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Comparative QSPR Studies with Molecular Connectivity, **Molecular Negentropy and TAU Indices**. **Part 2**. **Lipid–Water Partition Coefficient of Diverse Functional Acyclic Compounds**

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Comparative QSPR Studies with Molecular Connectivity, **Molecular Negentropy and TAU Indices**. **Part 2**. **Lipid–Water Partition Coefficient of Diverse Functional Acyclic Compounds**#

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Abstract

Motivation. In late eighties, topochemically arrived unique (TAU) scheme was described by Pal *et al*. in valence electron mobile (VEM) environment and this index was claimed to have power to decode chemical information in somewhat better way than molecular connectivity index (MCI). The TAU scheme is unique in that it unravels specific contributions of functionality, branching, shape and size factors to the physicochemical property or biological activity while other indices give mainly a global contribution of the molecule. Subsequently, several papers described QSAR and QSPR with TAU index to show the diagnostic potential of the parameter, but much work has not been done on the index. Thus, a comparison among the relations involving these indices may explore relative suitability of the schemes in describing physicochemical parameters.

Method. The present communication attempts to correlate lipid–water partition coefficient (log *P*) of 168 diverse functional acyclic compounds with TAU indices and to compare those with relations involving molecular negentropy (*I*) and first order valence molecular connectivity $({}^1\chi^{\vee})$ indices to explore the diagnostic feature of TAU scheme.

Results. This study shows that TAU indices can unravel specific contributions of molecular bulk (size), functionality, branching and shape parameters to the lipophilicity of diverse functional compounds. In general, lipophilicity increases with increase in molecular bulk and skeletal index value, and decreases with increase in branching and functionality.

Conclusions. The TAU index is an important tool in exploring structure–property relationship studies in view of its potential to unravel specific contribution of different structural parameters like molecular bulk, shape factors, branching, functionality and carbon skeletal structure.

Keywords. Quantitative structure–property relationships; QSPR; topochemically arrived unique (TAU) scheme; molecular connectivity index; structural descriptor; topological index; molecular graph; molecular negentropy; octanol–water partition coefficient; log *P*.

[#] Dedicated to Professor Haruo Hosoya on the occasion of the 65^t*^h* birthday.

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1 INTRODUCTION

Physico–chemical properties of organic compounds change in a systematic way with change in chemical structure, which actually determine the type and magnitude of intramolecular and intermolecular interactions [1]. The chemical information coded in the two dimensional structural representation (topology) are decoded to some extent by assigning a value (index) to a compound or fragment of a compound considering features like adjacency, branching, unsaturation, hetero–atom variation, cyclicity, perturbation effects etc [2–18]. Such indices have been found to be helpful in exploring quantitative structure–property, structure–activity and activity–property relationship studies [5–18]. Quantitative relations generated from such studies help in hypothesizing important contributions of specific structural aspects or chemical interactions in modifying physicochemical properties and biological activities and also in predicting properties or activities of untested (and even not synthesized) compounds.

Among the topological indices, molecular connectivity index of Kier and Hall has received maximum popularity [19, 20]. It has been correlated with several physicochemical properties and biological activities of diverse organic compounds. In late eighties, topochemically arrived unique (TAU) scheme was described by Pal *et al*. [21, 22] in valence electron mobile (VEM) environment and this index was claimed to have power to decode chemical information in somewhat better way than molecular connectivity index (MCI). Subsequently, a few papers [23–27] described QSAR and QSPR with TAU index to show the diagnostic potential of the parameter, but much work has not been done on the index. TAU scheme is unique in that it unravels specific contributions of functionality, branching, shape and size factors to the physicochemical property or biological activity while other indices give mainly a global contribution of the molecule. Thus, a comparison among the relations involving these indices may explore relative suitability of the schemes in describing physicochemical parameters.

