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# Quantum Chemical QSPR Study of The Best Parameters Influences on Intrinsic Viscosity of Polyisoprene Solution

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#### Abstract

Quantitative Structure-Property Relationship (QSPR) analysis for intrinsic viscosity [n] of Polyisoprene solution have been conducted. The study was done by using molecular modeling. The calculation was performed by the (PBE)exchangecorrelation functional method for (Perdew, Burke, and Ernzerhof) at 6-31G(d) basis set. The relationship analysis between intrinsic viscosity [n] of polyisoprene solution and physicochemical properties of four solvents [Benzene, Toluene, Hexane, and Isooctane] under study was done by multiple linear regression (MLR) analysis to produce the equation that relates the structural features to the intrinsic viscosity [n] models with one and two parameters linear properties. The results show good equations. The best model predicted in this study was the eq.3, with excellent statistical fit as evident from its R2=0.99, F=218144 and S=0.129, the model including the descriptors [LUMOs and HOMOs], which showed insignificant role in the intrinsic viscosity [n] of polyisoprene solution. And this could potentially offer a new opportunity in the design of novel properties of polymers or extended to other polymer composite.

Keywords: Intrinsic Viscosity, QSPR model.

### Introduction

Experimental measurements of thermodynamic parameters some involve experimental difficulties and they are not always feasible, and the corresponding methods possess real drawbacks <sup>(1)</sup>. Intrinsic viscosity  $[\eta]$  is extensively used for analysis characterization of synthetic polymers macromolecules (5,6) biological colloids<sup>(7)</sup>. nanoparticles, and Consequently, it is necessary to resort to theoretical calculation of these parameters. This option accessible because an important, fruitful

and current field of research in contemporary chemistry is the model and prediction of physical-chemistry properties of molecules<sup>(8)</sup>.

Quantitative structure–property/ activity relationships (QSPR/QSAR) are tools to estimate physico-chemical and biochemical parameters and reduce the coast, time and efforts <sup>(9,12)</sup>. (QSPR/QSAR) study is an important section in computational chemistry and uses frequently for predicting physico-chemical and biological activity of organic compounds.

To establish the relation between the structural characteristics of molecule and its properties the mathematical methods can be used. The basic strategy of QSPR employed to find the optimum quantitative relationship, that can then be used for the predication of the properties of molecular structures including those unmeasured or even unknown Antreaset al, have been investigated QSPR model for prediction of intrinsic viscosity in polymer solution by using the multiple liner regression technique on a database that consists of 65 polymersolvent combination involving

# Geometry Optimization & Experimental

Full geometric optimization calculations for four solvents(benzene, toluene, hexane and isooctane) were performed using PC Gamess (20).

### **Results and Discussion**

The relationship between intrinsic viscosity of polyisoprene solution and various descriptors of organic solvents (Physiochemical and alignmentindependent) were established sequential multiple regression analysis (MLR) in order to obtain QSPR models. The best multilinear regression (BMLR) procedure [13,14] was used to find the best correlation models from the selected noncollinear descriptors. The Table which descriptors 1. significant for experimental data, were

different polymer (16). In this research the QSPR model was produced by using the linear regression multiple (MLR) technique on a database that consists of one polymer in four organic solvent for the prediction of intrinsic viscosity of polyisoprene solution, and focus of some of the descriptors in deriving predictive QSPR models, (17,19). The descriptors used in the composition of models represents the solvent properties are two types theoretically such as HOMOs, LUMOs and Total Energy (T.Es), and practically such as the viscosity of the solvent and the dielectric constant (E) of the solvent.

Geometry optimizations were performed using 6-31G (d) basis set and PBE method <sup>(21)</sup>. Physical properties calculated involve total energy, LUMO<sub>s</sub> energy and HOMO<sub>s</sub> energy.

selected by QSPR-contingency module. To establish the statistical correlation, the physicochemical parameters were taken as independent variables and intrinsic viscosity as dependent variable. The best model was selected on the basis of statistical parameters viz observed with high correlation coefficient (R), sequential Fischer test (F), and low standard error of estimate (S), were employed to judge the validity of regression equation and evaluate the obtained OSPR models (22,25).

Table 1.Physic-chemical descriptors of the solvents.

NO	Solvent	<sup>1*</sup> Exp, [η]	$T.E_s$	$LUMO_s$	$HOMO_s$	<sup>2*</sup> η	<sup>3*</sup> Es
1	benzene	186	-231.964	0.009	-0.2574	0.649	2.283
2	toluene	184	-271.233	0.0093	-0.2457	0.632	2.385
3	hexane	91	-276.044	0.1037	-0.309	0.313	1.89
4	isooctane	110	-315.306	0.0856	-0.3037	0.502	1.94

 $<sup>^{1*}</sup>$ Exp[ $\eta$ ]=Ref=16,  $^{2*}\eta$ & $^{3*}$ Es= Ref = 26

Definition of Descriptors Used in This Study.

