QSAR Analysis of Amphipathic 3-Phenyl-7-propylbenzisooxazoles as PPAR_γ, PPAR_δ and PPAR_α Agonist

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A quantitative structural activity relationship (QSAR) study on a series of analogs of amphipathic 3-phenyl-7-propylbenzisooxazoles with agonist activity on PPAR₇, PPAR₅ and PPAR_α has been performed using combination of various thermodynamic, electronic and spatial descriptors. Several regression expressions are obtained using multiple linear regression analysis. The best QSAR is further validated by leave-one-out cross validation method. The study reveals that thermodynamic parameters were found to have overall significant correlationship with PPAR₇, PPAR₈ and PPAR_α agonist activity. Thus QSAR brings important structural insight to aid the design of potent new molecules.

Key Words: QSAR, 3-Phenyl-7-propyl-benzisooxazoles as PPAR, PPARs and PPARa.

INTRODUCTION

Lifestyle interventions and pharmacological treatments of metabolic disturbance, collectively known as metabolic syndrome, are only partially efficient and new therapeutic approaches are urgently needed. Because of their wide range of actions on glucose homeostasis, lipid metabolism and vascular inflammation, peroxisome proliferator-activated receptors (PPARs) are promising targets for the development of new drugs for the treatment of metabolic disorders such as diabetes, dyslipidemia and atherosclerosis¹. The peroxisome proliferator-activated receptors (PPAR) are lipid-activated transcription factors belonging to the nuclear receptor super family, which includes the receptors for steroid hormones, retinoids, thyroid hormone and Vitamin $D^{2.3}$. Three different isoforms PPAR_{α}, PPAR_{γ}, PPAR_{δ} of PPARs,

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which differs by their target tissue and physiological functions^{4,5}. The hypolipidemic fibrates and the insulin thiazolidinediones are believed to be acting through activation of the PPAR_{α}^{6,7} and PPAR_{γ}^{8,9} subtypes, respectively. Several recent studies suggest that PPAR_{δ} emerged as a powerful metabolic regulation in diverse tissues including fat, skeletal muscle and the heart^{10,11}. Its transcriptional program enhances fatty acid catabolism and energy uncoupling, resulting in decreased triglyceride stores, improved endurance performance and enhanced cardiac contractility respectively. Compounds that simultaneously activate the three peroxisome proliferatoractivated receptor (PPAR) subtypes α , γ and δ hold potential to address the adverse metabolic and cardiovascular conditions associated with diabetes and the metabolic syndrome¹².

QSAR studies represent an attempt to correlate structural or property descriptors of compounds with activities. It explains the reasons of observed variations caused by the change of the substituents. Thus QSAR studies have predictive ability and simultaneously provide deeper insight into the mechanism of drug-receptor interactions¹³ even before their synthesis. The aim of the present work is to study the QSAR of PPAR_{γ}, PPAR_{δ} and PPAR_{α} agonist and to optimize their physio-chemical properties. Thus may be helpful in designing new potent molecules.

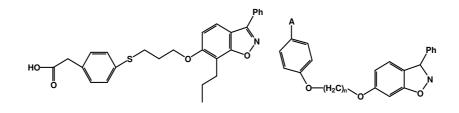
EXPERIMENTAL

A series of amphipathic 3-phenyl-7-propylbenzisooxazoles (Table-1) as PPAR_{γ}, PPAR_{δ} and PPAR_{α} agonists were taken from literature¹⁴ excluding compounds (ND = not run, max % at 3 μ M) with biological activity numerically not well defined. The transactivation data EC₅₀ (nM) value was converted to negative logarithmic dose for QSAR analysis. Thus correlating the data linear to free energy change and reducing the skewness of the data set. The molecular modeling studies were carried out using CS Chem-Office¹⁵ software version 6.0 (Cambridge soft) running on a P-IV processor. Structures of all the compounds were sketched using builder module of the program. Then the structure was subjected to energy minimization using molecular mechanics (MM2) until the root mean square (RMS) gradient value becomes smaller than 0.1 kcal/mol Å. Energy minimized molecule was subjected to re-optimization via Austin model-1¹⁶ (AM1) method until the root mean square (RMS) gradient attains a value smaller than 0.0001 kcal/ mol Å using MOPAC. The geometry optimization of the lowest energy structure was carried out using Eigenvector following (EF) routine. The descriptor values for all the molecules were calculated using compute properties module of the program. The various structural and physio-chemical descriptors considered for 3D QSAR studies are given in Table-2.

