Quantitative structure-activity relationship studies of series of chalcones derivatives as inhibitors of tumor necrosis factor-alpha

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ABSTRACT

Quantitative structure-activity relationship (QSAR) technique was used to predict the biological activity of a series of chalcones compounds as anti-inflammatory. 26 physicochemical descriptors are tested in QSAR equations configuration to predict biological effectiveness of compounds under study. The geometries of the compounds under investigation were initially optimized at level (PM3) in accordance to the semi-empirical theory, and subsequently through the B3LYP procedure at the 6-31G(d) basis set in accordance to the DFT theory. The values of correlation coefficients (R^2) in Eqs (1-3) ranged from 0.794-0.873, the Fisher ratios (F) values ranged from 14.161-26.206 and the standard errors (S) values ranged from 0.262-0.334. The results demonstrated good models based on Eq.3, along with high of R^2 , F and minimum S by employing three parameters $r(C3-C5)$, (LUMO+1) and (LUMO+2). This signifies that these parameters play a significant role in determining anti-inflammatory characteristics.

MSC: 30C45, 30C50

1. Introduction

Chalcone is regarded as a key class of natural product that is broadly distributed in vegetables, fruits, soya-based foodstuff, spices and tea. Its pharmacological activities have recently garnered much interest [1]. Chalcones, or 1,3- diaryl-2-propen-1-ones, fall under the flavonoid family. Chemically, they are made up from open-chain flavonoids, in which a three-carbon α , β -unsaturated carbonyl system joins the two aromatic rings. The most commonly occurring chalcones in nature are usually polyhydroxylated in the aryl rings. Many chalcones' phenolic groups possess radical quenching properties and have gained popularity in using the chalcones rich plant extract or the compounds as food preservatives or drugs [2].

Quantitative structure-activity relationship (QSAR) defines how a difference can occur in a known biological activity in terms of a function of molecular descriptors that have been derived

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based on the chemical structure pertaining to a set of molecules. Many physiological activities associated with a molecule could be linked with their structure as well as composition. Molecular descriptors, also referred to as numerical depictions pertaining to the molecular structures, are employed to perform QSAR analysis [3]. QSAR studies generally involve two steps: first, descriptors (physical-chemical parameters) are generated which encode for chemical structural information; and second, a statistical regression method correlates changes in structure with changes in chemical properties or biological activity [4].

Recently, many efforts have been made to employ QSAR models in medical-chemistry with a focus on robust models to allow easy prediction of different biological activities, ease practical efforts and optimise certain lead candidates with chemical entities like designing of chemicals in terms of Or11 receptor antagonists. This also includes subjecting 24 of the imidazole derivatives on QSAR [5], designing enzyme and hormone inhibitors that are more effective [6-8], enhancing antifungal and antibacterial activities for certain predicted chemicals [9-11], developing QSAR models pertaining to specific steroidal derivatives that have been presented as being anti-inflammatory agents [12] as well as considerable anti-cancer activities [13]. In the previous year, QSAR models were evaluated for anti-tubercular activity to fight against *M. tuberculosis* as well as anticancer agents to fight with hepatic cancer cell lines, HepG2 [14,15]. This work aims to evaluate QSAR on 15 chalcones derivatives as inhibiting LPS-induced TNP- α production that possess potent antioxidant characteristics.

Theory:

The electronegativity, χ and hardness η can be defined as below [16]:

$$\chi = \frac{1}{2}(I + A) \quad (1)$$

$$\eta = 1/2(I - A) \quad (2)$$

$$I = -\varepsilon_{HOMO} \quad \text{and} \quad A = -\varepsilon_{LUMO}$$

Here, A and I signify electron affinity and ionisation potential, respectively, for the molecule. HOMO represented the energy of Highest Occupied Molecular Orbital; LUMO represented the energy of Lowest Unoccupied Molecular Orbital. The chemical potential pertaining to cation and anion could be described as follows [17]:

