

Dedicated to Professor Victor-Emanuel Sahini  
on the occasion of his 80th anniversary

## QSAR ANALYSIS OF A SERIES OF NON-IMIDAZOLIC DERIVATIVES ACTING ON THE HISTAMINE H3 RECEPTOR

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A QSAR study using multiple linear regression (MLR) is reported for a series of non-imidazolic antagonists of the histamine H3 receptor. One of the best models contains four descriptors:  $DM_H$  (sp hybridization component of the AM1 dipole moment), HYE (hydration energy), logP and S\_110 (number of  $-SO_2-$  groups) and has statistical indices:  $n = 30$   $r^2 = 0.955$   $r^2_{adj} = 0.948$   $r^2_{CV(LOO)} = 0.925$   $see = 0.215$   $F = 132.32$   $press = 1.155$ . The antihistaminic activity of the ligands is influenced by their electronic, hydrophobicity and thermodynamic properties.

### INTRODUCTION

Histamine is an endogenous amine that acts as neurotransmitter. It is implicated in gastric secretion of HCl and pepsin, in bronchoconstriction, in pain and prurit. It produces different effects in the circulatory and the central nervous systems. Physiological effects of histamine are due to the activation of four receptor subtypes, namely H1 – H4. The H3 subtypes act both as presynaptic autoreceptors<sup>1</sup> and postsynaptic receptors<sup>2</sup> in central nervous system (CNS). The H3 antagonists have potential therapeutic applications in different CNS diseases: epilepsy, schizophrenia, insomnia, sleeping disease, bullimia, Alzheimer's disease, temporary loss of memory and incapacity of learning.<sup>3-10</sup> The search for novel histamine H3 receptor ligands with drug-like properties has resulted in the discovery of an increasing number of imidazole-free structures with improved pharmacokinetic and pharmacodynamic properties. We report here the results of a classical quantitative structure-activity relationships (QSARs) study for a series of non-imidazolic compounds synthesized and tested by Hancock et al<sup>11</sup> as antagonists on the guinea pig brain H3 receptor.

### METHODS

The structure and activities obtained by Hancock et al<sup>11</sup> for a series of 31 pyrrolidinic derivatives, together with a series of 14 derivatives of 2,5-diazabicyclo[2.2.1]heptane are displayed in Table 1. Antihistaminic activity has been expressed as the negative logarithm of inhibition constant (molarity units), pKi.

Molecular geometries of derivatives from both series have been optimized using the AMBER99 force field and then the AM1 Hamiltonian.<sup>12</sup> Optimization was performed with Polak- Ribiere conjugate gradient algorithm, method which is implemented in HyperChem7.52.<sup>13</sup> The geometry was optimized till the RMS gradient was equal to or less than 0.01 kcal/Åmol. For some representative structures a conformational search has been carried out<sup>14,15</sup> with the Conformational Search module also available in Hyperchem7.52.

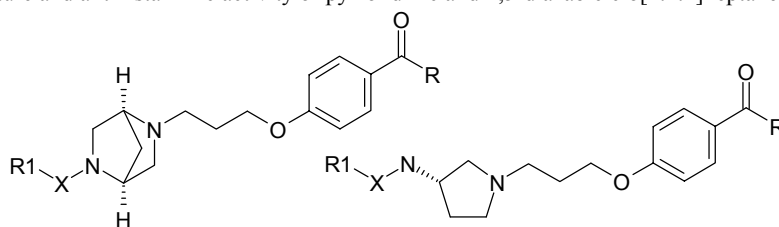
Lowest AM1 energies are obtained for conformers with two neighbor carbonylic groups in *anti* positions. Similarly, the lowest energy was obtained for conformers with the H and carbonyl in the amidic group in *anti* positions. For example, for compound **5**, the most stable conformer with *anti* positions both for carbonylic groups and

amidic hydrogens the  $\Delta H_f$  is -213.755 kcal/mol. A conformer with the amidic H and carbonylic group in *syn* position has a formation energy of -210.656 kcal/mol. The conformer of compound **5** with the lowest energy is presented in Fig 1. The possibility of stabilization of conformers by intramolecular H bonds was checked. Such stabilization could not be evidenced neither for the conformers obtained with the AMBER99 force field nor for those obtained with the AM1 Hamiltonian.

For calculating the descriptors to be tested in multiple linear regressions, the QSAR properties module from the HyperChem7.52 and DRAGON3.0<sup>16</sup> softwares have been used. From the AM1 method the energies of the highest

occupied molecular orbital,  $E_{HOMO}$  and of the lowest unoccupied molecular orbital  $E_{LUMO}$ , the total dipole moment and the *sp* hybridization component of the dipole moment have been obtained. The absolute electronegativity ( $\chi$ ) and the chemical hardness ( $\eta$ ) have been calculated as  $\chi \approx (IP + EA)/2$  and  $\eta \approx (IP - EA)/2$  assuming that the ionization potential,  $IP \approx -E_{HOMO}$  and the electron affinity,  $EA \approx -E_{LUMO}$ .<sup>17,18</sup> From the AM1 method resulted also a thermodynamic property, namely, the energy of formation,  $\Delta H_f$ . As descriptors have also been tested the energy terms (bond, angle, torsion and van der Waals energies) resulted for geometries of conformers optimized with the AMBER99 force field.

Table 1  
The structure and antihistaminic activity of pyrrolidinic and 2,5-diazabicyclo[2.2.1]heptane derivatives



No.	structure B		structure A			pKi
	Struct	R	X	R <sub>1</sub>		
5	A	CH <sub>3</sub>	-CO-	Boc L-Ala	6.05	
6	A	CH <sub>3</sub>	-CO-	Boc D-Ala	6.51	
7	A	CH <sub>3</sub>	-CO-	Boc L-Ser	6.23	
8	A	CH <sub>3</sub>	-CO-	Boc D-Ser	6.51	
9	A	CH <sub>3</sub>	-CO-	L-Ser	6.82	
10	A	CH <sub>3</sub>	-	-	5.76	
11	A	CH <sub>3</sub>	-	-	6.17	
12	A	CH <sub>3</sub>	-CO-	2-pyrazinyl	6.69	
13	A	CH <sub>3</sub>	-CO-	3-pyridyl	7.43	
14	A	CH <sub>3</sub>	-CO-	4-thiazol[2-(3-pyridyl)]yl	7.26	
15	A	CH <sub>3</sub>	-CO-	p-CN-phenyl	7.4	
16	A	Cyclopropyl	-CO-	4-thiazol[2-(3-pyridyl)]yl	7.3	
31	A	CH <sub>3</sub>	-SO <sub>2</sub> -	p-CN phenyl	8.22	
32	A	Cyclopropyl	-SO <sub>2</sub> -	p-CN phenyl	8.4	
33	A	Cyclopropyl	-SO <sub>2</sub> -	Phenyl	8.52	
34	A	Cyclopropyl	-SO <sub>2</sub> -	o-F phenyl	8.57	
35	A	Cyclopropyl	-SO <sub>2</sub> -	m-F phenyl	8.8	
36	A	Cyclopropyl	-SO <sub>2</sub> -	p-F phenyl	8.46	
37	A	Cyclopropyl	-SO <sub>2</sub> -	o-Cl phenyl	8.43	
38	A	Cyclopropyl	-SO <sub>2</sub> -	m-Cl phenyl	8.52	
39	A	Cyclopropyl	-SO <sub>2</sub> -	p-Cl phenyl	8.42	
40	A	Cyclopropyl	-SO <sub>2</sub> -	o-CN phenyl	7.96	
41	A	Cyclopropyl	-SO <sub>2</sub> -	m-CN phenyl	8.55	
42	A	Cyclopropyl	-SO <sub>2</sub> -	o-CH <sub>3</sub> phenyl	7.96	
43	A	Cyclopropyl	-SO <sub>2</sub> -	m-CH <sub>3</sub> phenyl	8.57	
44	A	Cyclopropyl	-SO <sub>2</sub> -	p-CH <sub>3</sub> phenyl	8.38	
45	A	Cyclopropyl	-SO <sub>2</sub> -	p-OCH <sub>3</sub> phenyl	8.54	
46	A	Cyclopropyl	-SO <sub>2</sub> -	4 (t-butyl)phenyl	8.28	
47	A	Cyclopropyl	-SO <sub>2</sub> -	p-Br phenyl	8.3	
48	A	Cyclopropyl	-SO <sub>2</sub> -	4-(ethyl) phenyl	8.34	
49	A	Cyclopropyl	-SO <sub>2</sub> -	N-CH <sub>3</sub> -imidazol-4-yl	8	
17	B	Cyclopropyl	-CO-	Boc D-Ala	6.51	
18	B	Cyclopropyl	-CO-	Boc L-Ala	5.66	
19	B	Cyclopropyl	-CO-	D-Ala	6.38	

Tab. 1. (continued)

No.	structure B		structure A		pKi
	Struct	R	X	R <sub>1</sub>	
20	B	Cyclopropyl	-CO-	L-Ala	6.63
21	B	Cyclopropyl	-CO-	cyclohexyl	6.87
22	B	Cyclopropyl	-CO-	Phenyl	6.46
23	B	Cyclopropyl	-CO-	p-F phenyl	6.59
24	B	Cyclopropyl	-CO-	2-pyridyl	6.72
25	B	Cyclopropyl	-CO-	2-furyl	6.68
26	B	Cyclopropyl	-CO-	CH <sub>2</sub> C(CH <sub>3</sub> ) <sub>3</sub>	5
27	B	Cyclopropyl	-CO-	Boc β-Ala	7.15
28	B	Cyclopropyl	-CO-	β-Ala	7.77
50	B	Cyclopropyl	-SO <sub>2</sub> -	Ethyl	6.92
51	B	Cyclopropyl	-SO <sub>2</sub> -	Phenyl	7.72

Statistical data were obtained with STATISTICA 5.0. The descriptor elimination was carried out using the forward stepwise regression method included in STATISTICA 5.0. Outliers are

considered compounds for which the pKi calculated values are outside of the interval of ±2 sigma.

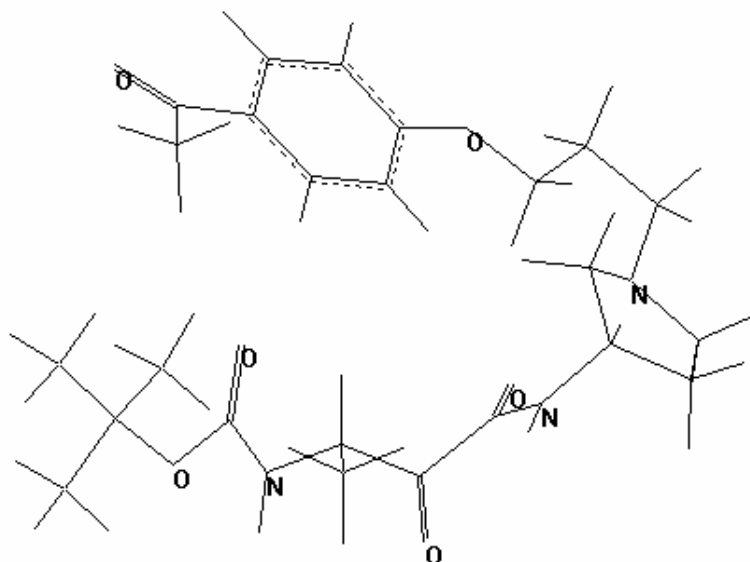


Fig 1 – The structure of the best conformer (with the lowest energy) of compound 5 resulted from AM1 calculations; the two vicinal carbonyl groups in *anti* positions together with the hydrogen atom and the carbonylic oxygen of the amidic group.

**RESULTS**

Statistical parameters of some monolinear regressions after the elimination of outliers are presented in Table 2. From this table one can see

that some structures (27, 28, 18, 22) are eliminated in many regressions and structure 26 is outlier in almost all regressions.

Table 2

Statistical values of monoparametric regressions after elimination of outliers

Descriptors	r	s	F	n	Outliers
E <sub>BOND</sub> (kcal/mol)	0.977	0.190	535.54	28	5,7,8,9,10,11,13,17,19,20,25,26,27,28,46,49,50
DM <sub>H</sub> (debye)	0.963	0.233	349.34	29	5,8,10,11,13,14,18,19,20,24,26,28,31,32,47,50
MOR26M	0.971	0.194	461.03	30	5,6,7,8,9,10,18,19,22,26,28,33,35,39,49
NS	0.974	0.215	580.71	33	5,10,13,14,15,16,18,26,27,28,50,51
S <sub>110</sub>	0.974	0.215	580.71	33	5,10,13,14,15,16,18,26,27,28,50,51

Tab. 1. (continued)

MOR26V	0.928	0.315	193.70	33	5,6,7,8,10,12,15,18,22,26,28,49
MOR11U	0.967	0.248	456.38	34	6,11,12,21,22,23,26,27,28,43,46
R2E	0.953	0.267	316.36	34	5,10,11,15,18,19,26,27,46,49,51
MOR20M	0.952	0.321	312.92	34	10,11,12,13,15,21,22,23,27,28,47*
R1M_	0.947	0.279	303.68	37	5,10,11,18,26,27,28,47
HOMT	0.807	0.594	80.43	37	14,16,22,23,26,27,28,49
MOR11P	0.948	0.296	319.57	38	7,10,12,17,26,27,28
MOR11E	0.947	0.316	314.94	38	12,21,23,26,27,28,46
MAXDN	0.898	0.380	149.43	38	5,7,10,18,26,28,50
N_072	0.939	0.331	276.68	39	12,19,22,26,27,50
MOR06M	0.908	0.390	172.78	39	5,13,18,26,28,50
MOR11M	0.943	0.323	306.65	40	12,17,26,27,28
HATS1M	0.888	0.430	141.92	40	10,26,28,46,47
MOR26P	0.887	0.430	139.96	40	7,10,26,28,49
R1M	0.914	0.378	196.91	41	18,26,28,47
BEHP1	0.900	0.409	166.11	41	13,18,26,51
logP	0.736	0.618	46.02	41	18,22,23,26
HYE (kcal/mol)	0.737	0.683	47.63	42	7,8,9*
RDF040M	0.862	0.485	118.67	43	26,28
E <sub>BEND</sub> (kcal/mol)	0.564	0.756	19.16	43	18,26
E (kcal/mol)	0.563	0.757	19.04	43	18,26,
MW	0.394	0.842	7.53	43	18,26,
DM (debye)	0.693	0.682	38.90	44	26
E <sub>HOMO</sub> (eV)	0.627	0.738	27.21	44	26
E <sub>LUMO</sub> (eV)	0.627	0.738	27.15	44	26
NMB	0.666	0.750	34.19	45	-
X (eV)	0.637	0.775	29.40	45	-

**r** – correlation coefficient; **see** – standard error of estimate; **F** – F ratio; **n** – number of cases; **E** – potential energy of a conformer calculated with the AMBER99 method; E<sub>BOND</sub> and E<sub>BEND</sub> – stretching and bending energy terms calculated with AMBER99 force field; E<sub>HOMO</sub> and E<sub>LUMO</sub> – AM1 energies of the highest occupied and of the lowest unoccupied molecular orbitals; DM – AM1 total dipole moment; DM<sub>H</sub> – sp hybridization component of AM1 dipole moment; HyE – hydration energy; absolute electronegativity,  $\chi = \frac{1}{2}(\text{IP}+\text{EA})$ ; chemical hardness,  $\eta = \frac{1}{2}(\text{IP}-\text{EA})$ ; log P – log of the n-octanol-water partition coefficient; MW – molecular weight; NMB – number of multiple bonds; NS – number of sulfur atoms; MAXDN – maximal electropological negative variation; BEHP1 – highest eigenvalue n.1 of Burden matrix/weighted by atomic polarizabilities; HOMT – total HOMA (harmonic oscillator model of aromaticity index); RDF040M – radial distribution function – 4.0/weighted by atomic masses; Mor11u – 3D-MoRSE - signal 11 unweighted; Mor06m – 3D-MoRSE - signal 06/weighted by atomic masses; Mor11m – 3D-MoRSE - signal 11/weighted by atomic masses; Mor20m – 3D-MoRSE - signal 20/weighted by atomic masses; Mor26m – 3D-MoRSE - signal 26/weighted by atomic masses; Mor11v – 3D-MoRSE - signal 11/weighted by atomic van der Waals volumes; Mor26v – 3D-MoRSE - signal 26/weighted by atomic van der Waals volumes; Mor11e – 3D-MoRSE - signal 11/weighted by atomic Sanderson electronegativities; Mor11p – 3D-MoRSE - signal 11/weighted by atomic polarizabilities; ; Mor26p – 3D-MoRSE - signal 26/weighted by atomic polarizabilities; R1m – R autocorrelation of lag 1/weighted by atomic masses; R1m\_ – R maximal autocorrelation of lag1/weighted by atomic masses; R2e – R autocorrelation of lag 2/weighted by atomic Sanderson electronegativities; N\_072 – number of RCO-N</>N-X=X groups; S110 – number of R-SO2-R groups.

Some descriptors can model by themselves the pKi values (Table 2). However, associating the descriptors with physical meaning the resulted multiple linear regressions have better chances to model the histaminic activity.

The regression (1) results by associating the E<sub>BOND</sub>, DM, DM<sub>H</sub>, SASA, HYE after elimination of outliers 7, 18, 21, 28:

$$\text{pKi} = 2.4375(0.6793) - 0.6410(0.1906)\text{E}_{\text{BOND}} + 0.3068(0.0403)\text{DM} + 0.4925(0.1470)\text{DM}_{\text{H}} + 0.0039(0.0008)\text{SASA} - 0.1216(0.0274)\text{HYE} \quad (1)$$

$$n = 41 \quad r^2 = 0.899 \quad r^2_{adj} = 0.885 \quad \text{see} = 0.333 \quad F = 62.40 \quad \text{press} = 3.89$$

This model suggests that the interactions of the derivatives with the histamine H3 receptor could be significantly influenced by: thermodynamic stability of conformers (HYE), electrostatic effects (DM), solvent effects (SASA, HYE) and

possibility of H bond formation (DM<sub>H</sub>). The cross-correlation matrix in Table 3 shows that there is no significant statistical cross-correlation between the descriptors of the regression (1).

