



Dipartimento di Chimica e Tecnologie del Farmaco Ph.D. in Pharmaceutical Sciences (XXIII cycle 2007 – 2010)

Rational Design of Novel Antiviral Compounds Through Computational Approaches

Mentor: Prof. Rino Ragno

> Candidate: Ira Musmuça







Advantages

- 1. Orders of Magnitude Cheaper and Faster
- Offers the Possibility to Predict Molecular Behaviours that Cannot be Elucidated in any Other Way
- Simulation of Complex Molecular Environments, Widening the Applicability of *in silico* Studies from the Interactions of Small Molecules with Key Protein Residues, to the Simulation of the Dynamic Evolution of Complex Biological Systems with Atomic Resolution



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Computational Drug Discovery



Broad Classification

- **1. Structure Based Approaches**
- 2. Ligand Based Approaches



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Development and Application of Medicinal Chemistry Computational Methods in the Research Area of QSAR, 3D-QSAR, Molecular Docking and Virtual Screening

- Orally Active Interferon Inducers
- > Allosteric Inhibitors of HCV NS5B RNA-dependent RNA Polymerase



Small-Weight Interferon Inducers Orally Active vs Hepatitis C Virus

Aim of the Work

Description of How Steric, Electrostatic, Hydrophobic and Hydrogen-Bonding Interactions Might Influence the Biological Activity of a Published Set of 176 IFN Inducers, Using a Ligand-Based 3D-QSAR Approach

Interferon Inducers



Virus Organism **Antiviral State** Immune System Cell

Alpha Interferon



Current Opinion in Immunology 2007, 19, 17-23



- Polynucleotides
- Fluorenones
- Pyrimidinones
- Anthraquinones

Small Weight Molecules with in vitro and in vivo alpha-IFN Inducing Activity



✤ 1*H*-imidazo-[4,5-*c*]quinolines





- Polynucleotides
- Fluorenones
- Pyrimidinones
- Anthraquinones

Small Weight Molecules with in vitro and in vivo alpha-IFN Inducing Activity



- ✤ 1*H*-imidazo-[4,5-*c*]quinolines
- 8-hydroxyadenines



Hirota et al. *J. Med. Chem.* **2002**, *45*, 5419-5422 Isobe et al. *J. Med. Chem.* **2006**, *49*, 2088-2095. Isobe et al. *Bioorg. Med. Chem.* **2003**, *11*, 3641-3647 Kurimoto et al. *Bioorg. Med. Chem.* **2003**, *11*, 5501-5508 Kurimoto et al. *Bioorg. Med. Chem.* **2004**, *12*, 1091-1099 Gerster et al. *J. Med. Chem.* **2005**, *48*, 3481-3491



Relationships Between Chemical-Physical Properties of Chemical Substances and their Biological Activities to Obtain a Reliable Statistical Model for Prediction of the Activities of New Chemical Entities

Primary Aims of QSAR

- Optimization of the Existing Leads so to Improve Their Biological Activities.
- Prediction of the Biological Activity of Untested and Sometimes yet
 Unavailable Compounds





Training Set Selection

- Molecular Modeling
- Molecular Alignment
- Molecular Interaction Fields
- Statistical Analysis
- External Validation
- GRID Plot Interpretation

¹ Hirota et al. *J. Med. Chem.* **2002**, *45*, 5419-5422

- ² Isobe et al. J. Med. Chem. 2006, 49, 2088-2095.
- ³ Isobe et al. *Bioorg. Med. Chem.* 2003, *11*, 3641-3647
- ⁴ Kurimoto et al. *Bioorg. Med. Chem.* 2003, 11, 5501-5508
- ⁵ Kurimoto et al. *Bioorg. Med. Chem.* **2004**, *12*, 1091-1099
- ⁶ Gerster et al. J. Med. Chem. 2005, 48, 3481-3491





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¹ Schuttelkopf et al. PRODRG: a tool for high-throughput crystallography of protein-ligand complexes. *Acta Crystallogr. D Biol. Crystallogr.* **2004**, *60*, 1355-1363

² Van Aalten et al. J. Comput. Aided Mol. Des. 1996, 10, 255-262

³ Berendsen et al. GROMACS: A message-passing parallel molecular dynamics implementation. *Comput. Phys. Commun.* **1995**, *91*, 43-56





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¹ Jain, A. N. Ligand-based structural hypotheses for virtual screening. J. Med. Chem. 2004, 47, 947-961



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Wade, R. C. Molecular Interaction Fields, In: 3D QSAR in Drug Design. Theory, Methods and Applications, Kubinyi, H. Ed.; ESCOM, Leiden, Netherlands, 1993, pp. 486-505.

