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QSAR of antioxidant activity of some novel sulfonamide derivatives

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Abstract

Development in discovery new drugs as antioxidants is still take a large interest between the scientists because oxidative stress has been caused a wide range of diseases such as chronic obstructive pulmonary disease (COPD), atherosclerosis, cancer and Alzheimer disease. In this article, we have been chosen benzene sulphonamide derivatives which had been synthesized and evaluated as anti-oxidants for QSAR study. We made mathematical relationship between (AOs) and three descriptors $log(\omega)$ electrophilicity, SCF and Mol Refractivity (IC₅₀ = - 6.38 - 9.26 log (ω)- 0.00173 SCF - 0.235 Mol Refractivity). By this equation, the values of experimental (IC₅₀) and estimated (IC₅₀) using the QSAR model have the convergence between the two values. By using QSAR model we predict new compounds have effectiveness as antioxidants. For example, **2g** and **2j** which have (calculated IC₅₀ = -0.33578 and 0.3412), respectively.

Keywords: QSAR, antioxidant activity, sulphonamide derivatives, molecular descriptors

1. Introduction

2. The microbial invasion can produce Reactive-Oxygen-Species (ROS)[1] Excess ROS can lead to oxidative stress. [2] Oxidative stress has been caused a wide range of diseases such as chronic obstructive pulmonary disease (COPD), atherosclerosis, cancer and Alzheimer disease. [3-5] Oxidative stress occurs due to an imbalance between production of oxidants and antioxidant defences and it damages biological systems. There are many mechanism explain oxidants contribute to cellular damage. [5, 6] So the scientists are trying to discover new antioxidants which are molecules that fight free radicals in the cell and

minimise oxidative stress. [7] From these compound sulphonamide derivatives (SO2-NH-), which are very important class in drug. The different functional groups in these compounds are responsible for their activity. They are quite stable and well tolerated in humans beings.[8] sulphonamides are very important several types of antibacterial,[9-11] in hypoglycaemic,[12] diuretic,[13, 141 antibacterial,[15] anticarbonic anhydrase,[16] antithyroid vitro in and in vivo, antiinflammatory,[17-19] anticancer,[20, 21] Anti HIV, antioxidant,[23] anticonvulsant,[24] [22] antitrypanosomal, [25] antitrypanosomal, [25] and antihypertensive.[26]

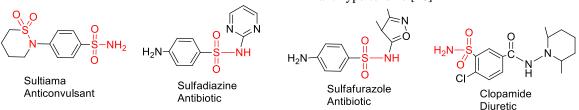


Figure 1: Examples of drugs synthesized from sulphonamides derivatives.

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Meanwhile, One of the most extensively utilized computational methods for designing pharmaceuticals and predicting pharmacological actions is quantitative relationships structure activity (QSARs).[27] (QSARs) are mathematical linear equations that aid in discovering and utilizing the relationship between molecular features and chemical. biological. toxicological, or other sorts of activities or attributes..[28] The target attribute of novel related compounds can be predicted using QSAR. As a result, strong and well-validated QSAR models can aid in the rational design of compounds with better activity, selectivity, safety, and/or physicochemical profile. [29, 30]

1- Computational methodology:

In this research, a data set of benzene sulphonamide derivatives which had been synthesized and evaluated as anti-oxidants has been selected for QSAR study.[31] In order to find factors that is directly related to their biological activity, through which mathematical models have been designed and the prediction of new compounds with biological effect.

In order to get on the best results of properties, one must follow the different steps in calculations including draw the structure of compounds. All molecules after built up by 2D Chemdraw then to ChemOffice Software were minimize their energy by using molecular mechanic MM2, then MMFF94. The reduction process is repeated until the root mean square (RMS) gradient value falls below 0.1 kcal/mol. Consequently, optimizations have been completed for those conformers through employing a semi-empirical method based on the Parameterized Model number3 (PM3). For optimization, the RHF wave function (restricted Hartree–Fock: closed shell) has been used until the root mean square (RMS) gradient was less than 0.01 kcal/mol.

Afterward, optimization structure and energy of the most stable conformer of the molecules, the next step was calculating of many of physical properties of compounds and also have been calculated the ϵ HOMO, ϵ LUMO, band gap Eg, hardness η , chemical potential μ , softness S, electronegativity χ , electrophilicity ω , and many descriptors.