Table 3

Values of partial and cross-correlation coefficients for regression (1)

	E <sub>BOND</sub>	DM	DM <sub>H</sub>	SASA	HYE
DM	-0.223				
DM <sub>H</sub>	-0.496	0.426			
SASA	0.067	0.072	0.138		
HYE	0.624	-0.060	-0.353	-0.098	
pKi	-0.651	0.649	0.696	0.336	-0.607

If in regression (1) instead of SASA is used as descriptor log P, results regression (2). After

elimination of outliers 18, 26, 28, regression (2) is:

$$\text{pKi} = 5.0731(0.3773) - 0.4474(0.1550) E_{\text{BOND}} + 0.1645(0.0378) \text{DM} + 0.4724(0.1217) \text{DM}_H - 0.1222(0.0220) \text{HYE} + 0.3764(0.0503) \log P \quad (2)$$

$$n = 42 \quad r^2 = 0.918 \quad r^2_{adj} = 0.907 \quad \text{see} = 0.279 \quad F = 80.87 \quad \text{press} = 2.81$$

The cross-correlation matrix for regression (2) is displayed in Table 4. Cross-correlation coefficients and other statistical parameters (r, see, F) are better than those in equation (1). log P in

equation (2) suggests that for these two series of compounds the transmembranar transport (high hydrophobicity) is important for the antihistaminic activity.

Table 4

Values of partial and cross-correlation coefficients for regression (2)

	E <sub>BOND</sub>	DM	DM <sub>H</sub>	HYE	log P
DM	-0.252				
DM <sub>H</sub>	-0.444	0.447			
HYE	0.580	-0.111	-0.234		
log P	-0.149	0.556	0.379	0.045	
pKi	-0.609	0.695	0.672	-0.500	0.689

To validate equation (2), the compounds have been divided into two series. Test series contained

compounds: 6,8,14,19,32,36,37,38. The regression for the learning series is:

$$\text{pKi} = 4.9973(0.429) - 0.4523(0.186) E_{\text{BOND}} + 0.1892(0.050) \text{DM} + 0.5062(0.175) \text{DM}_H - 0.1128(0.026) \text{HYE} + 0.3543(0.061) \log P \quad (3)$$

$$n = 34 \quad r^2 = 0.906 \quad r^2_{adj} = 0.890 \quad \text{see} = 0.303 \quad F = 54.18 \quad \text{press} = 2.57$$

The cross-correlation matrix corresponding to regression (3) is displayed in Table 5 and the pKi values of the test compounds predicted by regression (3) are presented in Table 6. The calculated activities are shifted to weaker potency than the experimental one and differ with more than 0.5 log units.

By associating the DM<sub>H</sub>, HYE, logP and S<sub>110</sub> descriptors, another statistical significant regression was obtained. After eliminating the outliers 5,18,19,25-28,40, regression (4) resulted:

$$\text{pKi} = 4.8352(0.1738) + 0.7305(0.1037) \text{DM}_H - 0.1358(0.0171) \text{HYE} + 0.3748(0.0441) \log P + 0.4347(0.1229) S_{110} \quad (4)$$

$$n = 37 \quad r^2 = 0.957 \quad r^2_{adj} = 0.951 \quad \text{see} = 0.201 \quad F = 176.82 \quad \text{press} = 1.292$$

Table 5

Values of partial and cross-correlation coefficients for regression (3)

	E <sub>BOND</sub>	DM	DM <sub>H</sub>	HYE	log P
E <sub>BOND</sub>	1.000				
DM	-0.177	1.000			
DM <sub>H</sub>	-0.478	0.529	1.000		
HYE	0.587	-0.168	-0.281	1.000	
log P	-0.004	0.515	0.441	0.119	1.000
pKi	-0.542	0.711	0.751	-0.492	0.636

Table 6

Experimental and predicted pKi values by regression (3) for the test set compounds

no	pKi <sub>exp</sub>	pKi <sub>pred</sub>	diff
6	6.513	6.387	0.126
8	6.509	6.041	0.468
14	7.260	7.364	-0.105
19	6.384	6.553	-0.169
32	8.398	8.325	0.073
36	8.456	8.397	0.059
37	8.432	8.265	0.167
38	8.523	8.516	0.007

Diff – difference: pKi(exp) – pKi(pred)

Regression (4) suggests that the number of the -SO<sub>2</sub>- groups influence favorably the antihistaminic potency; the greater the number of -SO<sub>2</sub>- groups,

the higher the activity. Cross-correlation and partial correlation coefficients are displayed in Table 7.

Table 7

Values of partial and cross-correlation coefficients for equation (4)

	DM <sub>H</sub>	HYE	logP	S_110
HYE	-0.367			
logP	0.419	0.143		
S_110	0.814	-0.411	0.563	
pKi	0.841	-0.519	0.648	0.889

The predictive potential of the regression was verified by the usual procedure. The 37 compounds have been divided into learning and test sets. By

eliminating the test set compounds: 7,12,14,31,42,43,51, the regression (5) was obtained.

$$\text{pKi} = 4.8030(0.2085) + 0.8092(0.1638) \text{DM}_H - 0.1354(0.0197) \text{HYE} + 0.3694(0.0499) \log P + 0.3627(0.1644) \text{S}_{110} \quad (5)$$

$$n = 30 \quad r^2 = 0.955 \quad r^2_{\text{adj}} = 0.948 \quad r^2_{\text{CV(LOO)}} = 0.925 \quad s = 0.215 \quad F = 132.32 \quad \text{press} = 1.155$$

pKi values calculated with equation (5) are shown in Table 8. Although two descriptors strongly

cross-correlate the predicted pKi values are unexpectedly good.

Table 8

pKi values calculated with regression (5) for test series

no	DM <sub>H</sub>	HYE	logP	S_110	pKi <sub>exp</sub>	pKi <sub>pred</sub>	Diff
7	0.332	-7.71	0.72	0	6.231	6.381	-0.150
12	0.609	-6.29	1.20	0	6.695	6.590	0.104
14	1.999	-5.45	1.04	0	7.260	7.542	-0.283
31	0.874	-10.85	2.08	1	8.222	8.110	0.112
42	1.508	-6.12	1.80	1	7.959	7.879	0.080
43	1.535	-6	3.20	1	8.569	8.402	0.167
51	1.244	-5.3	2.54	1	7.721	7.828	-0.107

## CONCLUSIONS

Antihistaminic activity of substituted pyrrolidinic derivatives can be modeled by many descriptors. pKi is influenced by electronic, hydrophobicity, steric and thermodynamic descriptors. The majority of models are based on stretching energy and SASA descriptors. Models with predictive potential give some suggestions regarding the influence of different properties on the antihistaminic activity. Stretching energies, hydration energies and number of peptidic bonds are inverse proportionally related to the antihistaminic potency and the rest of descriptors used in predictive models are directly proportional related to pKi. Models obtained suggest that antihistaminic activity is influenced by conformational stability of ligands, their degree of unsaturation, their electrostatic properties and hydrophobicity.

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*Dedicated to Professor Victor-Emanuel Sahini  
on the occasion of his 80th anniversary*

## CONFORMATIONAL ANALYSIS WITH AMBER AND AM1 METHODS OF A PARTIAL AGONIST, (R)-2-[(4-PHENYL-BUTYLAMINO)-METHYL]- CHROMAN-7-OL ACTIVE ON THE DOPAMINIC D2 RECEPTOR

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2-[(4-Phenyl-butylamino)-methyl]-chroman-7-ol, a high active partial agonist acting on the dopamine D2 receptor was chosen for a conformational analysis using the AMBER99 force field and the AM1 semiempirical quantum-chemical methods. The geometric properties of its conformers that occupy the low minima on the potential energy surface have been evaluated. Of 100 conformers retained from conformational search by optimization with AMBER force field had resulted only 38 distinct conformers which were distributed on an energy domain of 1.72kcal/mol above the lowest energy conformer. The conformers occupy a large space around the chromane moiety, but according to rotation barriers the ligand can modify its conformation to one suitable for efficient interaction with the amino acid residues from the active site of the D2 receptor.

### INTRODUCTION

Molecular cloning techniques have led to the characterization of a number of dopamine receptor subtypes which can be divided into the D1-like and the D2-like families. Whereas the D1-like family comprises D1 and D5 subtypes the D2-like family consists of D2, D3 and D4 receptors. According to neuropathological and genetic studies, selective dopamine D2 and D4 receptor agonists, partial agonists or antagonists might be of interest for the treatment of neuropsychiatric disorders including attention deficit hyperactivity, mood disorders and Parkinson's disease.<sup>1-3</sup>

On the other hand, the D3 receptor appears an important target for the development of candidate drugs since it is selectively expressed in the brain limbic system, which is thought to be involved in mood and cognitive disturbances as well as drug dependence. Thus, studies with respect to D2, D3 and D4 affinity, selectivity and ligand efficacy are of special current interest, as well as studies regarding ligand properties responsible for the biological activity.

Among synthesized ligands we choose to study the influence of steric factors onto biological activity of the 2-[(4-phenyl-butylamino)-methyl]-chroman-7-ol ligand having  $K_{i-high}$  and  $K_{i-low}$  values determined for both the high- and low-affinity agonist states of the dopamine D2 receptor ( $D_2^{high}$  and  $D_2^{low}$ ).<sup>4</sup> In this study we use two different molecular modelling methods for obtaining the geometry of conformers occupying the low minima on potential energy surface of this high potent partial agonist. The data can help in emphasizing the importance of steric effects, hydrogen bonds and van der Waals interactions of conformers with different residues from the active site of the D2 receptor.

### METHODS

For the purpose of modelling of the ligand geometry we use one molecular mechanics method, namely the AMBER99 force field, and a semiempirical quantum chemical method, AM1.<sup>5</sup>

The starting 2D-geometry of the 2-[(4-phenyl-butylamino)-methyl]-chroman-7-ol ligand was transformed into a 3D-geometry using the Model Builder module from Hyperchem7.52 package.<sup>6</sup>

The conformational analysis was performed with the Conformational search module from Hyper7.52 using the AMBER99 force field. This ligand has 7 rotatable bonds. Therefore seven dihedrals: O<sub>1</sub>-C<sub>2</sub>-C<sub>11</sub>-C<sub>12</sub> (D1), D2 C<sub>2</sub>-C<sub>11</sub>-N<sub>12</sub>-C<sub>13</sub> (D2), C<sub>11</sub>-N<sub>12</sub>-C<sub>13</sub>-

C<sub>14</sub> (D3), N<sub>12</sub>-C<sub>13</sub>-C<sub>14</sub>-C<sub>15</sub> (D4), C<sub>13</sub>-C<sub>14</sub>-C<sub>15</sub>-C<sub>16</sub> (D5), C<sub>14</sub>-C<sub>15</sub>-C<sub>16</sub>-C<sub>17</sub> (D6) and C<sub>15</sub>-C<sub>16</sub>-C<sub>17</sub>-C<sub>18</sub> (D7) have been varied between 0° and 360° with steps of 60°. The structure of the ligand and the numbering of heavy atoms is depicted in Fig 1.

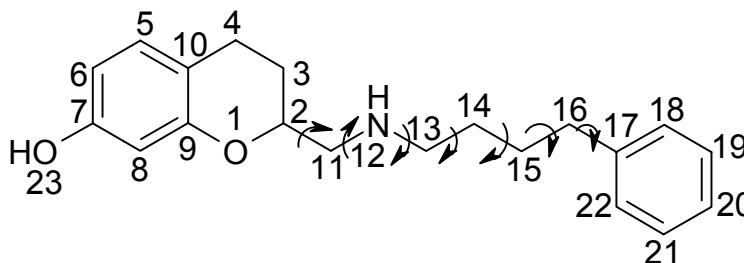


Fig. 1 – Numbering of atoms in (R)-2-[(4-phenyl-butylamino)-methyl]-chroman-7-ol and the dihedrals varied in conformational analysis.

In order to choose the starting conformations the conformer analysis was performed with Usage Direct method. The acceptance energy criterion was set at 100 kcal/mol above the energy of the best conformer. Standard preoptimization tests were carried out and the conformers that changed their chirality have been discarded. Two optimization criteria have been used: the maximum number of optimization cycles was set to 5000 and the RMS gradient norm was set to 0.01 kcal/Åmol. Search stop was achieved when the number of optimized conformers was 5000.

The 100 lowest energy conformations resulted from conformational search were further completely optimized with the AM1 semiempirical MO method as implemented in HyperChem7.52. The criteria set in AM1 optimization were: a RMS gradient norm of 0.01kcal/Åmol and a SCF convergence of 10<sup>-5</sup>. Optimization was performed with the algorithm Polak-Ribiere conjugate gradient.

## RESULTS AND DISCUSSION

Of 100 conformers given by conformational search with AMBER force field only 38 distinct conformers resulted. Their energy and dihedral values are presented in Table 1.

The distribution of conformers is not uniform. In the thermal energy domain (namely 0-0.6 kcal/mol) there are only two conformers with potential energy above the energy of the best conformer. In the energy domain 0.5 -1 kcal/mol there are four conformers, in the energy domain 1-1.5kcal/mol there are 19 conformers and in the energy domain 1.5 – 1.72 kcal/mol there are 12 conformers.

The superposition of the conformers in the energy domain 0 – 1 kcal/mol is presented in Fig. 2, while the superposition of the conformers in the energy domain 0 – 1.5kcal/mol is presented in Fig. 3.

Analyzing Figs 2 and 3 one can see that the space occupied by the side chain bound at 2-position of the chromane moiety is very large.

Table 1

Energy and dihedral values for 100 lowest energy conformers resulted from conformational search with AMBER99 method

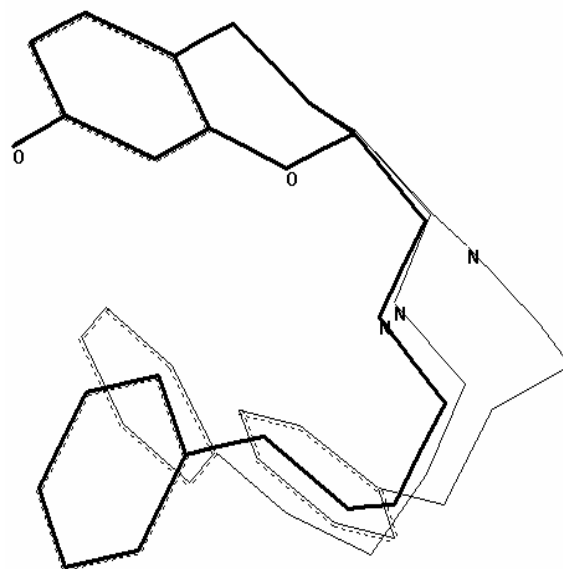
No	E	D1	D2	D3	D4	D5	D6	D7
1	3.816	53.819	-179.208	175.530	69.642	-69.141	-57.790	118.298
5	3.914	-53.349	-179.235	-174.387	53.690	65.619	-74.182	-62.194
11	4.355	68.154	179.994	-176.450	-83.299	58.526	53.550	-141.706
15	4.562	61.258	-170.271	177.242	-44.333	-57.577	174.407	-75.345
19	4.629	55.777	-178.960	176.003	-52.861	-69.909	68.291	60.876
23	4.717	-174.507	176.995	-177.748	49.431	57.949	-81.919	117.145
25	4.746	-58.189	177.777	-176.697	-72.215	63.896	55.668	-125.503
31	4.894	-60.642	-77.201	174.886	-67.754	-83.133	69.619	75.700
35	4.915	60.197	178.496	-176.529	-73.639	65.007	55.944	-127.773
37	4.964	-58.822	173.654	-63.368	-56.005	157.852	-69.364	-69.730
38	4.967	-80.554	68.159	172.974	57.084	68.307	-80.198	-64.593
41	5.075	68.049	-173.473	70.735	45.051	58.260	-82.603	-64.341

Table 1 (continued)

42	5.080	56.697	-178.889	175.782	60.462	-138.804	65.284	65.286
43	5.080	67.919	-173.555	70.559	45.179	58.404	-82.231	116.256
44	5.082	63.347	-176.579	178.915	-66.368	127.861	-69.667	103.995
45	5.083	63.282	-176.685	178.856	-66.353	127.823	-69.312	-76.431
46	5.087	57.057	-179.119	175.761	60.510	-137.959	65.510	-112.785
49	5.148	67.536	-173.504	70.655	45.246	58.329	-82.223	-63.956
50	5.154	67.563	-173.626	70.588	45.203	58.418	-82.138	115.973
54	5.168	61.623	-172.551	-179.953	-49.180	89.799	-165.509	109.842
56	5.213	-52.338	173.797	-172.163	48.633	60.135	-171.556	-78.396
58	5.224	-64.429	-175.782	177.967	-49.497	-58.141	82.074	63.874
59	5.225	-174.743	175.354	-177.238	-53.593	-62.807	84.147	-114.932
64	5.264	-64.981	-175.520	177.714	-49.123	-57.773	83.155	-116.894
68	5.302	-56.070	177.437	-172.138	53.660	70.158	-71.447	-59.475
69	5.324	60.470	-179.140	-179.977	60.636	179.660	62.919	73.789
73	5.359	60.970	-179.752	-178.973	60.461	-179.576	179.574	91.820
74	5.363	62.791	-169.476	175.503	-42.432	-57.180	174.893	105.039
77	5.418	59.676	-177.380	-179.649	-61.897	-178.851	-179.479	-92.625
80	5.449	59.736	-179.165	-179.823	60.623	-179.449	63.286	73.269
84	5.463	60.821	-171.358	179.476	-48.428	89.917	-165.254	-66.707
86	5.473	73.671	176.062	-172.521	-51.975	-62.369	83.220	65.863
88	5.482	60.659	-179.164	-179.100	61.039	-178.161	-63.263	-76.261
89	5.484	-58.309	178.368	-177.776	73.292	-66.837	-59.063	118.370
90	5.484	59.870	-177.633	-179.400	-62.220	-177.565	-63.008	-74.652
91	5.486	-59.782	177.930	179.709	62.433	178.738	179.678	-87.868
94	5.493	-85.386	63.275	177.353	59.454	76.257	-76.249	-68.174
100	5.538	60.242	-177.634	-179.191	-62.446	-179.965	63.615	75.688

E – AMBER energy (kcal/mol); see Fig 1 for dihedrals' notation: D1 – O<sub>1</sub>-C<sub>2</sub>-C<sub>11</sub>-N<sub>12</sub>, D2 – C<sub>2</sub>-C<sub>11</sub>-N<sub>12</sub>-C<sub>13</sub>, D3 – C<sub>11</sub>-N<sub>12</sub>-C<sub>13</sub>-C<sub>14</sub>, D4 – N<sub>12</sub>-C<sub>13</sub>-C<sub>14</sub>-C<sub>15</sub>, D5 – C<sub>13</sub>-C<sub>14</sub>-C<sub>15</sub>-C<sub>16</sub>, D6 – C<sub>14</sub>-C<sub>15</sub>-C<sub>16</sub>-C<sub>17</sub> and D7 – C<sub>15</sub>-C<sub>16</sub>-C<sub>17</sub>-C<sub>18</sub>

Fig. 2 – Superposition of three AMBER conformers with energies in a thermal energy range of 0-0.5 kcal above the energy of the most stable conformer (thick lines).



