THEORETICAL BIO-SIGNIFICANCE EVALUATION OF QUINAZOLINE ANALOGUES

ABSTRACT

Structure of nine quinazoline derivatives were optimized using density functional theory method in order to probe into the bioactivities of the compounds. The obtained descriptors which described the anti-human neuroepithelioma activity of the compounds were selected and used to develop a model using partial least square method. The developed model replicated the experimental IC_{50} indicative of the predicting power of the model. In addition, ligand-receptor interactions are reported and 2-((E)-2-(4-Bromo-phenyl)-vinyl)-3H-quinazolin-4-one (A₄) showed the greatest affinity to bind on the active site of human neuroepithelioma cell line.

Keywords: QSAR, DFT, quinazoline analogues, docking, neuroepithelioma cell line.

1.0 INTRODUCTION

Several natural and synthetic quinazoline based products as antimicrobial agents have been successfully developed over the last four decades (Carlos et al. 2013 and Majumdar et al. 2011). Quinazoline, a heterocyclic compound with molecular formula, $C_8H_6N_2$, has a double-ring structure consisting of a benzene ring fused with pyrimidine ring, and exists as yellow crystals (Connolly et al. 2005 and Abida et al. 2011). Quinazoline and its derivatives have shown various biological activities as antidepressant anti-inflammatory, antimalarial, anticancer, antifungal, antimicrobial, antiviral, anti-tubercular, anti-protozoan, anticonvulsant (El-Messery et al. 2012; Berest et al. 2011; Hu et al. 2015 and Ammerdorffer et al. 2017).

Bio-significance of molecular compounds is determined by the structure and the electronic properties of the molecules. Thus, in this work, the optimised structure and electronic properties of 2-((E)-Styryl)-3H-quinazolin-4-one (A₁), 2-((E)-2-(2-Methoxy-

phenyl)-vinyl)-3H-quinazolin-4-One (A_2), 2-((E)-2-(3-Methoxy-phenyl)-vinyl)-3Hquinazolin-4-one (A_3), 2-((E)-2-(4-Bromo-phenyl)-vinyl)-3H-quinazolin-4-one (A_4), 4-Chloro-2-((E)-styryl)-quinazoline (B_1), 4-Chloro-2-((E)-2-(2-methoxy-phenyl)-vinyl)quinazoline (B_2), 4-Chloro-2-((E)-2-(3-methoxy-phenyl)-vinyl)-quinazoline (B_3), 4-Chloro-2-((E)-2-(4-methoxy-phenyl)-vinyl)-quinazoline (B_4) and 4-Chloro-2-((E)-2-(2,4-dimethoxyphenyl)-vinyl)-quinazoline (B_5) were computed using quantum chemical method with a view to theoretically investigate the anti-human neuroepithelioma activity of the compounds through the development of a quantitative structural activity relationship, QSAR and molecular docking study. This set of nine quinazoline derivatives have been experimentally shown to have expressive bioactivity against human neuroepithelioma (Anna et al. 2010).

2.0 MATERIALS AND METHODS

2.1 Optimization

The studied compounds (Figure 1) were optimized using density functional theory via $6-31G^*$ basis set. The density functional theory method used in this work was done with B3LYP density functionals which include Becke's gradient exchange (Becke 1993) and Lee, Yang and Parr correlation (Lee et al. 1988). The accuracy of density functional theory calculations is based on the selected basis set. Among the calculated descriptors are the highest occupied molecular orbital (E_{HOMO}) energy, lowest unoccupied molecular orbital (E_{LUMO}) energy, dipole moment, chemical potential, chemical hardness, etc. The quantum chemical calculations were done using Spartan '14 software by wavefunction Inc.



A₁: H, A₂: 2-OMe, A₃: 3-OMe, A₄: 4-Br, B₁: H, B₂: 2-OMe, B₃: 3-OMe, B₄: 4-OMe, B₅: 2, 4-OMe.

Figure 1: Schematic structures of the studied Quinazolines

2.2 Descriptor selection and QSAR model development

The study of quantitative structure activity relationship (QSAR) require the use of appropriate molecular descriptors (with high simplicity and high predicting power) from the entire calculated descriptors (Leardi et al. 1992 and Pourbasheer et al. 2009). Therefore, QSAR model was developed via multiple linear regression method. The predicted bioactivity was calculated using partial least square method. The software used for partial least square method was XLSTAT 2018.

2.3 Quality and Validation of QSAR Model

The evaluation of the quality and validation of developed QSAR model were done by considering cross validation (Cv.R²) and adjusted R² (R_{adj}^2) [Equation 1 and 2].

$$CV.R^{2} = 1 - \frac{\sum(Y_{obs} - Y_{cal})^{2}}{\sum(Y_{obs} - \bar{Y}_{obs})^{2}} - \dots - (1)$$

The adjusted R^2 could be calculated using equation (2)

$$R_a^2 = \frac{(N-I) \times R^2 - P}{N-1 - P}$$
(2)

The developed quantitative structural activity relationship model could be considered prognostic, if C_v . $R^2 > 0.5$ and $R_{adj}^2 > 0.6$ (Golbraikh et al. 2002; Marrero et al. 2004 and Oyebamiji et al. 2018).