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TABLE-1 ANALOGS OF AMPHIPATHIC 3-PHENYL-7-PROPYLBENZISOOXAZOLE



Structure of compound 1

Basic structure of compound 2-29

 $EC_{_{50}}\,(nM)^*$ or MAX % (3 $\mu M)$

npd.	А	Isomer	n
	COOLI		2

Compd.	А	Isomer	n	EC_{50} (nM)* or MAX % (3 μ M)		
compu.				$PPAR_{\gamma}$	$PPAR_{\delta}$	$PPAR_{\alpha}$
1	-COOH	para	3	4	3	3
2	-COOH	para	4	ND	ND	ND
3	-COOH	meta	4	193	34%	8
4	-CH ₂ -COOH	para	3	194	147	3
5	-CH ₂ -COOH	para	4	60	86%	8
6	$-C(CH_3)_2$ -COOH	para	4	10	59%	2
7	-CH ₂ -COOH	meta	3	ND	ND	ND
8	-CH ₂ -COOH	meta	4	6	20	5
9	$-C(CH_3)_2$ -COOH	meta	4	52	400	50
10	-CH ₂ -COOH	ortho	3	ND	ND	ND
11	-CH ₂ -COOH	ortho	4	ND	ND	ND
12	-CH ₂ -CH ₂ -COOH	para	2	ND	ND	ND
13	-CH ₂ -CH ₂ -COOH	para	3	99%	74%	42%
14	-CH ₂ -CH ₂ -COOH	para	4	485	74%	429
15	-C(CH ₃) ₂ -CH ₂ -COOH	para	4	37	18%	53
16	$-CH_2-C-(CH_3)_2-COOH$	para	4	20	860	4
17	-CH ₂ -CH ₂ -COOH	meta	3	ND	ND	ND
18	-CH ₂ -CH ₂ -COOH	meta	4	ND	ND	ND
19	-CH ₂ -CH ₂ -COOH	ortho	3	ND	ND	ND
20	-CH=CH-COOH	para	3	254	148	56%
21	-CH=CH-COOH	para	4	29	640	13
22	O-CH ₂ -COOH	para	3	75	16	4
23	O-CH(CH ₃)-COOH	para	3	27	13	2
24	O-CH(CH ₃)-COOH	para	4	16	117%	10
25	O-CH(CH ₃) ₂ -COOH	para	4	2	172	2
26	O-CH ₂ -COOH	meta	4	87	127%	248
27	O-CH(CH ₃) ₂ -COOH	meta	4	100	85%	92
28	-CH ₂ -CH ₂ -CH ₂ -COOH	para	2	72%	69%	33
29	-CH ₂ -CH ₂ -CH ₂ -COOH	para	3	156	183	3

A = Acid proximal structure, $*EC_{s0}$ = The concentration yielding a 50 % response relative to the standard, ND = Not run.