$$\mu^- = -1/4(3I + A) \quad (3)$$

$$\mu^+ = -1/4(I + 3A) \quad (4)$$

$$\mu = -(I + A)/2 \quad (5)$$

2. Experimental

Modelling and Geometry Optimisation

15 chalcones derivatives were considered to carry out quantum chemical calculations. The semi-empirical PM3 method pertaining to the Hyperchem program 8.0 (www.hyper.com) was employed to get the final geometries. The B3LYP method was employed with the 6-31G(d) basis set that was used in the PCgamelss program, and further single point calculations were done for the obtained optimised geometries [18]. The inhibition of LPS-induced TNP- α production by 15 compounds that were studied was all taken from the reference [19]. Fig. 1 displays the structures of 15 compounds.

4a

4b

4c

4d

4e

4f

4g

4h

Figure 1. Chemical Structures of compounds

3. Results and Discussion:

The theoretical molecular descriptors proceeds have been derived from the compounds' chemical structure. Table 1 lists out the set of prepared molecules. For each type, three QSAR models were generated; overall, wherein 3 models would represent 15 chalcones derivatives as well parametrically depicted Eqs. 1-3 along with correlation coefficient ($R^2=0.794-0.873$). It is already known that a high predictive ability could be achieved when the MLR model possesses low standard error (S), high R^2 and Fisher ratio (F) values and the minimum number of descriptors [20, 21].

The first model relied on just three descriptors, (Q_{C5}), (Q_{Br}) [21] and (μ^-), which were found to be a good model, wherein the correlation coefficient R^2 values were found to be 0.794, as Eq. 1.

$$pIC_{50} = 9.918(Q_{C5}) - 54.782(Q_{Br}) - 2.169(\mu^-) + 3.288 \dots (1)$$

The obtained equation's statistical features:

$$R^2=0.794, R_a^2 = 0.738, S=0.334, F=14.161, Q^2=0.796$$

Fig 2. Present the relationships between experimental pIC_{50} and predicted pIC_{50} by Eq 1.

Fig. 2 shows the relationship between the experimental pIC_{50} data and predicted pIC_{50} by this model 1

Comp.	μ	Table 1. Descriptors as the independent variables used for QSAR analysis of											QBr
4a	-3.657	-2.7045	-4.6095	0.103	1.465	5.562	1.752	3.657	3.81	-5.562	-1.752	124.05 ₂	0.157308
4b	-3.4445	-2.56825	-4.32075	0.037	1.473	5.197	1.692	3.4445	3.505	-5.197	-1.692	123.36 ₉	0.148331
4c	-3.4405	-2.50925	-4.37175	0.071	1.464	5.303	1.578	3.4405	3.725	-5.303	-1.578	123.76	0.155762
4d	-4.193	-3.162	-5.224	0.035	1.469	6.255	2.131	4.193	4.124	-6.255	-2.131	123.88 ₃	0.16132

4e	-4.204	-3.1155	-5.2925	0.069	1.476	6.381	2.027	4.204	4.354	-6.381	-2.027	123.21 9	0.166109
4f	-3.959	-2.9195	-4.9985	0.094	1.472	6.038	1.88	3.959	4.158	-6.038	-1.88	122.89 4	0.17749
4g	-3.9585	-2.89325	-5.02375	0.069	1.493	6.089	1.828	3.9585	4.261	-6.089	-1.828	123.01 3	0.181339
4h	-4.129	-3.078	-5.18	0.074	1.494	6.231	2.027	4.129	4.204	-6.231	-2.027	122.87 5	0.185489
4i	-4.0715	-3.05025	-5.09275	0.099	1.471	6.114	2.029	4.0715	4.085	-6.114	-2.029	123.45 3	0.179761
4j	-4.187	-3.157	-5.217	0.058	1.48	6.247	2.127	4.187	4.12	-6.247	-2.127	123.04 3	0.185387
4k	-4.2855	-3.25025	-5.32075	0.095	1.466	6.356	2.215	4.2855	4.141	-6.356	-2.215	123.80 8	0.167155
4m	-4.106	-2.9905	-5.2215	0.072	1.473	6.337	1.875	4.106	4.462	-6.337	-1.875	122.51 5	0.184552
4n	-4.0595	-3.38075	-4.73825	0.077	1.475	5.417	2.702	4.0595	2.715	-5.417	-2.702	122.85 2	0.174444
4o	-3.526	-2.545	-4.507	0.087	1.442	5.488	1.564	3.526	3.924	-5.488	-1.564	125.67 7	0.146784
4p	-3.917	-2.67	-5.164	0.082	1.473	6.411	1.423	3.917	4.988	-6.411	-1.423	122.17 3	0.187728