2.4 Molecular docking studies

Molecular docking studies were performed by using the following softwares; Discovery Studio 4.1, Autodock tools 1.5.6, Autodock vina 1.1.2, and pymol 1.7.4.4. The receptor used in this work was obtained from protein data bank and adjusted for docking purpose. For the purpose of accuracy in docking, the residues (water molecules and crystallized drug-like molecules) were removed from the protein (**4bjx**) (Wyce et al. 2013). The treated receptor was subjected to autodock tool 1.5.6 for location of binding site and development of a grid box to cover the neuroepithelioma cell line binding site. Autodock vina software was employed to simulate the ligand into the active site of the receptor (protein) in order to calculate the binding energy of the ligand-receptor complexes. Also, inhibition constant was calculated using equation 3.

 $K_i = e^{\frac{-\Delta G}{RT}}$ (3)

3.0. *Results and Discussion*

-5.91

-5.96

-5.78

-5.72

 \mathbf{B}_2

B₃

B₄

B₅

-2.37

-2.41

-2.37

-2.35

3.54

3.55

3.41

3.37

4.05

3.72

4.35

2.93

4.71

4.71

4.71

4.58

3.1 Electronic Descriptors and QSAR Studies

The calculated descriptors include E_{HOMO} , E_{LUMO} , dipole moment (DM), Log P, molecular weight (MW), polar surface area (PSA), polarizability and ovality (Table 1).

	HOMO	LUMO	BG	DM				PSA	HBD	HBA		
Comp	(eV)	(eV)	(eV)	(Debye)	LOGP	MW	OVALITY	$(Å^2)$			POL	IC ₅₀
A ₁	-6.04	-1.70	4.34	4.12	3.39	248.285	1.40	32.085	0	2	61.55	1.39
A ₂	-6.02	-1.74	4.28	4.82	3.26	278.311	1.42	39.584	0	4	63.56	2.88
A ₃	-5.98	-1.73	4.25	5.71	3.26	278.311	1.46	39.100	0	3	63.77	5.13
A ₄	-6.05	-1.80	4.25	2.16	4.22	327.181	1.44	32.079	0	2	63.04	2.85
B ₁	-6.16	-2.28	3.88	3.10	4.83	266.731	1.41	13.636	0	3	62.18	4.96

297.757

296.757

296.757

326.783

1.46

1.46

1.46

1.51

0

0

0

0

20.169

20.935

20.938

26.775

3

8

2

3

64.41

64.40

64.43

66.64

0.58

5.28

4.34

1.75

Table 1: The calculated molecular descriptors obtained for quinazoline derivatives

The calculated descriptors were used as independent variables and the experimental IC_{50} (Table 2), served as dependent variables in the development of QSAR model using multiple linear regression. This was used to select the molecular parameters that perfectly describe anticancer activity of quinazolines. The developed QSAR model (Equation 4) was used to predict the bioactivity of the studied compounds (Table 2). The calculated squared correlation coefficient (R²) (0.982) and the adjusted R² (0.856) revealed the predicting power

and quality of the model developed as shown in Table 3. The model reproduced the observed IC_{50} as depicted by the residual values (Observed IC_{50} – Predicted IC_{50}) (Table 2).

 $IC_{50} = -450.637 - 11.660(E_{HOMO}) + 20.202(E_{LUMO}) - 0.945(MW) + 87.435(LOGP) + 133.365(OVALITY) + 5.227(PSA) ------(4)$

	Observed IC ₅₀	Predicted IC ₅₀	Residual
A_1	1.390	1.564	-0.174
A_2	2.880	2.638	0.242
A_3	5.130	5.178	-0.048
A_4	2.850	2.947	-0.097
\mathbf{B}_1	4.960	4.606	0.354
B_2	0.580	0.868	-0.288
B_3	5.280	5.592	-0.312
\overline{B}_4	4.340	4.317	0.023
B_5	1.750	1.450	0.300

Table 2: Observed IC₅₀ and Predicted IC₅₀

Table 3: Statistical parameters for developed QSAR model

N	p	R ²	R ² _{adj}
9	6	0.980	0.920

This developed QSAR model as shown in equation 4 shows a positive input of E_{LUMO} , Log P, Ovality and polar surface area (PSA). However, E_{HOMO} and molecular weight contributed negatively to the bioactivity and suggesting that, as E_{HOMO} or molecular weight decreases, biological activity of Quinazoline derivatives increases and vice-versa for E_{LUMO} , Log P, ovality and polar surface area. The QSAR model showed that the anticancer activity of the compounds was directly linked to these molecular parameters. Figure 2 demonstrates the quality of the developed QSAR while as shown in Figure 3, the residual values were observed on both positive and negative sides of the graph and this indicated that there was no systemic inaccuracy in the developed QSAR model.