TABLE-2 PHYSIOCHEMICAL DESCRIPTORS

S. No.	Descriptors	Туре	Descriptions
1	BP	Thermodynamic	Boiling point
2	СР	Thermodynamic	Critical pressure
3	СТ	Thermodynamic	Critical temperature
4	HF	Thermodynamic	Heat of formation
5	HLC	Thermodynamic	Henry's law constant
6	IGTC	Thermodynamic	Ideal gas thermal capacity
7	log P	Thermodynamic	Logarithmic partition coefficient
8	MP	Thermodynamic	Melting point
9	MR		Molar refractivity
10	SGP	Thermodynamic	Standard Gibb's free energy
11	VDW-1,4	Thermodynamic	Van der waals force
12	PARTCOFF	Thermodynamic	Partition coefficient for water/octanol
13	N-1,4-VDW		Non 1,4-Van der Waals force
	STERG	Thermodynamic	
	STBERG		Stretch bend energy
	TORERG	Thermodynamic	
	TOTERG	Thermodynamic	
18	CAA	Steric	Connolly accessible surface area
	CMA	Steric	Connolly molecular surface area
20	CSEV	Steric	Connolly solvent-excluded volume
	EM	Steric	Exact mass
22	MW	Steric	Molecular weight
-	OVAL	Steric	Ovality
	PMI-X	Steric	Principal moments of inertia-X axis
	PMI-Y	Steric	Principal moments of inertia-Y axis
	PMI-Z	Steric	Principal moments of inertia-Z axis
	DIPOLE-1	Electronic	Dipole moment - X axis
	-	Electronic	Dipole moment - Y axis
	DIPOLE-3	Electronic	Dipole moment - Z axis
	DIPOLE-4	Electronic	Resultant dipole moment
	EERG	Electronic	Electronic energy
	HOMO	Electronic	Energy of highest occupied molecular orbital
	LUMO	Electronic	Energy of lowest unoccupied molecular orbital
	REPLERG		Repulsion energy
		Electronic	Bending energy
36	DDERG	Electronic	Dipole-dipole energy

Sequential multiple regression analysis method was used to perform QSAR analysis employing in-house VALSTAT¹⁷ program. The auto-correlated parameters were eliminated depending on their individual correlation with the biological activity in order to avoid serious multicollinearity as well as the intercorrelation between the parameters with in acceptable range¹⁸ have been selected for the study. Use of more than one variable in

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the multivariate equation was justified by autocorrelation study. The statistical parameter *viz.*, correlation coefficient (r), standard error of estimate (s), sequential Fisher test (F) were considered to compare the generated QSAR models. The robustness and applicability of QSAR equation as best model, on the structural analogs was further confirmed, using various QSAR validation technique like leave-one-out (LOO) validated square correlation coefficient $Q^2 (r_{cv}^2)$ using cross validation method¹⁹, boot strapping square correlation coefficient (r_{bs}^2).

RESULTS AND DISCUSSION

In the present study, an attempt has been made to find structural requirement for PPAR_{γ}, PPAR_{δ} and PPAR_{α} agonist activity using QSAR Hansch approach on amphipathic 3-phenyl-7-propylbenzisooxazole analogs and statistically significant equations were obtained after removal of few compounds as outlier, the reason for outlying behaviour is not immediately apparent. For PPARy agonist, the statistically significant eqn. 1, with coefficient of correlation (r) = 0.736 and exclusion of compound 8 as outlier was considered as best model. The model showed overall internal statistical significance level better than 99 % as it exceeded the tabulated $F_{(2,15 \alpha 0.001)} =$ 11.34. The selected model explains more than 76 % variance in the biological activity. The intercorrelation within the parameter (ICWP) is significantly low (less than 0.2) suggested the non-dependency of parameters on each other. The model was subjected to leave one out (LOO) cross validation method, the values of $Q^2 \ge 0.3$ in cross validation method corresponds to a confidence limit greater than 95 % which minimize the risk of finding significant explanatory equation for biological activity just by mere opportunity. The value of cross-validated squared correlation coefficient ($Q^2 = 0.646$), predictive residual sum of square ($S_{PRESS} = 0.486$) and standard error of predictivity ($S_{DEP} = 0.443$) suggested good predictive ability of the biological activity. The boot strapping $r^2 (r_{bs}^2 = 0.748)$ is at par with conventional squared correlation coefficient indicating that no single compound much more or less contributed to the model.

$$BA = [58.878 (\pm 19.101)] + HE [6.795 (\pm 2.134)] + TOE [-0.057 (\pm 0.0513)]$$

n = 18, r = 0.874, r² = 0.764, variance = 0.158, std = 0.397,
F = 24.271, ICWP < 0.2 (1)

From the above study, for the PPAR_{γ} agonist activity HOMO energy contributing positively while Torsion energy (TOE) contributing negatively. HOMO energy is an electronic parameter. When a molecule acts as an electron pair donor, electron from its HOMO is supplied. The term indicates the importance of hydrogen bonding interactions. The positive contribution demonstrates that electron rich functional group may be favourable for

ligand receptor interaction. TOE is a thermodynamic parameter, which denotes the energy associated with deforming torsion angel in the molecule from their ideal value. The negative contribution suggested that bulky substituents are not favourable for activity. A plot of observed activity *vs.* calculated pEC₅₀ values of PPAR^{γ} activity using eqn. 1 is given in Fig. 1.

