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# **QSPR/QSAR Modelling of the Antioxidant Properties of Some Flavonoids**

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# *Authors' contributions*

*This work was carried out in collaboration among all authors. All authors read and approved the final manuscript.*

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# **ABSTRACT**

Several methods exist when seeking to experimentally evaluate the antioxidant properties of a natural bioactive substance. In the case of flavonoids, the methods used are mainly based on the experimental determination of the percentage of inhibition (IC50) or the redox potential (E). In the present work, a prediction study of the redox potential *E* and the inhibitory concentration *LogIC50* was carried out, using the AM1 and HF/6-311G(d,p) method. At the end of this study, three (03) QSPR models were validated and retained, one (01) for the prediction of the redox potential and four (02) for the prediction of the inhibitory concentration : **The Redox Prediction Model**, developed at the AM1 approximation level, for which 96.43 of the experimental variance is explained by the descriptors :  $E= -0.29 + 0.22E_{Homo} + 0.11E_{Lumo} - 0.05\overline{\omega}$  **The Inhibitory Concentration Prediction Models**, developed at the AM1 level, for which 96.35P of the experimental variance is explained by the descriptors :  $LogIC_{50} = -4,92 + 11,37E_{Homo} + 34,36E_{Lumo} + 0,67\overline{\omega}$  **The Inhibitory Concentration Prediction Model**, developed at the HF/6-311G level (d, p), for which 99.96P of the experimental variance is explained by the descriptors.  $LogIC_{50}$  =

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62,40 + 80,25 *EHomo* - 28,44*Elumo* + 52,01*S* - 71,26 *η* - 6,11*μ*

The development of these QSPR models represents a significant advance in predicting the antioxidant properties of bioactive molecules such as flavonoids based on descriptors calculated by quantum chemical methods.

*Keywords: Antioxidant; properties; QSAR/QSPR; quantum descriptors.*

# **1. INTRODUCTION**

The field of investigation is vast when it comes to experimentally evaluating the antioxidant properties of flavonoids. The methods used are mainly based on the experimental determination of the percentage inhibition (IC50) or the redox potential (E).

Indeed, several results published in the literature have shown that experimental parameters such as redox potential and inhibitory concentration allow to evaluate the antioxidant powers of bioactive molecules [1,2,3,4,5] (Jorgensen et *al.* 1999; Volikakis et *al.* 2000; Yamamura, 2003 and and Dragan AMI et *al.* 2017).

In the present work, QSAR/QSPR models for predicting antioxidant properties were developed. The aim is to find a linear relationship that will predict the experimental parameters E or IC50 as a function of quantum descriptors such as : electronic affinity (EA), ionisation energy (EI), hardness (η), softness (S), electronegativity (χ), electrophilic index (ω), energy (HOMO), energy (LOMO), energy gap (HOMO-LUMO), electrophilic index (ω), dipole moment (μ*),*  electron donor  $\overline{(\omega)}$  and electron acceptor  $\overline{(\omega)}$ , used in our previous studies.

From the results obtained, the redox potential (E) and the inhibitory concentration (IC50) of the flavonoids will be predicted and consequently their antioxidant power.

# **2. MATERIALS AND METHODOLOGY**

# **2.1 Materials**

In our work, 29 flavonoids with known experimental redox potential (E) and inhibitory concentration (IC50) values were selected as the structural basis for study. These flavonoids are classified into two groups or series: the learning series with 19 molecules ( $\approx$  2/3 of the base molecules) and the test series with 10 molecules (≈ 1⁄3 of the base molecules). The choice of molecules for the constitution of the groups is

arbitrary. These molecules are coded  $M_i$  in order to simplify their notations.

The molecules of the learning set will be used to develop predictive models of redox potential and inhibitory concentration from quantum descriptors, and those of the test set for validation of the developed models. The experimental values of the redox potential and the inhibitory concentration of the molecules are taken from the literature [6].

# **2.2 Methodology**

For the determination of the quantum descriptors on which the prediction models of the redox potential E and the inhibitory concentration IC50 of the studied molecules were developed, different levels of calculations were used. These are : AM1 and HF/6-311G (d, p). These levels of theory were chosen in view of the size of the molecules and the different quantum parameters to be evaluated.

### **2.2.1 Statistical analysis**

All molecules were optimised using the GAUSSIAN 09 program**.** Two software packages were used, according to their specificities, to perform the statistical analysis of the results and to plot the graph, i.e. XLSTAT and MATLAB**.** The choice of the quantum descriptors is based on two fundamental criteria.

