Intrinsic dimensionality of chemical space: Characterization and applications

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- US Environmental Protection Agency
- United States Air Force, Office of Scientific Research
- Agency for Toxic Substances and Disease Registry, Center for Disease Control and Prevention
- USDA
- •~ US 7.5 Million dollars since 1987



Property/ Activity/ Toxicity = f(S)



What is structure ?

- The structure of an assembled entity, e. g., a molecule can be looked upon as the relationship among its constituent parts
- A graph, G= [V, E] is an adequate representation of molecules where V is the set of atoms and E is the set of bonds or edges



Reality

Model Object

Mathematical Model

Method, Model and Matter, by Bunge



Representation of Molecular Structures by Graphs



Let V = (1, 2, 3, 4, 5)V x V = { $(1,1), (1,2), (1,3), (1,4), (1,5) \dots$ }

 $\mathbf{R}_1 = \{(1,5), (5,1), (2,5), (5,2), (3,5), (5,3), (4,5), (5,4)\}$

A binary relation on the set V



Methane

H

C

Н

H

H

Silicon tetrahydride

Η

Η

Si

H

Н

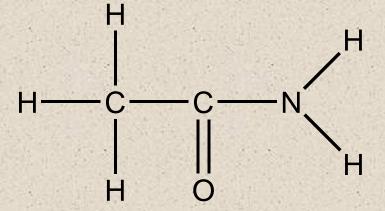


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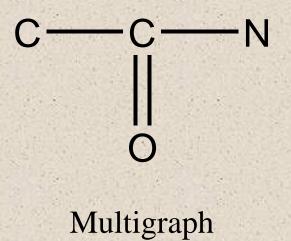
Neopentane

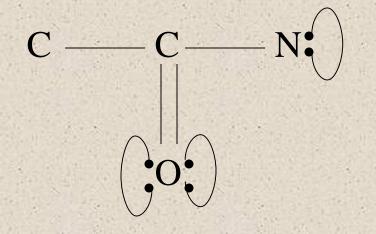
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Molecular Structure

Simple Graph





Pseudograph



Structure is a complex idea



Hierarchical Approach to Chemical Structure Representation

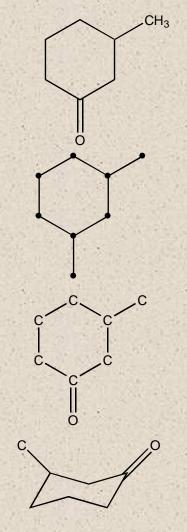
3-methylcyclohexanone

Topostructural Model

Topochemical Model

Geometrical Model

Quantum Chemical Model



 $H\Psi = E\Psi$

Chemist's representation of structure

Simple graph: Purely structural representation

Chemical graph: Contains chemical and valence information

3-Dimensional: Based on chemical graph

Based on principals of quantum mechanics



Characterization of Molecular Graphs Using TIs



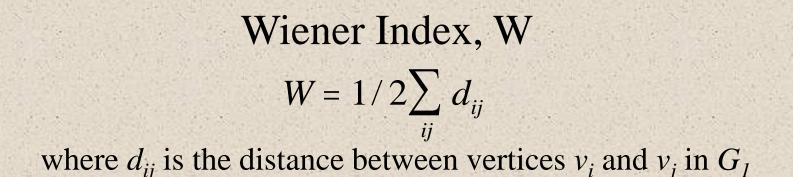
Molecular graphs can be characterized using numerical graph invariants or topological indices (TIs)

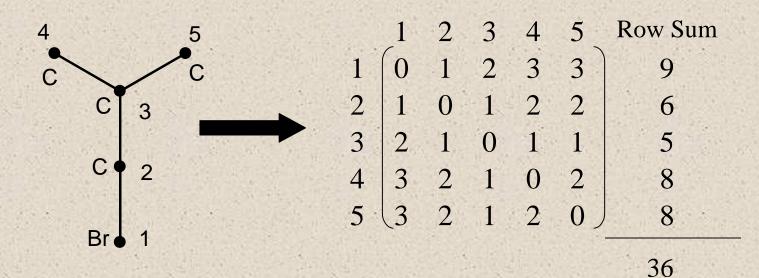
- Simple graph
- Multigraph
- Weighted graphs

