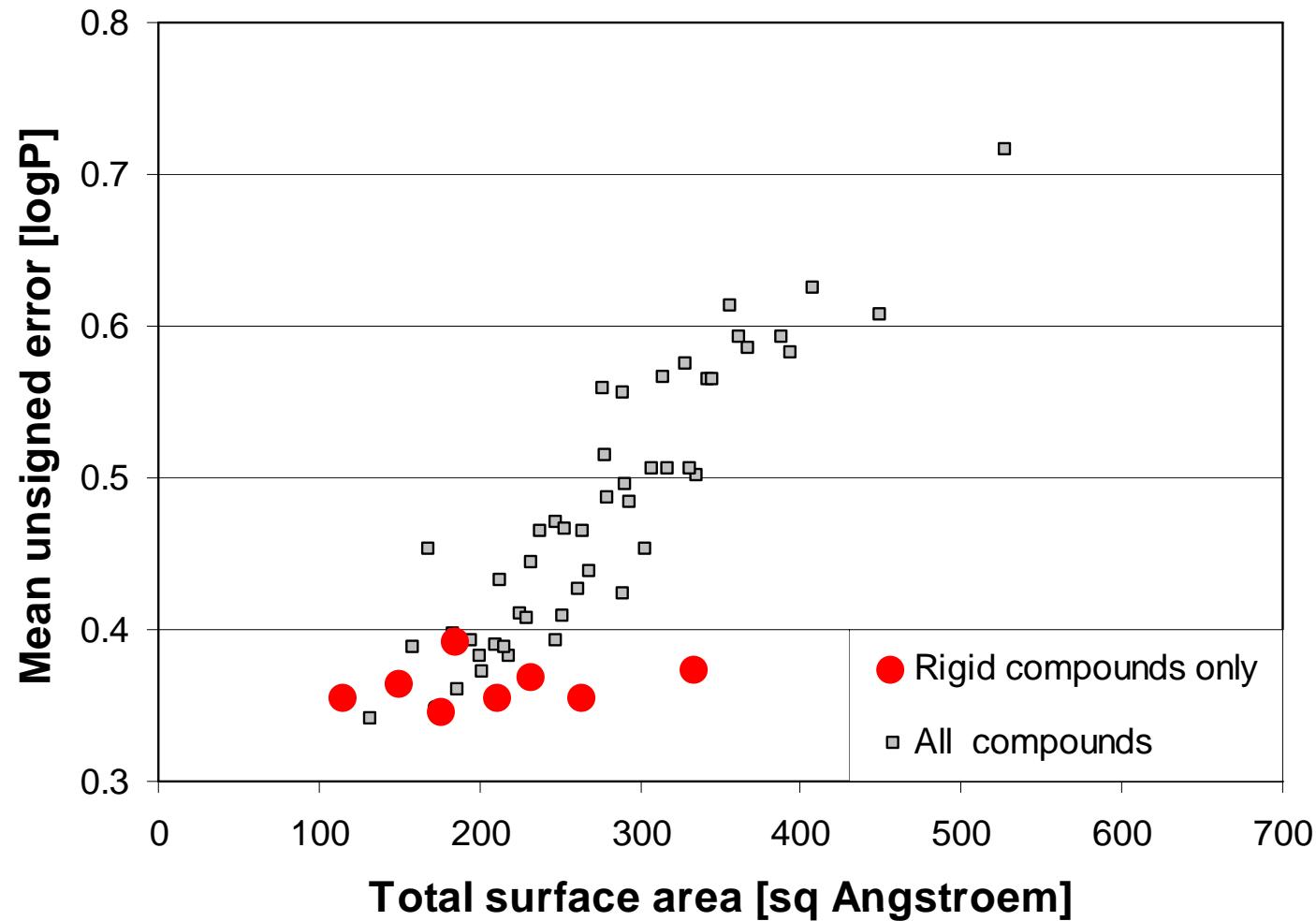
Model	Public Validation Set	In-house Validation Set
	RMSE	RMSE
SIM-logP	0.64	1.10
ClogP (BioByte)	0.52	1.14
SlogP (MOE)	0.68	1.19
AlogP (TSAR)	0.86	1.24
logP_o/w (MOE)	0.77	1.28
ACDlabs logP	0.45	1.36

A Surface-Integral Model for logP_{OW}, C. Kramer, B. Beck and T. Clark, J. Chem. Inf. Model., 2010, 50, ASAP (DOI: 10.1021/ci900431f).

Which Properties are Important?



Effect of Conformation



Multi-Conformation logP

Compound	Experiment	Model	CORINA conformation	Boltzmann weighted	
				Gas-phase	SCRF
Omeprazole	2.23	All compounds	2.94 (0.56-0.71)	2.97	2.65
	2.38	Rigid only	2.92	2.68	2.51 (0.13-0.28)↓
Risperidone	3.04	All compounds	2.48 (0.46)	3.20	3.14
		Rigid only	2.78	3.17	3.03 (0.01)↑
Haloperidol	4.30	All compounds	4.52 (0.22)	4.23	4.19
		Rigid only	4.17	4.20	4.18 (0.12)→

Reactivity

- The local ionization energy and electron affinity are not only related to donor-acceptor interactions, but also to nucleophilic and electrophilic properties, respectively.
- They are therefore ideally suited for predicting reactivity.
- One example is the lability to oxidation by cytochrome P450 enzymes

CypScore: Quantitative Prediction of Reactivity toward Cytochromes P450 Based on Semiempirical Molecular Orbital Theory, M. Hennemann, A. Friedl, M. Lobell, J. Keldenich, A. Hillisch, T. Clark and A. H. Göller,
ChemMedChem, 2009, 4, 657-669.

- ... predicts the likely sites of Cytochrome P450 metabolism in small molecules
- ... is a model for a hypothetical CYP P450 super-enzyme
 - “Simulates” oxidation reactions by P450 enzymes
 - Currently no active-site accessibility restrictions
- ... is based AM1 and ParaSurf
 - Describes oxidation lability of centers of chemical reactivity
 - Models available for most frequent oxidation reactions
 - All Models weighted to a consensus range
 - Restricted to Phase I reactions