Optimization with AM1 semiempirical MO method of the geometries presented in Table 1 resulted in the same number of conformers, but they are spread on a larger energy range ( $\Delta H_f$  is between -62.008 and -56.773 kcal/mol) as can be

seen from results depicted in Table 2, where conformers are ordered in the ascending order of their formation energy. The order of AMBER conformers according to their energy is modified for the AM1 conformers.

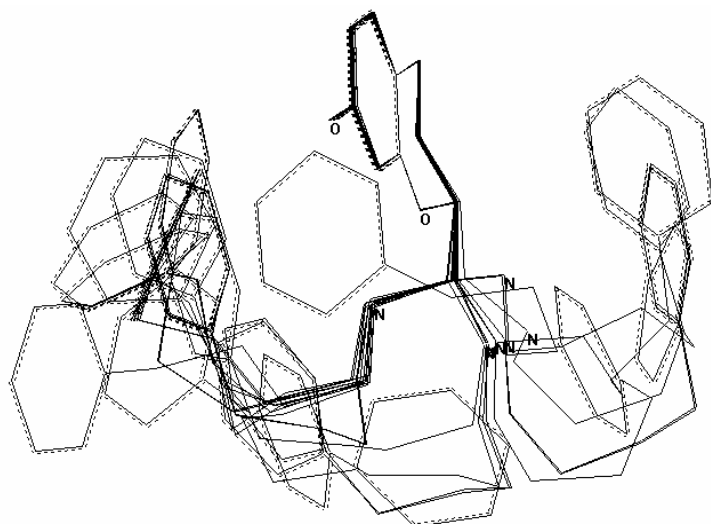


Fig. 3 – Superposition of 27 AMBER conformers with energies in a range of 1.5 kcal above the energy of the most stable conformer.

Table 2

Heat of formation, dipole moment and dihedral values for 100 lowest energy conformers resulted from geometry optimization with AM1 method

No	$\Delta H_f$	DM	D1	D2	D3	D4	D5	D6	D7
39	-62.008	2.154	-68.290	173.507	-69.412	-61.004	-172.753	-70.483	-65.345
92	-61.931	2.308	-69.751	170.009	-172.421	72.238	174.061	177.875	-72.851
95	-61.694	1.743	65.871	-167.150	172.872	-71.808	-173.600	-177.671	72.797
62	-61.646	2.619	-66.945	84.210	-163.583	70.833	64.561	171.811	74.805
78	-61.494	1.636	65.137	-169.129	167.432	65.025	176.377	177.807	-70.926
71	-61.022	2.374	-68.269	157.843	-162.123	63.931	66.363	-169.339	72.511
100	-60.996	1.658	65.412	-170.762	170.267	-73.119	-171.728	72.081	73.667
83	-60.964	1.751	65.819	-168.695	165.806	-77.955	177.122	-76.681	-78.261
48	-60.947	3.096	47.058	76.367	-113.494	-71.684	-175.238	72.118	73.044
84	-60.927	1.839	66.099	-167.355	174.546	-67.334	106.054	178.949	-74.273
47	-60.914	3.040	42.895	72.571	-106.138	-75.442	-175.826	72.001	75.413
97	-60.885	3.214	-62.593	-72.661	170.552	64.578	-100.650	-69.801	-74.951
88	-60.829	1.778	64.866	-167.112	167.284	64.946	-174.330	-72.345	-70.795
98	-60.829	1.779	64.840	-167.025	167.281	64.991	174.375	-72.339	-70.954
18	-60.771	1.905	63.491	-155.400	157.798	-65.405	-67.396	170.839	-70.577
15	-60.769	1.896	63.529	-156.155	157.071	-64.843	-66.498	171.133	-70.691
93	-60.755	1.555	66.482	-167.244	171.843	-72.897	-174.478	-72.482	-68.674
61	-60.739	1.883	62.416	-164.265	161.601	-65.388	106.479	-167.003	82.525
45	-60.719	1.536	65.405	-168.906	165.139	-77.807	178.121	-76.221	-75.682
53	-60.677	1.733	66.248	-167.516	168.376	65.972	178.254	73.519	70.378
82	-60.665	1.748	65.226	-170.262	165.150	62.964	175.468	71.242	66.692
42	-60.649	1.724	65.790	-167.492	166.792	65.492	179.648	73.620	79.248
54	-60.635	1.819	64.744	-165.169	170.411	-65.747	106.006	-172.184	78.797
58	-60.519	2.813	-62.553	-73.566	172.239	-77.415	-100.111	73.214	69.129
31	-60.514	2.830	-63.888	-75.310	172.515	-79.583	-92.288	78.721	70.716
43	-60.296	1.485	65.963	-168.466	71.738	70.916	73.649	-95.222	-64.309
27	-59.938	1.757	65.288	-171.182	146.983	-93.953	71.880	64.165	57.666
29	-59.888	2.445	-68.198	171.325	-163.648	-66.672	95.158	65.137	76.239
5	-59.707	1.968	-69.287	172.731	-170.245	77.631	77.228	-89.786	-64.191
4	-59.680	1.810	65.984	-166.937	168.738	70.231	-91.811	-66.152	-77.541
13	-59.490	1.632	65.475	-169.806	154.191	-95.485	70.266	63.109	55.662
34	-59.470	1.464	66.370	-171.006	172.970	-77.593	-77.548	90.231	65.073
87	-59.362	1.521	65.369	-163.195	174.390	-75.051	-73.892	91.137	61.990
24	-58.526	1.729	-166.290	179.690	-175.170	76.969	70.746	-99.996	-64.272
66	-58.419	2.741	-164.204	178.783	174.216	-76.181	-69.978	100.594	64.274
99	-58.170	3.488	75.586	-179.292	-169.786	-65.462	-75.723	178.271	-72.107
94	-57.332	1.867	-84.204	60.891	162.936	70.504	86.010	-85.804	-69.625
68	-56.774	3.228	-76.225	-173.957	177.385	71.819	88.125	-75.970	-61.663

$\Delta H_f$  – heat of formation (kcal/mol), DM – dipole moment (Debye), for the significance of the dihedrals see footnote of Table 1.

The distribution of conformers above the lowest energy conformer, **39**, is the following: four conformers are in the range of thermal energy (0 – 0.5 kcal/mol), six conformers are in the range of 1 kcal/mol, 24 conformers are in a range of 1.5 kcal/mol, 26 conformers are in a range of 2 kcal/mol, and 31 conformers are in a range of 2.5 kcal/mol above the lowest energy conformer and all (38) conformers are distributed in a range of 5.24 kcal/mol.

Comparing Fig 2 and Fig 4, displaying the superposition of the AMBER and AM1 conformers in the energy domain of thermal energy, it results

that the AM1 method gives a larger conformational space than AMBER does for the conformers in the same energy domain.

In Fig. 5 the superimposition of the lowest energy conformers resulted from the two methods is displayed. One can observe that the AM1 conformer has a more extended conformation than that of the AMBER conformer. As can be seen in Table 3 the lowest energy AMBER conformer has smaller distances between unbounded atoms than the distances displayed by the lowest energy AM1 conformer distances between atoms are expressed in pm.

Table 3

Comparison of dihedral and distance values in the lowest energy conformers obtained with AMBER and AM1 methods

No\dihedral	Energy	D1	D2	D3	D4	D5	D6	D7	
AMBER 1	3.816	53.819	-179.208	175.530	69.642	-69.141	-57.790	118.298	
AM1 39	-62.008	-68.290	173.507	-69.412	-61.004	-172.753	-70.483	-65.345	
No\distance		C <sub>8</sub> -C <sub>11</sub>	C <sub>8</sub> -N <sub>12</sub>	C <sub>8</sub> -C <sub>13</sub>	C <sub>8</sub> -C <sub>14</sub>	C <sub>8</sub> -C <sub>15</sub>	C <sub>8</sub> -C <sub>16</sub>	C <sub>8</sub> -C <sub>17</sub>	C <sub>8</sub> -C <sub>18</sub>
AMBER 1	446.4	472.8	602.2	653.0	705.6	620.3	542.1	405.3	
AM1 39	471.5	506.3	625.6	640.5	587.0	666.7	646.5	544.8	

Fig. 4 – Superposition of the AM1 conformers having energies in a range of 0-0.5kcal/mol above the energy of the most stable conformer, **39** (thick line).

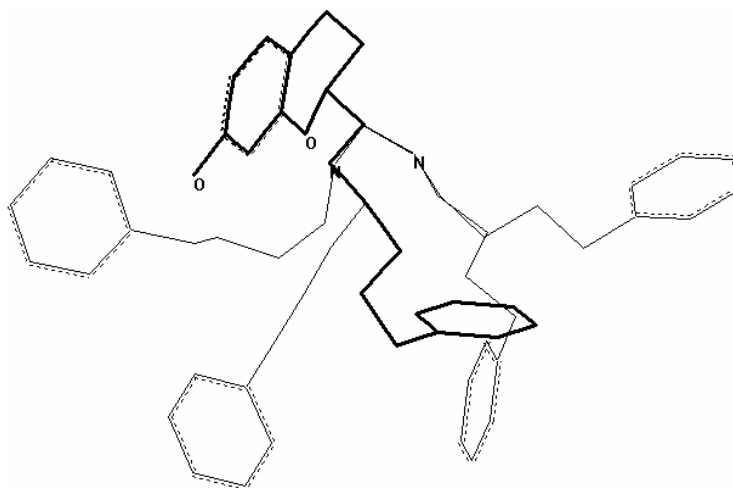
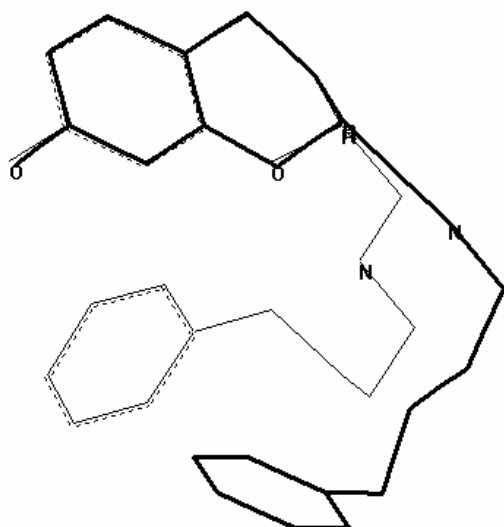


Fig. 5 – Superposition of the lowest energy conformers resulted from AMBER, conformer **1** and AM1 conformer **39** (thick line).

Values of the conformer energy suggest that a large number of conformers can populate the low energy levels on the potential energy surface and this number of conformers could be obtained from synthesis. It is interesting to find out what are the rotation energies of flexible bonds of these conformers and if these conformers could be able to surmount the rotation barriers energy for achieving other conformations more appropriate for interactions with the amino acid residues from

the D2 receptor active site. In Fig. 6 the rotation barrier energies of the dihedrals:  $O_1-C_2-C_{11}-N_{12}$ ,  $C_{11}-N_{12}-C_{13}-C_{14}$ ,  $N_{12}-C_{13}-C_{14}-C_{15}$ ,  $C_{14}-C_{15}-C_{16}-C_{17}$ , and  $C_{15}-C_{16}-C_{17}-C_{18}$  are displayed. Rotation of the  $C_2-C_{11}$ , and  $C_{15}-C_{16}$  bonds needs energies lower than 8 kcal/mol, while rotation of the  $C_{11}-N_{12}$ ,  $C_{13}-C_{14}$ , and  $C_{16}-C_{17}$  bonds (corresponding to the  $N_{12}-C_{13}-C_{14}-C_{15}$ ,  $C_{14}-C_{15}-C_{16}-C_{17}$ , and  $C_{15}-C_{16}-C_{17}-C_{18}$  dihedrals) needs energies over 15 kcal/mol.

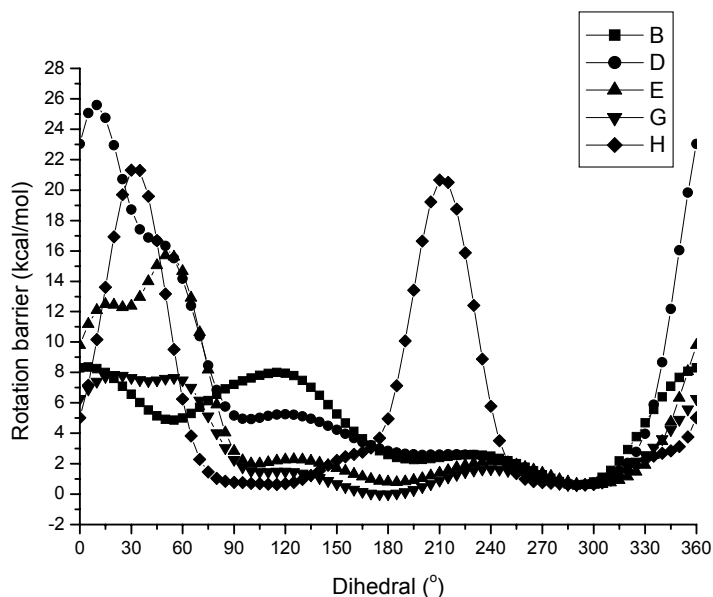


Fig. 6 – Rotation barriers for dihedrals of the lowest energy conformer obtained from conformational search using the AM1 method. Notation of dihedrals in the legend is: B –  $O_1-C_2-C_{11}-N_{12}$ , D –  $C_{11}-N_{12}-C_{13}-C_{14}$ , E –  $N_{12}-C_{13}-C_{14}-C_{15}$ , G –  $C_{14}-C_{15}-C_{16}-C_{17}$ , H –  $C_{15}-C_{16}-C_{17}-C_{18}$ .

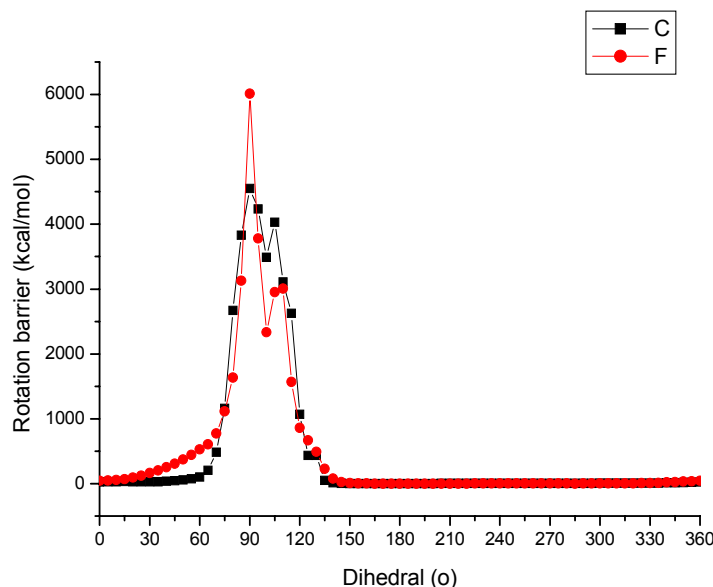


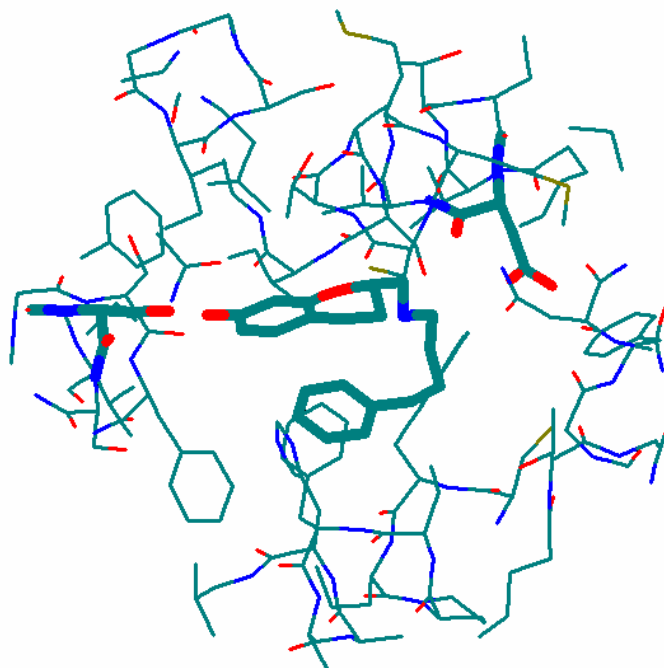
Fig. 7 – Rotation barriers for dihedrals of the lowest energy conformer obtained from conformational search using the AM1 method. Notation of dihedrals in the legend is: C for  $C_2-C_{11}-N_{12}-C_{13}$  and F for  $C_{13}-C_{14}-C_{15}-C_{16}$ .