Molecular Graph

Molecular descriptor



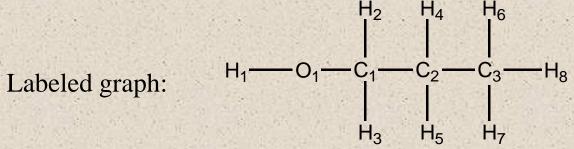




W = 36 / 2= 18



Calculation of IC, SIC & CIC



First Order Neighborhoods:

Subsets: I II III IV V VI

 (H_1) (H_2-H_8) (O_1) (C_1) (C_2) (C_3)

Probability:

1/12 7/12 1/12 1/12 1/12 1/12 (\mathbf{p}_i)

(Basak, Roy, Magnuson, and Harriss)



Equivalence Relation

- Can partition the vertex set, V(G), into disjoint subsets based on topological neighborhoods of vertices up to the *r*th order neighbors and provide indices of neighborhood complexity
- Is reflexive, symmetric, and transitive



Measures of Complexity and Redundancy

Information Content (IC₁)

 $IC_{1} = -\sum p_{i} \log_{2} p_{i}$ = 5 × ½ × log_2 ½ + ½ × log_2 ½ = 1.950*bits*

 $S/C_1 = IC_1/\log_2 12 = 0.544$

 $C/C_1 = \log_2 12 - IC_1 = 1.635 bits$

Basak, Roy and Ghosh, *Proc. 2nd Intl. Conf. Math.Modelling*, pp. 851-856, **1979**.
Roy, Basak, Harriss and Magnuson, *Mathl. Modelling Sci. Technol.*, pp. 745-750, 1984.
Basak and Magnuson, *Arzneim. Forsch./Drug Res.*, **33**, 501-503, 1983.
Raychaudhury, Ray, Ghosh, Roy and Basak, *J.Comput. Chem.*, **5**, 581-588, 1984.



QSAR and Molecular Descriptors



Strategies

- Laboratory experiments
- Property-property correlations $[P_1 = f(P_2)^a]$
- Structure-property correlations $[P = f(S)^b]$
 - QSAR / QSPR
 - Molecular Similarity

^a Experimentally determined^b Calculated



POLLY The Upjohn Company Glaxo (USA) US Army NIH, NINDS US Environmental Protection Agency

> **APProbe** The Upjohn Company Glaxo (USA)



Data Reduction via Principal Components Analysis 3,692 chemicals; 90 diverse TIs

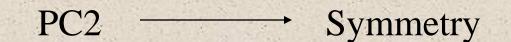
Principal	and the second second	Percent of	Cumulative
Component (PC)	Eigenvalue	variance	percent
1	39.6	44.0	44.0
2	14.6	16.2	60.2
3	9.9	11.0	71.2
4	6.4	7.1	78.3
5	3.3	3.7	82.0
6	3.2	3.5	85.5
7	1.9	2.1	87.6
8	1.8	1.9	89.5
9	1.5	1.7	91.2
10	1.2	1.3	92.5

(Basak, Magnusson, Niemi, Regal, and Veith, 1987)

- 10 PCs with Eigenvalues greater than 1
- First 10 PCs explain 92% of the variance within the data
- First 4 PCs account for 78% of the variance within the data



PC1 \longrightarrow Size



PC3 → Branching

PC4 Cyclicity

S.C. Basak, V.R. Magnuson, G.J. Niemi, R.R. Regal Discrete Applied Mathematics 19 (1988) 17 No. 19