**Criterion 1**: By nature, the dependence of *Y* on  $X_i$  is assumed to be linear. Therefore the

absolute value of the linear correlation coefficient

between the property  $Y$  and the variables  $X_i$ must be greater than 0.50: **|R| ≥0.50**

**Criterion 2**: The different samples *Yi* are assumed to be independent of each other. For two descriptors to be independent, the partial correlation coefficient (<sup>a<sub>ij</sub>)</sup> between these two descriptors i and j must be strictly less than 0.70 :*aij***<** 0.70

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## **Table 1. Structure of the flavonoids studied**





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## **Basic structure of flavonoids**<br>Molecules <sub>C2</sub> R3 **Molecules C2 R3 R5 R6 R7 R8 R12 R13 R14 R15 Names** M24 C=O OH OH H OH H OH H H H Morin M25 C=O OH OH H OH H H H H H Galangin M26 C=O Rutinose H H H H H OH OH H Rutin M27 C=O Glucose OH H OH H H OH H H Hyperoside M28 C=O OH OH OH H H H H H H Baïcalein M29 C=O OH OH Glucose H H H H H H Baïcalin



#### **Table 2. Molecules in the learning and test series**

**Chart 1. Quantum descriptors used. Debye (D)***; Electron-volt (eV)*

<b>Quantum descriptors</b>	Rating	<b>Expression</b>	Unit
Dipole moment	μ		(D)
Energy of the HOMO	$E_{Homo}$		(eV)
Energy from the LUMO	$E_{Lumo}$		(eV)
Electronic affinity	AE	$AE = -E_{Lum0}[12]$	(eV)
Ionisation energy	IE	$IE = -E_{Homo}([13])$	(eV)
GAP (HOMO-LUMO)	Gap	Gap = $E_{Lumo}$ . $E_{Homo}$	(eV)
Electronegativity	X	$X = \frac{(IP + EA)}{2}$ [14]	(eV)
Hardness.	η	$\eta = \frac{(IP - EA)}{2}$ [15]	(eV)
Softness	S	$S = \frac{1}{2\eta} [16]$	$(eV)^{-1}$
Electrophilic Index	ω	$ω = \frac{x^2}{2η} [17]$	(eV)
Donor electron power	$\overline{\omega}^-$	$\overline{\omega}^{\scriptscriptstyle -} = \frac{(3 \text{.EI} + \text{AE})^2}{16(\text{EI} - \text{AE})}$ [18]	(eV)
Electron acceptor power	$\overline{\omega}^+$	$\bar{\omega}^+ = \frac{(EI+3AE)^2}{1.00 \times 10^{10}}$ [19] $16$ $(EI-AE)$	(eV)

The predictive power of a model is also based on the Tropsha criteria. If the three fifths (3/5) of the criteria are verified then the model has a good predictive power. Normality tests were also carried out to verify the quality of the confidence interval obtained. These are the Shapiro-Wilk and Durbin-Watson tests.

## **2.2.2 Theoretical descriptors**

These are quantum descriptors calculated by quantum chemical methods. Chart 1 shows the quantum descriptors used in this study [7,8,9,10,11] [20-24].

#### **2.2.3Contribution of an explanatory variable to the prediction of a property**

The contribution of an explanatory variable  $X_i$ noted  $C_{X_i}$  to the prediction of a property of *Y* is based on the statistical parameter t\_test which indicates the significance of an explanatory variable in a model [25].

$$
C_{X_i} = \frac{|t\_test(X_i)|}{\Sigma |(t\_test(X_i)|} \times 100
$$

The contribution is expressed as a percentage (%), where  $|t\_test(X_i)|$ , the absolute value of the t test of the variable  $X_i$ ;  $\Sigma | (t\_test(X_i))|$  the sum of the absolute values of the t\_test of all the variables  $X_i$  of the model. The higher  $C_{X_i}$  is, the greater the contribution of the explanatory variable  $X_i$  in the model developed [26].

## **3. RESULTS AND DISCUSSION**

In Tables 3 to 6, the values of the calculated quantum descriptors and the values of the redox potential E and the inhibitory concentration IC50 of the molecules of the training and test series are recorded.

# **3.1 Selection of Quantum Descriptors for the Prediction of the** *LogIC50*  **Inhibitory Concentration**

The results for the final selection of the predictive quantum descriptors for the *LogIC50* inhibitory concentration are reported in Tables 8 and 9.

The results in Table 8 allow us to consider two groups of predictive quantum descriptors of *LogIC50* for the AM1 level:

- **Group 3**: LUMO Energy (*ELumo*), HOMO Energy (*EHomo*) and Electron Donor Power  $(\bar{\omega}^-)$  :
- **Group 4:** Energy of the HOMO (*EHomo*), Electron Acceptance Power  $(\bar{\omega}^+)$  and Electron Donor Power ( $\overline{\omega}$ <sup>-</sup>).