In Fig. 7 the rotation barrier energies of the dihedrals:  $C_2-C_{11}-N_{12}-C_{13}$ , and  $C_{13}-C_{14}-C_{15}-C_{16}$ , is displayed. The conformer energies for the  $C_{11}-N_{12}$  and  $C_{14}-C_{15}$  bonds are very high for the dihedral values between 60 and 120°.

The ligands of this series have been docked in the active site of a D2 receptor 3D model,

constructed by homology with the X-ray structure of rhodopsin (pdb entry 1F88). We compared the active conformation of the ligand in the receptor site with the conformers obtained through conformational analysis (Fig. 8).

Fig. 8 – The AMBER lowest energy conformer in the D2 receptor active site.



Superposition of the lowest energy AMBER conformer on the lowest energy conformer docked in the D2 receptor active site revealed that the first one can occupy in the situs a position similar to that of the docked conformer. This observation suggests that the D2 receptor ligand binding domain is large enough to allow conformational changes of the ligand in order to achieve optimum ligand-receptor interactions.

We performed the docking of S conformers at C2 position on the chromane moiety, this leads to an interesting observation: the hydrogen bond between Ser144 and chromane OH group are not possible anymore. This observation suggests the stereoselective behaviour of the receptor interacting with the chromane derivatives.

## CONCLUSIONS

Conformational analysis of a potent ligand, 2-[(4-phenyl-butylamino)-methyl]-chroman-7-ol, acting as partial agonist on the D2 receptor showed that the conformational space occupied by the

ligand is large. The AM1 conformers can be included in a parallelepiped with edges of 12.68, 12.88 and 13.97 Å respectively.

For ligands having flexible bonds, the AM1 method gives conformers in more extended conformations than those obtained with the AMBER method.

According to the small energy barriers between different conformers the ligand can adopt easily conformations suitable for interaction with residues from the D2 receptor ligand binding domain.

The active site of the D2 receptor is large enough to allow conformational changes of the ligand in order to achieve the optimum ligand-receptor interactions.

Docking studies revealed that D2 receptor had a stereoselective behaviour, preferring (2R)-chromane derivatives.

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*Dedicated to Professor Victor-Emanuel Sahini  
on the occasion of his 80th anniversary*

## QSAR STUDY FOR A SERIES OF IMIDAZO[1,5-*A*]- QUINOXALINE AND PYRAZOLO[1,5-*C*]QUINAZOLINE DERIVATIVES ACTING ON GABA<sub>A</sub>/BENZODIAZEPINE RECEPTOR

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Some classical QSARs on a large series of 4,5-dihydro-imidazo[1,5-*a*]quinoxaline and pyrazolo[1,5-*c*]quinazolines derivatives acting at GABA<sub>A</sub>/benzodiazepine receptor are presented. The MTD method has been applied and the MTD descriptor has been used together with other descriptors in MLR models. The best model includes ten descriptors: one Galvez topological charge index, seven BCUT descriptors mediated on atomic polarizabilities or on Sanderson atomic electronegativities and two 2D autocorrelation descriptors. This model suggests that the quinoxaline and quinazoline derivatives could interact with amino acids residues of the receptor by electrostatic or charge transfer processes.

### INTRODUCTION

Since 1960 benzodiazepines that act as agonists at the GABA<sub>A</sub>/benzodiazepine receptor have been used clinically for their anxiolytic, sedative and anticonvulsive effects. The disadvantages of these drugs consist in difficulty to separate these actions which, in time, lead to various side effects such as ataxia and amnesic-like effects, ethanol and barbiturate potentiation, sedation, as well as tolerance and dependence. There has been some extensive structure–activity studies aimed for developing new potent ligands having lower side effects, acting on GABA<sub>A</sub>/benzodiazepine receptor. The aim of this work was to obtain quantitative structure-activity relationships in order to evidence the structural properties of quinoxaline and quinazoline derivatives that influence significantly their biological activity at the GABA<sub>A</sub>/benzodiazepine receptor.

### METHODS

Structure and biological activities determined by Carter et al<sup>1-3</sup> for 128 of 4,5-dihydro-

imidazo[1,5-*a*]quinoxaline and pyrazolo[1,5-*c*]quinazolines derivatives have been used to obtain regressional models between the biological activity (expressed as the reciprocal logarithm of inhibition constant, pKi) and minimum topological difference, MTD, and some other descriptors. Backbone structures are presented in Scheme 1.

Molecular geometries of 4,5-dihydro-imidazo[1,5-*a*]quinoxaline and pyrazolo[1,5-*c*]quinazolines derivatives have been optimized using the AMBER99 force field and then the AM1 Hamiltonian.<sup>4</sup> Optimizations were performed using the Polak-Ribiere conjugate gradient algorithm. All methods are implemented in HyperChem7.52.<sup>5</sup> Geometry was optimized till the RMS gradient was equal to or less than 0.01 kcal/Åmol. For some representative structures a conformational search has been carried out using the Conformational Search module from Hyperchem7.52.

MTD method<sup>6,7</sup> is based on the hypothesis that a potent ligand fits into the receptor site as a key in its lock. Such a ligand should have the complementary shape of the receptor site. MTD method defines the minimum topological difference between a “standard” (a good model of

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a high active ligand) and each compound from the series. The less the MTD value is, the better the activity is. In order to calculate the MTD values, a hypermolecule is built through maximum atom by atom superposition of the  $i = 1 \dots N$  molecules from the series. Hypermolecule contains a number  $j = 1 \dots M$  of vertices. A vertex is composed from all superposed atoms which are placed at a distance no greater than 0.5 Å. In hypermolecule, for each vertex a vector  $\varepsilon_j$  and an index  $x_{ij}$  are assigned. The  $\varepsilon_j$  vector can take the values 1 (detrimental vertices), 0 (irrelevant vertices) or -1 (beneficial vertices). The  $x_{ij}$  index represents the occupancy of the  $j$  hypermolecule vertices by the atoms of the " $i^{\text{th}}$ " molecule. It can take two values  $x_{ij} = 1$  for occupied and  $x_{ij} = 0$  for unoccupied vertices.

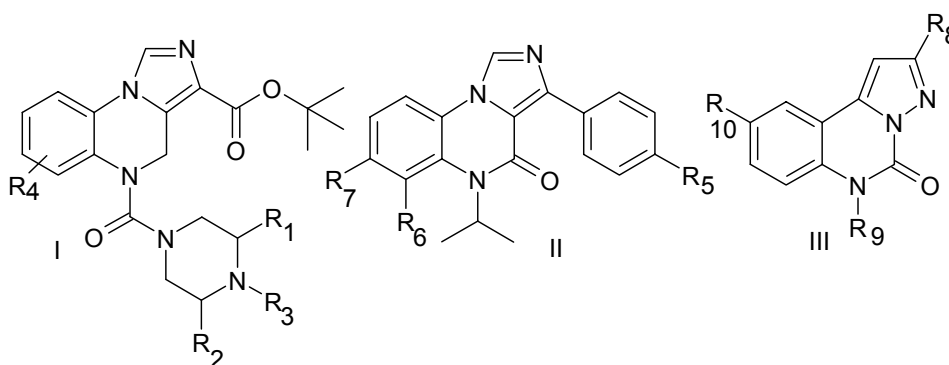
MTD<sub>*i*</sub>, the MTD value for the " $i^{\text{th}}$ " molecule from the series is calculated using the equation:

$$\text{MTD}_i = s + \sum_{j=1}^M \varepsilon_j x_{ij} \quad (1)$$

where  $s$  represents the number of vertices with  $\varepsilon_j = -1$  of a "standard" molecule; this can be the most active compound or a generic molecule which contains predominantly atoms or groups of atoms from the most active structures. The assignment of  $\varepsilon_j$  values is a cyclic process that ends when the best correlation coefficient is obtained for the equation:

$$A_i = a + a_1 X_{i1} + a_2 X_{i2} + \dots - b \text{MTD}_i \quad (2)$$

$X_{i1}, X_{i2}, \dots$  are other descriptors, the 'a' and 'b' values are the regression coefficients.



Scheme 1. Backbone structures of the quinoxaline and quinazoline derivatives.

A hypermolecule (Fig 1) with  $j=78$  vertices was obtained through atom by atom maximum superposition of molecules. After the building of the hypermolecule, the initial set of 128 quinoxaline and quinazoline derivatives was split in learning (99 compounds) and test sets (29 compounds). The splitting was performed manually by randomly choosing (for the test set) compounds having biological activities as diverse as possible, so that the distribution of compounds in the learning and test sets to be as similar as possible. In order to check the stability of data set, two different splittings have been carried out. Figures 1 and 2 display the distribution of compounds in the pair of test sets. As one can see the first test set has a worse distribution than the second test set.

Applying MTD method, the MTD descriptors have been calculated and they have been used in different multiple linear regressions (MLR). Apart of MTD, a number of descriptors have been calculated with the QSAR properties module from HyperChem7.52 and DRAGON3.0 softwares.<sup>8</sup>

From the AM1 method the energies of the highest occupied molecular orbital,  $E_{\text{HOMO}}$  and of the lowest unoccupied molecular orbital  $E_{\text{LUMO}}$ , total dipole moment and  $sp$  hybridization component of dipole moment have been calculated. Also, absolute electronegativity ( $\chi$ ) and chemical hardness ( $\eta$ ) have been calculated as  $\chi \approx (\text{IP} + \text{EA})/2$  and  $\eta \approx (\text{IP} - \text{EA})/2$  assuming that ionization potential,  $\text{IP} \approx -E_{\text{HOMO}}$  and electron affinity,  $\text{EA} \approx -E_{\text{LUMO}}$ .<sup>9,10</sup> From AM1 method resulted also a thermodynamic property, energy of formation,  $\Delta H_f$ . Also, as descriptors have been tested energy terms (bond, angle, torsion and van der Waals energies) resulted for geometries of conformers optimized with the AMBER99 force field. In order to select variables the forward and backward methods from STATISTICA5.0<sup>11</sup> have been applied.

## RESULTS

In order to apply the MTD method seven starting receptor maps have been built, where the

assignment of vertices was done assuming that the beneficial vertices should be present especially in the highest active compounds and the detrimental

vertices should be present mainly in the lowest active compounds. The optimized receptor map that gave the best correlation coefficient was:

$$S^* \begin{cases} \varepsilon_j = -1 : 3 - 10, 13, 15, 16, 22, 44, 54, 58, 63, 66, 67 \\ \varepsilon_j = 0 : 14, 17, 19 - 2, 23, 25, 27, 29 - 31, 34 - 43, 45 - 51, 57, 59, \\ \varepsilon_j = 1 : 1, 2, 11, 12, 18, 24, 26, 28, 32, 33, 52, 53, 55, 56, 60 - 62, 64, 68 - 78 \end{cases}$$

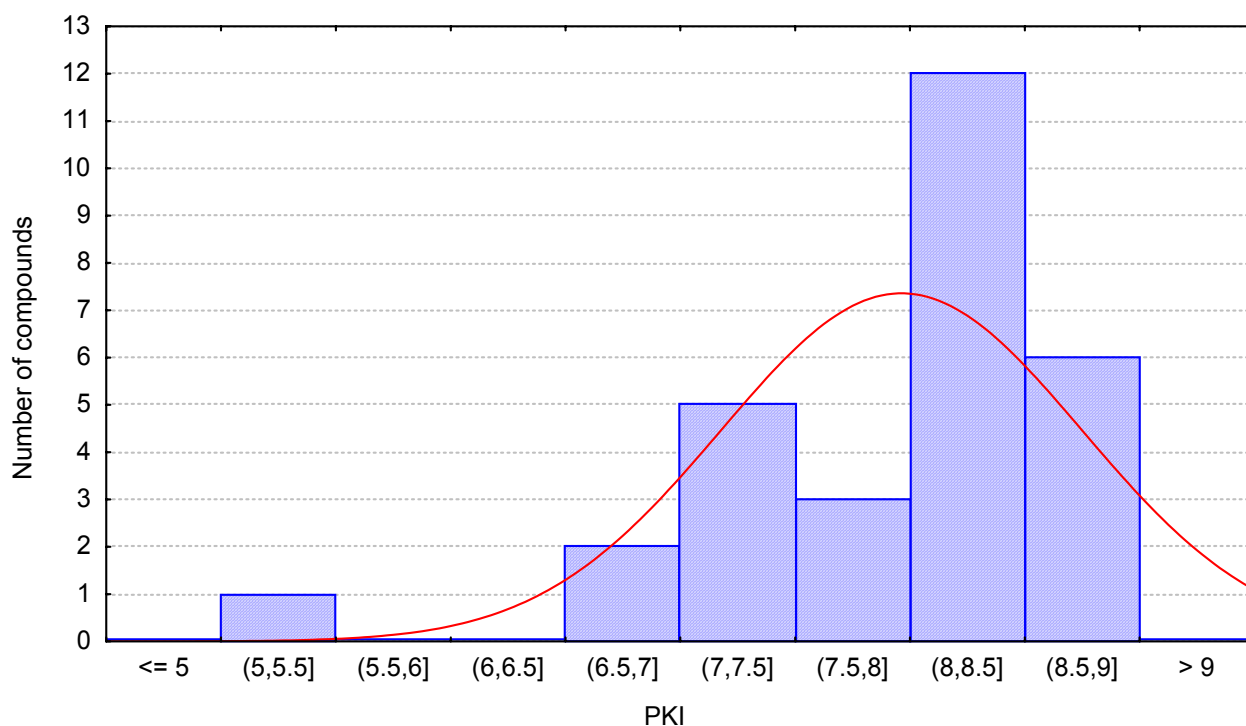


Fig. 1 – Distribution of the compounds from the first test set against their biological activity, pKi.

This map, S\*, resulted for the second learning set, was used to calculate the MTD values for both sets of compounds from the first and second learning set and for the two test sets associated

with the learning sets. The MTD models obtained with this map are significant, but very modest. The MTD regression model for the first learning set is:

$$pKi = 17.1429(\pm 0.6784) - 0.4623(0.0327) \text{ MTD} \tag{3}$$

$$n = 99 \quad r = 0.821 \quad s = 0.524 \quad F = 200.16 \quad r^2_{cv} = 0.502 \quad \text{press} = 40.61$$

and the MTD regression model for the second learning set is:

$$pKi = 15.5463 (\pm 0.481365) - 5072 (0.030865) \text{ MTD} \tag{4}$$

$$n = 99 \quad r = 0.858 \quad s = 0.454 \quad F = 270.04 \quad r^2_{cv} = 0.645 \quad \text{press} = 26.91$$

Using the regressions (3) and (4) the pKi values of the test sets have been calculated. The pKi calculated values, the experimental pKi values and the differences between them for the compounds from the two test sets are displayed in Table 1. Although the differences obtained with the

regression (4) are slightly better than those obtained with regression (3), for some compounds the differences are over one logarithm unit. Given that the standard error of estimate was 0.524 and 0.454, respectively for the regressions (3) and (4) these results were expectable.

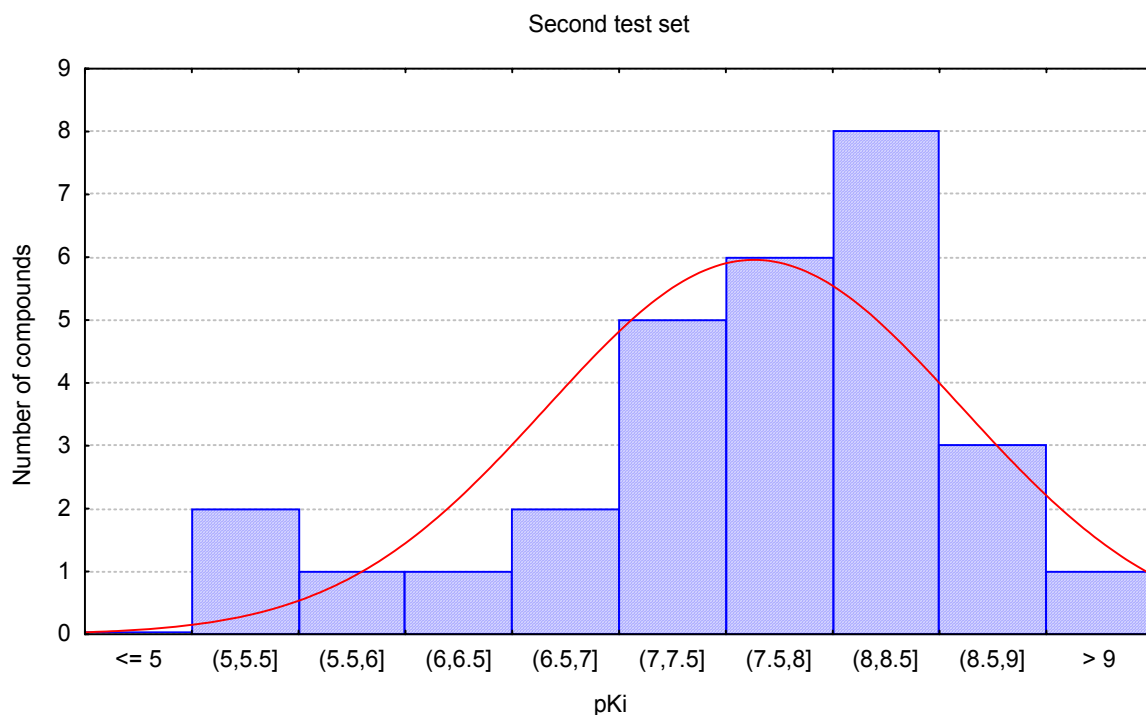


Fig. 2 – Distribution of compounds from the second test set against their biological activity, pKi.