Topological Indices: Their Nature and Mutual Relatedness

Subhash C. Basak, Alexandru T. Balaban, Gregory D. Grunwald, and Brian D. Gute Natural Resources Research Institute, University of Minnesota--Duluth, Duluth, Minnesota 55811, and Organic Chemistry Department, Polytechnic University Bucharest, Splaiul Independentei 313, 77206 Bucharest, Romania

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Hierarchical QSAR



Hierarchical QSAR

Biodescriptors

Relativistic ab initio

Solvation state ab initio

In vaccuo ab initio

In vaccuo semi-empirical

Geometrical / Chirality Parameters

Topochemical Indices

Topostructural Indices

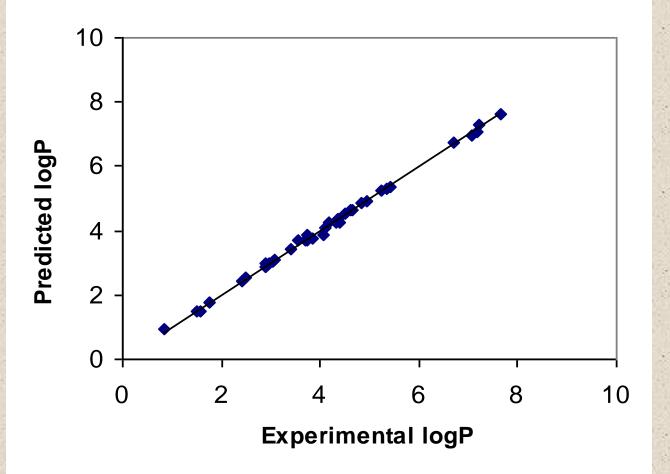
Complexity





Cost

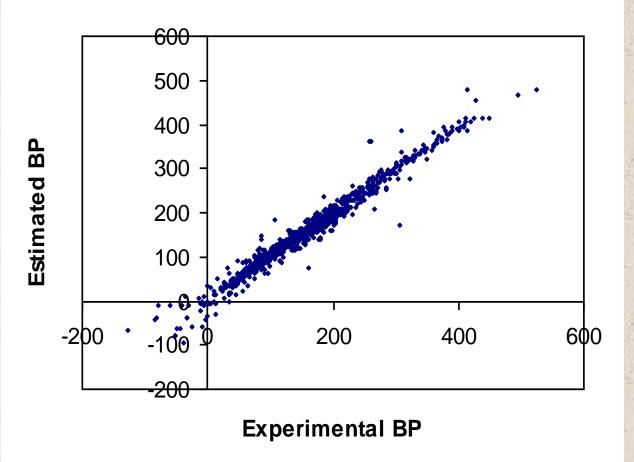
Blood:Air Partition Coefficient Model (TC) Developed on 39 Diverse Chemicals



S. C. Basak, D. Mills, H. A. El-Masri, M. M. Mumtzaz, and D. Hawkins Environ. Toxicol. Pharmacol., 16, 45–55 (2004). RESEARCH INSTITUTE

Normal Boiling Point for 1015 Diverse Chemicals

 $n = 1015, R^2 = 0.97, s = 15.7, F = 4014$



Basak, S. C. and Mills, D. MATCH, 2001, 44, 15-30.



Graph Theoretic vs Quantum Chemical Descriptors for the Prediction of Vapor Pressure

- 121 chlorinated chemicals
- Supercooled liquid VP at 298K
- Graph theoretic descriptors: $q^2 = 0.988$
- Polarizability (DFT, B3LYP): $q^2 = 0.974$

Basak, S. C.; Mills, D. SAR QSAR Environ. Res., in press.



Improvement in Predictive Models Upon Inclusion of Quantum Chemical Descriptors?