The analysis of Table 9 reveals that the quantum descriptors selected for the prediction of *LogIC50 at* the HF/6-311G level (d, p) are: EHomo, *ELumo*, S, η, χ and μ**.** These quantum descriptors allow us to consider two groups:

- **Group 5**: HOMO energy **(***EHomo*)**,** LUMO energy **(***ELumo***),** Softness **(***S*), Hardness (η**)**  and Density **(μ)**;
- **Group 6:** LUMO energy **(***ELumo)***,** Softness **(S),** Hardness (η**),** Electronegativity (χ), and Dipole moment **(μ).**

Code		<b>Quantum descriptors</b>											
	EHomo	ELumo	Gap	X	η	S	ω	$\mu$	$\overline{\omega}^+$	$\bar{\omega}^-$	E	LogIC	
												50	
M <sub>1</sub>	$-0.33$	$-0.00$	0.33	0.17	0.17	2.94	0.09	3.34	0.02	0.02	$-0.030$	-4.98	
M <sub>2</sub>	-0 .33	$-0.02$	0.31	0.18	0.16	3.13	0.10	2.61	0.03	0.21	0.020	$-5.07$	
M3	$-0.33$	$-0.00$	0.33	0.17	0.17	2.94	0.09	3.12	0.02	0.02	0.030	$-4.68$	
M4	$-0.33$	$-0.02$	0.31	0.18	0.16	3.13	0.10	4.11	0.03	0.21	0.080	$-4.72$	
M <sub>5</sub>	$-0.33$	$-0.01$	0.32	0.17	0.16	3.13	0.09	0.57	0.03	0.20	0.105	$-4.64$	
M6	-0 .33	$-0.01$	0.32	0.17	0.16	3.13	0.09	1.21	0.03	0.20	0.280	$-4.86$	
M7	-0.32	$-0.01$	0.31	0.17	0.16	3.13	0.09	2.78	0.02	0.19	0.180	-4.03	
M8	$-0.33$	$-0.00$	0.33	0.17	0.17	2.94	0.09	3.30	0.02	0.02	0.185	$-4.23$	
M9	$-0.33$	$-0.00$	0.33	0.17	0.17	2.94	0.09	1.56	0.02	0.02	$-0.060$	$-5.20$	
M10	$-0.33$	$-0.01$	0.32	0.17	0.16	3.13	0.09	4.13	0.02	0.20	0.080	$-4.70$	
M11	$-0.33$	$-0.04$	0.29	0.19	0.15	3.33	0.12	3.03	0.04	0.23	0.180	$-4.58$	
M12	$-0.33$	$-0.03$	0.30	0.18	0.15	3.33	0.11	2.87	0.04	0.22	0.360	$\overline{\phantom{0}}$	
M <sub>13</sub>	$-0.34$	$-0.03$	0.31	0.19	0.16	3.13	0.11	2.35	0.04	0.22	0.500	$\overline{\phantom{0}}$	
M14	$-0.33$	$-0.02$	0.31	0.18	0.16	3.13	0.10	2.43	0.03	0.21	0.132	4.18	
M15	$-0.34$	$-0.02$	0.32	0.18	0.16	3.13	0.10	2.64	0.03	0.21	0.590	$\overline{\phantom{0}}$	
M16	-0.32	$-0.02$	0.30	0.17	0.15	3.33	0.10	1.98	0.03	0.20	0.500	$\overline{a}$	
M17	$-0.32$	$-0.02$	0.30	0.17	0.15	3.33	0.10	3.03	0.03	0.20	0.538		
M18	$-0.33$	$-0.03$	0.31	0.18	0.16	3.13	0.10	1.40	0.04	0.21	0.540		
M19	$-0.33$	$-0.00$	0.33	0.17	0.17	2.94	0.09	1.06	0.02	0.02	$-0.030$	$-4.53$	

**Table 3. Values of the quantum descriptors calculated at the AM1 level and the experimental values of the redox potential E and the inhibitory concentration IC50 of the training series**

Code		<b>Quantum descriptors</b>										<b>Experimental descriptors</b>	
	$E_{Homo}$	$E_{Lumo}$	Gap			S	ω		$\overline{\omega}^+$	$\overline{\omega}^-$	E	LogI $\mathsf{C}_{50}$	
M20	$-0.32$	-0 .04	0.28	0.18	0.14	3.57	0.12	1.06	0.04	0.22	$-0.035$	$-4.80$	
M21	$-0.32$	$-0.04$	0.28	0.18	0.14	3.57	0.12	2.14	0.04	0.22	$-0.020$	$-4.96$	
M22	$-0.32$	-0 .04	0.28	0.18	0.14	3.57	0.12	2.39	0.04	0.22	$-0.010$	$-4.89$	
M <sub>23</sub>	$-0.32$	.04 -0	0.28	0.18	0.14	3.57	0.12	0.49	0.04	0.22	0.040	$-4.89$	
M24	$-0.31$	$-0.03$	0.28	0.18	0.14	3.57	0.12	1.91	0.04	0.21	0.080	$-5.00$	
M25	$-0.32$	$-0.04$	0.28	0.18	0.14	3.57	0.12	2.20	0.04	0.22	0.082	$-4.60$	
M26	$-0.31$	$-0.03$	0.28	0.18	0.14	3.57	0.12	2.95	0.04	0.21	0.082	$-4.52$	
M27	$-0.32$	$-0.03$	0.29	0.18	0.15	3.33	0.11	1.62	0.04	0.34	0.092	$-4.42$	
M28	$-0.35$	$-0.03$	0.32	0.19	0.16	3.13	0.11	2.95	0.04	0.23	0.102	$-4.29$	
M29	$-0.32$	-0 .03	0.29	0.18	0.15	3.33	0.11	4.62	0.04	0 .21	0.450	$-4.01$	