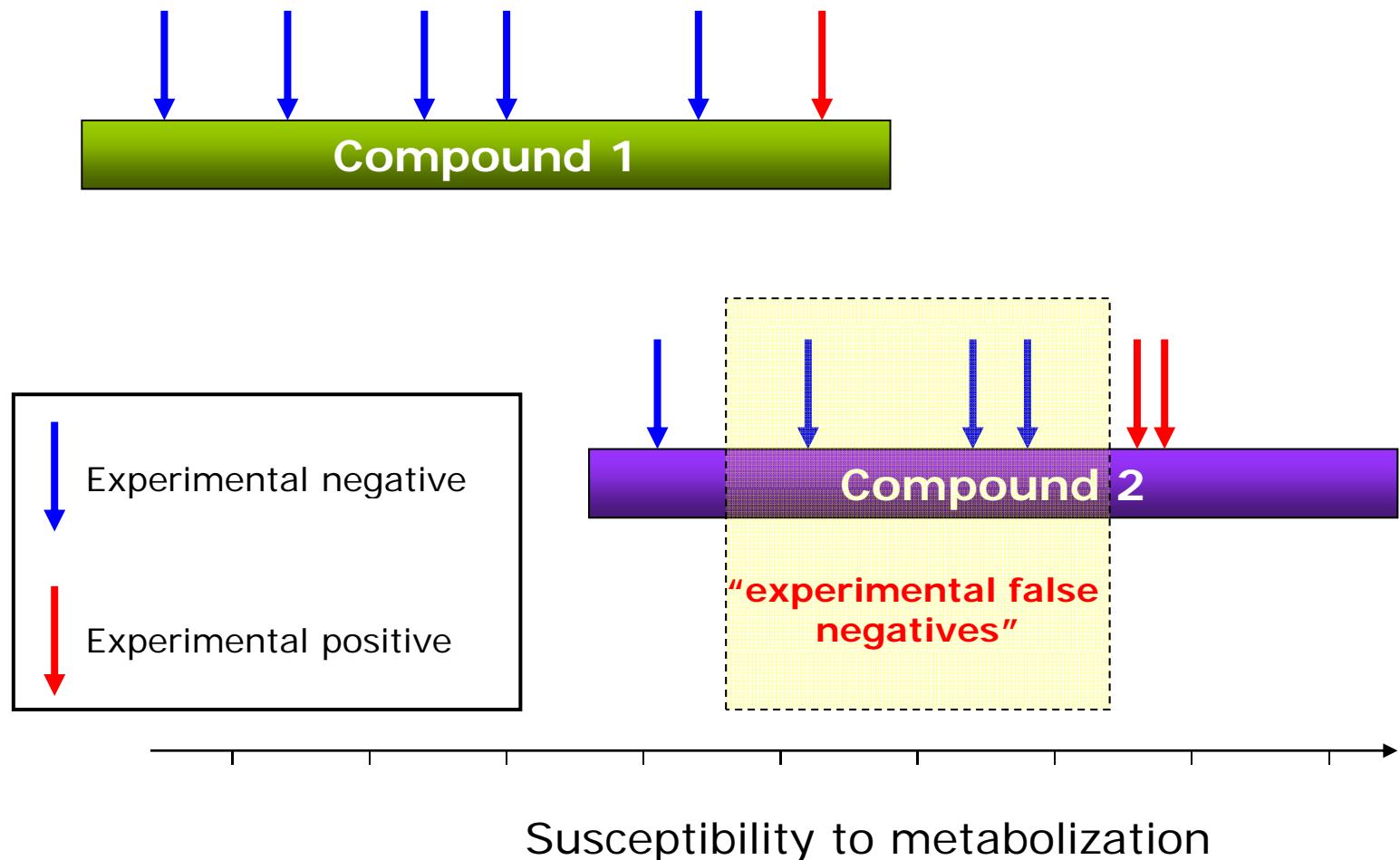


CEPOS Atomic Surface Descriptors (ParaSurf)

1. Triangulate surface
2. Calculate atomic contributions to the electron density at the corners
3. If two corners “belong” to the same atom, allocate the triangle to that atom
4. Atoms are described by the statistical characteristics (max, min, mean, variance etc.) of the distribution of the local properties on “their” surface.
5. Provides an excellent description of local reactivity

- **Dataset:**
 - MajorMetaboliteDB from Bayer with 950 compounds and 1750 reactions
- **Models:**
 - Individual models for the different types of reaction
 - All scaled to a common “metabolic strength” scale
 - All descriptor-based regression models
 - None with more than four descriptors
- **Regression philosophy:**
 - Data is binary (0,1), but we need a quantitative scale
 - Use “unsuitable” MLR and assume that the very weak positions will be detected by the descriptors
 - Better than a classification model (!) because of the structure of the data

Data



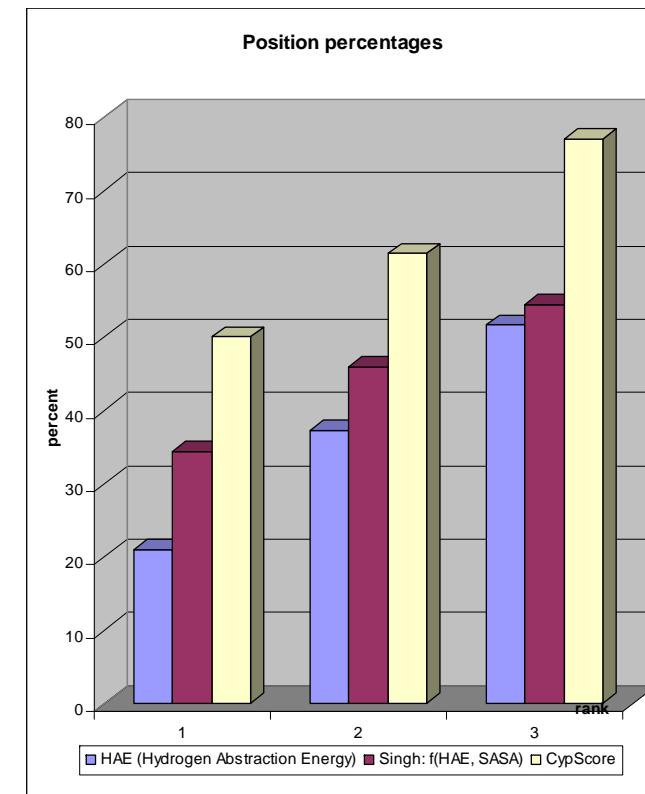
- Reaction centers:
 - Csp³
 - Csp² (Aryl)
 - Csp² (C=X)
 - N-oxidation
 - NR₂ and NR₃
- All centered and scaled
 - Scores range from -10 (very weak) to +10 (stable)
 - A score of zero marks equal probability of positive and negative

70 compounds, 41 from the Singh data set, plus 29 challenging ones

- 187 weak positions in 1797 heavy atoms (1 to 5 per molecule)
- all types of reactions
 - 124 aliphatic hydroxylations
 - 50 Double bond oxidations
 - 12 N-oxidations
- 73% of metabolic positions found



Bayer HealthCare



Results: in-house validation set

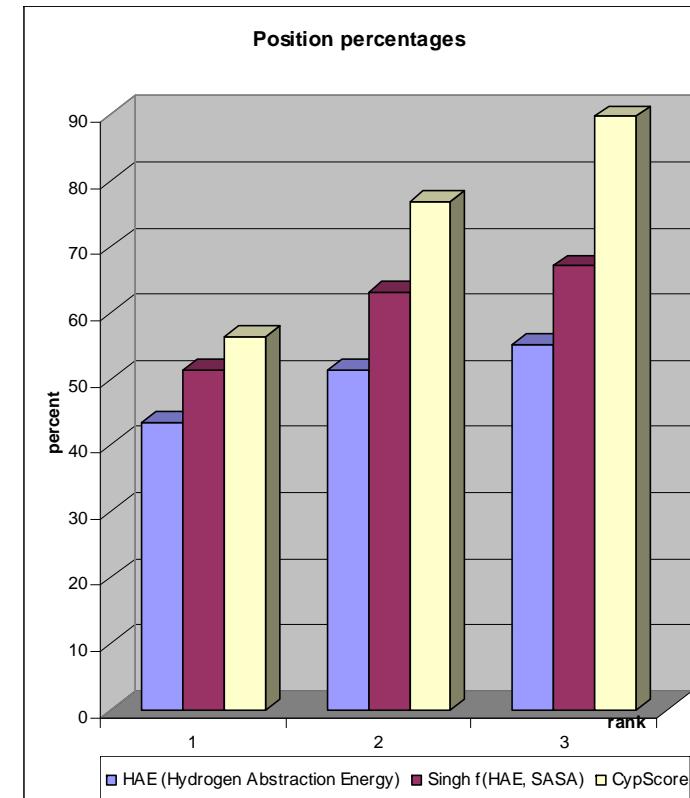
39 compounds from 12 in-house projects

- for these compounds the positions are clearly identified experimentally
- all types of reactions
- 90 % of metabolic positions found in 3 highest ranked positions



Bayer HealthCare

CypScore driven metabolic optimizations have been performed successfully at Bayer



