# A molecular structure descriptor derived from bond-disconnection: Application to quantitative structure property relationships

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A set of novel molecular structure descriptors, viz., vertex weighted walk parameter, has been derived by using a bonddisconnection protocol in hydrogen depleted vertex weighted molecular graph. The parameters are correlated with other topological indices by constructing the correlation matrix and by correlating with the principal components of these indices. The applicability of these parameters in quantitative structure property relationships has been investigated for boiling points, molar volume at 20 ºC, molar refractions at 20 ºC, heats of vaporization at 25 ºC, critical temperature and surface tension at 20 °C. Both the single parametric model and multiparametric model have been used for the QSPR studies. Successive exclusion of variables technique has been used to optimize the multiparametric models. The high collinearity of the observed and predicted properties reveals the applicability of the proposed parameters.

**Keywords:** Theoretical chemistry, Graph theory, Molecular descriptors, Topological indices, Connectivity parameters, Vertex weighted walk parameter, Structure-activity relationships

With the advent of information technology, chemical sciences is now replete with data pools for which, codification, classification and proper utilization of the data necessitate an upward growth. With the proliferated data set, selection of data for optimization has become a potential problem and suffers an enhanced risk of chance correlation. Rucker *et al.*<sup>1</sup> have used a randomization technique for validation of quantitative structure activity/property relationship (QSA/PR) models. Recently, JAVA language has been used to construct a program (OSIRIS) for a complete processing of drug discovery<sup>2</sup> which covers all information from compound synthesis via biological testing, secondary screening evaluation, chemistry-aware data visualization, physicochemical property prediction, three-dimensional pharmacophore comparisons, interactive modeling and computing grid based ligand-protein docking. Tripos Topomer Search Technology (TTST)<sup>3</sup>, a novel technique used as a measure of compound similarity, is found to be more promising than the classical QSAR<sup>4</sup> for pharmacophore modeling. Recently, Bender *et al.*<sup>5</sup> employed a diverse subset of the MDDR (Molecular Database Library Drug Data Report) database to present a large scale investigation

for virtual screening studies (where orthogonal descriptors are usually chosen for diverse virtual screening hit lists) and to understand descriptor behaviour.

 A new Monte Carlo variable selection (MCVS) method was proposed by Konovalov *et al.*<sup>6</sup> and was applied to the blood-brain barrier and human intestinal absorption problems using more than 1600 electronic remote versions of DRAGON molecular descriptors. Only a single descriptor TPSA (NO) (Topological Polar Surface Area using N and O polar Contribution) and ALOGP (atom-type summation logarithm of partition coefficient) could be interpreted as a casual biochemical QSAR relationship for the BBB (Blood Brain Barrier) and HIA (Human Intestinal Absorption) problems respectively yielding low P-values. The MCVS method is equally applicable to the multiple-linear-regression (MLR) based or non-MLR-based QSAR models.

Wester *et al.*<sup>7</sup> have enumerated graph representations of scaffold topologies for up to eight ring molecules and four-valence atoms, thus providing coverage of the lower portion of the chemical space of small molecules. They have examined scaffold topology distributions for several databases, viz.,

ChemNavigator $8$  and PubChem<sup>9</sup>, for commercially available chemicals, the Dictionary of Natural Products, a set of 2742 launched drugs, WOMBAT (World of Molecular Bioactivity) $10$ , a database of medicinal chemistry compounds, and two subsets of PubChem, "actives" and DSSTox (Distributed Structure-searchable  $Toxicity)$ <sup>11</sup> comprising toxic substances.

 The molecular descriptors generated by Shape Signatures method<sup>12</sup> was utilized by Chekmarev *et al.*13 with support vector machines (SVM) and Kohonen self-organizing maps  $(Kohonen SOM)^{14}$ techniques, which perform better in classification problems related to the analysis of highly clustered and heterogeneous property spaces. Such models are utilized to predict the potential for cardiotoxicity in drug discovery and elucidating the QSPR model for properties of pharmaceutical interest like aqueous solubility (Log *S*), melting point  $(T<sub>m</sub>)$ , and octanolwater partition coefficient  $(Log P)^{15}$ .