**Table 4. Values of the quantum descriptors calculated at AM1 and the experimental values of the redox potential E and the inhibitory concentration IC50 of the test series**

**Table 5. Values of the quantum descriptors calculated at the HF/6-311G level (d, p) and the experimental values of the redox potential E and the inhibitory concentration IC50 of the training set**

Code		<b>Quantum descriptors</b>										<b>Experimental descriptors</b>
	$E_{Homo}$	$E_{Lumo}$	Gap		n	S	ω		$\overline{\omega}^+$	$\overline{\omega}^-$	E	$LogIC_{50}$
M <sub>1</sub>	$-0.29$	$-0.06$	0.23	0.18	0.12	4.17	0.14	2.06	0.06	0.24	$-0.030$	$-4.98$
M <sub>2</sub>	$-0.29$	$-0.06$	0.23	0.18	0.12	4.17	0.14	2.61	0.06	0.24	0.020	$-5.07$
M <sub>3</sub>	$-0.29$	$-0.06$	0.23	0.18	0.12	4.17	0.14	2.45	0.06	0.24	0.030	$-4.68$
M4	$-0.29$	$-0.06$	0.23	0.18	0.12	4.17	0.14	1.02	0.06	0.24	0.080	$-4.72$
M5	$-0.30$	$-0.08$	0.22	0.19	0.11	4.55	0.16	2.73	0.08	0.27	0.105	$-4.64$
M6	$-0.33$	$-0.08$	0.25	0.21	0.13	3.85	0.17	8.10	0.08	0.27	0.280	$-4.86$
M7	$-0.29$	$-0.09$	0.20	0.19	0.10	5.00	0.18	4.21	0.10	0.29	0.180	$-4.03$
M8	$-0.30$	$-0.08$	0.22	0.19	0.11	4.55	0.16	3.68	0.08	0.27	0.185	$-4.23$
M9	$-0.32$	$-0.08$	0.24	0.20	0.12	4.17	0.17	3.70	0.08	0.28	$-0.060$	$-5.20$
M <sub>10</sub>	$-0.30$	$-0.08$	0.22	0.19	0.11	4.55	0.16	7.85	0.06	0.27	0.080	$-4.70$
M11	$-0.31$	$-0.07$	0.24	0.19	0.12	4.17	0.15	4.39	0.07	0.26	0.180	$-4.58$
M <sub>12</sub>	$-0.31$	$-0.06$	0.25	0.19	0.13	3.85	0.14	3.77	0.06	0.25	0.360	٠
M <sub>13</sub>	$-0.31$	$-0.07$	0.24	0.19	0.12	4.17	0.15	3.35	0.07	0.26	0.500	۰.
M14	$-0.31$	$-0.09$	0.22	0.20	0.11	4.55	0.18	3.02	0.10	0.30	0.132	4.18

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Code		<b>Quantum descriptors</b>			<b>Experimental descriptors</b>							
	$E_{Homo}$	$E_{Lumo}$	Gap				ω		$\bar{\omega}$	$\bar{\omega}^-$		LogI $\mathsf{C}_{50}$
M15	$-0.32$	$-0.09$	0.23	0.22	0.12	4.17	0.20	3.91	0.09	0.30	0.590	
M16	$-0.30$	$-0.09$	0.21	0.20	0.11	4.55	0.18	2.35	0.10	0.29	0.500	
M17	$-0.29$	$-0.08$	0.21	0.19	0.11	4.55	0.16	4.30	0.08	0.27	0.538	
M18	$-0.30$	$-0.09$	0.21	0.20	0.11	4.55	0.14	2.83	0.10	0.24	0.540	
M19	$-0.34$	$-0.08$	0.26	0.21	0.13	3.85	0.17	10.45	0.08	0.29	$-0.030$	$-4.53$

**Table 6. Values of the quantum descriptors calculated at the HF/6-311G level (d, p) and the experimental values of the redox potential E and the inhibitory concentration IC50 of the test series**







**Table 8. Selection of descriptors at the AM1 level of approximation**



# **3.2 Selection of Quantum Descriptors for the Prediction of Redox Potential**

The results for the final selection of the predictive quantum descriptors of the redox potential E are given in Tables 8 and 9.

The analysis of the results allows us to consider two groups of predictive quantum descriptors for the AM level1 :

- **Group 1**: LUMO energy (*ELumo*), HOMO energy (*EHomo*) and electron donor power  $(\overline{\omega}^-)$ ;
- **Group 2:** Energy of the HOMO (*EHomo*), electron acceptor  $(\bar{\omega}^+)$  and electron donor power  $(\overline{\omega}^-)$ .

For the HF/6-311G (d, p) method, no descriptor was retained, therefore there is no predictive model for redox potential at this level of calculation.