Table 1

pKi values calculated with regressions (3) and (4)

First test set; pKi calculated with eqn. (3)					Second test set; pKi calculated with eqn. (4)				
No	MTD	pKi exp	pKi calc	Diff	No	MTD	pKi exp	pKicalc	Diff
2	21	8.517	7.435	1.082	1	14	8.386	8.448	-0.062
3	22	8.176	6.972	1.204	7	14	7.964	8.448	-0.484
8	20	8.485	7.897	0.588	13	14	8.332	8.448	-0.116
9	21	8.349	7.435	0.914	14	14	8.022	8.448	-0.426
11	23	7.250	6.51	0.740	22	14	8.870	8.448	0.422
13	20	8.332	7.897	0.435	51	14	8.225	8.448	-0.223
14	21	8.022	7.435	0.587	53	14	7.747	8.448	-0.701
15	22	6.602	6.972	-0.370	54	14	8.262	8.448	-0.186
19	23	7.700	6.51	1.190	56	14	8.287	8.448	-0.161
20	23	7.401	6.51	0.891	69	18	7.402	6.42	0.982
21	22	8.280	6.972	1.308	72	17	5.772	6.927	-1.155
24	21	8.255	7.435	0.820	74	18	7.836	6.42	1.416
47	20	7.521	7.897	-0.376	78	16	8.467	7.434	1.033
61	25	7.415	5.585	1.830	79	15	8.529	7.941	0.588
62	23	7.499	6.51	0.989	80	15	7.955	7.941	0.014
65	24	6.939	6.048	0.891	82	13	9.013	8.955	0.058
78	22	8.467	6.972	1.495	87	15	8.783	7.941	0.842
79	21	8.529	7.435	1.094	88	17	6.666	6.927	-0.261
80	21	7.955	7.435	0.520	89	17	7.313	6.927	0.386
83	19	8.812	8.359	0.453	93	17	6.476	6.927	-0.451
85	20	8.233	7.897	0.336	95	15	7.456	7.941	-0.485
86	21	7.879	7.435	0.444	96	16	7.326	7.434	-0.108
87	21	8.783	7.435	1.348	98	15	7.688	7.941	-0.253
104	19	8.604	8.359	0.245	111	17	7.553	6.927	0.626
105	20	8.296	7.897	0.399	113	18	7.229	6.42	0.809
107	22	7.324	6.972	0.352	114	19	6.979	5.913	1.066
108	19	8.857	8.359	0.498	121	17	8.131	6.927	1.204
109	20	8.381	7.897	0.484	126	17	5.483	6.927	-1.444
128	23	5.182	6.51	-1.328	128	18	5.182	6.42	-1.238

Seven compounds, 13,14,78-80,87,128 belong to both test sets. Comparing their differences ( $pK_i \text{ exp} - pK_i \text{ calc}$ , exhibited in columns Diff) one can observe that the differences for the compounds in the second test set are slightly lower than those of the compounds from the first test set, however

globally, the results of the regression (4) are also very modest. In order to obtain better statistical models the MTD descriptor was associated with other partner descriptors. The indices of different statistical significant models (obtained with the first learning set) are collected in Table 2.

Table 2

Statistical parameters for some significant multiple linear regression

Descriptors		r	r <sup>2</sup>	r <sup>2</sup> adj	s	F
MTD,logP,RDF025p,RDF055p	(3)	0.853	0.727	0.716	0.486	62.72
MTD,logP,MoR32e	(4)	0.853	0.728	0.720	0.483	84.92
MTD,logP,JG11,GG11	(5)	0.859	0.737	0.726	0.477	65.95
MTD,logP,BELe1,BELe2,BEHp7,BELp1	(6)	0.872	0.761	0.745	0.460	48.84
MTD,logP,HATS0v,HATS1v,HATS4v,HATS5m	(7)	0.873	0.762	0.747	0.459	49.92
MTD, MoR09m, MoR10m, MoR12m, MoR18m, MoR21m, MoR25m	(8)	0.874	0.764	0.746	0.460	42.05
MTD,logP,X2A,X0v,X1v,X0Av,X1Av	(9)	0.875	0.765	0.747	0.459	42.31
MTD,logP,HATS2p,HATS4p,HATS5p,HATS7p,H0p,RTu,R4u	(10)	0.901	0.812	0.791	0.417	38.05

Log P – log 1-octanol/water partition coefficient calculated with QSAR properties module from HyperChem7.52; piPC04 – molecular multiple path count of order 04; RDF025p – radial distribution function -2.5 / weighted by atomic polarizabilities; RDF055p – radial distribution function -5.5 / weighted by atomic polarizabilities; MoR32e – 3D-MoRSE - signal 32 / weighted by atomic Sanderson electronegativities; JG11 – mean topological charge index of order 1; GG11 – topological charge index of order 1; BELe1 – lowest eigenvalue n. 1 of Burden matrix / weighted by atomic Sanderson electronegativities; BELe2 – lowest eigenvalue n. 2 of Burden matrix / weighted by atomic Sanderson electronegativities; BEHp7 – highest eigenvalue n. 7 of Burden matrix / weighted by atomic polarizabilities; BELp1 lowest eigenvalue n. 1 of Burden matrix/weighted by atomic polarizabilities; HATS0v – leverage-weighted autocorrelation of lag 0 / weighted by atomic van der Waals volumes; HATS1v – leverage-weighted autocorrelation of lag 1 / weighted by atomic van der Waals volumes; HATS4v – leverage-weighted autocorrelation of lag 4 / weighted by atomic van der Waals volumes; HATS5m – leverage-weighted autocorrelation of lag 5 / weighted by atomic masses; MoR09m – 3D-MoRSE - signal 09 / weighted by atomic masses; MoR010m – 3D-MoRSE - signal 10 / weighted by atomic masses; MoR12m – 3D-MoRSE signal 12 / weighted by atomic masses; MoR18m – 3D-MoRSE - signal 18 / weighted by atomic masses; MoR21m – 3D-MoRSE - signal 21 / weighted by atomic masses; MoR25m – 3D-MoRSE - signal 25 / weighted by atomic masses; X2A – average connectivity index chi-2; X0v – valence connectivity index chi-0; X1v – valence connectivity index chi-1; X0Av – average valence connectivity index chi-0; X1Av – average valence connectivity index chi-1; HATS2p leverage-weighted autocorrelation of lag 2 / weighted by atomic polarizabilities; HATS4p – leverage-weighted autocorrelation of lag 4 / weighted by atomic polarizabilities; HATS5p – leverage-weighted autocorrelation of lag 5 / weighted by atomic polarizabilities; HATS7p – leverage-weighted autocorrelation of lag 7 / weighted by atomic polarizabilities; H0p – H autocorrelation of lag 0 / weighted by atomic polarizabilities; RTu – R total index / unweighted; R4u – R autocorrelation of lag 4 / unweighted.

Even if the regressions from Table 2 also have low statistical indices, but the descriptors included in the models could give some information regarding the importance of various structural properties for the biological activity of quinoxaline and quinazoline derivatives. For example, logP is present in almost all models suggesting that the ligand hydrophobicity influences in a certain degree the  $pK_i$  values. The presence of descriptors weighted on atomic polarizabilities or atomic electronegativities Sanderson such as: radial distribution function, BCUT, GETAWAY descriptors could signify that  $pK_i$  is influenced by electronic or electrostatic properties of the compounds.

Because the MTD has a high partial correlation coefficient there are only few tolerated descriptors as partner descriptors. The higher the MTD partial correlation coefficient is, the fewer MTD partner descriptors are. Therefore for the second learning set the best obtained correlation had 0.89 for the total correlation coefficient. In these cases, besides steric information, MTD descriptor contains hydrophobicity and electronic information. Its steric nature is covered by other effects.

In order to obtain other models with better statistical indices we used a set of descriptors provided by DRAGON3.0. The outliers were eliminated and the following regression was obtained:

$$\begin{aligned}
 \text{pKi} = & 95.8351(\pm 10.101) + 3.052353(\pm 0.61579) \text{BELm6} - 3.356161(\pm 0.3067) \text{BEHv8} - \\
 & 71.29569(\pm 3.970161) \text{BELv3} + 9.463132(\pm 0.725744) \text{BEHe4} - 47.8192(\pm 4.820646) \text{BELe1} + \\
 & 42.05471(\pm 4.48708) \text{BEHp1} + 73.1973(\pm 3.94718) \text{BELp3} - 3.953559(\pm 0.295809) \text{GGI7} + \\
 & 0.047589(\pm 0.005007) \text{ATS8e} - 0.108897(\pm 0.010495) \text{ATS7p} \quad (5) \\
 n = & 83 \quad r = 0.962 \quad r^2 = 0.925 \quad r^2_{\text{adj}} = 0.915 \quad s = 0.197 \quad F = 89.15 \quad \text{press} = 2.80
 \end{aligned}$$

To verify the reliability of this regression model the calculated pKi values against the experimental ones have been plotted. A high

leverage point was observed for low pKi values. By eliminating this point (compound 106) the following model was obtained:

$$\begin{aligned}
 \text{pKi} = & -97.1579(\pm 10.4675) + 3.071(\pm 0.62) \text{BELm6} - 3.3509(\pm 0.3084) \text{BEHv8} - 72.7376(\pm 4.8618) \text{BELv3} + \\
 & 9.5754(\pm 0.7608) \text{BEHe4} - 48.0567(\pm 4.8668) \text{BELe1} + 42.4055(\pm 4.5604) \text{BEHp1} + 74.6249(\pm 4.8271) \text{BELp3} \\
 & 3.9647(\pm 0.2981) \text{GGI7} + 0.0478(\pm 0.005) \text{ATS8e} - 0.1095(\pm 0.0106) \text{ATS7p} \quad (6) \\
 n = & 82 \quad r = 0.955 \quad r^2 = 0.912 \quad r^2_{\text{adj}} = 0.899 \quad s = 0.198 \quad F = 73.34 \quad \text{press} = 2.79
 \end{aligned}$$

The new pKi values calculated with the regression model (6) have been plotted against the experimental pKi values. The regression line and the distribution of the points around the regression

line can be seen in Fig. 3. The model obtained has good statistical parameters: the value of the standard error of estimate (s) is around 0.2 logarithm units, in the experimental errors domain.

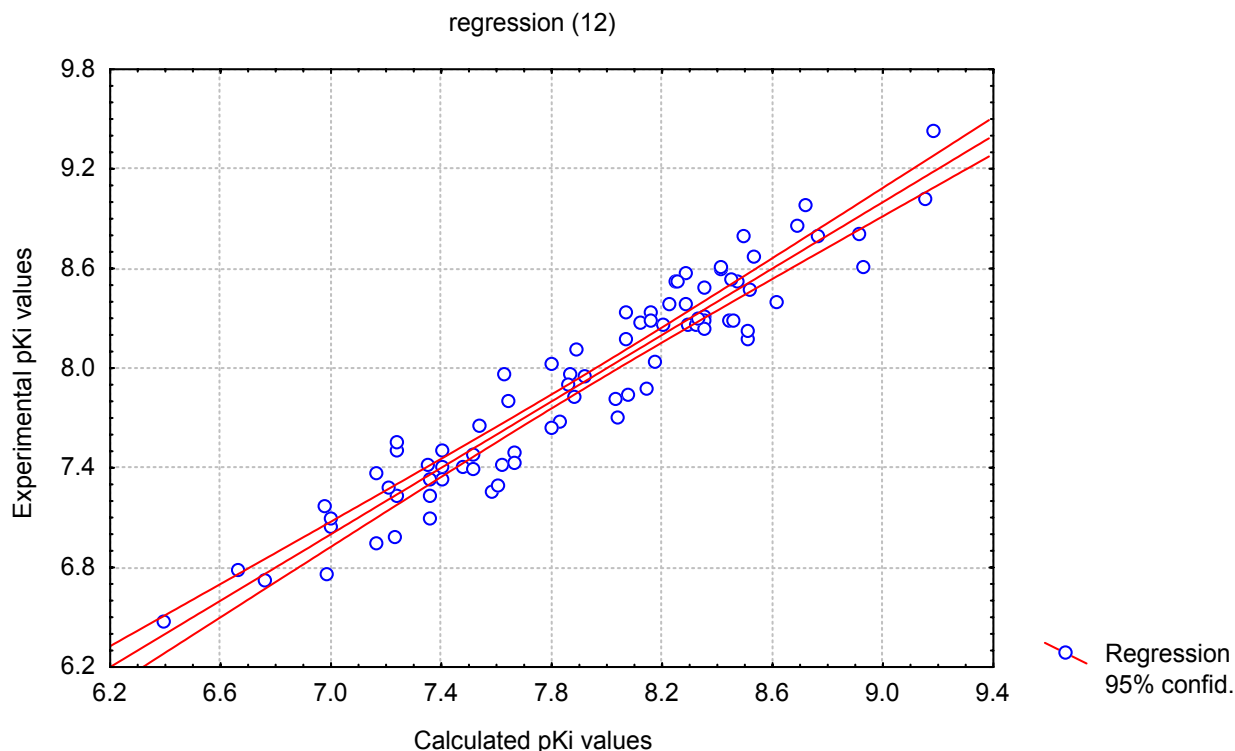


Fig. 3 – Calculated pKi against experimental pKi.

The first seven descriptors from the regression (6) are BCUT descriptors. GGI7 is Galvez topological charge index and the last two are 2D autocorrelation indices. In the BCUT group of descriptors there are three couples of descriptors weighted by: 1) atomic van der Waals volume, 2) atomic Sanderson electronegativities and 3) atomic polarizabilities. The two 2D autocorrelation indices are also weighted by atomic Sanderson

electronegativities and atomic polarizabilities. Thus, the regression model (6) is similar with the models presented in Table 2 and they suggest that pKi is influenced by electronic properties characteristic to the electrostatic and charge transfer processes. pKi is also influenced by steric properties characteristic to the unspecific interactions such as van der Waals and hydrogen bond interactions.

The results obtained should be related to the pharmacophore model proposed by Cook.<sup>12</sup> This pharmacophore contains three hydrophobic domains. In quinoxaline and quinazoline derivatives there are two regions defined by phenyl and *t*-Bu substituents on the saturate side of the 1,2,3,4-tetrahydroquinoxaline ring and another one on the opposite side, around the aromatic ring of quinoxaline. Also, the pharmacophore model contains three regions able to form hydrogen bonds, one can interact as donor and two regions as acceptors of hydrogen bonds.

### CONCLUSIONS

For a large set of quinoxaline and quinazoline derivatives active on the GABA/benzodiazepine receptor a classical QSAR study was performed using MTD and other descriptors calculated with DRAGON3.0 software.

MTD gives statistical significant but modest regression models with standard error of estimate around 0.5. Moreover, the MTD descriptor cannot be associated with other partner descriptors due to its high and complex information content.

The best regression model obtained contains ten descriptors that characterize the electrostatic and charge transfer interactions and also unspecific interactions such as hydrogen bond and van der Waals interactions.

The significance of the descriptors contained by the regression models suggests that the model is in accordance with the pharmacophore model proposed by Cook.

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Dedicated to Professor Victor-Emanuel Sahini  
on the occasion of his 80th anniversary

## THE PRIVILEGED STRUCTURES HYPOTHESIS FOR G PROTEIN-COUPLED RECEPTORS – SOME PRELIMINARY RESULTS

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Almost 20 years ago, the concept of privileged structures emerged as a potentially useful strategy in the discovery of novel biologically active compounds. Privileged structures are molecular fragments with versatile binding properties, i.e., structural chemotypes that are able to yield potent and often selective ligands for different biological targets through modification of functional groups. Thus, the identification of privileged structures becomes an important goal in the process of drug design. Herein we present a simple and efficient method for the identification of privileged structures targeting G-protein coupled receptor (GPCR).

### INTRODUCTION

Modern drug discovery is shifting focus from identifying suitable candidate drugs – which remains an essential but time-consuming goal – to identifying suitable candidate leads in order to maximize the cost-effectiveness and speed of the lead optimization process.<sup>1–5</sup> The industry is embarking in the process of efficient identification of small molecules, or small molecular fragments, which can further be optimized, in order to obtain lead structures. Though formulated almost two decades ago, the concept of privileged structures has received additional attention in recent years, in particular with the advent of combinatorial chemistry – where molecular libraries based on privileged structures are often given higher synthetic priority.

Briefly, this concept is based on the observation that certain chemical structural units occur in many ligands, i.e., “privileged (sub)structures”. Evans et al<sup>6</sup> noted that by chemical derivatization, such privileged structures can be used to develop new ligands. Often, privileged structures provide high-

affinity ligands for more than one receptor subtype,<sup>7</sup> e.g., the benzodiazepine class of drugs. The privileged structure (or scaffold) concept has been successfully used in discovering novel ligands for e.g., new growth hormone secretagogue ligands were identified by screening privileged structure compounds from other receptors.<sup>8,9</sup> Yang et al.<sup>10</sup> reported the synthesis and characterization of L-054,522 a small peptidomimetic agonist with high potency and selectivity for somatostatin receptor subtype 2. This compound was synthesized using the privileged structure dipeptide strategy, the coupling of a dipeptide to a privileged structure.<sup>11</sup> This strategy was used with good results for the generation of active and selective ligands on the melanocortin subtype-4 receptor.<sup>12–16</sup>

G-protein coupled receptors (GPCRs) are one of the most therapeutically significant classes of drug targets, and the identification of (novel) privileged GPCR scaffolds remains a priority. We hypothesized that certain privileged structures – if confirmed – might have a higher propensity to occur in drugs that target GPCRs. Such an example is phenoxymethyl-ethanolamine, which one can historically trace back to epinephrine and nore-

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pinephrine, the endogenous ligands that activate adrenergic receptors. Most of these structures act as antagonists on the adrenergic receptors (see Figure 1).

In this preliminary study, we examine the privileged structure hypothesis for GPCRs using a collection of literature compounds from the WOMBAT (World of Molecular BioAcTivity) database.<sup>17</sup> The choice for looking at GPCRs is related to the fact that at least 35% of the

therapeutically important drug targets are members of the GPCR family.<sup>18,19</sup> We applied the MDL ISIS software<sup>20</sup> to generate substructures and to query for their presence in the WOMBAT database.<sup>21</sup> In this paper, we summarize results based on over 600 substructures, used to query the 2004.2 version of WOMBAT, then highlight substructures that appear to be more specific to GPCRs.