 $T.E_s=Totol$ Energy in hartree solvent, HOMO<sub>s</sub>=The energy of Highest Occupied Molecular Orbital in hartree, LUMO<sub>s</sub>=The energy Lowest of Unoccupied Molecular Orbital in hartree, Es=Dielectric constsnt of solvent.  $^{1*}$ Exp[ $\eta$ ]=intrinsic viscosityof polyisoprene solution.<sup>2\*</sup> $\eta$ =Viscosity of solvent

The model of QSPR study has been build up with help of the descriptors T.E<sub>s</sub>, LUMO<sub>s</sub>, HOMO<sub>s</sub>, Viscosity and Eswas investigated. The predictive model of QSPR study has been built up with the help of the following

descriptors in Table 1. The best model derived from the (MLR) analysis was used to intrinsic viscosity polymer in the 4 organic solvents which using in the study [Benzene, Toluene, Hexane, and Isooctane]. Several equations were generated by using all the variables and the best statistically model that have obtained is one-two descriptor equation, table. 2 and 3, when depend on only one descriptor such as  $HOMO_s$ ,  $LUMO_s$ ,  $\eta_s$  and  $\epsilon_s$ . From table 2, The correlation between  $[\eta]$  and  $LUMO_s$  gave good correlation coefficient  $R^2$  values and low standerd error.

Table 2. Statistical parameters of the linear regressions models obtained by using one descriptor.

descriptor	$\mathbb{R}^2$	F	S
T.E,	0.450	1.640	44.8513
η,	0.873	13.779	21.542
٤s	0.960	48.548	12.0359
HOMO <sub>2</sub>	0.964	55.008	11.333
LUMO,	0.999	7125.924	1.0135

the best second model when depend on only two descriptor Table 3. That gave good model with high correlation coefficient  $R^2$  values.

Table 3. Statistical parameters of the linear regressions models obtained by using two descriptor.

descriptor	R2	F	S
$T.E_s \& \eta_S$	0.960	12.019	17.101
$\varepsilon_s$ &T. $E_s$	0.968	15.137	15.301
$HOMO_s\&T.E_s$	0.968	15.484	15.134
$HOMO_s\&\epsilon_S$	0.970	16.578	14.641
$\epsilon_{ m s} \& \eta_{ m s}$	0.981	26.342	11.679
$HOMO_s \& \eta_s$	0.997	203.799	3.482
$LUMO_s \& \eta_S$	0.999	2992.857	1.105
$T.E_s\&LUMO_s$	0.999	1803.275	1.424
$LUMO_s\&\epsilon_s$	0.999	37786	0.311
HOMO <sub>s</sub> & LUMO <sub>s</sub>	0.999	218144	0.129

The best model was chosen, whose specifications are presented in Table 3, we have chosen three equations as the excellent model which it has less standard error (SE) and high R<sup>2</sup> and F values. The resulting parametric models are depicted in Eq. 1 to 3, and figures

1 to 3. The model when depend on descriptors [LUMO<sub>s</sub> and T.E<sub>s</sub>], these gave good model with correlation coefficient  $R^2$  values for this model of 0.99, as equation 1.

# $[\eta]=195.013(+/-)522.009-988.039(+/-)1413.101LUMO_s+3.595(+/-)2.069T.E_s$ ----- Eq.1

From eq 1, the positive sign of  $T.E_s$  descriptor refers to a positive correlation with the intrinsic viscosity. In this model LUMO<sub>s</sub> have negative sign which suggest that intrinsic

viscosity decreases with increasing value of this parameter. The relationship between the experimental and predicted data shows in Fig 1.

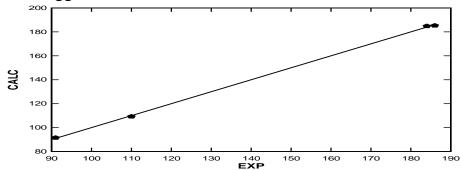


Figure 1.Plot of intrinsic viscosity prediction vsexperimental intrinsic viscosity using eq.1.

While when replace the descriptor  $T.E_s$  in eq 1, by the descriptor Es this lead to generated the eq 2. That gave model

with correlation coefficient R<sup>2</sup> values for this model of 0.99, with low less standard error (SE) comparable with eq.1.