## **3.3 QSPR Model of the Predictive Quantum Descriptors of the Redox Potential** *E* **of the Inhibitory Concentration** *LogIC50*

Based on **the** learning **set** and the selected predictive descriptors, the aim were to :

- Establish one or more QSPR model(s) for predicting the redox potential *E* and the inhibitory concentration *LogIC50* per calculation level.
- to carry out an analysis of the statistical parameters of the QSPR models developed.

The results of this work allowed the validation of the best models for predicting the redox potential

E and the *LogIC50* inhibitory concentration of flavonoids.

#### **3.3.1QSPR model of predictive quantum descriptors of redox potential by the AM1 method**

In order to select the group to be used for the regression equation of the QSPR model at the AM1 level of calculation, the Fisher coefficients of the two groups 1 and 2 compared and the most significant group in the Fisher sense selected.

Analysis of the results in Tables 10 and 11 shows that the Fisher coefficient  $(F_1)$  for Group 1 was higher than the Fisher coefficient (F2) for Group 2 :  $F_2 \nless F_1$ ; this means that the regression equation for Group 1 will be more significant than that for Group 2. Therefore, the quantum descriptors of group 1 can be preferred to establish the QSPR model of the AM1 level of redox potential.

The ANOVA table for Model 1, which was used to perform the analysis of variance, is shown in Table 10. This ANOVA table indicates that the pvalue (0.0005E-6) is less than  $α=0.05$ , showing that the regression equation of model 1 is significant in predicting redox potential.

The results of the multilinear regression obtained from the descriptors of group 1 are shown in Table 11.

The regression equation for model 1 is as follows :

$$
E = -0.29 + 0.22E_{\text{Homo}} + 0.11E_{\text{Lumo}} - 0.05\overline{\omega}^{-1}
$$

#### **3.3.2Contribution of the AM1 level quantum descriptors in the prediction of the redox potential E**

According to the absolute values of the t-test in Table 12, the importance of the quantum descriptors of the AM1 level in Model 1 is in the following order

$$
\overline{\omega}^- < E_{\text{Homo}} < E_{\text{Lumo}}
$$

Indeed, the contribution calculations show that the Lumo Energy *(ELumo)* makes a contribution of 48.35P in predicting the redox potential, the electron donor power  $(\bar{\omega}^{-})$  and the HOMO Energy (*EHomo*) make a contribution of 25.48 and 26.18P respectively. It is clear that the LUMO energy (*ELumo*) is the main predictive descriptor of the redox potential.

**Table 9. ANOVA table of the quantum descriptors of group 1 of the AM1 level**

	DS	SC	<b>MSC</b>	F <sub>1</sub>	P-value
Regression		0.58	0.19	10.68	$0.0005E-6$
<b>Residue</b>	15	0.27	0.01		
Total	18	0.86			



	DS	SC	<b>MSC</b>	F2	P-value
<b>Regression</b>		0.10	0.03	0.70	0.5634E-5
<b>Residue</b>	15	0.75	0.05		
<b>Total</b>	18	0.86			

**Table 11. Values of the regression coefficients of group 1 for model 1**



#### **3.3.2QSPR model of the predictive quantum descriptors of the** *LogIC50* **inhibitory concentration at the AM1 level: model 2**

The ANOVA tables (Tables 12 and 13**)** show that the Fisher coefficient (F4) of group 3 is higher than the Fisher coefficient (F5) of group 4: **F5 < F4 ;** therefore, the quantum descriptors of group 4 can be preferred to establish the QSPR model of the AM1 level for the prediction of the inhibitory concentration.

The results of the multilinear regression obtained from the descriptors of group 4 are shown in Table 14.

The regression equation for model 2 is as follows:

$$
LogIC_{50} = -4.92 + 11.37 E_{Homo} + 34.36 E_{Lumo} + 0.67\overline{\omega}^{-1}
$$

The ANOVA table for the model (Table 15) indicates that the p-value (0.00021E-7) is less than  $α = 0.05$ . Thus the regression equation of model 2 is significant in predicting the inhibitory concentration.

## **3.3.3Contribution of AM1 quantum descriptors to the prediction of the inhibitory concentration**

According to the absolute values of the t-test in Table 15, the importance of the quantum descriptors of the AM1 level in Model 2 was in the following order

 $\overline{\omega}$ <sup>-</sup>< $E_{\text{Homo}}$   $E_{\text{Lumo}}$ 

The contribution calculations show that the Lumo Energy ( $E_{Lumo}$ ) makes a contribution of 60.66 $\square$  in predicting the redox potential, the electron donor power  $(\bar{\omega}^{-})$  and the HOMO Energy (*EHomo*) make a contribution of  $37.71$  and  $2.25$ respectively. It is clear that the LUMO Energy (*ELumo*) is the main descriptor predicting the inhibitory concentration.