Description of Data Set and Property/Activity	Improvement
Acute toxicity of benzene derivatives	Minimal
Dermal penetration of PAHs	None
Mutagenicity of aromatic and heteroaromatic amines	None
Mutagenicity of 508 diverse compounds	None
Vapor pressure of 469 diverse compounds	None
Cellular toxicity of halocarbons	Minimal
Mosquito repellency of aminoamides	None
Mosquito repellency of DEET-related compounds	None
Blood and tissue:air partition coefficient for rat and human (blood, fat, brain, liver, muscle, and kidney)	None
Aryl hydrocarbon receptor binding affinity of dibenzofurans	None



Basak, S. C.; Mills, D.; Mumtaz, M. M.; Balasubramanian, K. Use of topological indices in predicting aryl hydrocarbon receptor binding potency of dibenzofurans: A hierarchical QSAR approach. *Indian J. Chem.*, 2003, 42A, 1385-1391.



Gute, B. D.; Grunwald, G. D.; Basak, S. C. Prediction of the dermal penetration of polycyclic aromatic hydrocarbons (PAHs): A hierarchical QSAR approach. SAR QSAR Environ. Res., 1999, 10, 1-15.



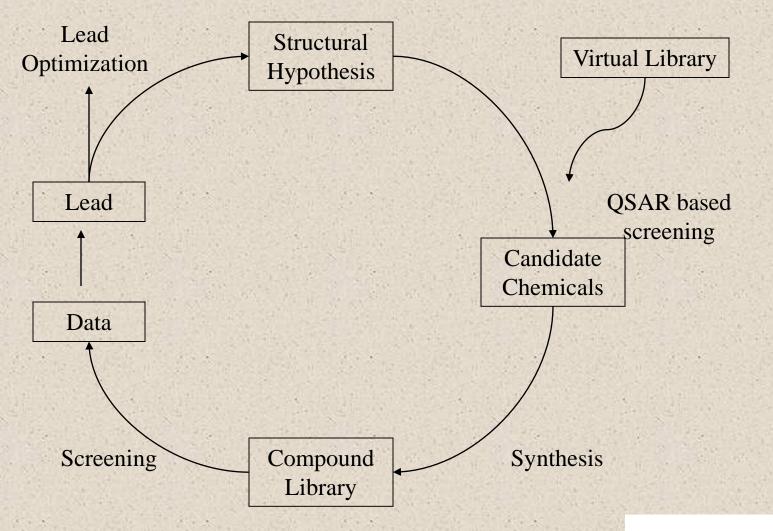
- Basak, S. C.; Gute, B. D.; Grunwald, G. D. Use of topostructural, topochemical, and geometric parameters in the prediction of vapor pressure: A hierarchical approach. J. Chem. Inf. Comput. Sci., 1997, 37, 651-655.
- Basak, S. C.; Grunwald, G. D.; Host, G. E.; Niemi, G. J.; Bradbury, S. P. A comparative study of molecular similarity, statistical, and neural network methods for predicting toxic modes of action of chemicals. *Environ. Toxicol. Chem.*, 1998, 17, 1056-1064.



Basak, S. C.; Gute, B. D.; Drewes, L. R. Predicting blood-brain transport of drugs: A computational approach. Pharm. Res., 1996, 13, 775-778. Mushrush, G. W.; Basak, S. C.; Slone, J. E.; Beal, E. J.; Basu, S.; Stalick, W. M.; Hardy, D. R. Computational study of the environmental fate of selected aircraft fuel system deicing compounds. J. Environ. Sci. Health, 1997, A32, 2201-2211.