 With the development of novel statistical tools and proliferated molecular descriptors, the QSAR or QSPR studies have become more complicated. However, the simple graph theoretical parameters have made significant contributions to solve the problems through simple statistical tools like multiple regression analysis. Based on the above, a novel topological parameter has been deduced from the structural graph of some hydrocarbons and has been used for QSPR of alkane isomers.

### **Theoretical**

## **Topological descriptors**

 The topological descriptors have been derived for C-3 to C-9 hydrocarbons by using the methods reported below.

The molecular connectivity index  $(\chi)$  as proposed by Randic<sup>16</sup>, Kier and Hall<sup>17</sup> was calculated from the hydrogen suppressed molecular graph. A valence 'δ' value was assigned to the constituting atom by considering the number of non-hydrogen atoms bonded to it.

The 'n' order connectivity index,  ${}^n\chi$ , is the sum of all connecting bonds,

$$
{}^{n}\chi = \Sigma C_{k} \qquad \qquad \ldots (1)
$$

where the connectivity value for first order (connecting atom i and j) is computed as  $C_k = 1/\sqrt{\delta_i \delta_i}$ , second order (connecting bonds i-j-k) as  $C_k = 1/\sqrt{\delta_i \delta_j \delta_k}$  and third order (connecting bonds i-j-k-l) as  $C_k = 1/\sqrt{\delta_i \delta_i \delta_k \delta_l}$ .

 In order to evaluate the molecular topological indices the following algebraic operations on adjacency [A], distance [D], reciprocal of distance [H], walk [W] and valence [V] matrices were carried  $out<sup>18-20</sup>$ .

$$
A_2 = \Sigma \Sigma [A^2] \qquad \qquad \dots (2)
$$

$$
S_D = \Sigma \Sigma [AD]_i \qquad \qquad \dots (3)
$$

$$
S_H = \Sigma \Sigma [AH]_{ij} \qquad \qquad \dots \tag{4}
$$

$$
S_W = \Sigma \Sigma [AW]_{ij} \qquad \qquad \dots \tag{5}
$$

$$
WW^{(1)} = [W]/2 \tag{6}
$$

$$
DW^{(1)} = [D]/2 \tag{7}
$$

$$
HW^{(1)} = [H]/2 \tag{8}
$$

The above defined equations were used to obtain the following molecular topological indices:

$$
{}^{1}\mathrm{W} = \frac{1}{2} \sum \sum d_{ij} \tag{9}
$$

where  $d_{ii}$  are the elements of [D].

$$
{}^{1}H = \frac{1}{2} \sum \sum h_{ij}
$$
 (10)

where  $h_{ii}$  are the elements of the matrix.

$$
R = \frac{1}{2} \sum \sum w_{ij} \tag{11}
$$

where  $w_{ii}$  are the elements of the walk matrix

$$
DI = \Sigma[DW^{(1)}(A+D)] \tag{12}
$$

$$
{}^{2}\mathrm{W} = \{\mathrm{DI} - (\mathrm{S}_{\mathrm{D}}/2)\}/2 \tag{13}
$$

$$
HI = \Sigma[HW^{(1)}(A+H)] \tag{14}
$$

$$
WI = \Sigma [WW^{(1)} (A+W)] \tag{15}
$$

$$
MTI = \sum \sum [A^2 + AD]_{ij}
$$
 ... (16)

$$
VD = \Sigma \Sigma (V \times D)_{ij} \tag{17}
$$

$$
VW = \Sigma \Sigma (V \times W)_{ij}
$$
 ... (18)

$$
VWI = \Sigma \Sigma [W x (V x W)]_{ij}
$$
 ... (19)

$$
VDI = \Sigma \Sigma [DW^{(1)} (V \times D)]_{ij} \qquad \qquad \dots \tag{20}
$$

$$
VMTI = \Sigma \Sigma [V^2 + V D]_{ij} \qquad \qquad \dots \tag{21}
$$

A new set of vertex weighted walk (VWW) descriptors considering the disconnection of bonds in the hydrogen depleted vertex weighted molecular graph has been derived.