The octanol–water partition coefficient (log *P*) representing the overall lipophilicity of a molecule has been extensively used in QSAR and QSPR models as hydrophobicity parameter [28]. Since experimental determination of partition coefficient values of large set of compounds is a tedious job, there are many approaches of calculating log *P* values, *e.g*., Ghose and Crippen's atom contribution method [29, 30], Bodor's quantum chemical method [31], Klopman's Multi–CASE method [32], Moriguchi's method [33] etc. Topological schemes also have been used to model log *P* values [34, 35]. The present communication attempts to correlate lipid–water partition coefficient (log *P*) of 168 diverse functional acyclic compounds (51 alcohols, 25 amines, 25 carboxylic acids and esters, 12 ethers, 6 halocarbons, 34 hydrocarbons and 15 ketones) with TAU indices and to compare those with relations involving molecular negentropy (*I*) [36] and first order valence molecular connectivity $({}^1\chi^{\nu})$ indices to explore the diagnostic feature of TAU scheme.

2 MATERIALS AND METHODS

The physicochemical parameter (log *P*) values were taken from the literature [37,38]. Molecular connectivity [1,19,20] and molecular negentropy [36,39,40] values were calculated according to the original references. Molecular negentropy (MN) is based on the information theory of Shannon [39,40] and obtained from total molecular graph. Atoms (vertices) of a molecular graph are partitioned into disjoint subsets of equivalent vertices and MN is obtained from the formula:

$$
I = -N \sum_{j} p_j \log_{10} p_j \tag{1}
$$

where *N* is the total number of vertices in total molecular graph and p is the probability of random selection of an element (vertex) belonging to subset *j* members of which are equivalent. The value of *p* is equal to n_j / N , n_j being the number of elements of subset *j*. Alternatively, MN can be conveniently calculated from the following:

$$
I = N \log_{10} N - \sum_{j} n_j \log_{10} n_j \tag{2}
$$

Molecular connectivity indices are obtained from hydrogen suppressed graphs of molecules and can be calculated according to the method of Kier and Hall $[1,19,20]$. $\chi^1 \chi$ and χ^2 are two important and simplest molecular connectivity indices. These are defined as follows:

$$
{}^{1}\chi = \sum_{e=1}^{N_e} [(\delta_i \delta_j)_e]^{-0.5}
$$
 (3)

where δ_i and δ_j are adjacency counts (number of atoms joined to the particular atom in the hydrogen–suppressed graph) of the vertices *i* and *j* joining the edge (bond) *e* and N_e is the total number of edges

$$
{}^{1}\chi^{\nu} = \sum_{e=1}^{N_e} [(\delta_i^{\nu} \delta_j^{\nu})_e]^{-0.5}
$$
 (4)

where δ_i^{ν} and δ_j^{ν} are valence δ values of the *i*th and *j*th vertices forming the edge *e*:

$$
\delta_i^{\nu} = (Z^{\nu} - h)/(Z - Z^{\nu} - 1)
$$
\n(5)

In Eq. (5), Z^v is the number of valence electrons, Z is the total number of electrons and h is the number of hydrogens attached to the atom. TAU [21–27] are *T*opochemically *A*rrived *U*nique indices developed in VEM (valence electron, mobile) environment. These include *T* (composite topochemical index), T_R (skeletal index), F (functionality index) and B (simple branching index). The topochemical composite index (*T*) is defined as:

$$
T = \sum_{i < j} E_{ij} = \sum_{i < j} (V_i V_j)^{0.5} \tag{6}
$$

where E_{ij} is VEM edge weight of the edge between i^{th} and j^{th} vertices. V_i represents VEM vertex

weight of the *i*th vertex, which may be calculated as the ratio of core count of the *i*th vertex λ_i to VEM count of the *i*th vertex θ_i . λ_i may be calculated as $(Z - Z^{\nu})/Z^{\nu}$ whereas θ_i may be calculated as $8 - (2h + 1.5v + n)$. When unsaturation is present, θ_i should be calculated as $0.5v + 2\pi$. The notations v, *n* and π represent the numbers of sigma bonds (other than hydrogen), nonbonded electrons and pi bonds associated with the atom in that order.

In case of a heteroatom, VEM edge weight of edge incident upon the heteroatom is assigned a negative value. The skeletal index T_R is the topochemical index of the reference alkane that may be obtained by replacing heteroatom with carbon and removing the multiple bonds that may be present. The derived indices *F* and *B* are easily calculated as $T_R - T$ and $T_N - T_R$ respectively where T_N is topochemical index of the corresponding normal alkane (for acyclic molecules).