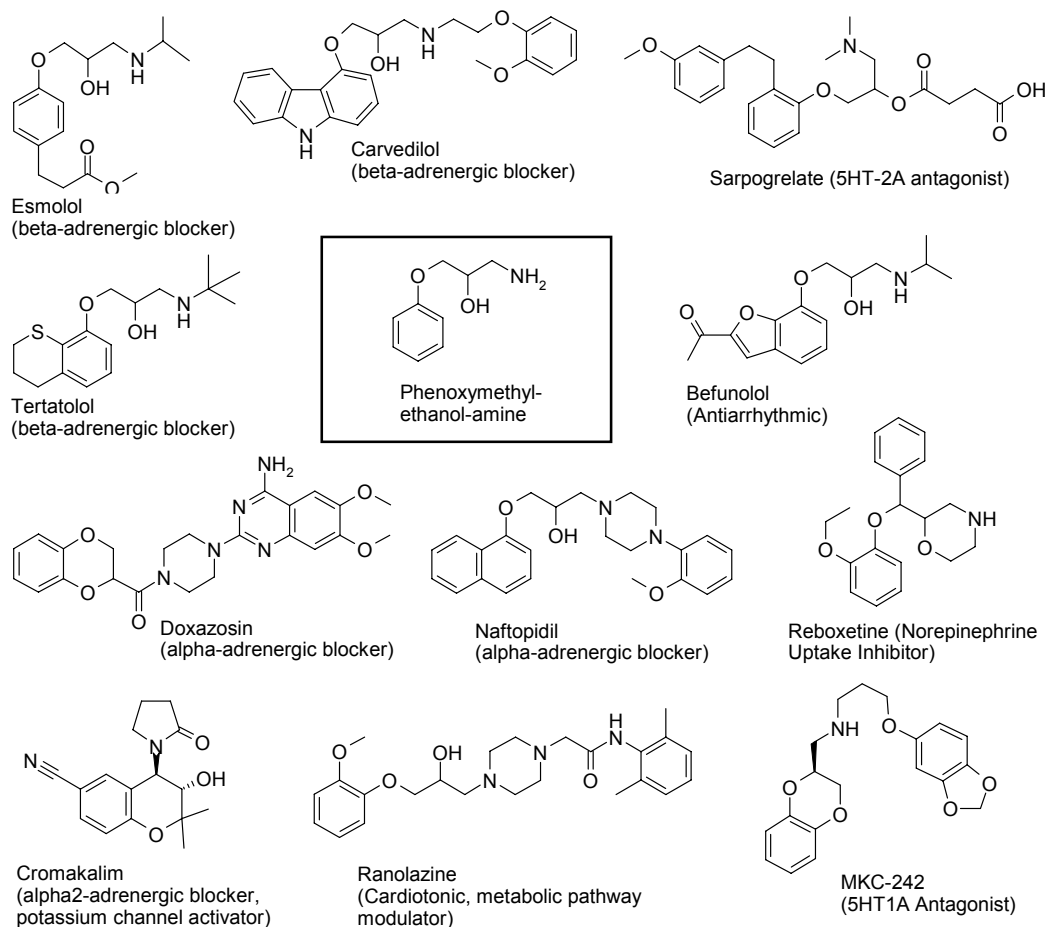


Fig. 1 – Drugs derived from phenoxyethyl-ethanolamine. These structure were derived from norepinephrine, where catechol replaces the phenyl moiety.

## METHODS

All chemical structures were manually sketched in ISISDraw2.3.1, and used as query. We then used substructure searching (fragments) in ISIS/Base2.3.1, using the WOMBAT 2004.2 database. This database is a collection of structures gathered from recent Chemistry literature, for the most part, the Journal of Medicinal Chemistry (1991-2003), and contains over 80,000 entries. Of these, over 30,000 entries were ligands tested on GPCR targets. The vast majority of these ligands were tested *in vitro*,

and approximately one third were tested on human (or recombinant human) GPCRs. Among the human-GPCR tested ligands, 1,230 have “generic names”, i.e., they are in a higher development phase in the drug discovery process.

To further validate the substructures found in WOMBAT, we used the MDDR (MDL Drug Data Report) 2002 database, in particular the subset consisting of Phase I-III and Launched drugs. MDDR database is available from MDL.<sup>22</sup> We recorded the frequency of occurrences for each substructure, comparing WOMBAT-GPCR with

MDDR and MDDR-GPCR (where “GPCR” refers to GPCR-specific entries in the two databases). We considered the MDDR-GPCR subset to be “biologically validated”, since compounds in Phase I or higher are already subject to intense scrutiny in terms of biological relevance for that particular GPCR target.

A number of 617 substructures were used for the query. Each substructure hypothesized to be privileged was stored separately in terms of occurrences in WOMBAT, in WOMBAT-GPCR, in MDDR and in MDDR-GPCR. Statistical processing of the results was performed in Microsoft Excel. For these 617 structures, the following notations were made: “F\_GPCR” represents a fragment’s occurrence in compounds from the WOMBAT-GPCR set, and “F\_DIF” represents the difference between the number of compounds indexed in the entire WOMBAT database and the number of compounds from the WOMBAT-GPCR subset, which contain within their molecule the queried structure. The selection of chemically privileged fragments was performed based on the following selection criteria:

The first selection criterion:  $F\_GPCR \geq F\_DIF$ ,

The second criterion:  $F\_GPCR > \text{median}$ . We calculated the median for F\_GPCR values in order to eliminate those fragments which have a poor contribution to the signal/noise ratio

The third criterion:  $F\_GPCR > \text{enrichment factor (EF)}$ ; where  $EF = 1 - (F\_DIF/F\_GPCR)$ . This criterion allows one to quickly highlight fragments that display a high frequency of occurrence.

## RESULTS AND DISCUSSION

Aimed at identifying privileged GPCR fragments, this preliminary study is focused on the hypothesis (are privileged substructure a valid concept?) and at cheminformatic tools that could formally be used to test this hypothesis. Several steps were observed:

Each fragment was queried separately on the two databases and their target subsets;

From these substructures, those that had relevant statistical values, i.e., F\_GPCR occurrence above zero ( $F\_GPCR > 0$ ) were considered for further study. Given this criterion, a number of 615 structures were maintained;

Criteria defined as above were applied to seek chemically privileged structures;

Biological validation of these “chemically privileged structures” was performed in MDDR.

First, we tested the:  $F\_GPCR \geq F\_DIF$  condition. From 615 fragments with  $F\_GPCR > 0$ , only 164 fragments met this criterion. These 164 structures were examined further; the F\_GPCR median value was found to be set at 15. Therefore, the second selection criterion was set as  $F\_GPCR > 15$ . Out of 164 fragments, a subset of 82 (50%) had frequency higher than the median, which was not unexpected. These 82 structures underwent further filtering, in order to quantitatively assess the *enrichment* of a particular fragment relative to the GPCR target class.

The frequency occurrence of a specific fragment in the non-GPCR set (F\_DIF) relative to its occurrence in the GPCR set was then used to determine the **enrichment factor, EF**. EF gives us a quantitative measure of the distribution of specific fragments in the GPCR subset relative to the entire database. EF values for each of the 82 significant substructures were calculated. We considered those fragments having EF values greater than the median of all EF values for the entire set of 82 fragments as “chemically privileged”. While empirical, this criterion was used to improve the signal-noise ratio and to ascertain that only meaningful substructures were examined. As expected, 50% of these structures (41 fragments) met the final criterion, and we report these in Figure 2.

The concept of “biological validation” was introduced to further discriminate “chemically privileged” substructures (i.e., those that are recurring motifs in medicinal chemistry literature) from those privileged scaffolds that have indeed passed the test of biological and perhaps clinical significance, in a setting more relevant for drug discovery.

We used MDDR for this purpose, in particular its subset of compounds in Phase I (or higher) clinical testing, as well as the subset of marketed drugs. These compounds already comply with a minimum set of therapeutic requirements (including Safety), hence structures present in this subset are considered to be “biologically validated”. After querying the 41 substructures in MDDR, the following observations were drawn. Only 6 fragments were biologically validated, namely F260, F457, F520, F543, F569, F596 in Figure 2 (these are labeled in **bold**). For some of these structures (F260, F520, F543) examples of drugs acting on GPCR which contain within their molecule the specified fragment (**in bold**) are displayed in Figure 3.

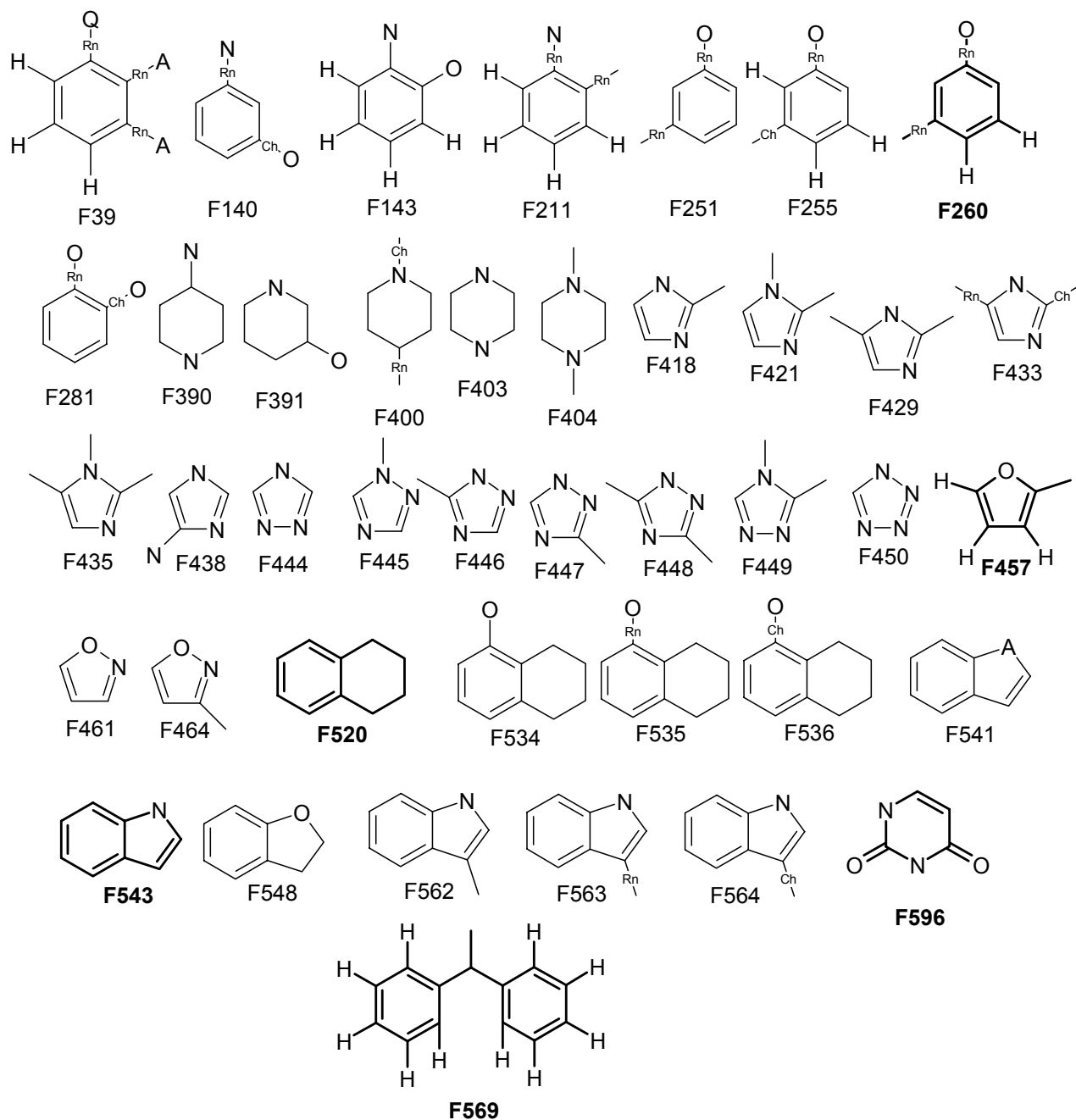


Fig. 2 – Chemically and biologically (bold) privileged structures.

Our preliminary results indicate that these 6 substructures are likely to be truly privileged scaffolds for GPCR targets. The remaining substructures remain, for now, “chemically privileged”, since they are not present in this collection of clinically relevant structures. However, we do not exclude the possibility that some of them may, indeed, turn out to be “biologically privileged” as well, for two reasons: (i) this MDDR subset is not exhaustive; (ii) new drugs are approved every year, therefore this study provides only a snapshot relevant for the 2002 drugs.

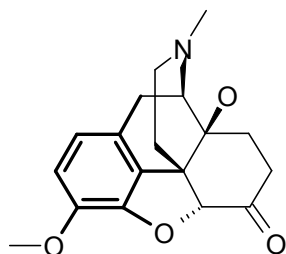
## CONCLUSIONS

Our preliminary study illustrates the application of the privileged structure concept on a rather limited number of chemical structures, and further identifies 41 substructures that are “chemically privileged”. Of these, only 6 appear to be biologically validated, which cautions one not to extend this concept onto a chemotype without first testing for clinical relevance. This study further suggests that the privileged substructure hypothesis is valid, at least with respect to those

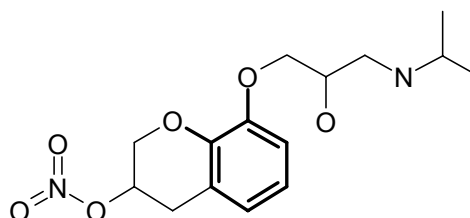
substructures that pass the “biological validation” criterion. Mining databases of biologically compounds using simple cheminformatic tools can thus become useful in the process of guiding combinatory library synthesis, which could further

yield potentially novel bioactive compounds for a specific category of biological targets.

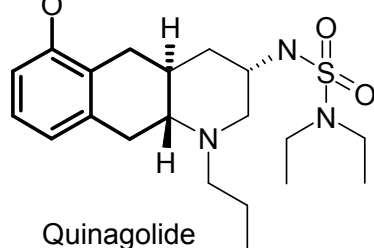
*Acknowledgment:* Part of this work was supported by the NIH Molecular Libraries Initiative award U54 MH074425-01 (TIO).



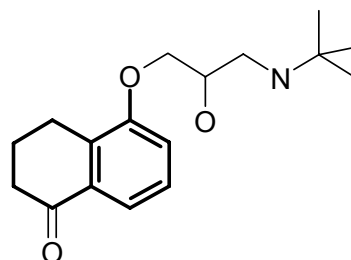
**Oxycodone**  
analgesic - narcotic  
 $\mu$ -,  $\delta$ -,  $\kappa$ - opioid receptors  
antagonist



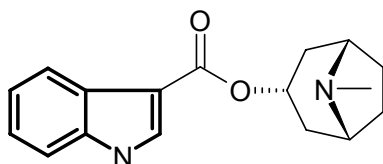
**Nipradilol**  
anti-glaucoma  
 $\alpha$ -,  $\beta$ - adrenergic receptors  
antagonist



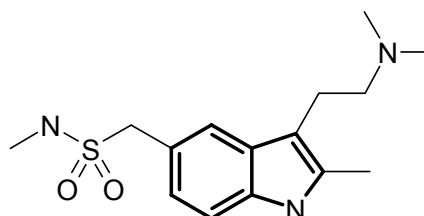
**Quinagolide**  
D2 receptor agonist



**Levobunolol**  
beta-adrenergic antagonist



**Tropisetron**  
antiemetic  
5HT3 antagonist  
5HT1 antagonist



**Sumatriptan**  
antimigraine  
5HT1B/D agonist  
5HT1F agonist

Fig. 3 – Examples of drugs acting on GPCR having in their molecules biologically (bold) privileged structures.

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  22. The MDDR database is available from <http://mdli.com/>

*Dedicated to Professor Victor-Emanuel Sahini  
on the occasion of his 80th anniversary*

## AM1 CONFORMATIONAL ANALYSIS OF THE (3S,5R,6R)-6-ACETYLAMIDOPENICILLANIC ACID

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Using the AM1 semiempirical MO method a conformational search study was performed for the conformers of the (3S,5R,6R)-6-acetylamidopenicillanic acid. The AM1 method gives 8 distinct conformers. The difference between the heat of formation of the minimum energy conformer and the one of the highest energy conformer is of 7.12 kcal/mol. The AM1 calculated bond lengths and angles for the minimum energy conformer are close to the experimentally determined values. The steric and electronic properties of the AM1 conformers are influenced by the orientation of the substituent attached to the amidic group bound at the lactamic ring.

### INTRODUCTION

The (3S,5R,6R)-6-acetylamidopenicillanic acid is one of the most known antibacterials of the penicillin class. As all penicillins, its molecular structure contains two fused rings (a  $\beta$ -lactamic four-membered ring and a thiazolidinic five-membered ring) having three chiral centers<sup>1</sup> noted with an asterisk in Fig. 1. Due to these chiral centers there are  $2^3=8$  possible diastereoisomers: 3R,5R,6R; 3S,5R,6R; 3R,5S,6R; 3R,5R,6S; 3S,5S,6R; 3S,5R,6S; 3R,5S,6S; 3S,5S,6S.<sup>2</sup> The 3S,5R,6R diastereoisomer is the natural product. Its two-fused rings represent the general structure of all antibacterials of the penicillin class.<sup>3</sup> Among the hypotheses on the action mechanism of penicillins the experimental data confirmed Strominger's hypothesis. He demonstrated that the D-Ala-D-Ala sequence from the general structure of penicillins is responsible for their antibacterial activity.<sup>3-7</sup> The thick line in Fig. 1 shows the key atoms forming the D-Ala-D-Ala sequence of the penicillin backbone.<sup>8</sup> The molecular and electronic structures of the (3S,5R,6R)-natural isomers of penicillins have been extensively studied with different quantum-chemical methods.<sup>9-15</sup> All used quantum-chemical methods give a pyramidalized

structure for the amidic nitrogen atoms, but in none of these works is mentioned the degree of pyramidalization of the N14 amidic nitrogen atom. This effect influences the results of the quantumchemical methods applied to structures with rotatable bonds.<sup>16</sup>

The present study analyzes the possibility to obtain a maximum number of distinct conformers of the (3S,5R,6R)-6-acetylamidopenicillanic acid by performing a conformational search using the AM1 semiempirical MO method. The influence of the substituents on the molecular and electronic structure of these conformers is also analyzed.