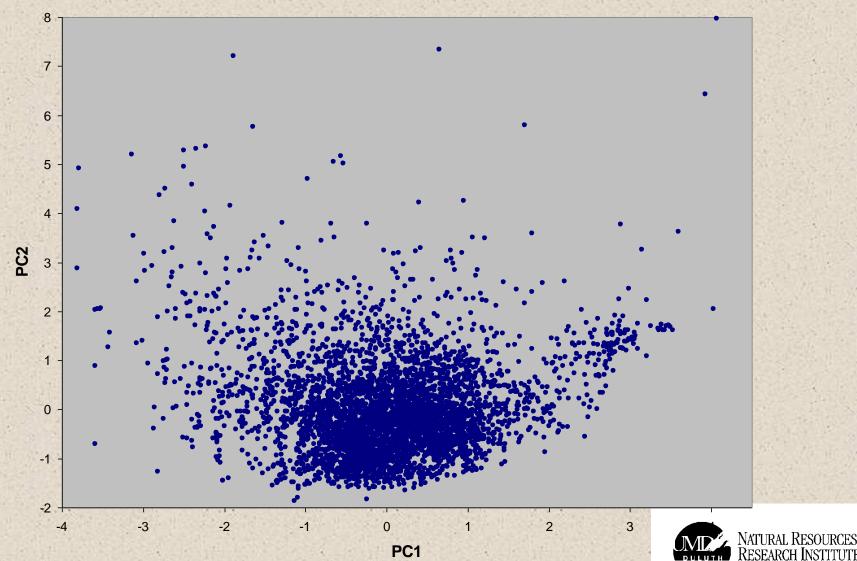


Combinatorial Chemistry & QSAR





PC_1 vs. PC_2 for **4,453** chemicals based on the correlation matrix of 98 variables

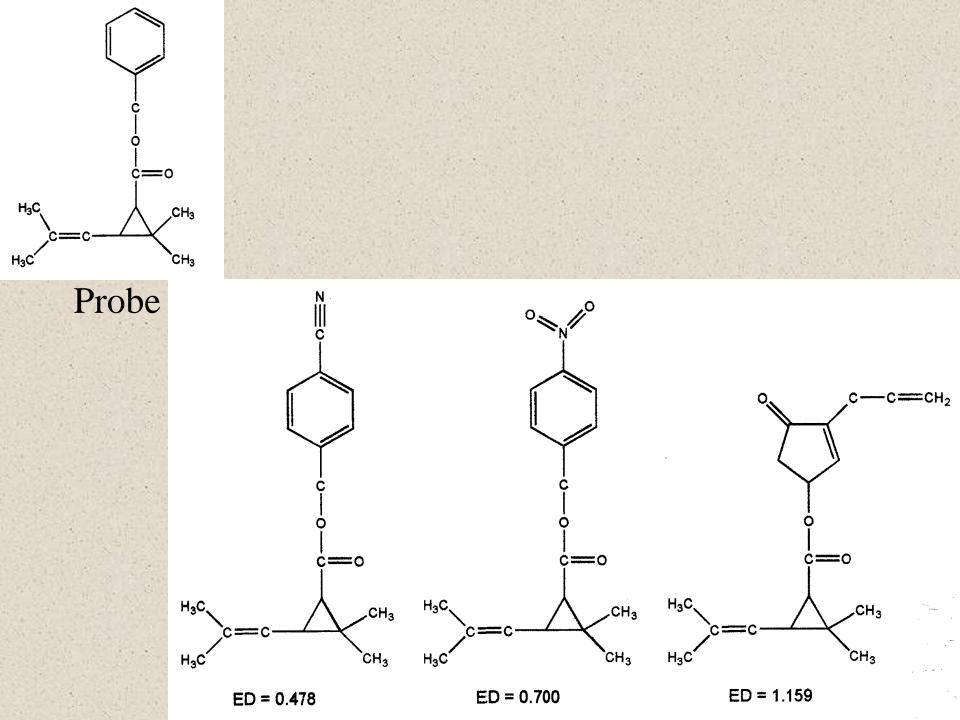


Euclidean Distance

ED_{ij} =
$$[\sum_{k=1}^{n} (D_{ik} - D_{jk})^2]^{1/2}$$

where n = the number of dimensions and D_{ik} and D_{jk} equal the data values of the kth dimension for chemicals i and j, respectively.





K-Neighbor Selection and Property Estimation

Intermolecular similarity of chemicals in each set was quantified using 3 to 5 distinct similarity methods.