Goodford, P. J. A computational procedure for determining energetically favorable binding sites on biologically important macromolecules. *J. Med. Chem.* **1985**, *28*, 849-857.



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Squared Correlation Coefficient

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

 $0 \leq r^2 \leq 1$

 r^2 measures of the 'simultaneous' variable variation r^2 = 0: the statistical model is not able to explain data r^2 =1: the statistical model is perfectly able to explain data

Baroni, M.; Costantino, G.; Cruciani, G.; Riganelli, D.; Valigi, R.; Clementi, S. Generating Optimal Linear PLS Estimations (GOLPE): An Advanced Chemometric Tool for Handling 3D-QSAR Problems. Quant. Struct. Act. Relat. 1993, 12, 9-20.

Cramer, R. D. III; Bunce, J. D.; Patterson, D. E.; Frank, I. E. Cross validation, bootstrapping and partial least squares compared with multiple regression in conventional QSAR studies. *Quant. Struct. Act. Relat.* **1998**, *7*, 18-25.



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Standard Deviation Error of Prediction

$$SDEP = \sqrt{\frac{\sum_{i=1}^{N} (Y_{\exp,i} - Y_{pred,i})^2}{N}}$$

Cramer, R. D. III; Bunce, J. D.; Patterson, D. E.; Frank, I. E. Cross validation, bootstrapping and partial least squares compared with multiple regression in conventional QSAR studies. *Quant. Struct. Act. Relat.* **1998**, *7*, 18-25.



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Squared Predictive Correlation Coefficient

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

Cramer, R. D. III; Bunce, J. D.; Patterson, D. E.; Frank, I. E. Cross validation, bootstrapping and partial least squares compared with multiple regression in conventional QSAR studies. *Quant. Struct. Act. Relat.* **1998**, *7*, 18-25.



Original Data Set Training Set Selection **Molecular Modeling** Molecular Alignment ۲ Molecular Interaction Fields **Training Set** Statistical Analysis **External Validation** ۲ **GRID** Plot Interpretation ۰ **PLS Models Test Set**

Compare the Test Set Compounds Y_{exp} -values with the Predictions made by the PLS model





Statistical results for (M1-M6) 3D -QSAR global models obtained from diverse GOLPE PLS analysis.															
	-	2		LOO			LTO		LSO-5		LHO			Test set	
IVI	۲	r-	q²	SDEP	PC	q²	SDEP	PC	q²	SDEP	PC	q²	SDEP	PC	SDEP _{ext}
M1	OH2	0.62	0.47	0.71	2	0.46	0.71	2	0.43	0.73	2	0.39	0.76	2	1.08
M2	ОН	0.73	0.61	0.61	2	0.61	0.61	2	0.60	0.61	2	0.56	0.64	2	1.05
M3	DRY	0.89	0.64	0.58	5	0.64	0.58	5	0.61	0.60	5	0.55	0.65	5	0.9
M4	N1	0.69	0.58	0.65	2	0.56	0.65	2	0.55	0.65	2	0.52	0.67	2	1.09
M5	0	0.72	0.60	0.61	2	0.60	0.61	2	0.60	0.62	2	0.57	0.64	2	1.05
M6	DRY+ OH	0.70	0.56	0.64	2	0.56	0.64	2	0.55	0.65	2	0.53	0.67	2	1.09
[*] M: model name; P: GRID probe; LOO: Leave One Out Cross-validation; LTO: Leave Two Out Cross-validation; LSO-5: Leave-Some-Out Cross- validation using 5 groups; LHO: Leave Half Out; <i>r</i> ² : conventional square correlation coefficient; <i>q</i> ² : cross-validation correlation coefficient; SDEP: cross-validated standard error of prediction; PC: optimal number of Principal Components; SDEP _{ext} : Standard Error of Prediction for the external test set.															



- Training Set Selection
- Molecular Modeling
- Molecular Alignment
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- GRID Plot Interpretation



Blue regions	A favorable (negative) interaction INCREASES activity.				
	A unfavorable (positive) interaction DECREASES activity.				
Yellow regions	A favorable (negative) interaction DECREASES activity.				
	A unfavorable (positive) interaction INCREASES activity.				







GRID/GOLPE PLS Coefficients contour maps for the M2 and M3 3D-QSAR models (contour levels 0.0049 yellow, -0.0049 cyan; contour levels 0.00452 yellow, -0.00452 cyan, respectively). To aid interpretation only the highest active compound **175** (in red) and one of the lowest active compounds **36** (in green) are shown. For the sake of clarity hydrogen atoms are omitted.