$$r(\theta, \phi) = \sum_{l=0}^N \sum_{m=-l}^l a_{lm} y_{lm}(\theta, \phi)$$

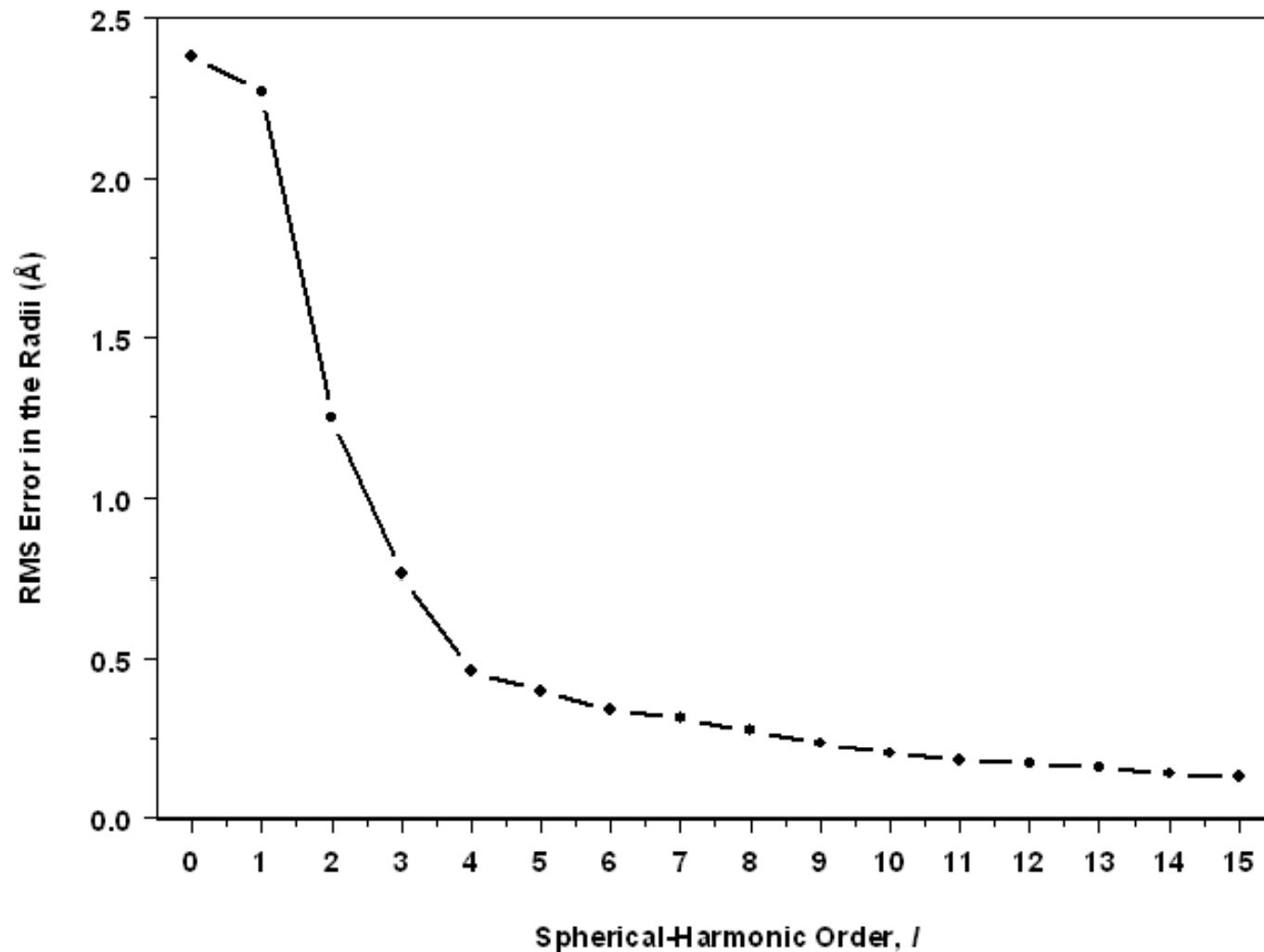
$$y_{lm}(\theta, \phi) = \{ N_{lm} P_{lm}(\cos \theta) \cos |m| \phi, m \geq 0 \}$$

and

$$y_{lm}(\theta, \phi) = \{ N_{lm} P_{lm}(\cos \theta) \sin |m| \phi, m < 0 \}$$

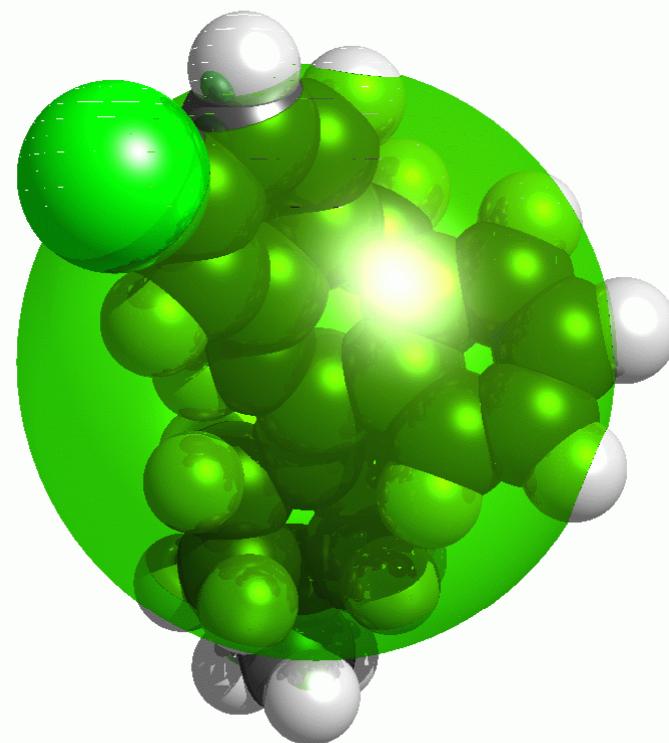
An Analytical, Variable Resolution, Complete Description of Static Molecules and Their Intermolecular Binding Properties, J.-H. Lin and T. Clark, J. Chem. Inf. Model., 2005, 45, 1010–1016.