Observed activity vs. Calculated activity

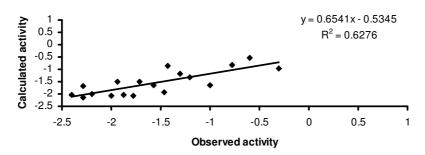


Fig. 1. A plot of observed vs. calculated pEC₅₀ values of PPAR_y activity using eqn. 1

For PPAR₈ agonist, the statistically significant eqn. 2 with coefficient of correlation (r) = 0.947 was considered as best model. The compound **8** was removed as outlier. The model showed overall internal statistical significance level better than 99 % as it exceeded the tabulated $F_{(2.7 \alpha 0.01)} = 12.25$. The selected model explains more than 90 % variance in the biological activity. The intercorrelation within the parameter (ICWP) is significantly low (less than 0.4). The value of cross-validated squared correlation coefficient ($Q^2 = 0.794$), predictive residual sum of square ($S_{PRESS} = 0.408$) and standard error of predictivity ($S_{DEP} = 0.341$) suggested good predictive ability of the biological activity. The boot strapping r² ($r_{bs}^2 = 0.930$) is at par with conventional squared correlation coefficient.

 $BA = [-8.110 (\pm 2.82109)] + CP [0.196 (\pm 0.236] + SE [0.290 (\pm 0.114)]$ n = 10, r = 0.947, r² = 0.897, variance = 0.083, std = 0.288, F = 30.587, ICWP < 0.4 (2)

From the eqn. 2, it is apparent that the thermodynamic properties are highly correlated with PPAR_{δ} agonist activity. Critical pressure (CP) is the minimum pressure that must be applied to liquefy the structure at the critical temperature. The positive contribution may facilitate attractive intermolecular forces in drug-receptor interaction. Stretch energy, the energy contribution associated with the deformation of a bond from its equilibrium bond length. The positive contribution reveals that the substituents, which increase the flexibility will increase the activity. A plot of observed activity *vs*. cal culated pEC₅₀ values of PPAR_{δ} activity using eqn. 2 is given iin Fig. 2.

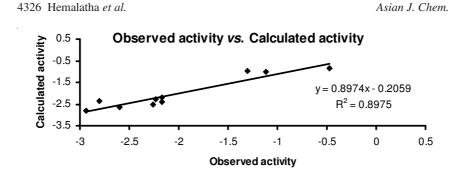


Fig. 4. A plot of observed vs. calculated pEC $_{50}$ values of PPAR $_{\delta}$ activity using eqn. 2

For PPAR_{α} agonist, among the various equations the statistically significant eqn. 3 with coefficient of correlation (r) = 0.852 was selected after removal of compound **5** and **8** as outlier. The model showed overall internal statistical significance level better than 99 % as it exceeded the tabulated $F_{(2,14\alpha0.01)} = 6.51$. The selected model explains more than 73 % variance in the biological activity. The intercorrelation within the parameter (ICWP) is significantly low (less than 0.2). The value of cross-validated squared correlation coefficient (Q² = 0.573), predictive residual sum of square (S_{PRESS} = 0.441) and standard error of predictivity (S_{DEP} = 0.400) suggested good predictive ability of the biological activity. The boot strapping r² (r²_{bs} = 0.790) is at par with conventional squared correlation coefficient.

$$BA = [-2.391 (\pm 3.648)] + CSEV [0.002 (\pm 0.009)] + NVDE [-0.233 (\pm 0.083)] n = 17, r = 0.852, r2 = 0.727, variance = 0.125, std = 0.353, F = 18.600, ICWP < 0.2 (3)$$

The above eqn. 3 explains, for PPAR_{α} agonist activity (Fig. 3), connollysolvent excluded volume a steric parameter contributes positively and non 1,4-vander Waals force a thermodynamic parameter contributes negatively.