In a vertex weighted graph, a valence  $\delta^v$  value for each atom can be assigned as:

$$
\delta_i^{\ \nu} = \sigma_i + p_i + n_i \tag{22}
$$

where  $\delta_i^v$  is the number of non-hydrogen valence electrons contributed by atom i, and  $\sigma_i$  and  $\pi_i$  are the number of sigma and pi bonds, and  $n_i$  is the number of non-bonding electrons<sup>21,22</sup>.

 Various order of VWW\* can be calculated by considering generation of fragments (ζ) after disconnection of a single bond (i,j), two consecutive single bonds  $(i,k)$ , three consecutive single bonds  $(i,l)$ , four consecutive single bonds (i,m), etc.,

 $\zeta_{i,j/k/l/m} = \left(\prod_{i=1}^n\right)$  $\prod_{i=1}^n \delta_i^{\ v})$  $\ldots$  (23) 'n' being the number of atoms in that fragment.

$$
C_p = \prod_{i=1}^n \zeta_{i,j/k/l/m} \qquad \qquad \dots \tag{24}
$$

$$
VWW^* = \sum_p \qquad \qquad \dots \tag{25}
$$

p = number possible bond disconnection and accordingly  $* = 1, 2, 3$  or 4 for one, two, three and four consecutive bond disconnections respectively or order of VWW. An example of calculation of  $VWW<sup>1</sup>$ has been presented in Table 1.

 The possible disconnections on the valenceweighted molecular graph are presented in Fig. 1.

From Table 1, VWW<sup>1</sup> =  $\Sigma$  C<sub>p</sub> = 12.272.

 The possible disconnections for second, third and fourth order VWW are represented in Figs  $2 - 4$ .





Fig. 1―First order VWW graph.



Fig. 2―Second order VWW graph.



Fig. 3―Third order VWW graph.



Fig. 4―Fourth order VWW graph.

The VWW<sup>1</sup>, VWW<sup>2</sup>, VWW<sup>3</sup> and VWW<sup>4</sup> have been calculated for C-3 to C-9 hydrocarbons and are listed in Table 2.

## **Physico-chemical properties of alkanes**

 The physico-chemical properties selected for the present investigation are boiling points (BP), molar volume (MV) at 20 ºC, molar refractions at 20 ºC (MR), heats of vaporization at 25 ºC (HV), critical temperature  $(T_c)$  and surface tension at 20 °C (ST)<sup>23</sup>.

### **Methodology**

 During the quantitative structure property/activity relationship (QSP/AR) studies, various indices are usually incorporated successively into the regression model followed by analysis for optimization. In the present work, a successive exclusion of variable (SEV) technique has been suggested where, from a basic regression model, successive regression models were derived by exclusion of variable having the minimum Student-'t' value. The resultant statistical



parameters like  $R^2$ , F and RMS (residual mean square) were analyzed. With decrease in independent variable,  $R^2$  value decreases and *F* and RMS value increases if the exclusion leads to optimization. The regression model with maximum *F* and minimum RMS values were considered for the optimized regression model.

### **Results and Discussion**

 The VWW parameters have been derived with a view to quantify the interaction of one atom with other atom within the molecule present at varying distances which may contribute or be able to explain the experimental factors like inductive effect, mesomeric effect, NOE (Nuclear Overhauser Effect), etc. We are considering up to fourth order of the VWW since the magnitude of the interaction of the atoms becomes negligible with increase in the number of bonds<sup>24</sup>. These interactions of various bond distances contribute to the physical and chemical properties of the compound.