STIMS are Simplest Topological Integers from Molecular Structures, that were derived by Pal *et al*. [23–26], as a subset of TAU to obtain an easy tool for predicting various properties and activities of simple molecules directly from the molecular graphs of their reference alkanes. These include *NP* (number of methyl carbons), N_I (number of methylene carbons), N_Y (number of tertiary carbons), N_X (number of quaternary carbons) and N_B (number of branched carbons).

The vertex count (N_V) of the hydrogen–suppressed molecular formula is purely a constitutional parameter because it may be obtained directly from the molecular formula. Even structural formula is not needed for obtaining the value of N_V . Obviously, any index showing better correlation with physicochemical or biological activity than that shown by N_V will have significance in the context of QSPR / QSAR studies.

The first order VEM molecular index T_R is considered as the index for lipophilicity while N_B , N_X and *NY* represent shape parameters. The functionality contribution and bulk parameter are represented by F and N_V respectively [21]. All TAU indices are basically derived by sequentially partitioning the composite index *T* into different factors. *T* may initially be factored into two components, T_R (skeletal index) and F (functionality). Subsequently, T_R may be partitioned into B (branching) and N_V (bulk). N_V can be partitioned into N_P , N_I and N_B . N_B may further be factored into *NX* and *NY*. During development of QSAR equations with TAU parameters, these hierarchical relations were followed. For obvious reasons, *B* and N_B (both represent branching) or N_P and N_B (both have interrelation) [24] or N_V and N_I (N_I may be considered as trimmed counterpart of N_V) [24] were not used in the same equation.

Mutiple linear regression analyses were done using a software program *RRR98* developed by one of the authors [41]. Statistical quality of the equations [42] was judged by examining the parameters like R_a^2 (adjusted R^2 , *i.e.*, explained variance), *r* or *R* (correlation coefficient), *F* (variance ratio) with *df* (degree of freedom), *s* (standard error of estimate) and *AVRES* (average of absolute values of residuals). Significance of the regression coefficients was judged by the *t* test. In case that intercept of an equation was statistically insignificant and omission of the same did not affect the

quality of the equation, exclusion of the intercept gave statistically more acceptable equation. A compound was considered as an outlier for a particular equation when the residual exceeded twice the standard error of estimate of the equation. The robustness of the best equations under different series was checked with leave–one–out (LOO) technique [43,44] using programs *KRPRES1* and *KRPRES2* [41]. Two LOO parameters, Q^2 (crossvalidation R^2 or predicted variance) and SDEP (standard deviation of error of predictions), were used to compare the equations.