Conclusions





Musmuca, I.; Simeoni, S.; Caroli, A.; Ragno, R. Small-Molecule Interferon Inducers. Towards the Comprehension of the Molecular Determinants Through Ligand-Based Approaches. J. Chem. Inf. Model. 2009, 49,1777-1786

Conclusions



 These features are fully in agreement with several anti-HCV derivatives able to stimulate interferon release in PBMC (peripheral blood mononuclear cell), recently reported by Pryde et al.¹



¹ Pryde et al. The discovery of a novel prototype small molecule TLR7 agonist for the treatment of hepatitis C virus infection *Med. Chem. Commun.*, 2011, Advance Article



HCV NS5B RNA-dependent RNA polymerase

HCV NS5B RdRp



Hepatitis C virus (HCV), the agent responsible for most cases of blood-borne hepatitis, was discovered by Choo et al. 20 years ago¹



¹ Choo, Q.L.; Kuo, G.; Weiner, A. J.; Obverby, L. R.; Bradley, D. W.; Houghton, M. Isolation of a cDNA clone derived from a blood-borne non-A, non-B viral hepatitis genome. *Science* 1989, *244*, 359-362.

² Chander, G.; Sulkowski, M. S.; Jenckes, M. W.; Torbenson, M. S.; Bass, H. F. Treatment of chronic hepatitis C: a systematic review. *Hepatology* **2002**, *36*, S135-S144.

HCV NS5B RdRp





- Essential Enzymatic Activity for a Correct Viral Replication
- Possibility to Design Selective Inhibitors *versus* the Only Infected Cells
- Available Structural Data (Bresanelli et al., PNAS, **1999**)

HCV NS5B RdRp





Ribbon show of the overall structure of NS5B RdRp with domains colored according to thumb (green), palm (yellow), and fingers (orange). Three allosteric binding sites surfaces are also shown. Dark red colored surface corresponds to the thumb allosteric bs, dark violet colored surface corresponds to the palm allosteric bs and the dark blue colored surface corresponds to the allosteric binding site situated in the thumb domain, near but clearly distinct from the first one (dark red surface).



NS5B Inhibitors can be classified into two major groups:

Nucleoside Analogues

Non Nucleoside Analogues





NS5B Inhibitors can be classified into two major groups:

Nucleoside Analogues

Non Nucleoside Analogues



Thumb Allosteric Inhibitors





Palm Allosteric Inhibitors





Ligand-Based, Structure-Based and 3D-QSAR Protocol











1nhu

1nhv

1os5

1yvx



ΟН Hat

HaC





2d3u

2d3z

2d41

2gir



2hwh



2hwi



2i1r





2hai

205d

Ligand-Based, Structure-Based and 3D-QSAR Protocol



Polymerase-Inhibitor Complex Structures Preparation



Case et al. The Amber biomolecular simulation programs. J. Comput. Chem. 2005, 26, 1668-1688

Meng, E. C.; Pettersen, E. F.; Couch, G. S.; Huang, C. C.; Ferrin, T. E. Tools for integrated sequence-structure analysis with UCSF Chimera. *BMC Bioinformatics* **2006**, *7*, 339.

Polymerase-Inhibitor Complex Structures Preparation



Molecular Interaction Fields



PLS Analysis Results for the Thumb and the Palm Structure-Based 3D-QSAR Models. ^a							
Ν	GRID Probe	V	PC	r ²	q^2		
15	C1=	5133	3	0.99	0.69		
10	C1=	3848	3	0.99	0.55		

^a N, number of compounds in the training set; V, number of GOLPE variables; PC, optimal number of principal components; r^2 , conventional square correlation coefficient; q^2 , cross-validation correlation coefficient; SDEP, cross-validated standard error of prediction using the leave-five-out cross-validation method



Fitting and Cross-Validation Plots for the Thumb (left) and Palm (right) Training Sets.

Key Steps for the Assessment of SB and LB Alignments Processes



Assessment of Docking: Redocking

Assessment of the Autodock Program in the Redocking Stage. RMSD Values for the First Ranked Pose (Best Docked), the Lowest Energy Docked Conformation of the Most Populated Cluster, (Best Cluster) and the One Closest to the Experimentally Bound Conformation (Best Fitted Cluster).