Minitab release 14.1, was used for regression analysis of the calculated data. Firstly, a general correlation matrix that gathered all calculated properties has been done and that affected the biological activities properties. Really, such a matrix will be very useful in order to scrutinize if the existence of any combination between the calculated parameters with each other. Secondly, using this software, the varied combinations of descriptors were submitted to stepwise multiple linear regression analysis. The independent variables are individually added or eliminated from the model at each step of the regression based on the fisher ratio at the worth decided to enter and remove until the 'best' model is established in stepwise multiple linear regression analysis.

Thirdly, analysis of the results by using regression analysis based on the results of the stepwise to obtain the linear equation, and find the best models are summarized in statistical qualities of the models were reached by parameters like correlation coefficient (r), standard deviation SD, adjusted squared correlation coefficient (r2), and cross-validation (q^2) . Another point is using the best subsets regression statistical method which is regarded as an effective method for identifying models that accomplish the goals with the fewest possible predictors Subset models can estimate regression coefficients and predict future responses with less variation than full models with all factors. Then, using the multiple regression analysis in order to find the relationship between activity and two descriptors or more. Finally, after regression analysis, the relationship between experimental and calculated bioactivity (Fitted values) has been drawn.

Multilinear regression statistical analysis was performed to develop the optimum QSAR equation for predicting (IC₅₀). The analysis was carried out utilising the backward technique in the Minitab programme. The original data for sulfonamide derivatives was optimized in this study by selecting the independent and dependent factors that impact the QSAR equation. The independent variable was (IC₅₀) data, and the independent variables were a mixture of electrophilicity (ω), The self-consistent field (SCF), and Mol Refractivity. The output data is subjected to multilinear regression analysis using statistical metrics such as the correlation coefficient r, r², standard deviation (SE), and cross validation (q²). After considering the statistical data, the model is selected.

2- Results and Discussion:

Table 1, shown the chemical structures of benzene sulphonamide derivatives and experimental IC_{50} (mg/ml) as anti-oxidants.[31]

Table 2 shown the modelling for all the molecules and calculate descriptors using PM3 and HF 6-311 G (d,p) bases set calculations.

Comp. No.	Compounds structure	0.05 mg/mL (%)	0.10 mg/mL (%)	0.15 mg/mL (%)	0.20 mg/mL (%)	0.25 mg/mL (%)	IC ₅₀ (mg/mL)
1a		2.44	7.07	9.90	14.91	19.15	0.5358
1b		3.34	8.61	13.50	17.74	22.24	0.5080
1c	H ₃ C	1.54	6.94	10.28	13.37	23.14	0.3799
1d	CH ₃ CH ₃ CH ₃ CH ₃ CH ₃	2.57	7.46	10.93	15.94	20.18	0.5294
1e	H ₃ C	2.83	8.23	13.24	22.62	32.13	0.3287
1f	H ₃ CS HN CH ₃	3.31	7.76	10.58	14.31	18.91	0.4779
1g		2.06	4.76	7.58	11.43	13.88	0.8475
1h		2.19	6.17	9.25	14.91	20.18	0.5250
1i	H ₃ C	2.31	6.17	9.90	14.65	21.72	0.4653
1j	OH CH3 CH3	3.34	9.00	13.24	19.28	22.75	0.5826
* E == at al	Vitamin C 2019: Table 1 the chemical structures and	11.31	21.85	34.32	48.20	60.15	0.2090

*Eze et al 2019: Table 1 the chemical structures and IC_{50} (mg/ml) for the studied compound