Radial-Distance Fit



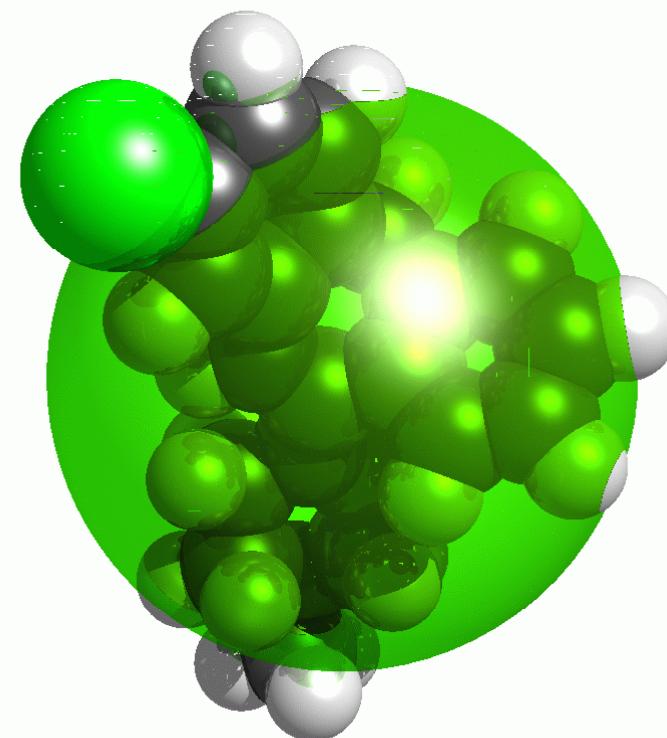
Shape Description

L=1



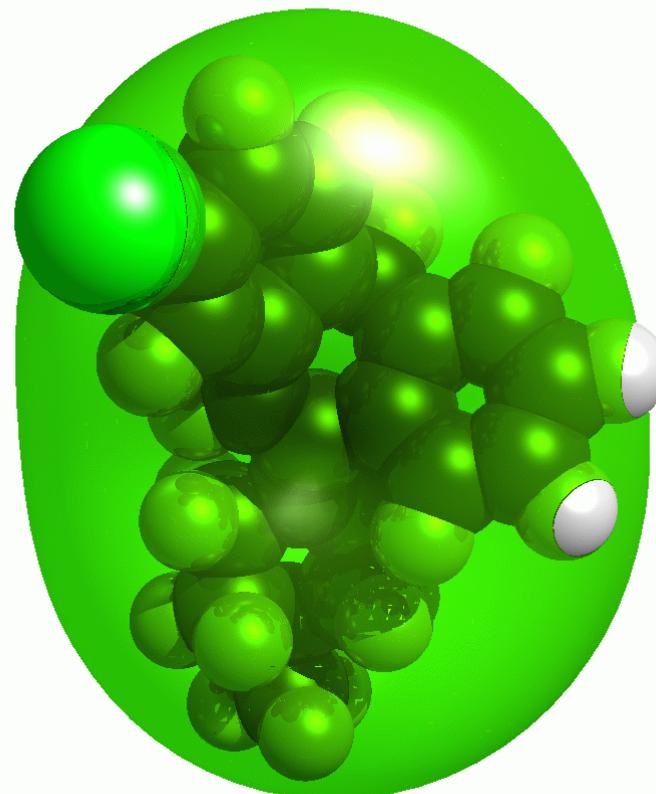
Shape Description

L=2



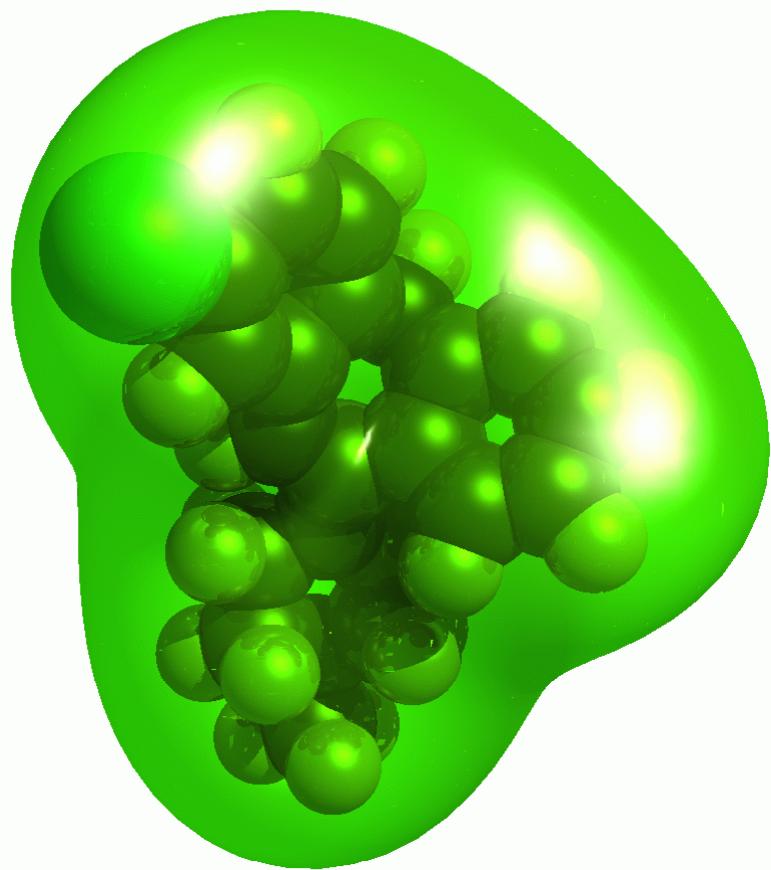
Shape Description

L=3



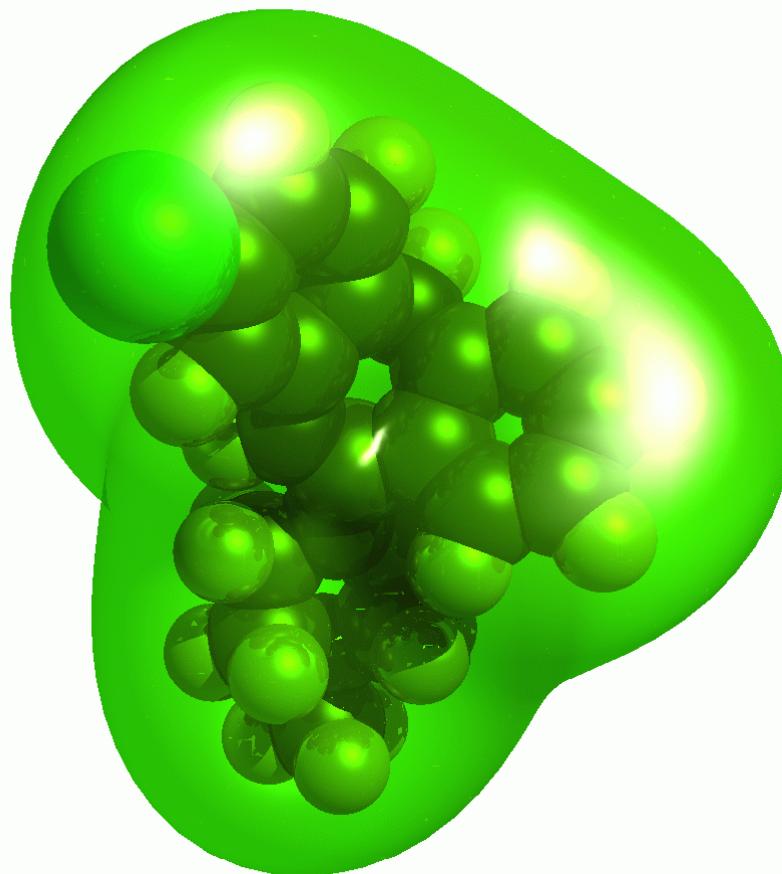
Shape Description

L=4