		Table 1. Topological indices of diverse functional aliphatic compounds					
No	Compound Name	α^{ν}	\boldsymbol{I}	\boldsymbol{T}	$\mathcal{T}_{\mathcal{R}}$	${\cal T}_N$	
$\mathbf{1}$	Methanol	0.447	3.238	-0.817	1.000	1.000	
$\boldsymbol{2}$	Ethanol	1.023	6.555	0.130	1.414	1.414	
3	n -Propanol	1.523	10.315	0.630	1.914	1.914	
$\overline{\mathbf{4}}$	n -Butanol	2.023	14.404	1.130	2.414	2.414	
5	n -Pentanol	2.523	18.755	1.630	2.914	2.914	
6	n –Hexanol	3.023	23.325	2.130	3.414	3.414	
7	n –Heptanol	3.523	28.081	2.630	3.914	3.914	
8	n –Octanol	4.023	33.001	3.130	4.414	4.414	
9	n -Nonanol	4.523	38.006	3.630	4.914	4.914	
10	Isopropanol	1.413	7.679	0.683	1.731	1.914	
11	Isobutanol	1.879	11.768	0.985	2.270	2.414	
12	tert-Butanol	1.724	7.622	1.092	2.000	2.414	
13	Isopentanol	2.379	16.120	1.485	2.769	2.914	
14	2-Methyl butanol	2.417	18.528	1.523	2.807	2.914	
15	1-Methyl butanol	2.451	18.528	1.721	2.769	2.914	
16	3-Pentanol	2.489	14.314	1.759	2.807	2.914	
17	3-Methyl-2-butanol	2.324	15.893	1.593	2.641	2.914	
18	2-Methyl-2-butanol	2.284	15.291	1.652	2.561	2.914	
19	2,2-Dimethyl-1-propanol	2.170	11.973	1.276	2.561	2.914	
20	2-Hexanol	2.951	23.098	2.221	3.269	3.414	
21	3-Hexanol	2.989	23.098	2.259	3.307	3.414	
22	3-Methyl-3-pentanol	2.845	18.054	2.213	3.121	3.414	
23	2-Methyl-2-pentanol	2.784	19.860	2.152	3.061	3.414	
24	2-Methyl-3-pentanol	2.862	20.462	2.131	3.179	3.414	
25	3-Methyl-2-pentanol	2.862	22.870	2.131	3.179	3.414	
26	4-Methyl-2-pentanol	2.807	20.462	2.076	3.124	3.414	
27	2,3-Dimethyl-2-butanol	2.667	17.225	2.034	2.943	3.414	
28	3,3-Dimethyl-1-butanol	2.670	16.543	1.776	3.061	3.414	
29	3,3-Dimethyl-2-butanol	2.624	16.316	1.894	2.943	3.414	
30	2-Methyl-2-hexanol	3.284	24.617	2.652	3.561	3.914	
31	3-Methyl-3-hexanol	3.345	27.025	2.713	3.621	3.914	
32	3-Ethyl-3-pentanol	3.406	17.005	2.774	3.682	3.914	
33	2,3-Dimethyl-2-pentanol	3.205	24.389	2.572	3.481	3.914	
34	2,3-Dimethyl-3-pentanol	3.228	24.389	2.595	3.503	3.914	
35	2,4-Dimethyl-2-pentanol	3.140	21.981	2.507	3.416	3.914	
36	2,4–Dimethyl–3–pentanol	3.234	16.563	2.503	3.551	3.914	
37	2,2-Dimethyl-3-pentanol	3.162	21.072	2.432	3.481	3.914	
38	2,2,3-Trimethyl-3-pentanol	3.534	25.162	2.902	3.811	4.414	
39	Cyclohexanol	2.575	17.674	2.345	3.393	3.414	
40	4-Penten-1-ol	2.133	16.858	0.894	2.914	2.914	
41	3-Penten-2-ol	2.080	16.403	0.917	2.769	2.914	
42	1-Penten-3-ol	2.115	16.630	1.062	2.807	2.914	
43	1-Hexen-3-ol	2.615	21.059	1.562	3.307	3.414	
44	2-Hexen-4-ol	2.618	20.832	1.455	3.307	3.414	

Table 1. Topological indices of diverse functional aliphatic compounds

j.

3 RESULTS AND DISCUSSION

The calculated topological indices of 168 compounds are given in Table 1. Tables 2–9 show relations of hydrophobicity (log *P*) with different topological indices. All regression coefficients and variance ratios of the reported equations are significant at 95% and 99% levels respectively unless otherwise stated (marked with *). Table 10 shows the literature log *P* values of the compounds [37,38] and also the calculated values according to the best equations of individual series and the composite set (*vide* foot note of Table 10).

Table 2. Relations of hydrophobicity (log *P*) of alcohols with various indices. Model equation, log $P = \sum \beta_i x_i + \alpha$

	Eq Index			Regression coefficient(s) and constant					Statistics	
	Type	β_1	β_2	β_3	β_4	α	Q^2	R_a^2	S	AVRES
		se	se	se	se	se	(SDEP)	$(r \text{ or } R)$	(F(df))	(n)
7	MCI	$0.907 \frac{1}{2} \chi^{V}$				-1.118	0.709	0.752	0.435	0.199
		0.073				0.194	(0.467)		(0.870) $(152.827(1, 49))$	(51)
8	MN(I)	0.112 I				-0.843	0.738	0.759	0.429	0.252
		0.009				0.169	(0.443)		(0.874) $(158.762$ $(1, 49)$	(51)
9	TAU	0.755 T				-0.097	0.711	0.764	0.424	0.273
		0.059				0.115			(0.466) (0.877) (163.039) $(1, 49)$	(51)
10	TAU	0.973 T_R -0.416 F				-1.186	0.744	0.815	0.376	0.203
		0.078	0.103			0.303			(0.438) (0.907) $(111.313 (2, 48))$	(51)
11	TAU	$-0.436 F -1.162 B$		$0.489 N_{V}$		-1.209	0.699	0.814	0.377	0.190
		0.107	0.304	0.039		0.281		(0.475) (0.908)	(73.713(3, 47))	(51)
12	- TAU			$-0.434 F$ 0.475 N_V $-0.292 N_R$		-1.075	0.735	0.812	0.379	0.233
		0.108	0.038	0.078		0.280	(0.446)	(0.907)	(72.886(3, 47))	(51)
13	TAU			$-0.409 F$ 0.463 N_V -0.144 N_P		-0.857	0.692	0.786	0.404	0.217
		0.114	0.041	0.056		0.314	(0.480)	(0.894)	(62.253(3, 47))	(51)
14	TAU			$-0.445 F$ 0.488 N_V -0.403 N_X -0.274 N_Y -1.116			0.689	0.813	0.378	0.202
		0.108	0.040	0.125	0.079	0.312		(0.483) (0.910)	(55.308(4, 46))	(51)