## **3.3.4QSPR model of the predictive quantum descriptors of the** *LogIC50* **inhibitory concentration at the HF/6-311G level (d, p): model 3**

The Fisher coefficients for groups 5 and 6 are provided by the ANOVA tables in Tables 16 and 17. From the analysis of the results, the Fisher coefficient (F11) of group 5 is higher than the Fisher coefficients (F12) of group 6**;** its regression equation will be more significant than that of group 6.

The ANOVA table for Model 3 indicates that the  $p$ -value ( $p$ -value =  $0.026E-8$ ) is smaller than  $\alpha$ =0.05. This shows that the regression equation of Model 3 is significant in predicting the inhibitory concentration of the molecules.

The results of the multilinear regression are shown in Table 17.







	DS	SC	<b>MSC</b>	F5	P-value
<b>Regression</b>		3.65	7.82	1.16	0.00356E-4
<b>Residue</b>	15	4.06	8.20		
Total	18	5.71			

**Table 14. Values of the regression coefficients of model 2 at AM1 level**



The regression equation for model 6 is as follows:

 $LogIC_{50} = 62.40 + 80.25 E_{Homo} - 28.44 E_{Iumo}$ + 52.01 *S* -71.26 *η* – 6.11*μ*

#### **3.4 Statistical Parameters of Model 1 of Models 1, 2 and 3**

The results of the statistical parameters are reported in Table 19.

The results show that :

- The redox potential is strongly correlated with the quantum descriptors of the AM1 level as  $R = 0.9820$ . In addition  $96.43P$  of the experimental variance of the redox potential is explained by the descriptors of model 1. It can be said that model 1 is validated and can be retained as a model for predicting the redox potential of the studied molecules.
- The *LogIC50* inhibitory concentration is correlated with the quantum descriptors at the AM1 and HF/6-311G levels (d, p). Indeed,  $96.35P$  and  $99.96P$  of the

experimental variance of the inhibitory concentration are explained by the descriptors of model 2 and model 3 respectively. It can be said that models 2 and 3 are validated and can be retained as a predictive model for the inhibitor concentration.

## **3.5 Internal LOO Validation of Models 1, 2 and 3**

The results are reported in Table 20. They indicate that :

- Model 1 has a very high predictive capacity las 94.9P of the molecules in the training set have their redox potential predicted.
- Model 3 has a very high predictive ability, 98.9P of the molecules in the training set have their predicted inhibitory concentration.
- The model has a very high predictive capacity ( $Q_{LOO}^2$  = 0,941) because 94.1P of the molecules in the training set have their predicted inhibitory concentration



**Table 15. ANOVA table of the quantum descriptors of group 5 of level HF/6-311G (d, p)**









#### **Table 18. Statistical parameters for the external validation of models 1, 2 and 3**



## **3.6 External Validation of Models 1, 2 and 3**

The results are reported in Table 20 and show that :

- Model 1, has high predictive power (Q2ext  $= 0.891$ ) as 89, 1.P of the molecules in the test series have their redox potential predicted. In addition 93, 50% of the experimental variance in redox potential is explained by the quantum descriptors of model1 at the AM1 level.
- Model 2, has a high predictive power. Indeed, 97.80P of the molecules in the test series have their redox potential predicted. In addition 98.30% of the experimental variance of the inhibitory concentration is explained by the quantum descriptors of model 2 at the AM1 level.
- Model 3 has high predictive power as 96.10P of the molecules in the test series have their redox potential predicted. Also, 97.70P of the experimental variance in the percentage of inhibition is explained by the quantum descriptors of model 3 at the HF/6-311G level (d, p).

# **3.7 Verification of Tropsha Criteria for Models 1, 2 and 3**

The analysis of the results is recorded in Table 21 and shows that :

- For model 1 only criteria 1, 2 and 4 are verified; i.e. the of Tropsha's criteria. The model is therefore efficient in predicting the redox potential
- For model 2, only criteria 1, 2 and 4 are verified while for model 3, all 5 criteria are verified. Models 2 and 3 therefore perform very well in predicting the inhibitory concentration.

#### **3.8 Normality Tests of Models**

## **3.8.1 Normality tests for model 1**

**Shapiro-Wilk test (***Epréd***)**

This test gives the following results:  $w = 0.390$ ; pvalue = 0.879; α =0.05

Model 1	<b>Internal</b>	n	<b>PRESS</b>	$\bm{Q}_{\bm{Loo}}^2$	$S_{Press}$	
		19	0.197	0.949	0.170	
	<b>Extern</b>	n	$R^2_{ext}$	<b>PRESS</b>	$Q2$ ext	<b>S</b> <sub>PRESS</sub>
		10	0.935	0.149	0.891	0.044
Model 2	<b>Internal</b>	$\mathbf n$	<b>PRESS</b>	$Q_{Loo}^2$	$S_{Press}$	
		19	0.221	0.989	0.134	
	<b>Extern</b>	n	$R^2_{ext}$	<b>PRESS</b>	$Q2$ ext	$S_{PRESS}$
		10	0.983	0.149	0.978	0.116
Model 3	<b>Internal</b>	$\mathbf n$	<b>PRESS</b>	$Q_{Loo}^2$	$S_{Press}$	
		19	0.256	0.941	0.322	
	<b>Extern</b>	n	$R^2_{ext}$	<b>PRESS</b>	$Q2$ ext	<b>S</b> <sub>PRESS</sub>
		10	0.977	0.198	0.961	0.143

**Table 19. Statistical parameters for internal and external validation of models 1, 2 and 3**





**Interpretation of the test**: Since the calculated p-value is above the alpha threshold significance level  $(0.879 > 0.05)$ , it is concluded that the predicted values of the redox potential by model 1 follow a normal distribution.