Table 1.Continued

Comp.	LUMO+1	LUMO+2	ω^-	ω^+	$\Delta\omega^+$	$r(\text{C}=\text{C})$	$r(\text{O}-\text{H})$	$r(\text{C}-\text{Br})$	$r(\text{C}=\text{O})$	η (ev)	QC2	QO4	QC3
4a	-0.634	-0.161	5.5767691	1.9197691	7.4965382	1.362	0.981	1.947	1.250	1.905	-0.12884	-0.42072	-0.1404
4b	-0.715	-0.195	5.3263568	1.8818568	7.2082136	1.355	0.98	1.95	1.248	1.7525	-0.08182	-0.38685	-0.13735
4c	-0.531	-0.078	5.1307914	1.6902914	6.8210829	1.361	0.981	1.946	1.251	1.8625	-0.13228	-0.41544	-0.10374
4d	-0.789	-0.571	6.6174045	2.4244045	9.0418089	1.358	0.981	1.949	1.251	2.062	-0.12589	-0.41619	-0.14786
4e	-1.123	-0.775	6.4332927	2.2292927	8.6625853	1.351	0.98	1.946	1.244	2.177	-0.12129	-0.38692	-0.13927
4f	-0.448	-0.174	6.0088991	2.0498991	8.0587981	1.354	0.981	1.935	1.246	2.079	-0.12798	-0.39921	-0.14062

4g	-0.514	-0.462	5.9230378	1.9645378	7.8875756	1.355	0.982	1.937	1.246	2.1305	-0.11485	-0.41733	0.079286
4h	-0.813	-0.726	6.382588	2.253588	8.636176	1.355	0.982	1.936	1.245	2.102	-0.11542	-0.41373	0.079548
4i	-0.473	-0.285	6.3491071	2.2776071	8.6267142	1.357	0.981	1.937	1.249	2.0425	-0.11913	-0.41177	-0.14559
4j	-0.991	-0.65	6.6060896	2.4190896	9.0251791	1.357	0.981	1.935	1.247	2.06	-0.14904	-0.41962	0.10377
4k	-0.742	-0.631	6.8366048	2.5511048	9.3877096	1.36	0.982	1.944	1.251	2.0705	-0.12644	-0.41832	-0.14629
4m	-2.193	-0.691	6.1102784	2.0042784	8.1145568	1.353	0.981	1.937	1.249	2.231	-0.1235	-0.40863	-0.14512
4n	-0.296	-0.027	8.2692497	4.2097497	12.478999	1.353	0.981	1.936	1.245	1.3575	-0.12573	-0.39495	-0.14523
4o	-0.454	-0.048	5.176618	1.650618	6.827236	1.364	0.98	1.95	1.253	1.962	-0.11902	-0.43295	-0.13239
4p	-2.356	-0.897	5.3462101	1.4292101	6.7754202	1.355	0.982	1.936	1.25	2.494	-0.12285	-0.412	-0.14404

Definition of Descriptors Used in This Study: μ =Chemical potential, μ^+ =Chemical potential for cation, μ^- =Chemical potential for anion, Q_{c5} : Charge of C5 atom, $r(C3-C5)$:Length bond between two carbon atoms three and five, I =ionization potential, A = electron affinity, χ =Electronegative, $E.Gap$ =Different between HOMO and LUMO is energy gaps in eV, HOMO=The energy of Highest Occupied Molecular Orbital in eV, LUMO= The energy of Lowest Unoccupied Molecular Orbital in eV, q_{Br} : Charge of bromine atom, $r(C-Br)$:Length bond between carbon atom and bromine atom, ω^- = Electro donating, ω^+ = Electro accepting, $\Delta\omega^-+$ = Electro accepting relative to the electro donating power, $r(C=C)$: Length bond between carbon atom and carbon atom, $r(O-H)$:Length bond between oxygen atom and hydrogen atom, $r(C=O)$: Length bond between carbon atom and oxygen atom, η : hardness, Q_{c3} : Charge of C 3 atom, Q_{c2} : Charge of C 2 atom, Q_{O4} : Charge of O 4 atom