 $[\eta] = 236.741 - 1076.715(+/-)1253.646LUMOs - 17.938(+/-)254.040Es...$ Eq.2

In eq 2, the negative sign of these descriptors refers to inverse correlation between LUMOs and Es with the

intrinsic viscosity. The relationship between the experimental and predicted data, Fig (2).

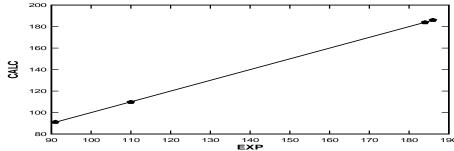


Figure 2.Plot of intrinsic viscosity prediction vs experimental intrinsic viscosity using eq2.

In equation 3.These build up by using the descriptors [LUMO $_{s}$  and HOMO $_{s}$ ]. This model generated by

replace the descriptor Es in eq 2, by descriptor HOMO<sub>s</sub> gave the best model

with correlation coefficient R<sup>2</sup> values for this model of 0.99, and very low standard error (SE) 0.129.

 $[\eta] = 157.104(+/-)213.910-150(+/-)868.827HOMO_s-1084.892(+/-)LUMO_s ----Eq.3$ 

Also in eq 3, the negative sign of HOMO<sub>s</sub> and LUMO<sub>s</sub> descriptors refers to inverse correlation between the intrinsic viscosity and these parameters.

The relationship between the experimental and predicted data was showed in Fig 3.

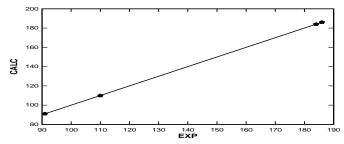


Figure 3.Plot of intrinsic viscosity prediction vs experimental intrinsic viscosity using eq 3.

The predicted intrinsic viscosity of polyisoprene and experimental intrinsic viscosit can be seen in table 4.

Table 4.Experimental and predicated of intrinsic viscosity of Polyisoprene by using Eqs 1-3.

	Exp	Calc	Calc	Calc
Solvent	[η]	eq.1	eq.2	eq.3
benzene	186	185.28	186.09	186.05
toluene	184	184.84	183.94	183.96
hexane	91	91.56	91.18	91.07
isooctane	110	109.3	109.77	109.91

## **Conclusion**

The linear model and multiple regressions were performed between intrinsic viscosity [η] of polyisoprene chemical solution and some parameters/descriptors organic of The study indicated that solvents. intrinsic viscosity for polymer can be modeled. The values of R<sup>2</sup>, S and F suggest that the QSPR models in Eqs from1 to 3 are predicted and validated. From all the results the Eq 3.have small

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value of S and large the values of F and R<sup>2</sup> that regard as better in QSPR model. The experimental and the predicted values using MLR model was excellent, and this model including [LUMOs and HOMO<sub>s</sub>] showed significant role in the intrinsic viscosity [n] of polyisoprene solution. This study may be helpful for chemists and researcher understanding parameters intrinsic viscosity of polyisoprene solution understudy.

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# دراسة نظرية وكمية بتقنيات QSPR لأفضل تأثير للمتغيرات على اللزوجة الجوهرية [η] لمحلول البولى ايزوبرين

صادق محمد حسن اسماعيل كوكب علي حسين حوراء كريم ضايف جامعة البصرة \_كلية التربية للعلوم الصرفة\_قسم الكيمياء.

#### الخلاصة

وصفت اللزوجة الجوهرية لمحلول البولي ازوبرين باستخدام تقنيات العلاقة التركيبية الخصائصية الكمية QSPR . وحد الحسابات بطريقة دالة العلاقة الاستبدالية Perdew, Burke, and Ernzerhof) وعند المستوى النظري (Perdew, Burke, and Ernzerhof) وعند المستوى النظري (Perdew, Burke, and Ernzerhof) [Benzene, Toluene, Hexane, and الجوهرية [ $\eta$ ] لمحلول البوليمر والخواص الفيزوكيميائية للمذيبات MLR لتكوين معادلة تربط مميزات التركيب Isooctane تحت الدراسة بأستخدام تحليل الخطي المتعدد MLR لتكوين معادلة تربط مميزات التركيب بخواص اللزوجة الجوهرية [ $\eta$ ]. أظهرت النتائج موديلات جيدة ذات علاقة خطية مع اثنين من المتغيرات. وكان افضل موديل للتنبأ في هذه الدراسة معادلة رقم 3، مع وسائط احصائية S والتي اظهرت دور مميز في التأثير على اللزوجة الجوهرية لمحلول البولي أيزوبرين. وهذه الدراسات ممكن أن تقدم فرص حديثة لتصميم خواص جديدة للبوليمرات او التوسع لمكونات بوليمرية اخرى.