#### **Durbin-Watson test (residuals) :**

This test gives the following results: U= 0.785; pvalue = 0.4860; α = 0.05

**Interpretation of the test**: Since the calculated p-value is above the alpha significance level  $(0.4860 > 0.05)$ , it is concluded that the residuals are not autocorrelated. Therefore, they do not contain any information that could influence the prediction of model 1.

## **3.8.2 Normality tests of model 2**

#### **Shapiro-Wilk test**

This test gives the following results:  $w = 0.239$ ; pvalue = 0.067; α =0.05

**Interpretation of the test**: Since the calculated p-value is above the alpha threshold significance level  $(0.067 > 0.05)$ , it is concluded that the predicted values of the inhibitory concentration by model 2 follow a normal distribution.

#### **Durbin-Watson test:**

This test gives the following results: U=0.463; pvalue = 0.1137; α =0.05

**Interpretation of the test**: Since the calculated p-value is above the alpha significance level, the residuals are not autocorrelated. They do not contain any information that could influence the prediction of model 2.

#### **3.8.3 Normality tests of model 3**

The results are as follows:

#### **Shapiro-Wilk test**

This test gives the following results: w=0.105; pvalue = 0.077; α =0.05

**Interpretation of the test**: Since the calculated p-value is above the alpha threshold significance level  $(0.077 > 0.05)$ , it is concluded that the predicted values of their inhibitory concentration by model 3 follow a normal distribution.

#### **Durbin-Watson test:**

This test gives the following results: U=0.993; pvalue = 0.1232; α =0.05

**Interpretation of the test**: Since the calculated p-value is above the alpha significance level, the residuals do not contain information that could influence the prediction of model 3.

## **3.8.4 Predicted model equations**

From the various statistical tests in Table 18 we can deduce that the equations of the models are as follows:

**Model 1**: Prediction of the redox potential, which is summarized as follows:

 $E$ = -0.29 + 0.22 $E$ <sub>Homo</sub> + 0.11 $E$ Lumo – 0.05 $\bar{\omega}$ <sup>-</sup>

*n***=19; R=0.9820 ; <b>R**<sup>2</sup> =0.643 ; **R**<sub>2</sub><sup>2</sup> = 0.9582 ; **S** = 0.0755 ; *F***=**4.3492 ; *FIT***=**0.230

**Model 2**: Prediction of the inhibitory concentration is summarised as follows at the AM1 level:

$$
LogIC_{50} = -4.92 + 11.37E_{Homo} + 34.36 E_{Lumo} + 0.67\overline{\omega}^{-1}
$$

 $n=19;$  **R** = 0.9816;  $R^2$  = 0.9635;  $R_{aj}^2$  = 0.8575; *S***=**0.0850; *F***=**4.3492 ; *FIT* **=** 0.0230

**Model 3:** The prediction regression equation is summarised as follows:

 $LogIC_{50} = 62.40 + 80.25 E_{Homo} - 28.44 E_{Iumo}$ + 52.01 *S* -712.6 *η* – 6.11*μ*

*n*=19; *R* =0.998;  $R^2$  =0.998;  $R_{aj}^2$  = 0.996; S= 0.24936; *F***=**1.4614; *FIT***=**0.08

## **3.9 Predicted Values of Redox Potential and Inhibitory Concentration of 29 Flavonoids by Models 1, 2, 3**

Table 21 shows the predicted values of redox potential and percentage inhibition of the 29 flavonoids by models 1, 2 and 3.

These results show that there is good agreement between the model values and the experimental values published in the literature.



#### **Table 21. Experimental and predicted values of redox potential and inhibitory concentration of the 29 flavonoids by the models**

## **4. CONCLUSION**

A prediction study of the redox potential *E* and the inhibitory concentration *LogIC50* was performed, using the semi-empirical methods AM1 and HF/6-311G (d, p).

- $\checkmark$  The application of the descriptor selection criteria made it possible to determine and retain 6 groups of quantum descriptors, including 2 groups of descriptors for the prediction of the redox potential *E* and 4 groups of descriptors for the prediction of the inhibitory concentration *LogIC50*. The antioxidant properties of the molecules depend strongly on these groups of descriptors.
- $\checkmark$  From the multilinear regression analysis, several prediction models (one model for redox potential and two for inhibitory concentration) were established from the quantum descriptors. The established

models are validated and perform well according to Tropsha criteria.