Binding	DDD	Linend Fratma	Best Docked	Best Cluster		Best Fitted		
Site	PDB	Ligand Entry	RMSD	Cluster N°	RMSD	Cluster N°	RMSD	
	1NHU	1	3.17	3	2.17	3	0.89	
	1NHV	2	4.13	2	4.75	26	1.86	
	1OS5	13	3.46	1	3.46	9	1.50	
	1YVX	3	3.81	1	3.81	6	1.58	
	1YVZ	4	3.74	4	1.92	4	0.78	
	2D3U	6	0.71	1	0.71	1	0.44	
	2D3Z	7	0.75	1	0.75	1	0.60	
	2D41	8	1.43	1	1.43	1	0.58	
Thumb	2GIR	5	5.70	2	1.12	2	0.70	
	2HAI	14	2.05	1	2.05	1	0.92	
	2HWH	9	9.86	2	2.13	2	0.79	
	2HWI	10	0.34	1	0.34	1	0.24	
	2l1R	11	5.84	2	1.67	2	0.73	
	2JC0	15	0.85	1	0.85	1	0.68	
	205D	12	5.74	3	2.78	7	1.28	
	Average F	RMSD	3.44		2.00		0.90	
	1YVF	17	3.38	4	1.12	4	0.93	
	1Z4U	16	0.89	1	0.89	1	0.57	
	2AWZ	18	3.53	3	1.72	3	1.11	
	2AX0	19	0.84	1	0.84	1	0.52	
	2AX1	20	3.24	2	0.99	2	0.61	
Palm	2FVC	21	1.04	1	1.04	1	0.74	
	2GC8	24	2.03	1	2.03	2	1.82	
	2GIQ	22	3.49	5	1.91	5	1.81	
	2JC0	15	0.65	1	0.65	1	0.42	
	2JC1	23	0.74	1	0.74	1	0.47	
	Average RMSD		1.98		1.19		0.90	

Assessment of Docking: Redocking Modeled

Checks more realistically Autodock's ability to reproduce binding mode conformations of molecules with no experimental binding data

Examples of Redocking Modeled



Superimposition of the Surflex-aligned conformer (carbon atoms in green) and the re-docked conformer (carbon atoms in magenta) to the experimental conformation (in orange) of one thumb NNI (6, on the left) and one palm NNI (16, on the right) within the NS5B (cyan colored ribbons). Atom bonds are in stick fashion. Hydrogen atoms are omitted for the sake of clarity.

				best cluster		best fitted cluster	
binding site	PDB	ligand entry	best docked (rmsd)	cluster N°	rmsd	cluster N°	rmsd
thumb	1NHU	1	2.25	1	2.25	3	1.87
	1NHV	2	4.37	14	3.68	19	2.54
	1OS5	13	7.59	5	3.61	32	1.81
	1YVX	3	3.51	2	1.77	10	1.58
	1YVZ	4	1.73	1	1.73	1	1.67
	2D3U	6	5.79	2	5.06	14	1.79
	2D3Z	7	5.36	2	3.75	8	1.22
	2D41	8	6.49	2	1.51	2	1.51
	2GIR	5	1.57	1	1.57	1	1.57
	2HAI	14	1.45	1	1.45	1	1.45
	2HWH	9	3.60	7	2.9	7	1.71
	2HWI	10	3.61	3	2.20	13	1.40
	2I1R	11	2.27	2	1.77	3	1.61
	2JC0	15	4.26	11	9.7	3	3.38
	205D	12	3.48	1	3.48	18	1.73
	average rmsd		3.82		3.1		1.79
				best clus	ster	best fitted	cluster
binding site	PDB	ligand entry	best docked (rmsd)	cluster N°	rmsd	cluster N°	rmsd
palm	1YVF	17	2.56	2	3.43	9	2.52
	1Z4U	16	2.26	2	3.87	12	1.78
	2AWZ	18	1.64	1	1,.64	1	1.64
	2AX0	19	1.85	1	1.85	3	1.58
	2AX1	20	1.68	1	1.68	1	1.68
	2FVC	21	1.88	1	1.88	3	1.06
	2GC8	24	11.13	2	2.08	6	1.84
	2GIQ	22	1.84	1	1.84	11	1.64
	2JC0	15	1.21	1	1.21	1	1.21
	2JC1	23	0.97	1	0.97	1	0.97
	average rmsd		2.7		2.09		1.59

Assessment of Docking: Cross-Docking

Ligand-Based Alignment: Surflex-Assessment

- 1. The Ligand-Based alignment of the molecules was achieved using Surflex-Sim
- 2. This method optimizes the pose of a query molecule to an object molecule in order to maximize 3D similarity



Examples of Surflex alignment. Superimposition of the modeled ligand conformations (carbon atoms in green) to the experimental ones (carbon atoms in orange) of three compounds of training sets (from left to right, compounds **6**, **12** and **16**). Atom bonds are in ball and stick fashion. Hydrogen atoms are omitted for the sake of clarity.