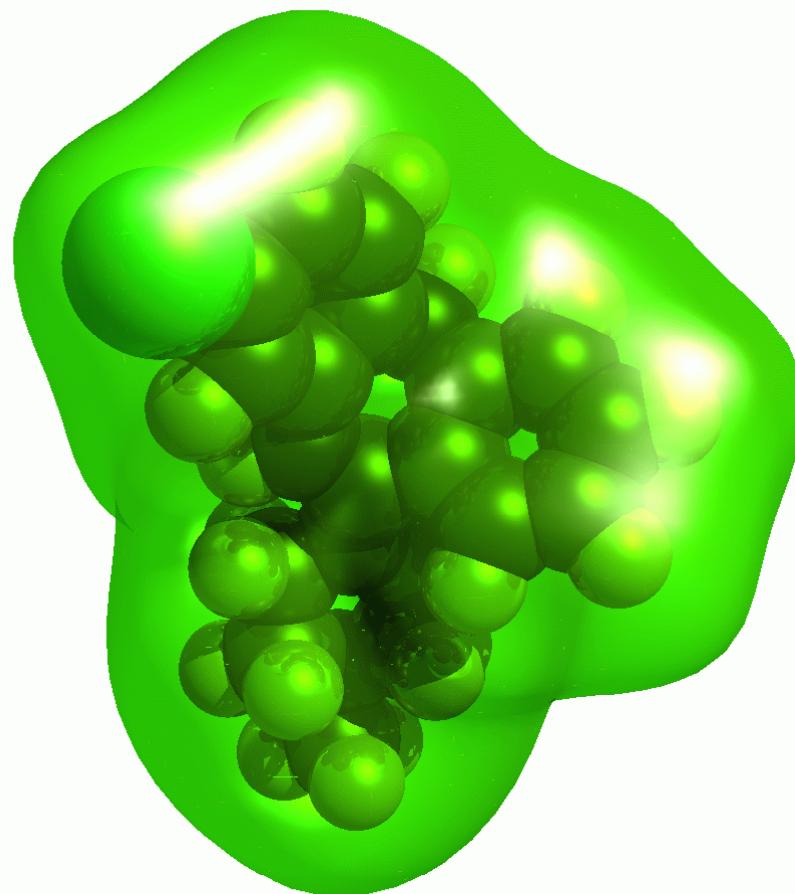
Shape Description

L=5



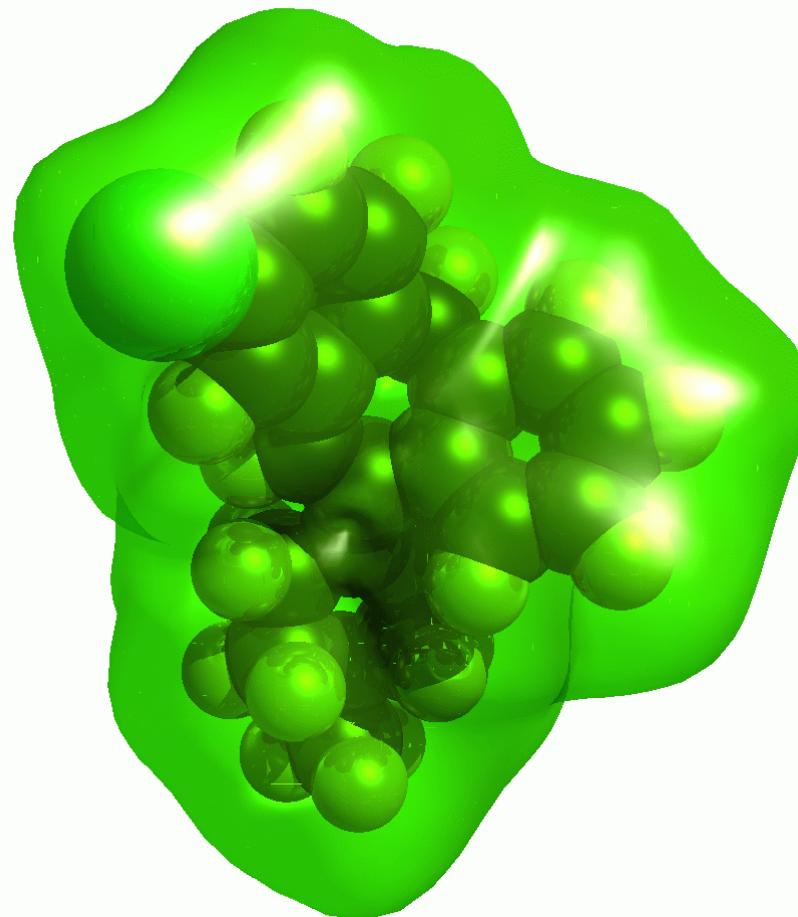
Shape Description

L=10



Shape Description

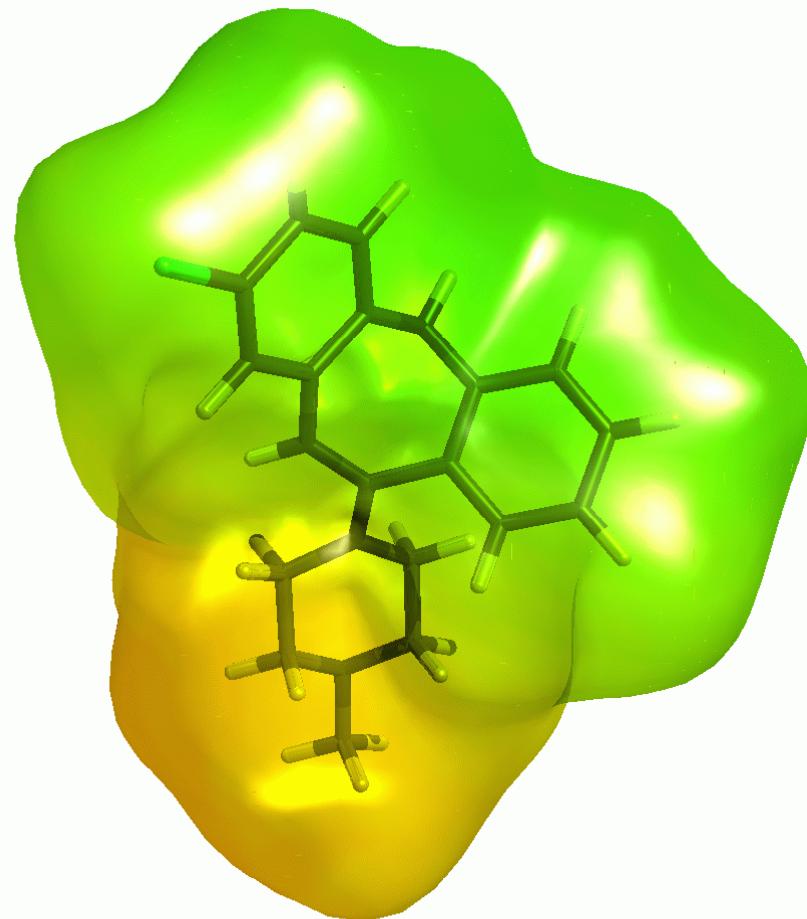
L=15



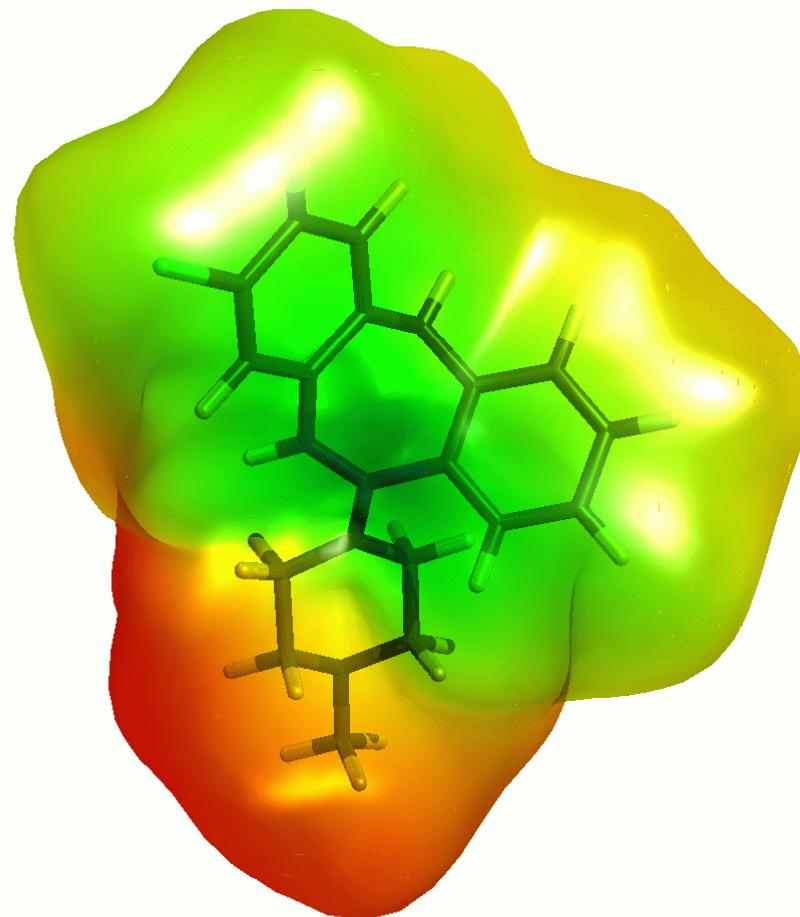
L=0



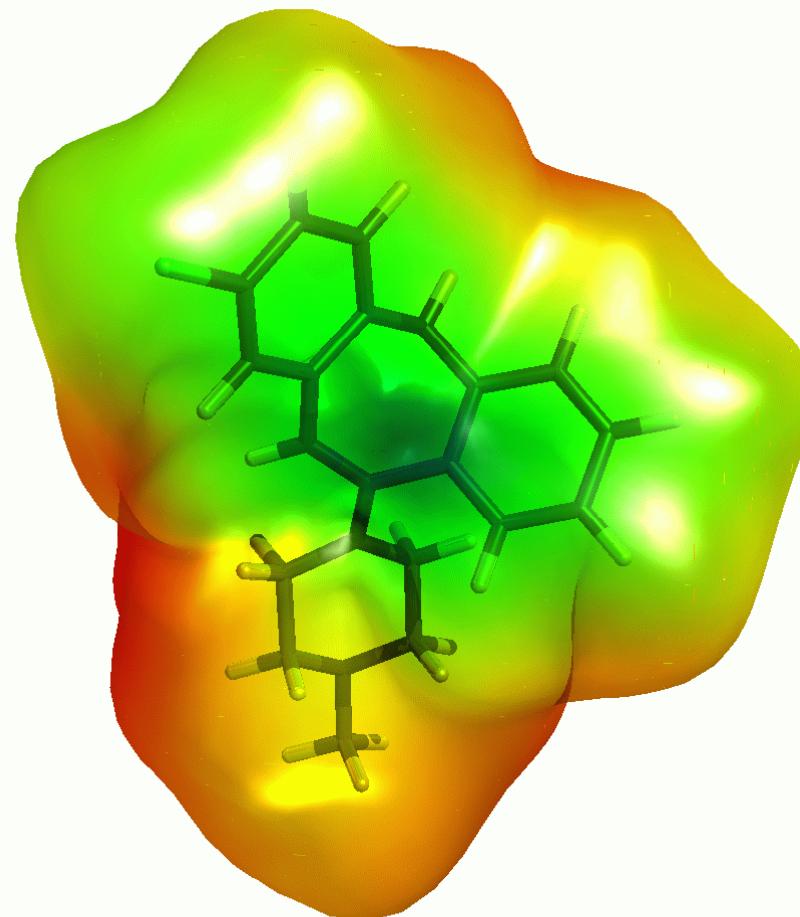