se = standard error; F values are significant at 99% level $[df = np, n - np - i, np = no$. of predictor variables; i = 1 if intercept is present; i = 0, otherwise]; *t* values of the regression coefficients and constants are significant at 95% level $[df = n - np - i]$

3.**1 QSPR of Alochols** (*n* **= 51**)

Eqs. (7)–(14) (Table 2) show the relations of log *P* of alcohols with different indices. First order valence molecular connectivity and molecular negentropy could explain the variance of the data set to the same extent (75.2% and 75.9% respectively) while the latter could predict the variance (73.8%) better than the former (70.9%) . TAU indices, Eqs. (9) – (14) , could explain up to 81.5% of the variance vis–a–vis predict up to 74.4% of variance and show specific contributions of carbon skeleton (T_R) , functionality (F) , bulk (N_V) , branching (B) and shape parameters $(N_X$ and $N_Y)$. The relations show positive impact of T_R and bulk (N_V) and negative impact of functionality, branching and shape parameters. Ethylene glycol and 2,3–butanediol show outlier behavior in all the models. Though equation 10 is of the best statistical quality, Eq. (14) carries more information about specific contributions of different shape and size factors, and hence, it has been used to calculate the log *P* values of the alcohols as reported in Table 10.

Eq	Index		Regression coefficient(s) and constant				Statistics			
	Type	β_1	β_2	β_3	β_4	α	Q^2	R_a^2	S	AVRES
		se	se	se	se	se	(SDEP)	$(r \text{ or } R)$	(F(df))	(n)
15	MCI	$0.943 \frac{1}{2} \chi^{\nu}$				-1.167	0.960	0.964	0.171	0.136
		0.037				0.100	(0.177)	(0.983)	(645.042(1, 23))	(25)
16	MN(I)	0.104 I				-0.628	0.878	0.894	0.295	0.217
		0.007				0142	(0.310)	(0.948)	(202.840(1, 23))	(25)
17	TAU	0.702 T				0.425	0.672	0.734	0.466	0.371
		0.086				0.134	(0.507)	(0.863)	(67.307(1, 23))	(25)
18	TAU	0.951 T_R				-1.512	0.942	0.951	0.200	0.154
		0.044				0.132	(0.206)	(0.976)	(468.424(1, 23))	(25)
19	TAU	$0.479 N_V$	$-0.144* N_R$			-1.600	0.952	0.958	0.184	0.143
		0.020	0.075			0.127	(0.194)	(0.981)	(277.630(2, 22))	(25)
20	TAU	$0.420 N_I$	$0.717 N_{Y}$			-0.481	0.788	0.825	0.379	0.228
		0.039	0.161			0.183	(0.408)	(0.916)	(57.383(2, 22))	(25)

Table 3. Relations of hydrophobicity (log *P*) of amines with various indices. Model equation, log $P = \sum \beta_i x_i + \alpha$

3.**2 QSPR of Amines** (*n* **= 25**)

The relations of log *P* of amines with different indices are shown in Eqs. (15)–(20) (Table 3). The variation of lipophilicity of the amines could be explained to the extent of 96.4% by first order valence connectivity index (predicted variance 96.0%) while it was only 89.4% for molecular negentropy (predicted variance 87.8%). In case of TAU indices, the composite topochemical index *T* could explain 73.4% (predicted variance 67.2%) while skeletal index T_R could explain 95.1% of the variance (predicted variance 94.6%). The functionality index *F* did not show any importance. Further, T_R was partitioned into size and shape factors and the best relation obtained, Eq. (19), could explain 95.8% of the variance (predicted variance 95.2%). However, the regression coefficient of *N_B* in Eq. (19) was significant at 90% level. Piperidine showed outlier behavior for this equation. For amines, bulk and T_R show positive impact and branching shows negative impact on lipophilicity.