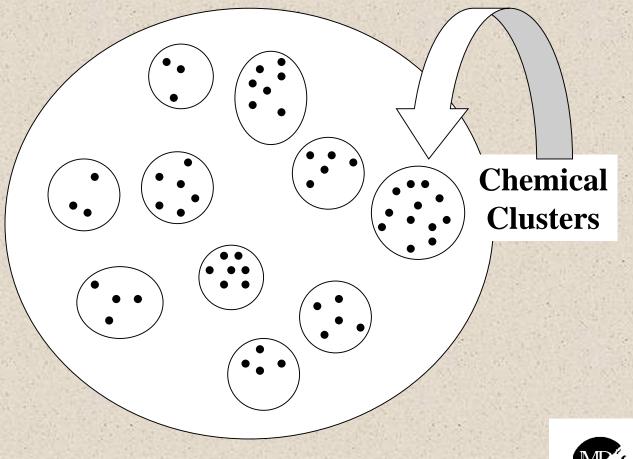
For each chemical, *K*-nearest neighbors were determined for K = 1, 2, ..., 10, 15, 20, 25.

Estimated property values are determined as the mean observed value of the *K*-nearest neighbors.

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	k	r	se	
	5	0.958	27.6	
KNN Estimation of	10	0.956	28.5	
Boiling Points for 1037	15	0.953	30.0	
Diverse Chemicals	20	0.949	31.5	
	25	0.946	32.9	
	50	CJAB7	NATURA L REGOURCES RESEARCH INSTITUT	

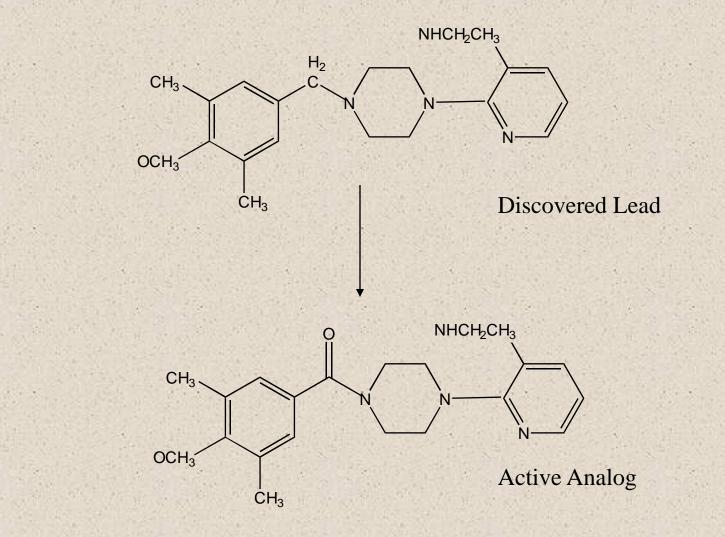
Structure Space

Chemical Space





HIV-I RT Inhibitor Discovered by Similarity Search



Discovered by Upjohn-Pharmacia



JP-8

(~230 chemicals, ~2,000 chemicals)

- Skin toxicity
- Immunosuppression
- Systemic toxicity

230! or 2,000! mixtures to be tested



230! or 2,000! mixtures to be tested

230

Clustered using TIs

10 - 15 clusters

230! → 15!



Quo Vadimus?



