



Ligand-Based Drug Design











1. Overview on LBDD

- A. Definition and Methods
- B. Pharmacophore Approach
- C.QSAR
- D.3-D QSAR

2. Application (and demonstration) of 3-D QSAR to medicinal chemistry

Drug Discovery Process





Introduction to Ligand-Based Drug Design



chair 1: blue H's axial Conformational Search Molecular Docking

Simulate Annealing

Ab Initio QM



Rino Ragno: Computational Medicinal Chemistry Applications to Epigenetic Targets Inhibitors

Introduction: Computational Techniques



Drug Design = Computational & Synthesis Tandem



Rino Ragno: Computational Medicinal Chemistry Applications to Epigenetic Targets Inhibitors

09-09-2015

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Introduction: Computational Techniques



Computational Medicinal Chemistry





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Introduction: Computational Techniques





In the absence of three-dimensional (3-D) structures of potential drug targets, ligand-based drug design is one of the popular approaches for drug discovery and lead optimization. 3-D structure activity relationships (3-D QSAR) and pharmacophore modeling are the most important and widely used tools in ligandbased drug design that can provide crucial insights into the nature of the interactions between drug target and ligand molecule and provide predictive models suitable for lead compound optimization.



"A pharmacophore is the ensemble of steric and electronic features that is necessary to ensure the optimal supramolecular interactions with a specific biological target and to trigger (or block) its biological response."

C.-G. Wermuth et al., Pure Appl. Chem. 1998, 70: 1129-1143

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Definition by IUPAC



A **pharmacophore** is the ensemble of steric and electronic features that is necessary to ensure the optimal supramolecular interactions with a specific biological target structure and to trigger (or to block) its biological response.

A **pharmacophore** does not represent a real molecule or a real association of functional groups, but a purely abstract concept that accounts for the common molecular interaction capacities of a group of compounds towards their target structure.

The **pharmacophore** can be considered as the largest common denominator shared by a set of active molecules. This definition discards a misuse often found in the **medicinal chemistry** literature which consists of naming as **pharmacophores** simple chemical functionalities such as guanidines, sulfonamides or dihydroimidazoles (formerly imidazolines), or typical structural skeletons such as flavones, phenothiazines, prostaglandins or steroids.



Interaction sites used to define a pharmacophore,

- H-bonding,
- hydrophobic
- Electrostatic

Defined by atoms, ring centers and virtual points.



The First Paper on Pharmacophoric Approach



The Conformational Parameter in Drug Design: The Active Analog Approach

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Introduction to Ligand-Based Drug Design

BDZ Cook Model





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Modern Pharmacophore Model







QSAR date back to the 19th century

In 1863, A.F.A. Cros at the University of Strasbourg observed that toxicity of alcohols to mammals increased as the water solubility of the alcohols decreased

In the 1890's, Hans Horst Meyer of the University of Marburg and Charles Ernest Overton of the University of Zurich, working independently, noted that the toxicity of organic compounds depended on their lipophilicity.

Louis Hammett (1938) Electronic Parameters (Sigma-Rho) Robert W. Taft (1952) Steric Parameters (Es) Corvin Hansch (1964) Hydrophobic Parameter (LogP - π)

The Hansch Equation





The Hansch Equation

$$\delta_{\mathbf{X}} \log k_i = \delta_{\mathbf{X}} \Delta G_{\text{hydrophobic}} + \delta_{\mathbf{X}} \Delta G_{\text{electronic}} + \delta_{\mathbf{X}} \Delta G_{\text{steric}}$$

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The first QSAR equations were based on the observation that partition coefficients, as expressed by log *P* values, are to some extent, correlated to certain biological endpoints.

$$log (1C) = k1 log P + k2\sigma + k3$$

Conc. of compound required to produce a standard response in a given t

Logarithm of the molecule's partition coefficient (1-octanol/water) Hammet Parameter (molecule's electronic characteristics)



'60: QSAR Golden Age



April 20, 1964 Correlation of Biological Activity and Chemical Structure

1617

[CONTRIBUTION FROM THE DEPARTMENT OF CHEMISTRY, POMONA COLLEGE, CLAREMONT, CALIFORNIA]

$\rho - \sigma - \pi$ Analysis. A Method for the Correlation of Biological Activity and Chemical Structure