In the first eq. 1 presented above, the negative value associated with the charge of bromide and chemical potential (μ^-) signifies that there is a decrease in the activity with rise in the charges values for bromide and chemical potential (μ^-) of the chalcones. On the other hand, positive values pertaining to charge of C5 as well as imply that there is a rise in activity with increase. The charge of atoms is responsible for the biological activity effects was reported in the literatures [22,23].

While in eq. 2, there were three descriptors, and a very good correlation coefficient R^2 could be established when the descriptor Q_{c5} was substituted with L (O4-C1-C2). Decrease in standard error could be observed when compared with Eq. 1. In this model, the activity of the chalcones was found to increase with the decrease in the charge of bromide as well as chemical potential (μ^-). In contrast, if there is an increase in angle (O4-C1-C2), it will increase the activity as well.

$$pIC_{50} = 0.444L(O4-C1-C2) - 26.493(Q_{Br}) - 1.873(\mu^-) - 54.151 \dots (2)$$

The obtained equation's statistical properties:

$$R^2 = 0.816, R_a^2 = 0.765, S = 0.316, F = 16.215, Q^2 = 0.818$$

Fig. 3 depicts the relationship existing between the predicted pIC_{50} via this model 2 and the experimental pIC_{50} data.

Fig.3, show the relationship between the experimental pIC_{50} data and predicted pIC_{50} by this model.2

In Eq. 3, three descriptors (LUMO+1), (LUMO+2) were employed and $r(C3-C5)$. This yielded the decrease standard error and increase correlation coefficient and the sequential Fischer test (F). The bond length, bond angle and LUMO molecular orbitals were applied to predict the biological activity [24].

$$pIC_{50} = 1.190(LUMO+1) - 3.149(LUMO+2) - 54.204r(C3-C5) + 84.939 \dots (3)$$

The obtained equation's statistical properties:

$$R^2 = 0.873, R_a^2 = 0.838, S = 0.262, F = 26.206, Q^2 = 0.876$$

Fig. 4 depicts the relationship existing between the predicted pIC_{50} via this model 3 and the experimental pIC_{50} data.

Fig.4, show the relationship between the experimental pIC_{50} data and predicted pIC_{50} by this model.3

Table 2 displays the variance inflation factors pertaining to descriptors employed in Eqs. 1, 2 and 3. In all equations, the VIF pertaining to the descriptors (Q_{C5}), (Q_{Br}), μ^- , L (O4-C1-C2), (LUMO+1), (LUMO+2) and r (C3-C5) were fairly large.

Table 2. The varied inflation factor (VIF) quantifies the severity of multicollinearity least squares regression analysis.

Equations Descriptor	Eq.1	Eq.2	Eq.3
Q_{C5}	3.454	-	-
Q_{Br}	3.187	4.511	-
μ^-	2.405	2.159	-
$L(O4-C1-C2)$	-	2.809	
HOMO+1	-	-	2.501
HOMO+2	-	-	3.196
$r(C3-C5)$		-	1.517

4. Conclusions

The calculated parameters based on Quantum chemical could be successfully employed in deriving and designing QSAR that can predict activity values of series of chalcones as anti-inflammatory. The study signified that modelling of the predicted anti-inflammatory for compounds under study could be done based on multiple linear regression (MLR). The good equation was derived based on twenty six parameters. The model was found to rely on Eq. 3. The best produced model is the MLR model that has a good statistical fit as clear from $Ra^2 = 0.838$, $R^2 = 0.873$, $S = 0.262$, and $F = 26.206$.

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