 When the TI values are the same for the two different molecular graphs, they are said to be degenerate. Balaban<sup>25</sup> observed a high degeneracy for TI values like  ${}^{1}\chi$  and  ${}^{1}W$ . In many cases the degeneracy can be explained by the similar pattern of molecular graphs. Furthermore, in alkane isomers the TI values,  ${}^{1}\chi$ ,  ${}^{1}W$ ,  ${}^{2}W$ , R, DI, WI, MTI, VD, VW, VDI and VWI decrease with increasing branching and the values of TI such as  ${}^{2}\chi$ ,  ${}^{1}H$ ,  ${}^{2}H$ , HI, VMTI increase with increase in branching. The analysis of VWW data reveals that  $VWW<sup>1</sup>$  decreases with increasing branching while  $VWW^2$  and  $VWW^3$  increase with increase in branching in the alkane chain. To check the interrelationships of the variables, the crosscorrelation matrix of the parameters of alkane isomers is represented in Table 3.

 Analysis of the cross-correlation matrix reveals the following:

- 1 A good inter-relationship is observed between almost all the molecular descriptors.
- 2 The new parameters,  $VWW<sup>1</sup>$  and  $VWW<sup>2</sup>$ , show high correlation coefficient with  $\mathrm{^1W}, \mathrm{^2W}, \mathrm{DI}, \mathrm{R},$ MTI, VD, VW, VDI, VWI and  ${}^{1}\chi$ , while its third and fourth order indices show comparatively lesser correlation coefficient.

 To have better insight towards the interrelationships, the TIs have been subjected to principal component analysis (PCA). PCA reduces the size of the descriptors and at the same time proposes some new orthogonal descriptors for the molecules. Each descriptor has more or less contribution from all the test descriptors with variable loading. The results of PCA are given in the Table 4.

Table 3―Simple correlation matrix table of TIs of alkane isomers <sup>1</sup>  $\mathrm{^1W}$   $\mathrm{^2W}$ W  $DI$   $H<sup>1</sup>$   $H<sup>2</sup>$ HI R WI MTI VD VW VDI VWI VMTI  $\frac{1}{\chi}$   $\frac{2}{\chi}$   $\frac{3}{\chi}$  $3\chi$  VWW<sup>1</sup> VWW<sup>2</sup> VWW<sup>3</sup> VWW<sup>4</sup>  $\mathbf{^1W}$  1  $2W$ 0.98 1 DI 0.98 1 1  $H^1$  0.92 0.83 0.84 1  $H<sup>2</sup>$  0.89 0.80 0.80 0.99 1 HI 0.89 0.80 0.80 0.99 1 1 R 0.98 0.99 0.99 0.82 0.77 0.77 1 WI 0.91 0.97 0.97 0.69 0.65 0.65 0.97 MTI 1 0.98 098 0.92 0.88 0.89 0.98 0.91 1 VD 0.99 0.99 0.99 0.90 0.87 0.87 0.98 0.93 0.99 1 VW 0.99 0.99 0.99 0.88 0.86 0.86 0.98 0.94 0.99 1 1 VDI 0.99 1 1 0.87 0.84 0.84 0.99 0.95 0.99 1 1 1 VWI 0.94 0.99 0.99 0.76 0.72 0.72 0.99 0.99 0.95 0.96 0.97 0.97 1 VMTI 0.96 0.93 0.93 0.96 0.96 0.96 0.9 0.82 0.96 0.97 0.96 0.95 0.87 1  $\alpha^1$ χ 0.97 0.91 0.91 0.94 0.9 0.9 0.91 0.81 0.96 0.94 0.91 0..93 0.85 0.92 1  $^{2}$  $\chi$ χ 0.63 0.53 0.53 0.81 0.82 0.83 0.51 0.41 0.63 0.61 0.62 0.57 0.48 0.75 0.59 1  $\alpha^3$ χ 0.65 0.54 0.55 0.81 0.83 0.83 0.52 0.38 0.64 0.62 0.59 0.59 0.45 0.72 0.76 0.52 1 VWW1 0.96 0.94 0.94 0.86 0.81 0.81 0.95 0.88 0.96 0.94 0.92 0.94 0.9 0.88 0.98 0.49 0.64 1  $VWW^2$  0.82 0.82 0.72 0.72 0.97 0.98 0.98 0.68 0.55 0.82 0.8 0.78 0.76 0.62 0.90 0.85 0.83 0.84 0.74 1  $\text{VWW}^3$  0.65 0.65 0.55 0.55 0.81 0.83 0.83 0.5 0.36 0.64 0.64 0.60 0.59 0.44 0.74 0.72 0.55 0.90 0.59 0.86 1  $VWW^4$  0.73 0.73 0.69 0.69 0.75 0.74 0.75 0.65 0.56 0.73 0.74 0.72 0.72 0.61 0.76 0.72 0.53 0.52 0.66 0.75 0.73 1

