



Short Communication

QSAR Study of Rabbit Aortic Angiotensin II Antagonists Compounds Using Different Descriptors

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Abstract

For present study, the various QSAR models have been developed to predict the activities in terms of log 1/C for 11 Rabbit Aorta Angiotensin II Antagonists compounds with the help of quantum chemical and energy descriptors viz. heat of formation, Gibbs free energy, molar refractivity, HOMO energy, LUMO energy, absolute hardness, Softness, Chemical potential and electronegativity. The parameter adopted in this calculation is the semi-empirical PM3 based. The QSAR model sixth provides a good arrangement between obs log 1/c and predicted activity.

Keywords: PM3, Absolute hardness, global softness, electronegativity, chemical potential, hardness, HOMO; LUMO, heat of formation (ΔH), Gibbs free energy (ΔS), molar refractivity (MR).

Introduction

The use of quantitative structure-activity relationships (QSAR) since their advent in 1962¹ has become increasingly helpful in understanding many aspects of chemical-biological interactions in drug and pesticide research as well as many areas of toxicology. With a properly designed set of congeners, carefully tested in almost any biological system, it has become easy to derive a QSAR by a steadily increasing number of computerized approaches. Getting a new QSAR no longer calls for rushing into print. What is called for is support for it from as many points of view as possible. In fact, there are so many fancy new programs that almost any set of chemicals acting on a given system can be correlated mathematically. Some wit has remarked that if you cannot derive a correlation equation it is a bad reflection on your library since there seems to be an almost unlimited selection of parameters. The real problem is to deduce when the result can be related to our general knowledge of chemistry and biology. For work in progress, one can test new molecules to check the equation, but for most published work, this is not possible. We are finding that lateral support is possible in a variety of ways²⁻¹⁰. From the beginning to present day, luck has played a major role in drug discovery¹¹. In the present report, we review nonpeptide angiotensin antagonists. The vasoactive hormone Angiotensin II produced by the renin-angiotensin system (RAS) plays an integral role in the pathophysiology of hypertension because it effects the regulation of fluid volume, electrolyte balance, and blood volume in mammals^{12,13}. Renin is a proteolytic enzyme produced mainly in the juxtaglomerular apparatus of the kidney, which acts on the circulating α -globulin angiotensinogen produced by the liver¹⁴.

In the present study we have taken structures of a set of compounds Angiotensin II and then compared to the numerical values of a biological activity. The goal here has been to find some numerical information for a molecule. This structure information and the measured property or activities are then converted into a mathematical model of relationship. From a quality model it is possible to predict and to design compounds for synthesis and testing that have a good possibility for activity. In this paper, the multi linear regression analysis has been applied for QSAR study. The relationship has been worked out between the Log1/C values of a series of compounds and certain quantum chemical descriptors.

Material and Methods

The compounds taken for study are Rabbit Aortic derivatives of Angiotensin II and shown in figure-1.

Figure-1

The Quantum Mechanical QSAR: The quantum chemical parameter based QSAR study was performed by the following important descriptors like Eigen value of highest occupied molecular orbital (EHOMO), Eigen value of lowest unoccupied

molecular orbital (ELUMO)¹⁵, Absolute Hardness (η)¹⁶, Chemical Potential (μ)¹⁷, Global Softness (S)¹⁸, Electronegativity (χ)¹⁹, Heat of formation (ΔH), Gibbs free energy (ΔS), Molar Refractivity (MR). The molecules were drawn by spartan06v110, software and the geometries were optimized at PM3 level in conjunction with molecular mechanics. The global hardness and electronegativities were calculated using frontier orbital energies obtained from PM3 results and reported in tables 2. Multiple linear regression analysis (MLR) is performed to establish the QSAR.

A data set of Rabbit Aortic of Angiotensin II compounds were taken with their observed activity is shown in table 1.

Results and Discussion

Multiple linear regression (MLR) analysis: MLR analyses were performed using Minitab 16 software. The quantum mechanical descriptors were used as independent variables and the Obsd log₁/C₅₀ values as the dependent variables. In the statistical analyses, the systematic search was performed to determine the significant descriptors. The correlation matrix was developed to minimize the effect of co-linearity and to avoid redundancy and the variables physically removed from the analysis, which shows exact linear dependencies between subsets of the variables and multi-co-linearity (high multiple correlations between subsets of the variables). The MLR equations of different QSAR models are as follows:

First QSAR model: MLR equation of this QSAR model P log 1/C is given by - Obsd log 1/C = 2.41 - 6.40 E LUMO, S = 0.520140, PRESS = 3.33043, r²= 39.2%.

Second QSAR model: MLR equation of this QSAR model P log 1/C is given by - Obsd log 1/C = - 12.3 - 4.76 E LUMO - 1.79 E HOMO, S = 0.431588, PRESS = 2.53884, r²= 62.8%.

Third QSAR model: MLR equation of this QSAR model P log 1/C is given by - Obsd log 1/C = 43.0 + 25.6 E LUMO + 24.2 E HOMO + 26.1 S, S = 0.439751, PRESS = 2.45686, r²= 38.65%.

Fourth QSAR model: MLR equation of this QSAR model P log 1/C is given by - Obsd log 1/C = 39.9 + 22.4 E LUMO + 21.9 E HOMO + 23.7 S - 0.0070 MR, S = 0.470021, PRESS = 4.97156, r²= 66.9%.

Fifth QSAR model: MLR equation of this QSAR model P log 1/C is given by - Obsd log 1/C = 68.3 + 33.9 E LUMO + 33.1 E HOMO + 34.3 S - 0.0223 MR + 0.00252 ΔH , S = 0.439634, PRESS = 8.14412, r²= 75.9%.