### METHODS

The gas phase equilibrium geometry of the conformers was obtained by AM1 calculations with an SCF convergence of  $10^{-5}$  and a RMS gradient of  $10^{-2}$  kcal/Å·mol. Conformational search was performed by varying the C2-C3-C11-C12, C5-C6-N14-C15 and C6-N14-C15-C16 dihedrals in the range  $0 \pm 180^\circ$  with steps of  $15^\circ$ . The energy criterion was set to 30 kcal/mol above the minimum energy conformer. The calculations have been performed with the Conformational Search module

from the HyperChem7.52 package.<sup>17</sup> In order to obtain all conformers, including the minimum energy conformer, for structures with more than two flexible bonds the conformational search is generally performed by the grid method.<sup>16-18</sup> For around 99% of the possible conformers we should

know approximately the minimum ratio between the number of starting grid points and the number of possible conformers. To find this ratio a plot of the dependence of number of conformers on the number of starting grid points is presented in Fig. 2.

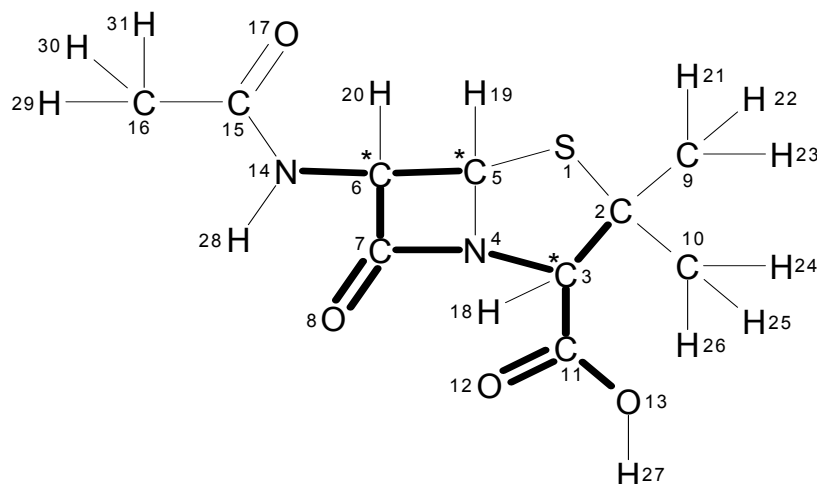


Fig. 1 – Atom numbering of the (3S,5R,6R)-6-acetylamidopenicillanic acid (the three chiral centres: C3, C5 and C6 are marked with asterisks).

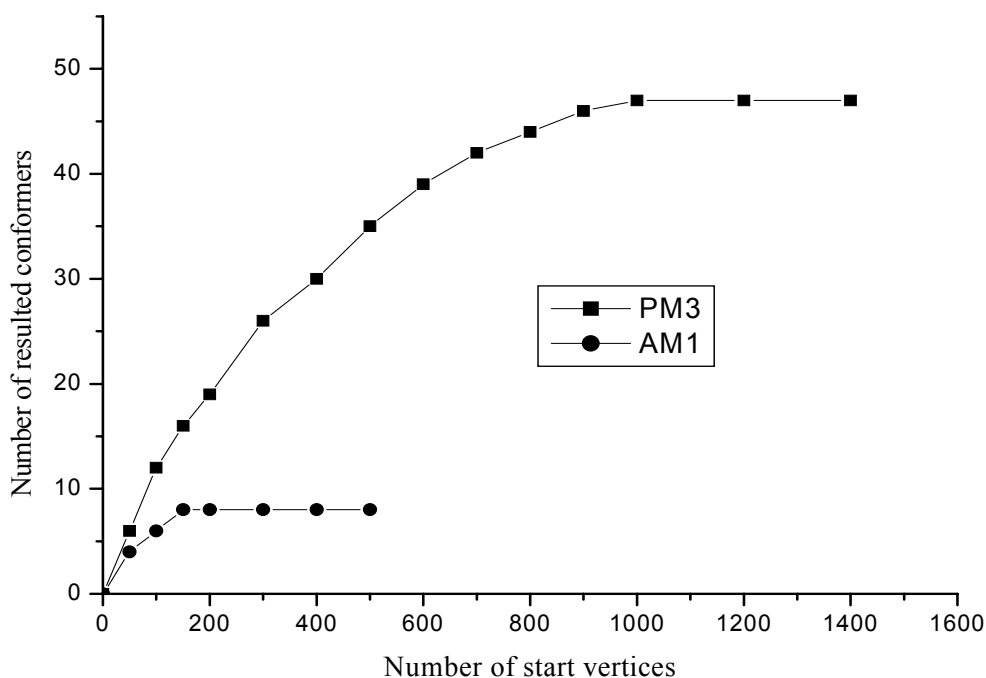


Fig. 2 – Functional dependence of the number of discovered conformers based on the number of starting geometries for the AM1 and PM3 semiempirical MO methods.

Fig 2 shows that for the PM3 method only after around 1000 starting geometries the total number of 47 distinct conformers remains constant, while for the AM1 method the total number of 8 distinct conformers remains constant after 150 starting

geometries. This data leads to the conclusion that for obtaining the maximum number of possible conformers the ratio between the number of starting points and the number of possible conformers should be around 20:1.



## RESULTS

The AM1 method gives structures with pyramidalized nitrogen atoms and consequently generates a pseudochiral configuration for the N14 atom in the penicillanic acid.<sup>19</sup> The C6-C15-H28-N14 improper dihedral was used to measure “conicity” of the out-of-plane substituents at the N14 nitrogen atom. This “conicity” could be considered a measure of the degree of pyramidalization of the N14 atom. For a planar structure of the substituents around the N14 atom, this dihedral is 0°. The amidic groups can have *anti* or *syn* relative positions of the O17 and H28 atoms. For all resulted conformers the AM1 method gives lower energy for structures with *anti* amidic groups than the energy of those with *syn* amidic groups. To be sure that the obtained conformers are minima on the potential energy surface and not saddle points, for each conformer was performed a calculation of normal vibrations. All vibrations are positive, including  $\nu_{\min} > 0$ . The wavelength number of the maximum vibration,  $\nu_{\max}$  and zero point vibrational energy (ZPVE) are given in Table 1.

The geometric and electronic data of the conformers resulted from conformational search are displayed in Tables 1-6. The total energy,  $E_{\text{tot}}$ , was used to order the conformers. To analyze the results given by the AM1 method, the geometry of conformers was compared with the X-ray structure<sup>20</sup> of the (2S,5R,6R)-3,3-dimethyl-7-oxo-6-phenylacetylamino-4-thia-1-aza-bicyclo [3.2.0]heptane-2-carboxylic acid 1-ethoxycarbonyloxy-ethyl ester by measuring the ring and the exocyclic bond lengths, the ring bond angles including those that measure the bending of the two fused rings, and some proper and improper dihedrals. By selecting the atoms in the same order in the X-ray geometry and in the geometry of each conformer, the structures have been superposed and the RMS fit errors were calculated.

The AM1 method gives 8 distinct conformers (Tab. 1). Between the heat of formation of the minimum energy conformer (1am1) and the heat of formation of the conformer with the highest energy (8am1) is a difference of 7.12 kcal/mol.

Table 1

Some structural and energetic characteristics of the conformers resulted from AM1 calculations for the (3S,5R,6R)-6-acetilamidopenicillanic acid, arranged according to the decreasing variation of the total calculated energy and heat of formation, compared to some experimental data

Energetic order	$E_{\text{tot}}$ (kcal/mol)	$\Delta H_{\text{form}}$ (kcal/mol)	$IP_1$ (eV)	$EA_1$ (eV)	$\mu$ (Debye)	Dihedral angles (°)		Improper angle (°)	Chirality N14	Position O17, H28
						5-1-2-3	5-2-3-4			
1am1	-77547.31	-122.404	-9.350	-0.042	5.203	5.688	18.157	-8.460	R	Anti
2am1	-77546.94	-122.037	-9.369	-0.052	4.704	5.211	17.922	-8.348	R	Anti
3am1	-77546.22	-121.316	-9.371	-0.100	1.779	7.073	18.065	-1.067	-	Syn
4am1	-77545.77	-120.871	-9.392	-0.112	2.563	6.596	17.651	-0.658	-	Syn
5am1	-77543.51	-118.604	-8.978	0.282	4.410	-1.606	8.824	7.579	S	Anti
6am1	-77543.06	-118.154	-8.994	0.274	5.632	-1.877	8.298	7.563	S	Anti
7am1	-77540.57	-115.667	-9.269	-0.091	2.434	-12.53	4.173	7.496	S	Syn
8am1	-77540.19	-115.284	-9.283	-0.102	2.776	-12.59	3.926	7.512	S	Syn
Expr.Data <sup>20</sup>	-	-	-	-	-	17.583	40.263	-1.145	-	Anti
Expr.AMPC <sup>21</sup>	-	-	-	-	-	40.3	-9.7	-	-	-
Expr.PN09 <sup>21</sup>	-	-	-	-	-	21.1	-144.6	-	-	-
Expr.PN14 <sup>21</sup>	-	-	-	-	-	-40.9	-2.9	-	-	-
Expr.PN39 <sup>21</sup>	-	-	-	-	-	16.9	36.6	-	-	-

The first four conformers in Table 1 have negative conicity and (R)-pseudochirality at the N14 atom, while the last four conformers have positive conicity and (S)-pseudochirality at the N14 atom. In the minimum energy conformer the amidic group atoms have *anti* orientation, while in the highest energy conformer the amidic atoms have *syn* orientation. First two conformers have a pronounced negative conicity at the N14 atom (around  $-8.5^\circ$ ), while the following two conformers have a conicity at the N14 atom near  $0^\circ$ . In these conformers the amidic group is almost planar. Although the conformers 3am1 and 4am1 have around N14 a “conicity” near  $0^\circ$  the atoms of this amido group have a *syn* orientation and therefore a lower steric likeness ( $0.932\text{\AA}$  respectively  $1.179\text{\AA}$ ) comparing to the experimental geometry. The last four conformers have a conicity around  $7.5^\circ$ . Energy differences of 1.088, 1.166, 2.937, and 2.870 kcal/mol resulted between the pairs of conformers 1am1 $\rightarrow$ 3am1, 2am1 $\rightarrow$ 4am1, 5am1 $\rightarrow$ 7am1 and 6am1 $\rightarrow$ 8am1 having the amidic bonds in *anti* and *syn* respectively. The energy difference between the two orientations of the carboxylic function is of 0.367 kcal/mol for the first two conformers and of 0.450 kcal/mol for the 5am1 and 6am1 conformers.

The superimposition of all conformers using the plane formed by the N4, C5 and C6 atoms is given in Fig. 3. According to this superimposition, the conformers can be classified in three groups with different puckering of the thiazolidinic ring: *i*) 1am - 4am1, *ii*) 5am1 - 6am1, and *iii*) 7am1 - 8am1. The data in Tab. 1 shows that the first four conformers have a positive puckering given by the 5-1-2-3 dihedral ( $5.21^\circ$  for conformer 2am1 and  $7.07^\circ$  for conformer 3am1), and by the 5-2-3-4 dihedral ( $17.65^\circ$  for conformer 4am1 and  $18.16^\circ$  for conformer 1am1). The second group of conformers (5am1 and 6am1) have a negative puckering almost close to zero for the 5-1-2-3 dihedral and around  $-8.5^\circ$  for the 5-2-3-4 dihedral. The last two conformers have an accentuated negative puckering of  $-12.5^\circ$  for the 5-1-2-3 dihedral and around  $-4^\circ$  for the 5-2-3-4 dihedral. The puckering is not generated by the rotation of the carboxylic group, but by the rotation around the C5-N14 and N14-C15 bonds. All the types of puckering are evidenced in Fig. 3. It should also be mentioned that in the experimentally determined structure<sup>21</sup> the thiazolidinic ring is also puckered.

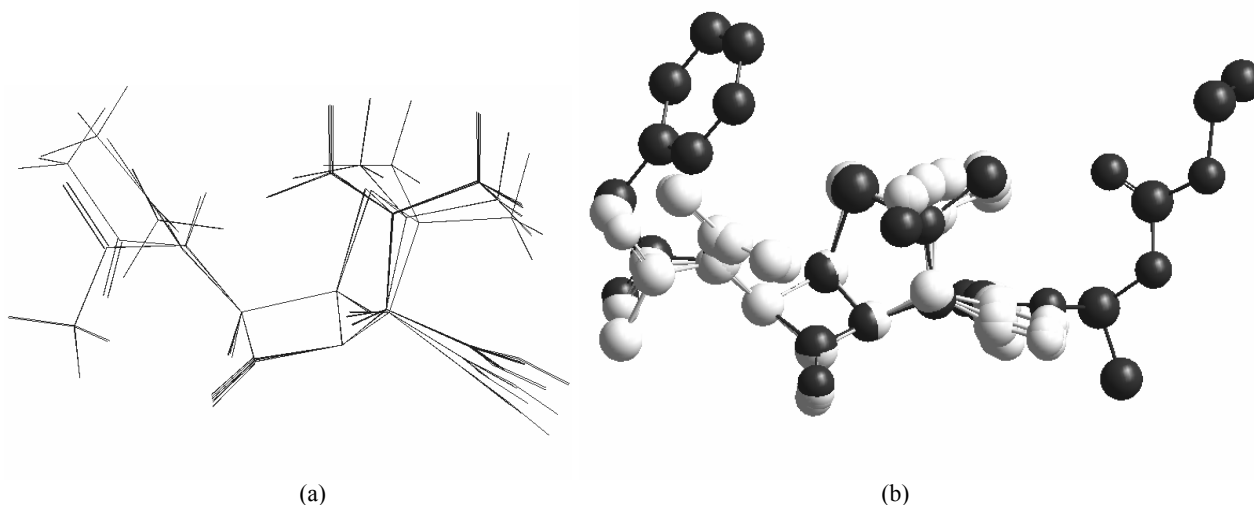


Fig. 3 – Superimposition of all AM1 conformers (a) and superimposition of the heavy atoms in the experimental geometry (black atoms),<sup>21</sup> over the heavy atoms of the 8 AM1 conformers (b).

There is no correlation between the energy or geometric data of the conformers and the ionization energies or the electron affinities (Tab. 1.). One can observe only that their values are ordered in *anti* - *syn* pairs of conformers.

For all the AM1 conformers both the highest occupied molecular orbital (HOMO =  $-IE_1$ ) as well as the lowest unoccupied molecular orbital (LUMO =  $-EA_1$ ) are dominated by the

contribution of the sulfur atom S1 (Fig. 4). At the HOMO level, beside the sulfur atom, the C5-C6, C5-H19 and C2-C9, C2-C10  $\sigma$  bonds have also an important contribution, while at the LUMO level the contribution of the C2 respectively C5 atoms are important. For the dipole moments the *syn* structures have values sensibly lower than the *anti* structures (Tab. 1.)

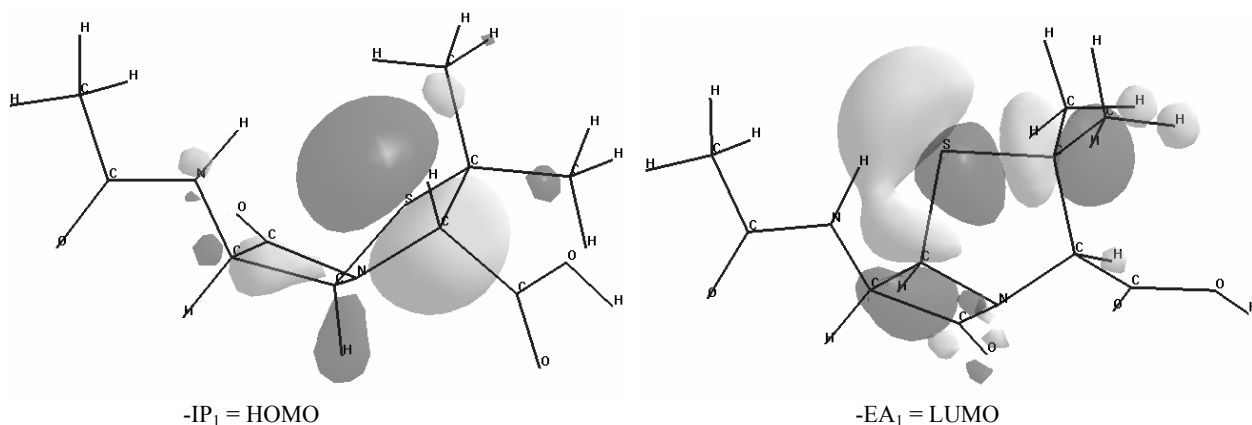


Fig. 4 – Sulfur atom predomination both in the HOMO ( $IP_1$ ) as well as in the LUMO ( $EA_1$ ).

The net charge on the sulfur S1 atom is slightly positive, while the net charges on the N4, N14, O8, O12, O13, O17 atoms are negative in all conformers (Tab. 2.). Although the two nitrogen atoms (N4, N14) are of the same type, namely amidic atoms, the N4 atom is the strongest pyramidalized due to its position between the two fused rings. It has, as expected, a net charge lower than that of the N14 exocyclic nitrogen atom. The net charge of the  $\beta$ -lactamic oxygen atom (O8) is

lower than the net charges of the other oxygen atoms (O12, O13, O17). There are no evident correlations between net atomic charges and energy or geometric data of the AM1 conformers.

The electrostatic potential isosurface for the lowest and highest energy conformers (1am1) and (8am1) respectively are shown in Fig. 5. As can be seen in Fig 5 the shapes of the isosurfaces are influenced mainly by the rotation of the exocyclic amidic bond.