 Analysis of the data of Table 4 reveals the following:

- 1 The PC1 can explain 79.1% of variance of the total descriptors. The cumulative percent of variance with second PC is found to be 89.1%. Subsequent addition of each principal component increases the cumulative percent of variance.
- 2 When the PC1 and PC2 of 73 alkane isomers are plotted against each other it is found that the alkanes are clubbed in specific domain according to their molecular size. All nonane isomers are clubbed in one region; similarly the octanes, heptanes, hexanes, pentanes, butanes and propane are also arranged (Fig. 5).
- 3 The contribution of each descriptor towards the PCs is also determined from the simple correlation coefficient with all the TIs and it is

Table 4―Eigen values, percent of variance and cumulative percent of variance used in principal component analysis





Fig. 5―Plot of first principal component versus second principal component.

found that all the TIs contribute their maximum towards the first principal component, while  $\frac{3}{\chi}$ and VWW<sup>3</sup> contribute moderately to the second PC.

4 The correlation coefficients of TIs with respect to the PC1 are plotted against the correlation coefficient of PC2 (Fig. 6). The plot further classifies the TIs into different classes.

### **Monovariable regression model**

 All the TIs are subjected to multiple regression analysis for a single parametric equation,

$$
Physical property = a TI + b \qquad \qquad \dots \tag{26}
$$

where 'a' is the sensitivity of the TI towards the physicochemical property of the testing molecules and 'b' is a constant. In monovariable regression model, the alkane isomers are found to have good regression coefficient values (Table 5).

The boiling point values correlate well with the  ${}^{1}\chi$ , having F value of 2379.4 with  $R = 0.98$  indicating the level of confidence of the regression model to be 99.9%. High correlation with properties like  $T_c$  $(R = 0.96, F = 948.04), HV (R > 0.9951, F = 8984.33),$ ST  $(R = 0.91, F = 303.62)$  have also been observed. Molar volume values correlate well with  $\mathrm{^{1}W}$  values  $(R = 0.97, F = 1142.5)$ , while molar refraction correlate well with <sup>1</sup>H values ( $R = 0.98, F = 1751.95$ ).



Fig. 6―Plot of the correlation coefficient of PC1 versus PC2 of TIs.

### **Optimization of regression model with multivariable equations**

Though the monovariable equations are found to have significant predictability, attempts have been taken to obtain equations with higher confidence level by using a multiparametric regression model.