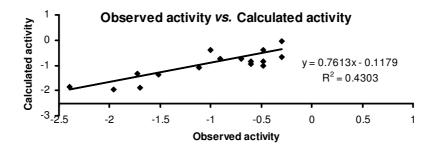


Fig. 5. A plot of observed vs. calculated pEC₅₀ values of PPAR_{α} activity using eqn. 3

Compd. –	PP	$PPAR_{\gamma}$		$PPAR_{\delta}$		$PPAR_{\alpha}$	
	Obs.	Calcd.	Obs.	Calcd.	Obs.	Calcd.	
1	-0.602	-0.546	-0.477	-0.851	-0.4771	-0.395	
2	ND	-	ND	-	ND	-	
3	-2.286	-1.679	ND	-	-0.9031	1.046	
4	-2.288	-2.156	-2.167	-2.212	-0.4771	-0.847	
5	-1.778	-2.080	ND	-	-0.9031	-0.740	
6	-1.000	-1.630	ND	-	-0.3010	Outlier	
7	ND	-	ND	-	ND	-	
8	-0.778	-0.814	-1.301	-0.981	-0.6989	-0.744	
9	-1.716	-1.513	-2.602	-2.649	-1.6989	-1.884	
10	ND	-	ND	-	ND	-	
11	ND	-	ND	-	ND	-	
12	ND	-	ND	-	ND	-	
13	ND	-	ND	-	ND	-	
14	-2.686	Outlier	ND	-	-2.6325	Outlier	
15	-1.568	1.626	ND	-	-1.7243	-1.329	
16	-1.301	-1.196	-2.935	-2.803	-0.6021	-0.838	
17	ND	-	ND	-	ND	-	
18	ND	-	ND	-	ND	-	
19	ND	-	ND	-	ND	-	
20	-2.405	-2.033	-2.170	-2.378	ND	-	
21	-1.462	-1.938	-2.806	-2.370	-1.1139	-1.096	
22	-1.875	-2.047	-1.204	Outlier	-0.6021	-0.954	
23	-1.431	-0.852	-1.114	-1.015	-0.3010	-0.653	
24	-1.204	-1.339	ND	-	-1.0000	-0.393	
25	-0.301	-0.981	-2.236	-2.282	-0.3010	-0.039	
26	-1.940	-1.517	ND	-	-2.3945	-1.859	
27	-2.000	-2.070	ND	-	-1.9638	-1.958	
28	ND	-	ND	-	-1.5185	-1.351	
29	-2.193	-2.002	-2.263	-2.529	-0.4771	-1.032	

 $TABLE-3 \\ OBSERVED AND CALCULATED PEC_{50} VALUES OF \\ PPAR\gamma, PPAR\delta AND PPAR\alpha AGONISTS$

ND = EC_{50} values are not well defined.

The positive contribution of connolly-solvent excluded volume suggests that bulky substituents can contact with large volume and it may facilitate drug receptor interaction. The non 1,4-vander Waals force refers the ability of molecules to interact with the receptor by non vander Waals force and hydrogen bonding during drug-receptor interaction. The negative contribution shows that the non-van der Waals interactions unfavourable for the activity.

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The present study reveals important structural insight in designing new molecule for PPAR_{γ}, PPAR_{δ} and PPAR_{α} agonist activity. The statistically significant equations show that the thermodynamic parameters have good correlation with biological activity. For PPAR_{γ} agonist activity nucleophilc non-bulkier substituents are favourable. For PPAR_{δ} agonist activity flexible conformer with attractive intermolecular force and for PPAR_{α} bulkier vander Waals bonding substituents are favourable for the activity. The study also shows that inverse relationship between PPAR_{γ} and PPAR_{α} agonist activity.

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