- Chemical structure
- DNA sequence, genomics
- Proteomics pattern

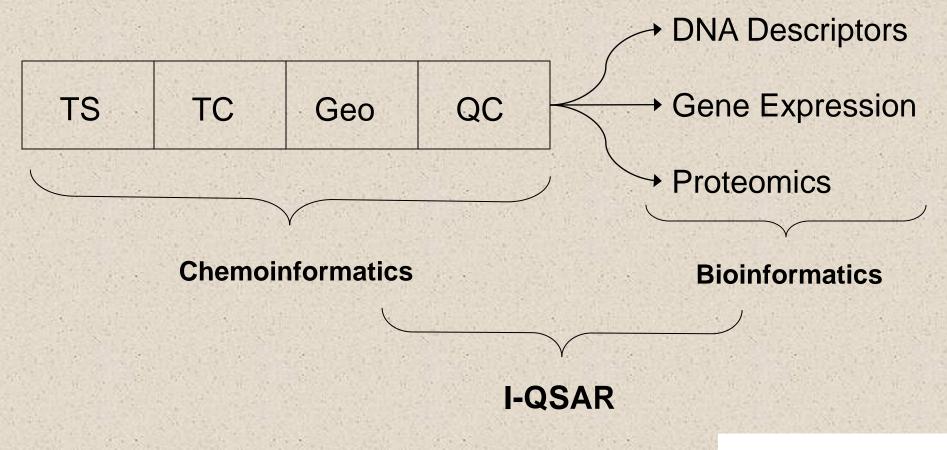
Matrix

Distance Adjacency D/D

Matrix invariants (Structural invariants or descriptors)



Integrated QSAR (I-QSAR)





Chemo-bioinformatics Guest editorial JCIM, 46, 1, 2006 Discrete mathematical chemistry has made important advances in the past twenty five years. This has been fueled primarily by two factors: a) formulation of new concepts and b) easy access to high speed computers. Methods developed in this field have found applications in pharmaceutical drug design and hazard assessment of environmental pollutants.



Chemo-bioinformatics Guest editorial JCIM, 46, 1, 2006 Interestingly, discrete mathematical concepts, originally developed for the characterization of chemical systems, are being extended to deal with explosion of data in the "omics" science, viz., genomics, proteomics, etc. A few of the papers from the Fourth Indo-US Workshop published in this issue of JCIM are outstanding examples of this expanding chemo-bioinformatics continuum.



The enormous landscape

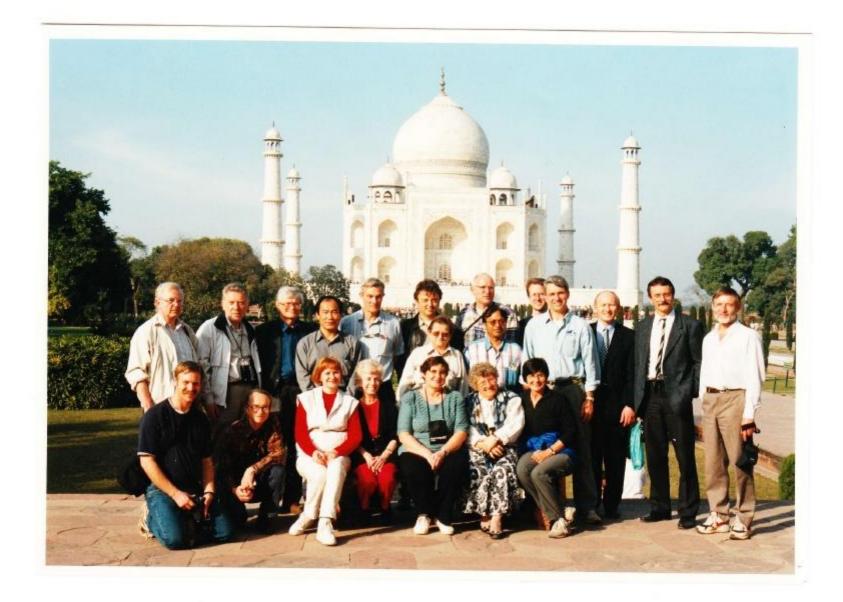
Even the same atoms of the same element, when they exist in different molecules, exhibit different behaviours. The chemical symbol H even seems to signify atoms of a completely different nature. In chemistry, this terrible individuality should never be avoided by "averaging," and, moreover, innumerable combinations of such atoms form the subject of chemical research" K. Fukui, Nobel Lecture, 1981



Isomorphic laws in Science Not only are general aspects and viewpoints are alike in different fields of science; we find also formally identical or isomorphic laws in completely different fields

L. von Bertalanffy, *British Journal for the philosophy of science*, 1950







In Santa Barbara, 1933

Life is like riding a bicycle. To keep your balance you must keep moving. —Albert Einstein, in a letter to his son Eduard, February 5, 1930¹