Com. No. Descriptors	1 a	1b	1c	1d	1e	1f	1g	1h	1i	1j
•	-	-	-	-	-	-	-	-	-	-
HOMO (ev)	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3	0.3
	398	573	543	599	203	210	483	460	517	531
	6	5	2	5	4	3	7	5	2	3
	0.0	0.0	0.0	0.0	-	0.0	0.0	0.0	0.0	0.0
LUMO (ev)	316	276	309	268	0.0	355	244	318	370	370
	8	6	8	5	352	4	5	1	7	3
	0.3	0.3	0.3		0.2	0.2	0.3	0.3	0.3	
GAP (ev)	081	296	233	0.3	851	854	239	142	146	0.3
	8	9	4	331	4	9	2	4	5	161
	2.3	1.8	3.3	2.8	2.4	1.9	1.5	1.0	1.2	0.7
Log P	63	76	21	34	21	34	51	63	32	45
	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1		0.1
Electrophilicity	119	124	147	122	108	113	072	135	0.1	203
Lieedopiinienty	82	03	83	89	3	37	76	9	201	93
	-	-	-	0)	-	-	-	-		-
Chamical not	0.1	0.1	0.1	-	0.1	0.1	0.1	0.1	-	0.1
Chemical pot.	857	925	926	0.1	777	782	864	889	0.1	950
	7	1	5	934	7	9	1	3	944	8
	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1	0.1
Hardness	540	648	616	665	425	427	619	571	573	580
	9	45	7	5	7	45	6	2	25	5
	6.7	3.7	3.2	3.5	5.4	5.0	2.0	4.6	5.9	4.3
Dipole (Debye)	229	273	396	368	07	558	854	809	402	013
RMS	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
(Kcal/Mol)	016	012	03	022	01	012	011	028	017	008
	-	_	_		_	_	_	_	_	_
SCF	111	103	128	-	115	105	155	150	154	143
(Kcal/Mol)	.96	.47	.07	120	.05	.28	.83	.79	.04	.72
Connolly	575	491	651	595	642	611		572	488	545
Acc. Area	.48	.81	.76	.99	.10	.28	576	.97	.33	.91
(A° ²)	6	2	5	7	5	2	.32	3	6	5
Connolly	315.7	274.6	397.5	328.5	355.9	337.2	314.1	310.7	263.6	291.6
Molec. Area $(A^{\circ 2})$	41	44	34	99	05	99	13	06	209.0 99	02
Connolly	293.1	288.2	405.2	307.4	327.4	310.7	298.8	274.8	226.2	253.5
Sol.Excl.Vol.(A ^{o3})	39	57	11	89	21	65	44	271.0	41	233.5
Source (III)	1.479	1.301	1.557	1.491	1.549	1.520	1.453	1.519	1.468	1.505
Ovality	56	47	88	52	22	23	1.435	96	62	32
Molar Refractivity	8.794		00	8.972	9.778	9.314	8.661	8.197	8.197	7.733
(cm^3/mol)	8	8.331	9.436	2	5	7	5	0.1 <i>) </i> 7	0.1 <i>) </i> 7	9
		2 (02	3.899	3.400	2.590	2.091	1.770	1.271	1.461	0.962
Partition Coeff.	3.192	2.693	2	2	2	2	4	4	4	4
Molecular Volume	3186.	2242.	2794.	3018.	3053.	3421.	2724.	2466.	2630.	2552.
(bohr**3/mol)	29	72	24	37	77	8	23	62	06	46
Entropy(Cal/Mol.	136.0	130.1	150.2	138.5	152.2	145.2	142.9	136.8	135.7	129.7
K)	95	91	51	5	43	9	9	7	74	81
Therm. Energy	271.1	251.6	305.6	285.3	287.0	267.6	269.9	250.7	249.9	230.9
(Kcal/Mol)	63	75	03	79	9	09	4	19	93	84
Zero-Point	259.8	241.1	292.5	272.7	274.0	255.4	257.7	239.4	238.8	220.7
Energy(Kcal/Mol)	87	66	84	86	93	75	92	62	29	01
	36360	28953	66482	53478	67497	54360	53796	42800	44350	34945
Balaban Index	4	4	1	3	9	7	0	9	0	2
Cluster Count	22	21	23	22	23	22	22	21	21	20
	8340	7259	9733	8540	9629	8448	8372	7285	7539	6524