By Corwin Hansch and Toshio Fujita¹

Received August 19, 1963

Using the substituent constant, σ , and a substituent constant, π , defined as $\pi = \log P_X - \log P_H (P_H \text{ is the partition coefficient of a parent compound and <math>P_X$ that of a derivative), regression analyses have been made of the effect of substituents on the biological activity of benzoic acids on mosquito larvae, phenols on gram-positive and gram-negative bacteria, phenyl ethyl phosphate insecticides on houseflies, thyroxine derivatives on rodents, diethylaminoethyl benzoates on guinea pigs, and carcinogenic compounds on mice.

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A Mathematical Contribution to Structure-Activity Studies

Spencer M. Free, Jr., and James W. Wilson

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where R was H or CH_8 ; X was Br, Cl, or NO_2 ; and Y was NO_2 , NH_2 , or CH_8CONH .

BIOLOGICAL ACTIVITY OF TEN TETRACYCLINES

Compound identification									Bio-
Com-	*	-R		-X-				Y	logical
pound	Н	CH_{3}	NO_2	C1	Br	NO_2	$\rm NH_2$	CH3CONH	activity
III	1		1			ţ			60
[V	1			1		1			21
V	1				1	1			1.5
VI	1			1			i		525
VII	1				1		1		320
VIII	Ł		I				I		275
IX		1	1				1		160
X		1	1					1	1.5
XI		ł			1		1		140
XII		i			1			1	75

CONTRIBUTION OF STRUCTURAL CHANGES^a

			iain posi	itions		
R		X		$\tilde{\lambda}$		
$a[\mathbf{H}]$	75	b [C1]	84	c [NH ₂]	123	
$a[CH_3)$	-112	$b[\mathbf{Br}]$	-16	$c[CH_3CONH-]$	18	
		$b[\mathrm{NO}_2]$	-26	$c[\mathrm{NO}_2]$	-218	





Full Papers

A Combined Hansch/Free-Wilson Approach as Predictive Tool in QSAR Studies on Propafenone-Type Modulators of Multidrug Resistance^[1]

Claudia Tmej^{a)}, Peter Chiba^{b)}, Mario Huber^{a)}, Elisabeth Richter^{b)}, Manuela Hitzler^{b)}, Klaus-Jürgen Schaper^{c)}, and Gerhard Ecker^{a)}*

Arch. Pharm. Pharm. Med. Chem.

© WILEY-VCH Verlag GmbH, D-69451 Weinheim, 1998 0365-6233/98/0708/0233 \$17.50 +.50/0





QSAR Limitations



- Congeneric Series
- Same Binding Mode
- Lack of 3-D Structural Information
- Statistical Limitation



- (i) Multivariable linear regression analysis (MLR)
- (ii) Principal component analysis (PCA)
- (iii) Partial least square analysis (PLS)

least square applications

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Squared Correlation
Coefficient
$$R^{2} \text{ or } r^{2} = 1 - \frac{\sum_{i=1}^{N} (Y_{exp,i} - Y_{calc,i})^{2}}{\sum_{i=1}^{N} (Y_{exp,i} - \overline{Y})^{2}}$$

Cross-Validated R²

 Q^2 or q^2 q^2

$$q^{2} = 1 - \frac{\sum_{i=1}^{N} (Y_{\exp,i} - Y_{pred,i})^{2}}{\sum_{i=1}^{N} (Y_{\exp,i} - \overline{Y})^{2}}$$

i=1

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The predictive ability of a model is estimated using a reduced set of structural data

 $-\infty \leq q^2 \leq 1$

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- K-fold cross-validation
- Repeated random sub-sampling validation
- Leave-one-out crossvalidation



HOW TO EVALUATE THE QUALITY OF A MODEL

Y-Scrambling

A statistical test of prediction tools, in which models are fitted for randomly reordered property/activity values and compared with the model obtained for the actual property/activity values.



A new model is obtained for such permuted data, R^2 and Q^2 are then recalculated.

This step is repeated for a sufficient number of times (iterations):

a good number being 50 to 100.

Values obtained in the above fashion are compared with the true values obtained for the model that was fitted on the real data.



External Test-Set

SDEP (Standard Deviation Error of Prediction)