L=1



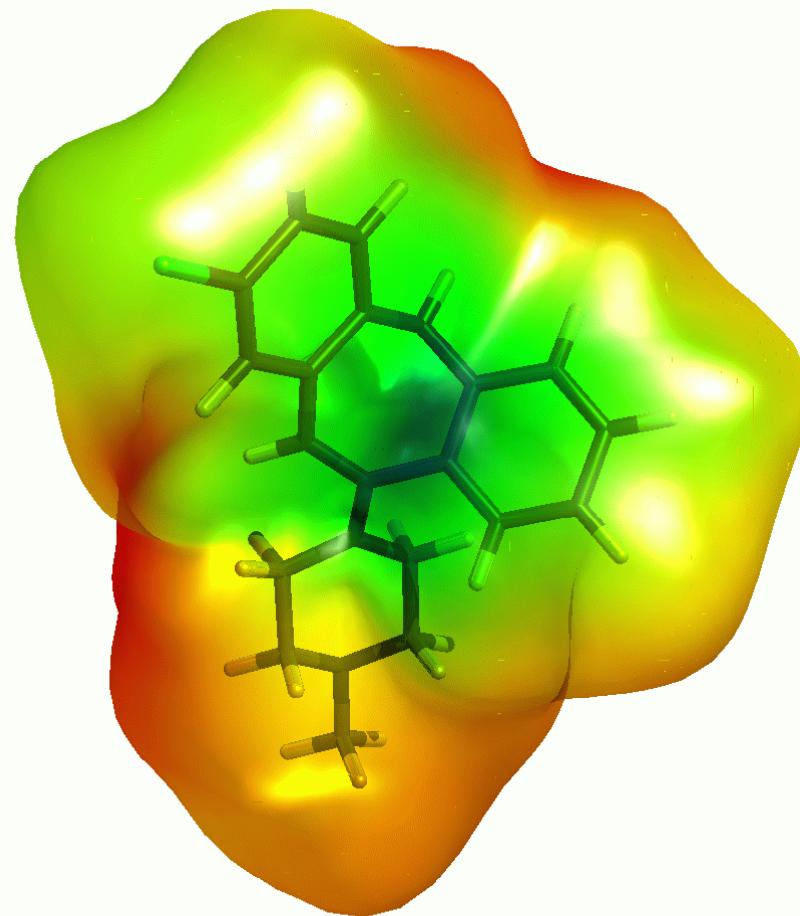
L=2



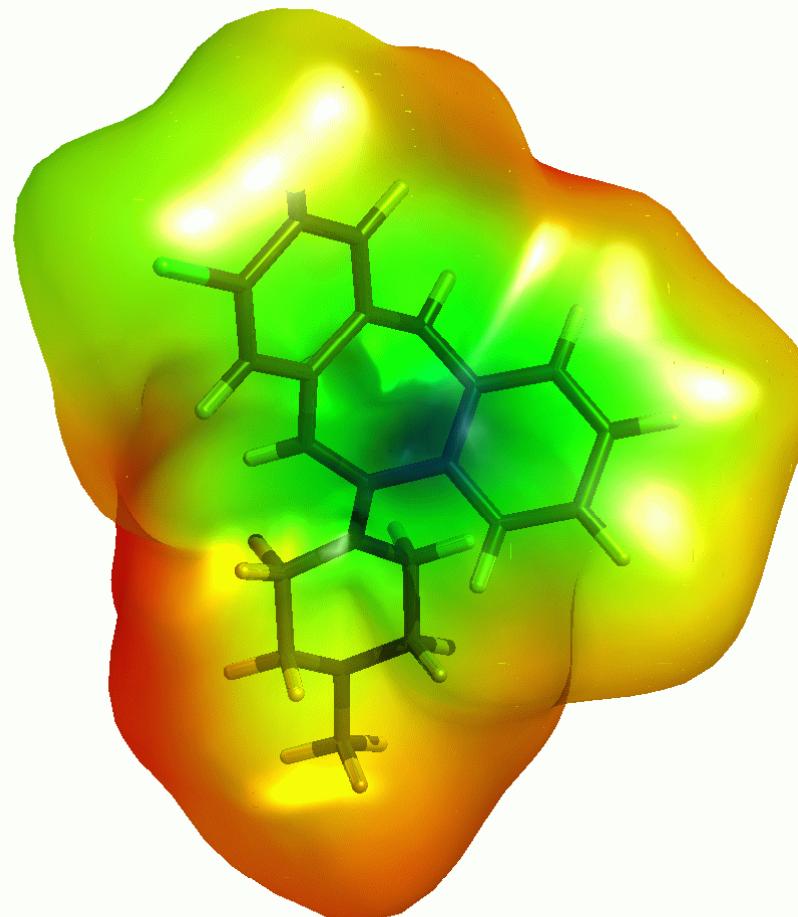
L=3



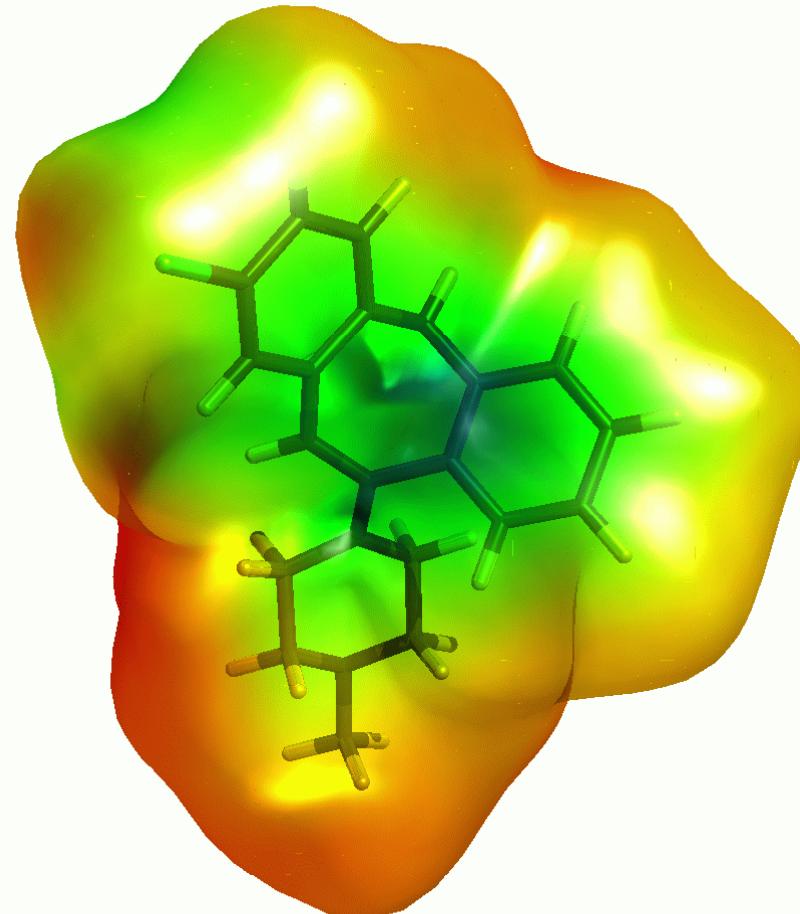
L=4



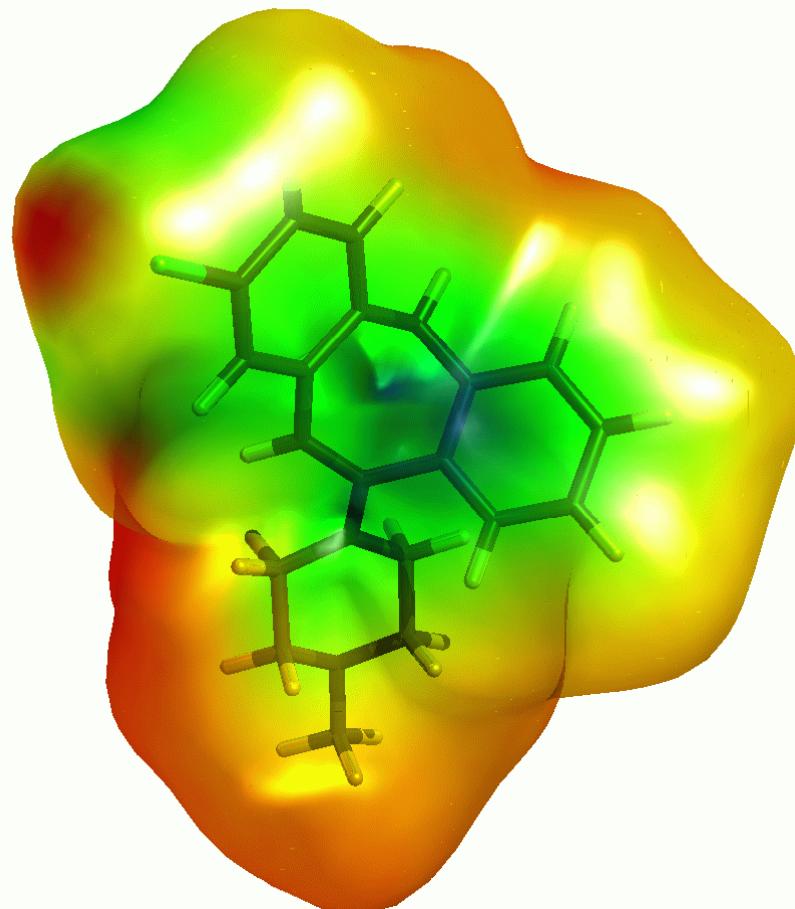
L=5



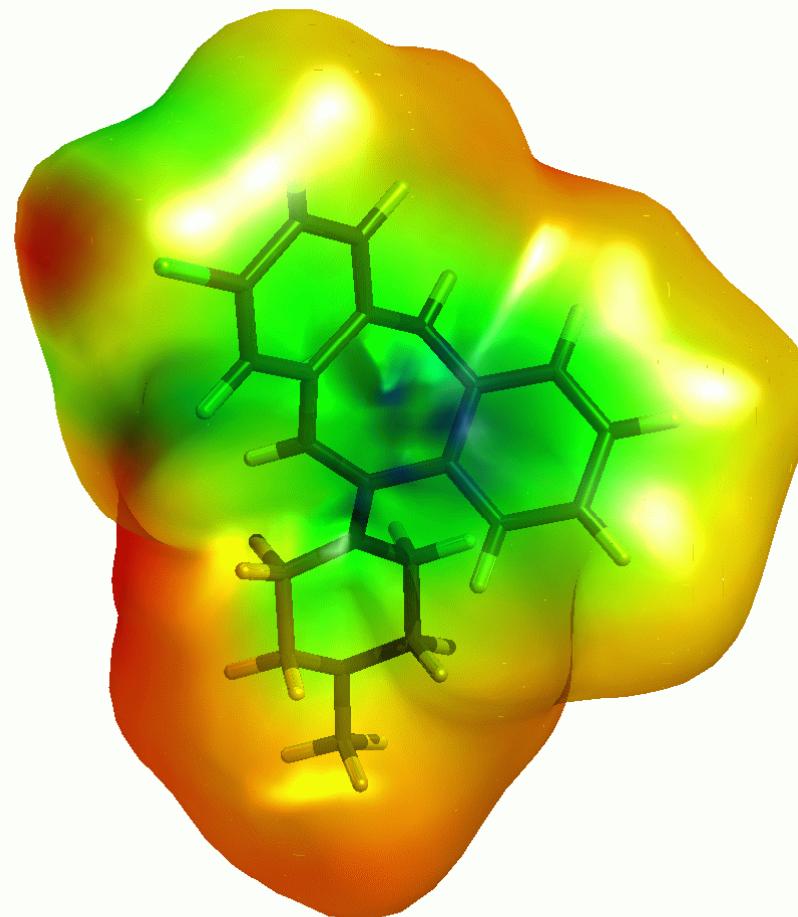
L=10



L=15



L=20



- “ParaLeap”