 Table 2: The descriptors calculated for the studied compounds

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							Index
20.04 20.04 10.04 10.04	75.27	75.27	75.27	75.27	66.48	66.48	Pol.Sur.Area (A°2)
20.04 20.04 19.04 19.04 19.05	20.04	21.04	20.04	21.04	19.04	20.04	C1 A 44 11 4
55 55 76 76 18.05	55	35	55	35	76	55	Shape Attribute
44 44 42 42 40	44	46	44	46	44	46	Sum of degrees
65.33 70.66 68.66 68.66 66.66	65.33	67.33	66.66	68.66	66.66	68.66	Sum of Val.
33 67 67 67 67	33	33	67	67	67	67	degrees
12 13 12 13 12	12	13	12	13	12	13	Topological Diameter (bonds)
0.001 0.001 0.001 0.001 0.002	0.001	0.000	0.001	0.001	0.000	0.000	Total
063 418 736 736 126	063	868	228	002	868	709	Connect.
1137 1125 982 1018 884	1137	1292	1118	1272	952	1092	Wiener Index
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	44 65.33 33 12 0.001 063	46 67.33 33 13 0.000 868	44 66.66 67 12 0.001 228	46 68.66 67 13 0.001 002	44 66.66 67 12 0.000 868	46 68.66 67 13 0.000 709	Sum of Val. degrees Topological Diameter (bonds) Total Connect.

By using the Minitab and statistical methods to analysis and simulate of all descriptors with the (AO) activity. We conclude that cannot find mathematical relationship between all of them. So we carry out the filtering process for these descriptors by relied on the matrix result.

The correlation matrix shows the relationship between two variables and the best measure performs when the variables are connected in a linear fashion. Table 3 lists the descriptors after filtering and we obtain the mathematical relationship between (AOs) and three descriptors as shown in eq. (1):

$IC_{50} = -6.38 - 9.26 \log (\omega) - 0.00173 SCF - 0.235$ Mol Refractivity ------(1)

n=8 r^2 =96.4 % q^2 =93.6 % SE = 0.0396 The eq. (1) denotes the best QSAR model for sulfonamide derivatives. The predictive power of QSAR models is calculated by internal and external validate-using the methods of leave-one-out (LOO).

Statistical metrics of cross validation coefficients also validated the training and test sets' prediction capacity (q2). Three descriptors employed in the model have a strong link with (AO) bioactivity. (IC50). The result shows that (IC50) depends on these descriptors, so the (ω) is calculated from the orbitals energy using the eq. (2) below:

$$\omega = \mu^2 / 2\eta$$
 -----(2)

Where μ is the chemical potential & it is calculating by eq. (3)

$$\mu = 0.5 (\varepsilon_{HOMO} + \varepsilon_{LUMO}) -----(3)$$

 η is the Hardness that calculate by eq. (4)

$$η = 0.5$$
 (ε_{HOMO} - ε_{LUMO}) -----(4)

From the model we can conclude that an electronic variable, (ω) , SCF represents The self-consistent field approach for calculating a quantum many-body system's ground state wave function. As well as the Molar Refractivity, which is a measure of a compound's overall polarizability and is affected by temperature, pressure, and index of refraction. The Molar Refractivity is one of the most influenced attributes on the (IC50).

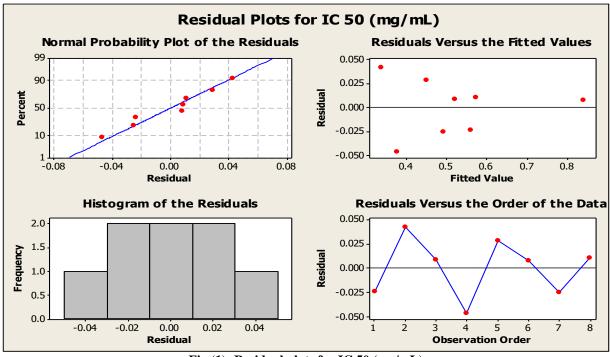
Table 4 indicates the values of experimental (IC50) and estimated (IC50) using the QSAR model. It shows

the convergence between the two values and the presence of a slight difference that can be neglected.[32]

Table4. Shown the experimental (IC_{50}) and calculated (IC_{50})