Sixth QSAR model: MLR equation of this QSAR model P log 1/C is given by - Obsd log 1/C = 48.1 + 19.2 E LUMO + 20.5 E HOMO + 21.1 S - 0.0287 MR + 0.00377 ΔH - 0.00267 ΔG , S = 0.257392, PRESS = 3.44928, r²= 93.4%.

Table-1

Comp.No.	X	Y	Obsd log 1/C
1	H	C ₃ H ₇	10.1
2	4-CH ₃	C ₃ H ₇	9.64
3	5-CH ₃	C ₃ H ₇	8.8
4	5-Cl	C ₃ H ₇	9.07
5	5-F	C ₃ H ₇	9.04
6	5-I	C ₃ H ₇	8.38
7	5-C ₆ H ₅	C ₃ H ₇	8.91
8	5-NO ₂	C ₃ H ₇	8.33
9	5-NH ₂	C ₃ H ₇	7.6
10	NHCOCH ₃	C ₃ H ₇	7.9
11	CH ₃	C ₃ H ₇	8.15

Table 2

Calculated numeric values of Rabbit Aortic derivatives of Angiotensin II by using different descriptors

E LUMO (e.v)	E HOMO (e.v)	η	S	χ	MR (cm ³ /mol)	ΔH (kJ/mol)	ΔG (kJ/mol)
-0.91440444	-9.18046674	4.1330312	8.086859	5.0474356	131.6	227.81	782.85
-1.02445517	-9.35097705	4.1632609	8.374848	5.1877161	130.31	232.71	762.5
-0.96486433	-9.1878434	4.1114895	8.151428	5.0763539	126.11	52.34	579.62
-0.94426421	-9.10625315	4.0809945	8.047228	5.0252587	142.8	304.68	840.97
-1.00749999	-8.98875476	3.9906274	7.996199	4.9981274	155.9	340.5	654.5
-1.00749999	-8.98875476	3.9906274	7.996199	4.9981274	132.6	224.5	945.78
-0.82840499	-9.06461794	4.1181065	7.857479	4.9465115	135.13	261.6	849.3
-0.99470202	-9.14545999	4.075379	8.140134	5.070081	143.47	74.32	740.61
-0.88977348	-9.18518271	4.1477046	8.061301	5.0374781	130.64	227.81	782.85
-1.00057022	-9.39019716	4.1948135	8.390767	5.1953837	126.84	252.34	579.62
-0.9076514	-8.73202537	3.912187	7.630281	4.8198384	132.84	95.59	677.85

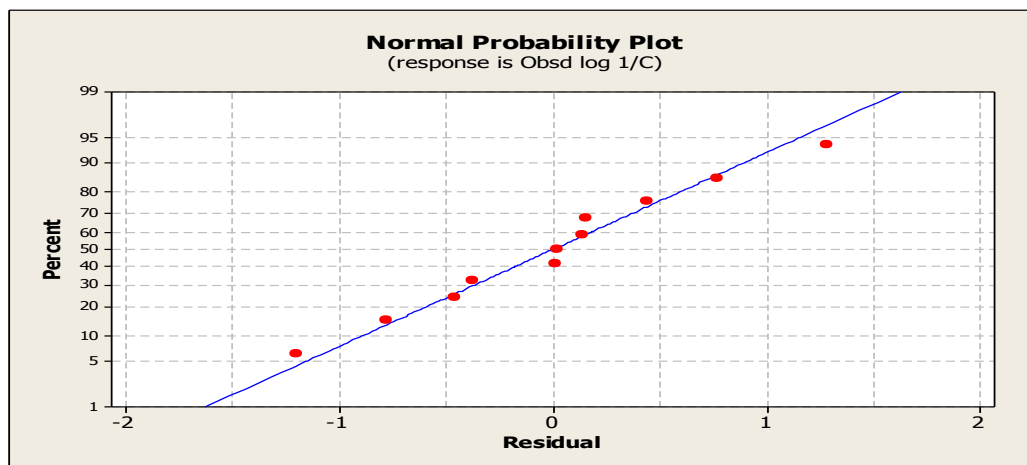


Figure- 2
Normal probability plot of responses of QSAR model Sixth

Conclusion

Values of the descriptors of the Angiotensin II Antagonist derivatives have been calculated using PM3 method and are given in table-2. With the help of these values of descriptors, six QSAR models have been developed using MLR analysis in different combinations of descriptors. The Chemical Potential (μ) and Absolute Hardness (η) descriptors have no predicting power and hence not included in the models. Best QSAR models is the model sixth listed below-

Sixth QSAR model: MLR equation of this QSAR model $P \log 1/C$ is given by $- \text{Obsd } \log 1/C = 48.1 + 19.2 E \text{ LUMO} + 20.5 E \text{ HOMO} + 21.1 \Delta S - 0.0287 MR + 0.00377 H - 0.00267 \Delta G$, $S = 0.257392$, $\text{PRESS} = 3.44928$, $r^2 = 93.4\%$.

This is one of the best QSAR model in all the six models and has been developed using E LUMO, ELUMO, Global Softness (S), Molar Refractivity (MR), Heat of reaction (ΔH) and Gibbs free energy (ΔG). This MLR equation is given by Value of regression coefficient is 93.4%, Prediction sum of squares coefficient (PRESS) is 3.44928 and Standard error of the regression (S) is 0.257392 which indicate the ability of predictive power of this QSAR model. QSAR model sixth can efficiently be used for the prediction of activity of any derivative of compound. The normal probability plot of responses is obsd log 1/C is shown in figure-2, which is clearly illustrates the high predictive power of the QSAR model six.

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