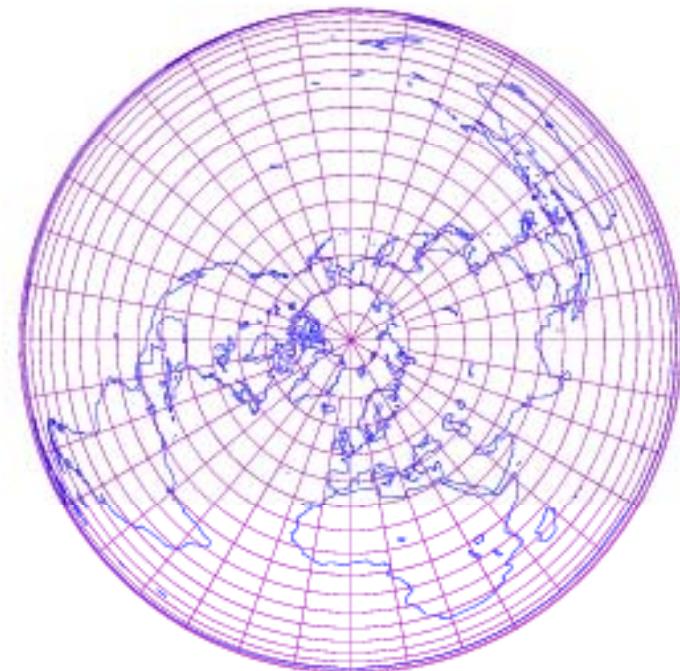
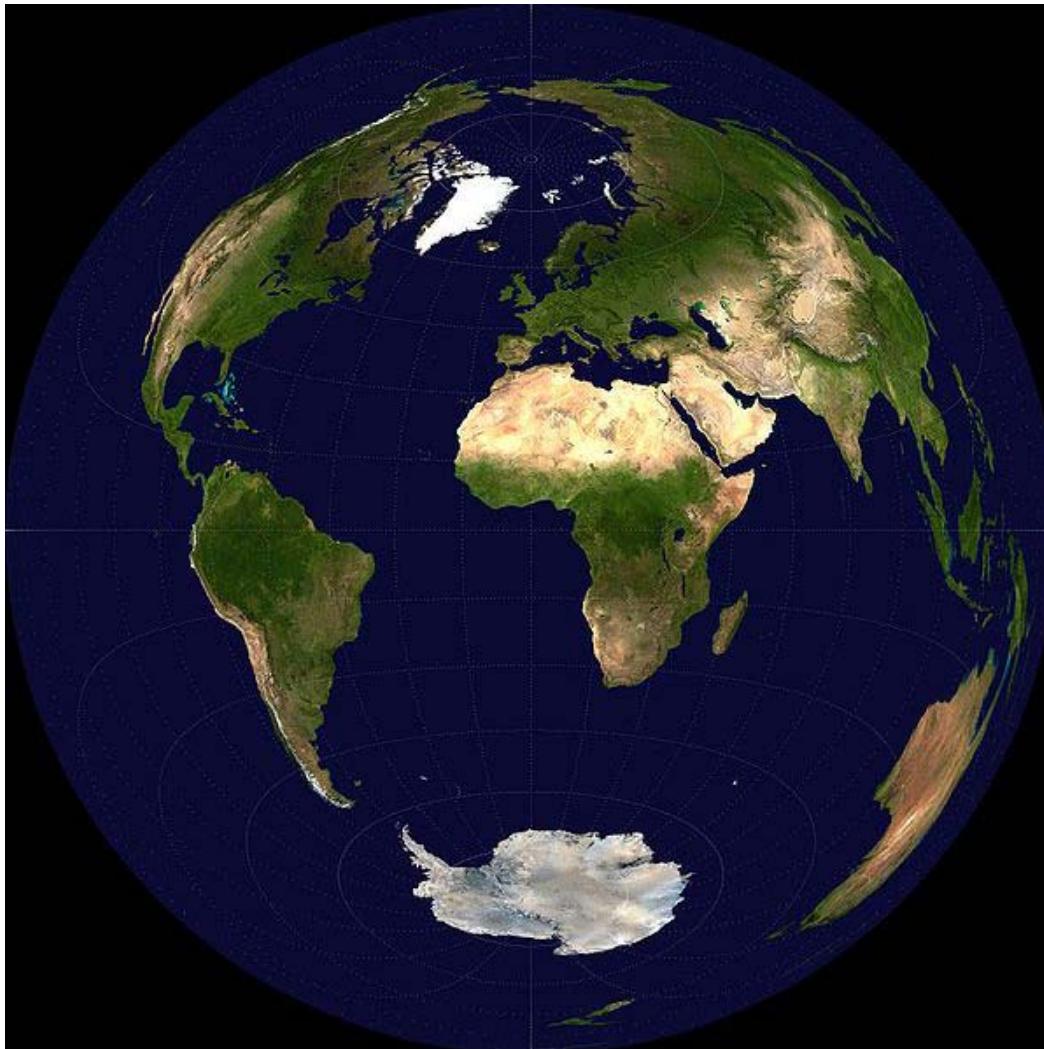
- High-throughput visualization of molecular similarity using Lambert azimuthal equal-area projections
- Joint development with Boehringer-Ingelheim, Biberach (Bernd Beck, Thomas Fox)
- Uses ParaSurf™ and ParaFit™ technology