 The BP and TC of 73 alkanes, MV, MR, HV of 69 alkanes and ST of 68 alkanes have been correlated with the 15 molecular descriptors  $(^1W, ^2W, D I, ^1H,$ 2 H, HI, R, WI, MTI, VD, VW, VDI, VWI, VMTI and VWW). The t-values were considered for optimization, and the variables having least t-value were excluded and then further subjected to regression analysis. Improved regression models were indicated by increasing  $F$  and  $R^2$  values and with

Table 5―Regression coefficient (*R*) of monovariable







 a Basic regression model: Explanatory variables: 15 molecular descriptors;  $N = 73$ ,  $R^2 = 0.9972$ ,  $F = 1336.6$ , RMS = 5.8711.

minimum residual mean square (RMS) value. A representative example of successive exclusion of variables for optimizing the regression model has been reported for BP in Table 6. For the BP, TC, MV, MR, HV and ST, respective optimized regression models represented by Eqs 27-32 have been proposed.

BP (°C) = 
$$
(14.2216 \pm 2.0954)^1 W + (0.3134 \pm 0.0385)^2 W
$$
  
+ $(13.2056 \pm 0.8453)^1 H - (3.8061 \pm 0.5187) MTI$   
+ $(0.0092 \pm 0.0014) VW - (0.1045 \pm 0.0121) VDI$   
+ $(6.5133 \pm 1.0587) VWW - (81.9501 \pm 3.5564)$  ... (27)

 $(N= 73, R^2 = 0.9964, F = 2569.9, RMS = 6.5251)$ 

 $T_c$  (<sup>o</sup>C) = (12.581±0.5113)<sup>1</sup>H - (0.4451±0.0375) MTI  $+(12.769\pm0.8514)$  VWW  $+(42.144\pm7.1847)$ ... (28)

$$
(N = 73, R2 = 0.9821, F = 1259.4, RMS = 45.772)
$$

$$
MV(cm3) = -(6.5208 \pm 0.7363)1 W -(0.0684 \pm 0.0073) DI + (18.937 \pm 1.6152)1 H -(1.1551 \pm 0.0978) HI + (1.7360 \pm 0.1646) MTI + (0.0399 \pm 0.0048) VDI + (0.2815 \pm 0.3727) VWW + (51.958 \pm 4.1902) ... (29)
$$

 $(N = 69, R^2 = 0.9984, F = 4763.5, RMS = 0.6060)$ 

MR = – (0.0011±0.0001) DI + (2.8728±0.1367)1 H – (0.1214±0.0072) HI + (0.0388±0.0033) MTI – (0.1183±0.0186) VWW + (12.146±0.3145) ... (30)

 $(N = 69, R = 0.9999, R^2 = 0.9997, F = 50279.2,$  $RMS = 0.0074$ 

HV (kJ mol-1) = (0.0487±0.0031) HI – (0.0225±0.0039) MTI + (1.7916±0.0930) VWW + (1.0612±0.6568) ... (31)

$$
(N = 69, R2 = 0.9890, F = 2038.7, RMS = 0.3168)
$$

ST (dyne cm-1) = (2.3229±0.2969) 1 W + (0.0233±0.0027) DI + (0.1018±0.0072) HI – (0.6560±0.0738) MTI + (0.0008±0.0002) VW – (0.0143±0.0018) VDI + (0.6183±0.1524) VWW + (9.3768±0.7224) … (32)

 $(N = 68, R^2 = 0.9755, F = 341.58, RMS = 0.1017).$ 



By using the proposed optimized equations, calculated physical properties have been plotted against the observed values. The linearity of the plots justifies the applicability of the regression models.

 The experimental values of MV, MR, HV and ST of propane, butane, 2-methyl propane and 2,2,3,3-tetramethylbutane are not available in the literature. Using the optimized regression equations (Eqs 29-32), the values above have been successfully predicted and are listed in Table 7.

## **Conclusions**

 The graph theoretical parameters discussed above are found to be capable of distinguishing chemical structures of various isomers of an alkane from one another. The principal components of the molecular descriptors in all the compounds are due to the contributions of individual descriptors of the concerned compounds. Though PCs seem to be abstract values, the plot of  $r(PC1)$  versus  $r(PC2)$ exhibits a distinct demarcation between various alkane series. Furthermore, the novel parameters VWW are found to correlate well with different molecular properties of the alkane series.

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