Figure 2: Plot of predicted IC₅₀ against observed IC₅₀



Figure 3: Plot of residual values against observed IC₅₀

4.0 Docking Studies

The study of molecular docking was executed in this research in order to comprehend the binding mode of drug-like molecule against the receptor (4bjx). First and foremost, Ramachandran plot was used to evaluate the orientation and conformation of human neuroepithelioma cell line (PDB ID: 4bjx), and it was discovered that almost all the residue in the studied receptor (i.e. 99.2% of the residues) was found in a favoured region and the entire residues (i.e. 100.0% of the residues) were in allowed regions. More so, there were no outliers as shown in Figure 4. Thus, the receptor used in this work was stable and of a good quality. As shown in Table 4, the dock scores for ligand-receptor complexes ranged from -8.2 to -7.6 kcal/mol and it was discovered that A_4 had the highest possibility to inhibit human neuroepithelioma cell. The residues involve in the interaction between studied compounds (A₁-B₅) and **4bjx** were VAL-87, CYS-136, LEU-92, TRP-81, PRO-82 for A₁, LEU-92, TRP-81, VAL-87, ASN-140, PRO-82, CYS-136, ILE-146 for A2, ASN140, CYS-136, PRO82, GLN-85, LEU-92, LYS-91, ILE-146, VAL-87 for A₃, ASN140 and PRO82 for A4, CYS-136, VAL-87, LEU-92, PRO-82, ILE-146, TRP-81 for B1, VAL-87, CYS-136, LEU-92, PRO-82, TRP-81 for B₂, MET-132, MET-105, PRO-82, TRP-81, VAL-87, ILE-146, CYS-136, LEU-92 for B₃, LEU-92, VAL-87, CYS-136, ILE-146, MET-105, TRP-81 for B₄ and CYS-136, ILE-146, VAL-87, GLN-85, LEU-92, TRP-81, TYR-97, LEU-94, TYR-139 for B₅. The ligand and receptor complex displayed in Figure 5 showing the residue that was involved in the interaction together with hydrogen bonds for A₄.

Also, Lipinski's rule of five was observed for the compounds used in this work and all the studied compounds proved to be drug-like compounds. As shown in Table 1, the molecular weight value were \leq 500, Hydrogen bond acceptor were \leq 10, Hydrogen bond donor were less than \leq 5 and calculated logP value were also \leq 5.



Figure 4: Ramachandran plot for human neuroepithelioma cell line (PDB ID: 4bjx)

Compound	Affinity (kcal/mol)	K _i	Interacting Residues
A ₁	-8.1	8.72 x 10 ⁵	VAL-87, CYS-136, LEU-92, TRP-81, PRO-82
A ₂	-8.0	7.37 x 10 ⁵	LEU-92, TRP-81, VAL-87, ASN-140, PRO-82, CYS-136,
			ILE-146
A ₃	-8.1	8.72 x 10 ⁵	ASN140, CYS-136, PRO82, GLN-85, LEU-92, LYS-91, ILE-
			146, VAL-87
A ₄	-8.2	1.03 x 10 ⁶	ASN140 and PRO82
B ₁	-8.1	8.72 x 10 ⁵	CYS-136, VAL-87, LEU-92, PRO-82, ILE-146, TRP-81
B ₂	-8.1	8.72 x 10 ⁵	VAL-87, CYS-136, LEU-92, PRO-82, TRP-81
B ₃	-7.8	5.25 x 10 ⁵	MET-132, MET-105, PRO-82, TRP-81, VAL-87, ILE-146,
			CYS-136, LEU-92
B ₄	-7.7	4.44 x 10 ⁵	LEU-92, VAL-87, CYS-136, ILE-146, MET-105, TRP-81
B ₅	-7.6	3.75 x 10 ⁵	CYS-136, ILE-146, VAL-87, GLN-85, LEU-92, TRP-81,
			TYR-97, LEU-94, TYR-139

Table 4: Ligand-receptor dock score of quinazoline derivatives with 4bjx.



Figure 5: Interactions of A_4 with the residue in the active gouge of human neuroepithelioma cell line (4bjx).

Conclusion

The anti-human neuroepithelioma activity of quinazoline analogues was observed using theoretical approach. In this study, density functional theory was use to optimized the molecular compounds and the descriptors obtained were used for QSAR analysis using partial least square method. The descriptors, E_{HOMO} , E_{LUMO} , Log P, ovality, PSA, and molecular weight were implicated in the anti-human neuroepithelioma activity of quinazoline derivatives and the developed QSAR model replicated the experimental IC₅₀. Also, 2-((E)-2-(4-Bromo-phenyl)-vinyl)-3H-quinazolin-4-one (A_4) inhibited the 4bjx most as revealed by the affinity and the inhibition constant (K_i).

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