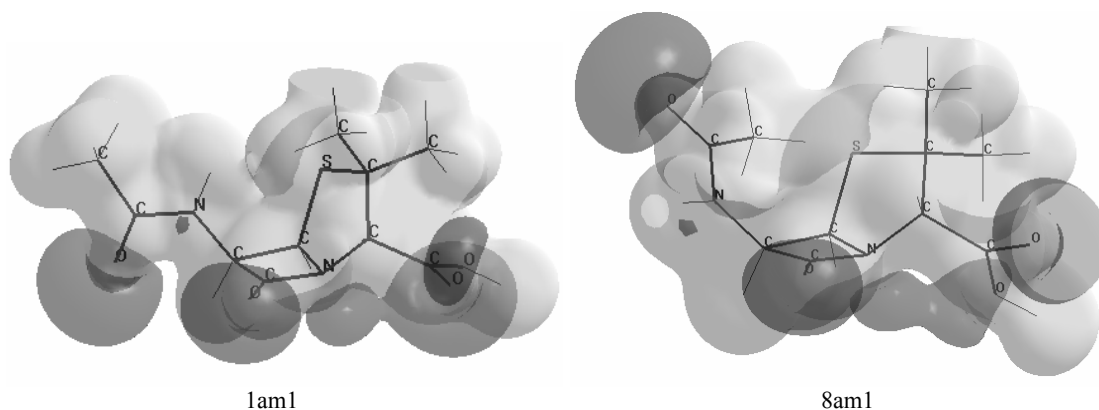


Fig. 5 – Electrostatic potential isosurfaces calculated with the AM1 method.

Table 2

Charge densities on some representative atoms, extreme vibration frequencies ( $v_{\min.}$ ,  $v_{\max.}$ ) and the vibration energy for the equilibrium state of the conformers resulted from AM1 computation, for the (3S,5R,6R)-6-acetilamidopenicilanic acid, arranged according to the decreasing variation of the calculated heat of formation

Energy order	$\Delta H_{\text{form}}$ (kcal/mol)	Net charge on atom							Vibration ( $\text{cm}^{-1}$ )		ZPVE (kcal/mol)
		S1	N4	N14	O8	O12	O13	O17	$v_{\min.}$	$v_{\max.}$	
1am1	-122.404	0.043	-0.258	-0.370	-0.239	-0.331	-0.310	-0.348	31.18	3439.91	154.299
2am1	-122.037	0.041	-0.258	-0.370	-0.238	-0.356	-0.292	-0.348	31.91	3440.34	154.325
3am1	-121.316	0.069	-0.256	-0.379	-0.242	-0.332	-0.318	-0.346	23.03	3422.70	154.440
4am1	-120.871	0.067	-0.256	-0.380	-0.241	-0.354	-0.293	-0.345	22.78	3413.17	154.464
5am1	-118.604	0.050	-0.260	-0.362	-0.229	-0.339	-0.318	-0.360	27.94	3481.12	154.529
6am1	-118.154	0.049	-0.259	-0.362	-0.228	-0.359	-0.294	-0.359	27.22	3481.59	154.562
7am1	-115.607	0.074	-0.263	-0.367	-0.248	-0.330	-0.318	-0.348	27.74	3426.53	154.734
8am1	-115.284	0.076	-0.262	-0.367	-0.247	-0.355	-0.290	-0.348	27.79	3426.68	154.777

In Fig. 6 it can be observed that the fundamental vibration  $\nu_{\min}$  is a backbone vibration where all atoms are implicated. The movement vectors are proportional with the AM1 calculated force constants. The maximum energy vibration  $\nu_{\max}$  is an elongation vibration of the N14 - H28 bond. For all AM1 conformers the nature of these two vibrations does not change. The equilibrium

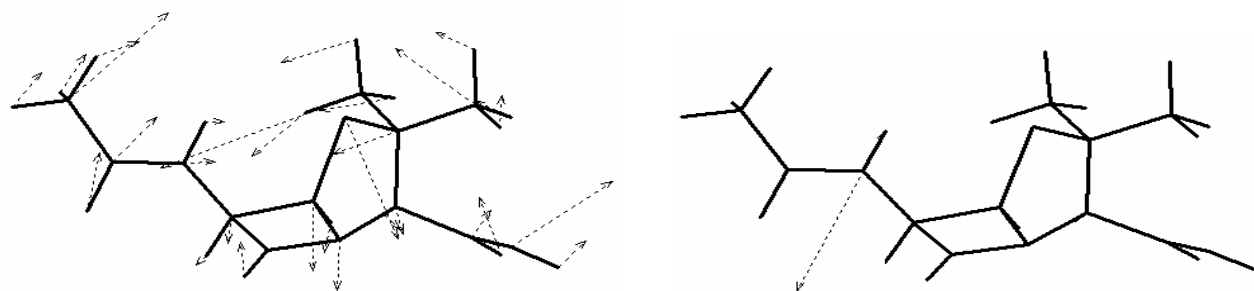


Fig. 6 – Movement vectors resulted from AM1 calculations for vibrations  $\nu_{\min} = 31.18 \text{ cm}^{-1}$  and  $\nu_{\max} = 3439.91 \text{ cm}^{-1}$  of the conformer 1am1.

A relatively good likeness (RMS error = 0.356Å) between the AM1 minimum energy conformer and the experimental geometry,<sup>3,20,21</sup> resulted from the comparison of the calculated and experimental values of bond lengths (Tables 3 and 4). Regarding the values calculated with semiempirical methods,<sup>9</sup> for other models of penicillins the differences resulted from the AM1 and PM3 semiempirical MO methods are of the same order, while the *ab initio* and DFT methods<sup>10</sup> give values much closer to the experimentally determined values (Tab. 3 and 4). The values of the S1-C2 bond lengths obtained in this paper are around 1.82Å, slightly higher than the S1-C5 bond lengths which are approximately 1.78Å. This difference between the above two bond lengths is also observed experimentally. Curiously, the advanced methods HF/6-31G\* and B3LYP/6-31G\*\*,<sup>10</sup> do not reproduce this difference (Tab. 3 and 4).

The C2-C3, C3-N4, and N4-C5 bond lengths for all AM1 conformers are comparable with the experimentally obtained values and in a good concordance with the advanced HF/6-31G\* and B3LYP/6-31G\*\* methods. For the bonds in the  $\beta$ -lactamic ring: C5-C6, C6-C7 and N4-C7, the calculated values are slightly greater than the experimentally obtained values. For distances between the S1 and C9 ( $C_\alpha$ ), respectively S1 and C10 ( $C_\beta$ ) unbound atoms the calculated values are slightly lower than those experimentally obtained.<sup>20</sup> From the values in Table 3 and 4 one can conclude that the geometry of the AM1 conformers is strongly influenced by the exocyclic function bound through the amidic group to the C6

zero point energy (ZPVE) is relatively constant around the value of 154.5 kcal/mol. It increases insignificantly from around 154.299 kcal/mol for the minimum energy conformer to 154.777 kcal/mol for the conformer of maximum energy. In a *syn-anti* pair of conformers the energy difference between a structure with an *anti* amidic group and a structure with a *syn* amidic group is of 0.1 kcal/mol.

atom. Actually, this function is the one that also determines the variation of the biological activity of the different penicillins.

The calculated bond angles of the cycles and the 4-5-6-7 improper angle which measures the planarity of the  $\beta$ -lactamic cycle compared to the experimentally obtained values,<sup>3,20,21</sup> or obtained by other authors,<sup>9,10</sup> are presented in Tab. 5 and 6.

The AM1 values for the 4-5-6-7 improper angle indicate that the  $\beta$ -lactamic cycle for all conformers is not plane, but has a slight positive puckering. This positive puckering can also be observed in experimental data. A grouping in pairs *anti-syn* can also be observed, which suggests that the rotation of the carboxylic group with 180° does not influence significantly the puckering of the  $\beta$ -lactamic ring. Puckering of the  $\beta$ -lactamic cycle could be rather influenced by the conformation of the substituent attached to the C6 atom. The values for the 2-1-5, 5-4-7, 4-5-6, 5-6-7, 6-7-4 angles are in the range of the experimentally obtained values, while the 1-2-3, 2-3-4 and 1-5-4 angles are greater than the experimental values, and the angle 3-4-5 is lower than the experimental value. The bending angles between the thiazolidinic and  $\beta$ -lactamic rings split the conformers into two distinctive groups: the (R) conformers (Tab. 5) and the (S) conformers (Tab. 6). The 1-5-6 bending angle for the (R) conformers is close to the experimental value, while for the (S) conformers the values are lower than the experimental ones. The 3-4-7 bending angle is greater than the experimental value both for the (R) as well as for the (S) conformers.

Table 3

Bond lengths of the cycles and the exocyclic distances from the sulphur atom of the metil groups for the conformers resulted from AM1 calculations, for the (3S,5R,6R)-6-acetilamidopenicilanic acid, with R pseudochirality at the N14 nitrogen atom, arranged according to the decreasing variation of the calculated heat of formation, compared to some experimental and computed data in literature

Enegetic order	$\Delta H_{\text{form}}$ (kcal/mol)	Bond lengths (Å)								Distance (Å)		RMS Fit
		S1-C2	C2-C3	C3-N4	N4-C5	C5-S1	C5-C6	C6-C7	C7-N4	S1-C9	S1-C10	RMS Error(Å)
1am1	-122.404	1.821	1.560	1.446	1.477	1.785	1.582	1.568	1.447	2.710	2.700	0.356
2am1	-122.037	1.821	1.560	1.445	1.477	1.785	1.582	1.568	1.447	2.710	2.699	0.811
3am1	-121.316	1.820	1.560	1.447	1.479	1.781	1.582	1.572	1.444	2.712	2.698	0.932
4am1	-120.871	1.821	1.560	1.446	1.479	1.781	1.582	1.572	1.444	2.711	2.697	1.179
Exp.Data <sup>20</sup>	-	1.860	1.573	1.458	1.465	1.824	1.550	1.547	1.376	2.779	2.749	-
Exp.AMPC <sub>21</sub>	-	1.854	1.556	1.466	1.466	1.812	1.528	1.528	1.382	-	-	-
Exp.AMPC <sub>3</sub>	-	1.855	1.573	1.463	1.470	1.791	1.554	1.540	1.360	-	-	-
Exp.PN09 <sup>21</sup>	-	1.859	1.564	1.468	1.476	1.815	1.561	1.541	1.367	-	-	-
Exp.PN14 <sup>21</sup>	-	1.841	1.554	1.449	1.486	1.835	1.551	1.536	1.389	-	-	-
Exp.PN39 <sup>21</sup>	-	1.857	1.567	1.450	1.457	1.838	1.546	1.542	1.378	-	-	-
Exp.PenV <sup>3</sup>	-	1.87	1.57	1.46	1.52	1.82	1.58	1.55	1.46	-	-	-
Exp.6APA <sub>21</sub> <sup>3</sup>	-	1.859	1.575	1.445	1.450	1.822	1.554	1.520	1.392	-	-	-
(ax.) HF/6-31G* <sup>10</sup>	-	1.831	1.539	1.446	1.456	1.835	1.546	1.526	1.388	-	-	-
(eq.) HF/6-31G* <sup>10</sup>	-	1.825	1.546	1.452	1.457	1.823	1.544	1.527	1.383	-	-	-
(ax.) B3LYP/6-31G** <sup>10</sup>	-	1.851	1.545	1.454	1.470	1.863	1.554	1.542	1.407	-	-	-
(eq.) B3LYP/6-31G** <sup>10</sup>	-	1.846	1.552	1.460	1.469	1.850	1.551	1.543	1.402	-	-	-
AM1 <sup>9</sup>	-	1.821	-	-	-	1.786	-	-	1.447	-	-	-
PM3 <sup>9</sup>	-	1.875	-	-	-	1.815	-	-	1.481	-	-	-

Table 4

Bond lengths of the cycles and the exocyclic distances from the sulphur atom of the metil groups for the conformers resulted from AM1 calculations, for the (3S,5R,6R)-6-acetilamidopenicilanic acid, with S pseudochirality at the N14 nitrogen atom, arranged according to the decreasing variation of the calculated heat of formation, compared to some experimental and computed data in literature

Enegetic order	$\Delta H_{\text{form}}$ (kcal/mol)	Bond lengths (Å)								Distances (Å)		RMS Fit
		S1-C2	C2-C3	C3-N4	N4-C5	C5-S1	C5-C6	C6-C7	C7-N4	S1-C9	S1-C10	RMS Error(Å)
5am1	-118.604	1.8240	1.5607	1.4466	1.4770	1.7759	1.5868	1.5651	1.4437	2.7163	2.6902	1.145
6am1	-118.154	1.8243	1.5609	1.4460	1.4772	1.7749	1.5869	1.5650	1.4438	2.7157	2.6894	1.330
7am1	-115.667	1.8211	1.5611	1.4470	1.4772	1.7777	1.5893	1.5631	1.4396	2.7073	2.6983	1.203
8am1	-115.284	1.8214	1.5614	1.4462	1.4774	1.7771	1.5893	1.5629	1.4399	2.7070	2.6975	1.374

Table 5

Bond angles of the cycles and the dihedral improper angle for the  $\beta$ -lactamic cycle of the conformers resulted from AM1 calculations for the (3*S*,5*R*,6*R*)-6-acetylamidopenicilanic acid with R pseudochirality at the N14 nitrogen atom, arranged according to the decreasing variation of the heat of formation, compared to some experimental and computed data in literature

Energetic order	Angle (°)													Dihedral (°)
	C2-S1-C5	S1-C2-C3	C2-C3-N4	C3-N4-C5	S1-C5-N4	C5-N4-C7	N4-C5-C6	C5-C6-C7	C6-C7-N4	S1-C5-C6	C3-N4-C7	N4-C5-C6-C7		
1am1	95.124	106.964	111.280	114.380	109.624	93.343	89.691	84.963	91.320	118.315	121.688	5.967		
2am1	95.103	107.007	111.270	114.458	109.610	93.328	89.698	84.955	91.324	118.300	121.730	6.012		
3am1	95.097	106.950	111.193	114.263	109.831	93.448	89.736	84.860	91.411	118.806	121.814	5.315		
4am1	95.090	107.000	111.199	114.357	109.827	93.421	89.755	84.854	91.430	118.800	121.783	5.294		
Expr. Data <sup>20</sup>	95.190	104.168	105.893	116.134	105.264	94.396	88.129	84.692	91.541	118.203	127.633	7.818		
Expr. AMPC <sup>21</sup>	90.4	103.1	106.0	117.9	103.8	93.7	88.5	85.7	91.6	118.9	126.1	4.3		
Expr. AMPC <sup>3</sup>	89.8	104.2	105.3	117.8	103.4	93.4	88.4	83.6	93.1	118.4	127.7	-		
Expr. PN09 <sup>21</sup>	95.2	103.9	105.6	115.8	105.5	94.2	87.7	84.4	92.6	118.9	126.7	7.1		
Expr. PN14 <sup>21</sup>	89.6	103.2	107.4	115.9	103.1	92.2	87.6	84.4	91.8	117.0	123.3	14.0		
Expr. PN39 <sup>21</sup>	94.9	105.0	106.0	117.2	105.8	93.8	88.8	84.2	92.0	119.3	126.7	7.0		
Expr. PenV <sup>3</sup>	96.0	105.0	104.0	120.0	103.0	88.0	92.0	83.0	96.0	122.0	129.0	-		
Expr. 6APA <sup>3,21</sup>	95.6	105.7	106.0	118.0	103.7	93.5	88.1	84.6	91.7	120.2	132.5	-		
(ax.) HF/6-31G* <sup>10</sup>	93.3	-	-	116.1	-	-	88.8	-	-	119.0	-	-		
(eq.) HF/6-31G* <sup>10</sup>	90.2	-	-	117.8	-	-	88.8	-	-	117.4	-	-		
(ax.) B3LYP/6-31G** <sup>10</sup>	92.7	-	-	116.4	-	-	89.1	-	-	119.2	-	-		
(eq.) B3LYP/6-31G** <sup>10</sup>	89.8	-	-	118.1	-	-	89.1	-	-	117.1	-	-		
AM1 <sup>9</sup>	95.1	-	-	-	-	-	89.7	84.9	-	-	-	6.2		
PM3 <sup>9</sup>	94.5	-	-	-	-	-	88.8	87.6	-	-	-	4.5		

Table 6

Bond angles of the cycles and the dihedral improper angle for the  $\beta$ -lactamic cycle of the conformers resulted from AM1 calculations for the (3*S*,5*R*,6*R*)-6-acetylamidopenicilanic acid with S pseudochirality at the N14 nitrogen atom, arranged according to the decreasing variation of the heat of formation, compared to some experimental and computed data in literature

Energetic order	Angle (°)													Dihedral (°)
	C2-S1-C5	S1-C2-C3	C2-C3-N4	C3-N4-C5	S1-C5-N4	C5-N4-C7	N4-C5-C6	C5-C6-C7	C6-C7-N4	S1-C5-C6	C3-N4-C7	N4-C5-C6-C7		
5am1	95.125	107.351	111.721	114.884	110.107	93.857	89.256	85.206	91.338	121.785	122.903	4.220		
6am1	95.112	107.363	111.709	114.937	110.123	93.824	89.288	85.193	91.372	121.813	122.799	4.092		
7am1	94.514	107.265	111.313	115.001	109.436	93.746	89.059	84.960	91.466	121.355	123.534	6.298		
8am1	94.501	107.261	111.295	115.034	109.456	93.715	89.085	84.953	91.497	121.351	123.455	6.222		

## CONCLUSIONS

The AM1 method gives only 8 distinct conformers of the (3S,5R,6R)-6-acetylamidopenicillanic acid and the energy difference between the lowest and highest energy conformers is of 7.12 kcal/mol. A good concordance between the geometry of the AM1 minimum energy conformer and the experimental geometry was obtained. There are no major differences between the results obtained with the AM1 semiempirical MO method and the ab initio and ab initio/DFT methods. The AM1 method gives three distinctive classes of bending of the thiazolidinic and  $\beta$ -lactamic rings and three classes of puckered thiazolidinic rings. The substituents at the C6 carbon atom influence significantly the number of conformers of the (3S,5R,6R)-6-acetylamidopenicillanic acid and the molecular and electronic structure of these conformers.

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