Lambert Azimuthal Equal Area Projection

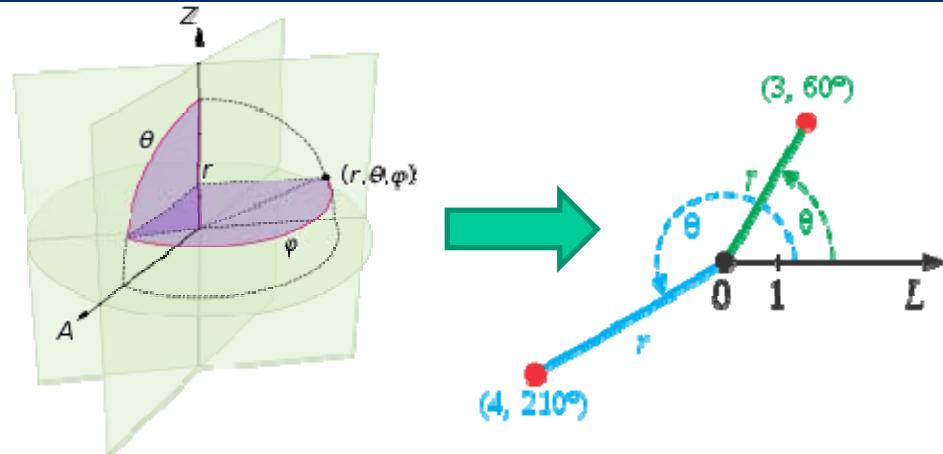
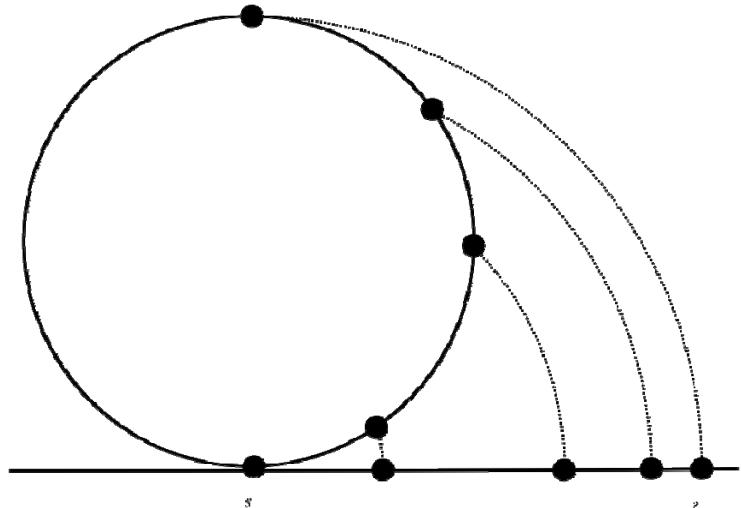


- **Johann Heinrich Lambert**
- 1728 – 1777
- Swiss mathematician, physicist and astronomer
- In 1764 Lambert was invited by Euler to join the Prussian Academy of Sciences in Berlin under the sponsorship of Frederick II of Prussia. He worked there until his death in 1777.
- Also known for the Lambert-Beer law

Lambert Projection of the Earth



Lambert Azimuthal Equal Area Projection



Spherical
coordinates,
 θ, ϕ

Polar
coordinates,
 R, Θ

$$(R, \Theta) = \left(2 \cos\left(\frac{\phi}{2}\right), \theta \right)$$

Lambert Azimuthal Equal Area Projection for Molecules

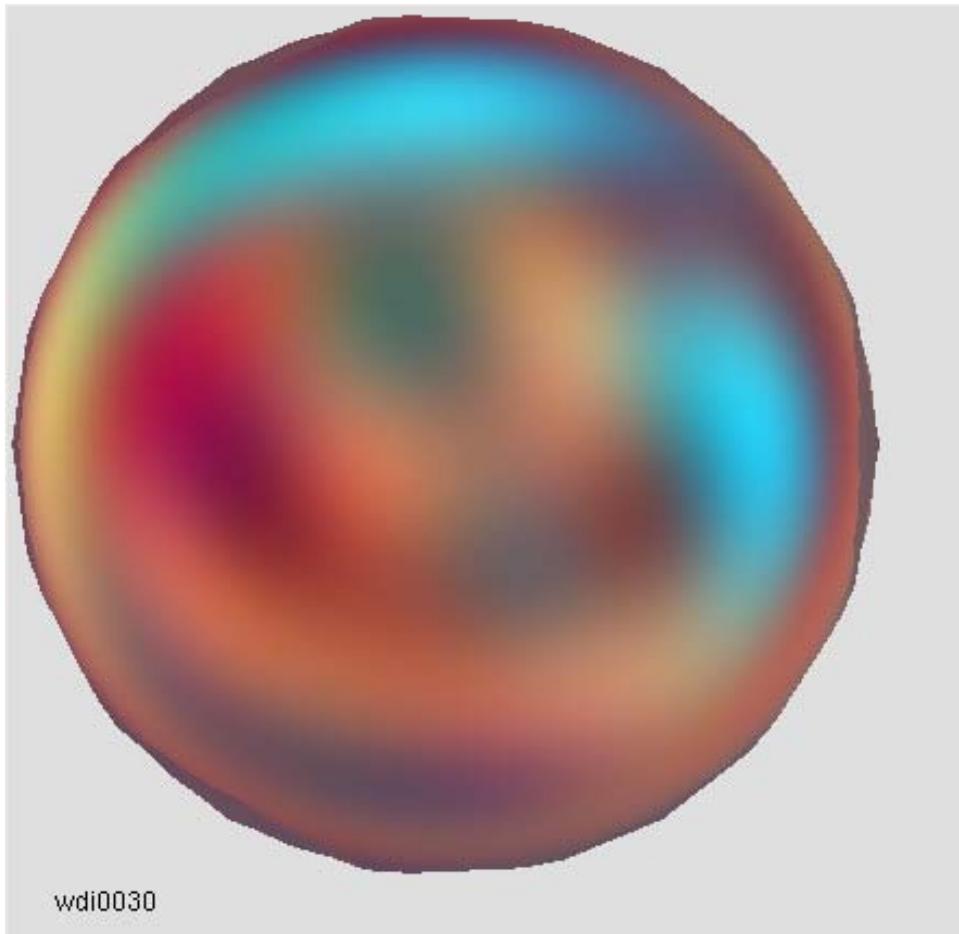
Lambert projection:

$$(R, \Theta) = \left(2 \cos\left(\frac{\phi}{2}\right), \theta \right)$$

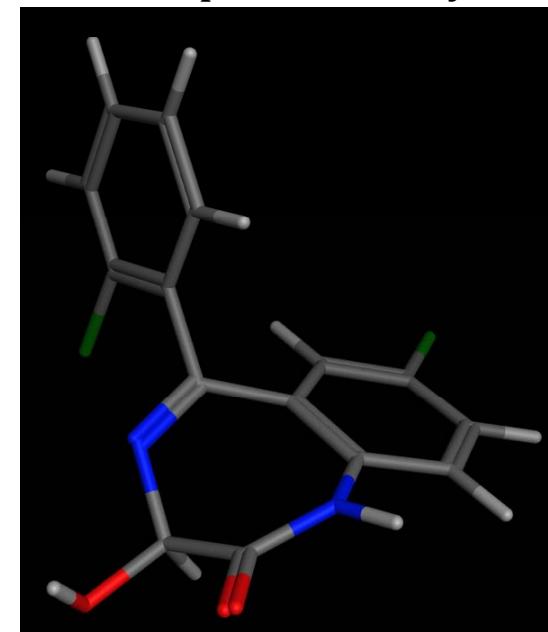
Spherical-harmonic
representation:

$$r(\theta, \phi) = \sum_{l=0}^N \sum_{m=-l}^l a_{lm} y_{lm}(\theta, \phi)$$

RGB-Encoding

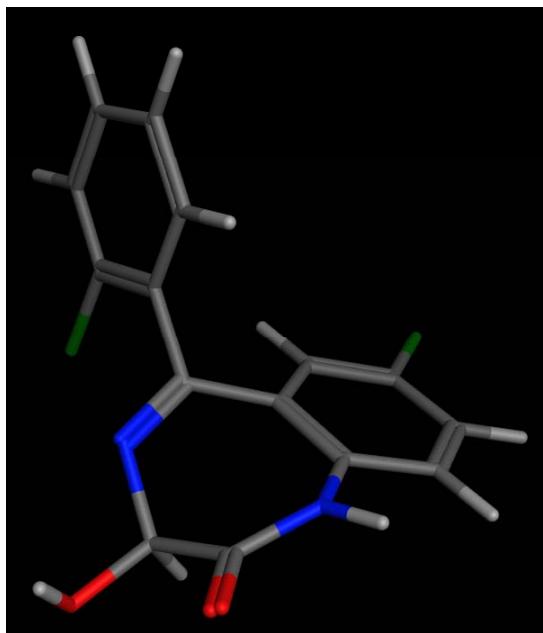


- Lambert projection
- Three properties encoded on the surface:
 - R=local ionization energy
 - G=local electron affinity
 - B=local polarizability



Similarities

Lead Compound



Similarities



Thank You

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