The development of these QSPR models represents a significant advance in the prediction of antioxidant properties of bioactive molecules such as flavonoids based on descriptors calculated by quantum chemical methods. This is a contribution to the database of the two main parameters (E and IC50) involved in the prediction of antioxidant properties of bioactive molecules.

# **COMPETING INTERESTS**

Authors have declared that no competing interests exist.

# **REFERENCES**

1. Hendrickson HP, Kaufman AD, Lunte CE. J. Pharm. Biomed Anal. 1994;12:325.

- 2. Jovanovic SV, Steenken S, Simic MG,<br>Hara Y. Antioxidant properties of Antioxidant properties of flavonoids: Reduction potentials and electron, transfer reductions of flavonoid radicals. In: Rice-Evans, C. A.; Packers, L., eds. Flavonoids in health and disease. New York: Marcel Dekker, Inc. 1996;137- 161.
- 3. Labuda J, Buková M, Heilerová L, Šilhár S, Štepánek I. Evaluation of the redox properties and anti/pro-oxidant effects of selected flavonoids by means of a DNAbased electrochemical biosensor. Anal Bioanal Chem. 2003;376:168-173.
- 4. José T, Alexandra G, Manuela E, Garrido JG, Fernanda B. Hydroxycinnamic acid antioxidants: An electrochemical overview, hindawi publishing corporation bio med research international. Article ID 251754. 2013;11.
- 5. Lien EJ, Ren S, Bui HH, Wang R. Free Radic. Biol. Med. 1999;26:285.
- 6. Bin Y, Akira K, Kensunke A, Fumiyo K. Estimation of the antioxidant activities of flavonoids from their oxidation potentials. Analytical sciences; the Japan society for analytical chemistry. 2001;1.
- 7. Pearson RG. Hard and soft acids and bases, J. Am. Chem. Soc. 85, 3533-3539, (1963),(2001).
- 8. R. G. Parr, R. A. Donnelly, M. Levy, W. E. Palke. J. Chem. Phys. 1978;68:3801-3807.
- 9. Geerlings P, De Proft F, Langenaeker W. Chem. Rev, 2003 ;103 :1793-1874.
- 10. Caro A, Zagal JH, Bedioui F, Adamo C, Cardenas-Jiron GI. J. Phys. Chem. A. 2004;108:6045-6051.
- 11. De Vleeschouwer F, Jaque P, Geerlings P, Toro-Labbe A, De Proft F. J. Org Chem. 2010;75:4964-4974.
- 12. Morrel H Cohen, Adam W. Journal of Statistical Physics, On Hardness and Electronegativity Equalization in Chemical Reactivity Theory; 2006.
- 13. Bonin KD. Kresin electric-dipole polarizabilities of atoms molecules and clusters. World Scientific, Singapore, VV; 1997.
- 14. Payán G, Sergio A, Norma Flores H, Antonino P, Manuel Piñón M, Daniel G. Computational molecular characterization of the flavonoid rutin. Chemistry Central Journal. 2010;4:12.
- 15. Laffly. Multiple regression: principles and application examples; 2006.
- 16. Golbraikh A. Tropsha, beware of q2 J. Mol. Graph. Model. 2002;20:269-276.
- 17. Dragan A, Duanka D, Drago B, Vesna R, Bono L, Nenad T. SAR and QSAR of the antioxidant activity of flavonoids. Current Medicinal Chemistry. 2007;14:827- 845.
- 18. Durbin J, Watson GS. Testing for serial correlation in least squares regression II. Biometrika. 1951;38(1-2):159-179.
- 19. Tetko V, Sushko I, Pandey AK, Zhu H, Tropsha A, Papa E, Oberg T, Todeschini R, Fourches D, Varnek A. J. Chem. Inf. Model. 2008;48:1733.
- 20. Jovanovic SV, Steenken S, Hara Y, Simic MG. Reduction potentials of flavonoid and model phenoxyl radicals which ring in flavonoids is responsible for antioxidant New York: Marcel Dekker, Inc. 1998;137- 161.
- 21. Jaworska J, Jeliazkova NN, Aldenberg T. Altern. Lab. Anim. 2005;33:445-459.
- 22. Jorgensen LV, Cornett C, Justesen U, Skibsted LH, Dragsted LO. Free Rad. Res. 1998;29:339.
- 23. Medjdoub GA. Contribution to the study of chemical reactivity using conceptual DFT, application to heterocycle chemistry; 2012.
- 24. Chattaraj PK, Perez P, Zevallos J, Toro-Labbe A. J. Phys. Chem. A. 2006 ;104 :4272-4283.
- 25. Rubalya V, S, Neelamegam P. Selective ABTS and DPPH- radical scavenging activity of peroxide from vegetable oils , International Food Research Journal. Journal homepage. 2015;22(1):289-294.
- 26. Shapiro SS, Wilk MB. An analysis of variance test for normality (Complete Samples). Biometrika. 1965;52(3 and 4):591-611.

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