TEST SET{Thumb Domain (81 molecules)Palm Domain (223 molecules)

PLS Analysis Results for the Thumb and the Palm Structure-Based 3D-QSAR Models.^a

Ν	GRID Probe	V	PC	r ²	q ²	SDEP _{ext}
15	C1=	5133	3	0.99	0.69	0.65
10	C1=	3848	3	0.99	0.55	1.05

^a N, number of compounds in the training set; V, number of GOLPE variables; PC, optimal number of principal components; r^2 , conventional square correlation coefficient; q^2 , cross-validation correlation coefficient; SDEP, cross-validated standard error of prediction using the leave-five-out cross-validation method

External Validation of the 3D-QSAR Models



Diagrams of number of compounds with an error of prediction greater than 1.5 (violet), 1.0 (dark red) and lower than 1.0 (pale yellow).



Virtual Screening

Molecular Structure and Antiviral Activity ^a of the Compounds Selected by VS Protocol								
Thum	b Domain	Palm Domain						
	NH_2 $N \rightarrow N$	$HO \rightarrow O \rightarrow$	HO					
NSC 123526	NSC 125626	NSC 169534	NSC 3354					
IC ₅₀ = 46.0 μM	IC ₅₀ = 73.3 μM	IC ₅₀ = 64.5 μM	IC ₅₀ = 54.3 μM					
^a The data represents ar	average of at least two indepe	endent experiments						

Binding Mode Analysis



Musmuca, I.; Caroli, A.; Mai, A.; Kaushik-Basu, N.; Arora, P. and Rino Ragno. Combining Structure-Based Three-Dimensional Quantitative Structure-Activity Relationship Analysis and Cross-Docking Procedures for in Silico Screening of Hepatitis C Virus NS5B Polymerase Inhibitors. J. Chem. Inf. Model. 2010, 50, 662-676.

Binding Mode Analysis



Musmuca, I.; Caroli, A.; Mai, A.; Kaushik-Basu, N.; Arora, P. and Rino Ragno. Combining Structure-Based Three-Dimensional Quantitative Structure-Activity Relationship Analysis and Cross-Docking Procedures for in Silico Screening of Hepatitis C Virus NS5B Polymerase Inhibitors. J. Chem. Inf. Model. 2010, 50, 662-676.

Binding Mode Analysis



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- Virtual Screening of <u>1990 compounds</u> from the NCI Diversity Set
- Structure-Based 3D-QSAR models used as external scoring function
- Selection of the most predictive molecules for biological assays against recombinant NS5BCΔ21
- Outcome of biological studies: <u>4 active compounds</u> versus our biological target
- <u>Binding mode analysis of 4 selected compounds within thumb subdomain</u>
- Selection of <u>NSC 123526</u> as our hit compound since:
 - i. Endowed with the lowest inhibitory activity (IC₅₀ = 46.0 μ M)
 - ii. Its docked conformer, best overlaps with the most active compound of the thumb training set (visual inspection of their binding modes)
 - iii. The most interesting from a medicinal chemistry point of view





De Bonis et al. Journal of Medicinal Chemistry 2008, 51, 1115-1125





In figure are reported on the left STLCs derivative **51** (light brown) as proposed by Autodock and overlapped to **NSC 123526** (green) as docked into HCV-NS5B, and on the right derivative **51** as proposed in by DeBonis el al. (*J. Med. Chem.* **2008**, *51*, 1115–1125) in the Human Mitotic Kinesin Eg5 (HMKEg, Pdb entry code 2fme). To some extent compound **51** seems to bind either NS5B or HMKEg making similar interactions.







	R ₃ (R ₃ (R ₂ (R ₁ (Sca		· I Harl Cham	$\begin{array}{c} R_4 \\ R_3 \\ R_2 \\ R_3 \\ R_4 \\ R_4 \\ R_5 \\ R_6 \\ Scaffold B \end{array}$				
8.	₽.	R.	B.	2000, 57, 1115 R.	-1125 X	¥	R.	
nq	152	ing .	154	196	^		NH ₂	
Ref. Cl	н	н	н	н	С	S	~~соон	
-сң,	-F		-OH protected	CI	N		∕~соон	
-CH-(CH _a) ₂	CH ₂						COOH	
÷			о:				Soso H₀	
-OH protected		н,с,°,	H ₃ C ^O					
-SH protected		O ₀ 0 HaC					~ ^у ^К сн,	
-NH ₂ protected								
-OCH			H ₃ CO				~8 Д-соон	
-CF3			HO				~°\C^°	
-CN			нус о́					
			н _я с. 0.80					
			Hz′ ¥2°					
			o, H o cHs					
¥У-сн,								
J. CH								





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