$(1C_{50})$									
Compou	Experime	Calcula	Resid	[Resid					
nd)	ntal	ted	ual	ual					
	(IC_{50})	(IC_{50})	error	error] ²					
	mg/mL	mg/mL							
1a	0.5358	0.5598	-	0.0005					
			0.024	760					
			0						
1b	0.3799	0.3378	0.042	0.0017					
			1	724					
1c	0.5294	0.5211	0.008	0.0000					
			3	689					
1e	0.3287	0.3760	-	0.0022					
			0.047	373					
			3						
1f	0.4779	0.4495	0.028	0.0008					
			4	066					
1g	0.8475	0.8396	0.007	0.0000					
			9	624					
1i	0.4653	0.4911	-	0.0006					
			0.025	656					
			8						
1j	0.5826	0.5722	0.010	0.0001					
			4	082					
PRESS			0.000787175						

By using QSAR model we can predict new compounds have effectiveness as antioxidants and apply all the steps of the theoretical calculation to calculate descriptors of the proposed model.[27] The new compounds are derived from compounds 1a-c and 1e by replacement CH3 in butyl group by NH2, OH or H to give compounds 2a-1 respectively. Table 5From table 5, there is a difference in the values of the theoretical (IC50) between the compounds. This difference is due to the different electrophilicity values, SCF, and the mol refractivity.





The compounds 2g and 2j showed the lowest values of IC50 (calculated IC50 =0.33578 and 0.3412), respectively, which proves and confirms the success of the mathematical model, where the lower the value of (IC50) meaning the least amount of the compound gives the highest effectiveness. This activity may be due to the presence of the amine group as a

compensator which has a medicinal character and biological activity. [33, 34] It is also possible to suggest many other compounds and calculate their effectiveness theoretically through the proposed model based on the available data.

NO ·	COM.	Єномо	ELUMO	Electro- philicit y (ω)	Log(ω)	Mol.Re f	SCF	IC50
2a	H ₃ C	0.3666 5	-0.085	0.18106 4	0.7421 7	8.6997	- 99.56	1.3797 2
2b		0.3558	0.0277	0.11207 3	0.9505	8.4841	145.1 5	0.6789 7
2c		0.3558	0.0273	0.11169 4	0.9519 7	8.331	106.7 2	0.6620 9

Table 5: The predicted compounds and their important descriptors with calculated (IC₅₀) by the model of QSAR

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	NH ₂							
2d		0.3565 8	0.0255 2	0.11025	0.9576 1	8.331	- 89.79	0.6850 4
2e		0.3579 4	0.0282	0.11307 5	0.9466 4	8.0203	135.3 8	0.7352 8
2f		0.3561 1	0.0280 5	0.11246 3	0.9489 9	7.8672	- 96.86	0.7264 1
2g	CH ₃ CH ₃ CH ₃ CH ₃ CH ₃	0.3451 2 8	0.0329	0.11443 6	- 0.9414 4	9.3409	- 111.6 7	0.3357 8
2h	H ₃ C	- 0.3494 1	0.0320 5	0.11462 7	- 0.9407 1	9.1253	- 157.4 7	0.4589 9
2i	H ₃ CH ₃ H ₃ H	- 0.3478 6	0.0321 1	0.11430 9	- 0.9419 2	8.9722	- 118.8 2	0.4392 7
2j	H ₃ C	0.3206	0.0357 9	0.11149 9	0.9527	9.6834	- 100.8 9	0.3412
2k	H ₃ CS HN	0.3215	0.0361	0.11203 7	0.9506 4	9.4678	146.4 3	0.4513 2
21	H ₃ CS HN	0.3264	0.0338 7	0.11092 9	0.9549 6	9.3147	108.9 2	0.4623 7

3- Conclusions

This study used a theoretical approach on a few molecular descriptors to create a model that might be used to link the structure of benzenesulfonamide derivatives as antioxidants to their biological activity. Internal and external validation using leave-one-out (LOO) approaches are used to determine the prediction potential of QSAR models. Statistical metrics of cross validation coefficients also validated the training and test sets' prediction capacity (q^2) .

There is strong relationship between three descriptors which used in the model and (AO) bioactivity (IC₅₀). The result shows that (IC₅₀) depends on these descriptors. We used QSAR to predict new compounds have high effectiveness as antioxidants. QSAR model is robust, reliable and significant. QSAR model will be useful in medicinal chemists and pharmaceutical to design and synthesize novel antioxidants drugs.

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