3.**3 QSPR of Carboxylic Acids and Esters** (*n* **= 25**)

Table 4 shows the relations of log *P* of carboxylic acids and esters with different indices, Eqs. (21)–(29). First order molecular connectivity index could explain 99.6% variance (predicted

variance 99.5%) while molecular negentropy explained 99.4% variance (predicted variance 99.3%). Under the TAU scheme, composite index *T* showed somewhat inferior relation which could explain 96.3% of the variance (predicted variance 95.9%). However, when the composite index was partitioned into skeletal, functionality, shape and size factors, the relations could explain up to 99.7% of the variance (predicted variance up to 99.6%). The relations showed positive impact of skeletal index T_R and bulk (N_V) , and negative impact of functionality (F) and branching (B) factors. This means that lipophilicity of carboxylic acids and esters increases with increase in molecular size, and it decreases as the molecule becomes more branched and as the functionality value rises. Decanoic acid showed outlier behavior in almost all the relations.

Table 4. Relations of hydrophobicity (log *P*) of carboxylic acids and esters with various indices

Eq	Index			Regression coefficient(s) and constant					Statistics	
	Type	β_1	β_2	β_3	β_4	α	Q^2	R_a^2	S	AVRES
		se	se	se	se	se	(SDEP)	$(r \text{ or } R)$	(F(df))	(n)
21	MCI	$0.998 \frac{1}{\chi^{\nu}}$				-1.190	0.995	0.996	0.087	0.061
		0.013				0.042	(0.093)	(0.998)	(6042.458(1, 23))	(25)
22	MN(I)	0.108 I				-0.651	0.993	0.994	0.106	0.085
		0.002				0.044	(0.116)	(0.997)	(4095.987(1, 23))	(25)
23	TAU	0.972 T				0.314	0.959	0.963	0.265	0.188
		0.039				0.079	(0.275)	(0.982)	(632.016(1, 23))	(25)
24	TAU	1.009 T_R	$-0.214 F$			-1.586	0.995	0.997	0.079	0.060
		0.012	0.051			0.126	(0.092)	(0.998)	(3630.220(2, 22))	(25)
25	TAU	$-0.211 F$	$-0.951 B$	$0.505 N_V$		-1.689	0.995	0.997	0.081	0.060
		0.054	0.275	0.006		0.142	(0.097)	(0.998)	(2315.067(3, 21))	(25)
26	TAU	$-0.183 F$	$0.509 N_V$	$-0.143 N_R$		-1.769	0.996	0.997	0.075	0.053
		0.049	0.006	0.034		0.126	(0.091)	(0.999)	(2699.400(3, 21))	(25)
27	TAU	$-0.183 F$	$0.509 N_V$	$-0.143 N_P$		-1.482	0.996	0.997	0.075	0.053
		0.049	0.006	0.034		0.157	(0.091)	(0.999)	(2699.400(3, 21))	(25)
28	TAU	$-0.183 F$	$0.509 N_I$	$0.875 N_{Y}$		-0.750	0.996	0.997	0.075	0.053
		0.049	0.006	0.035		0.123	(0.091)	(0.999)	(2699.402(3, 21))	(25)
29	TAU	$-0.183 F$	0.509 N _L	$0.875 N_P$		-2.501	0.996	0.997	0.075	0.053
		0.049	0.006	0.035		0.160	(0.091)	(0.999)	(2699.402(3, 21))	(25)

3.**4 QSPR of Ethers** (*n* **= 12**)

Eqs. (30)–(37) describing relations of log *P* of ethers with topological indices are shown in Table 5 which shows that 95.1% of the variance could be explained by first order molecular connectivity index (predicted variance 93.7%). However, only 63.3% could be explained by molecular negentropy (predicted variance 55.6%). Similarly, the composite topochemical index *T* could explain only 68.1% of the variance (predicted variance 58.2%). However, the topochemical skeletal index T_R explained 93.1% of the variance (predicted variance 91.6%). This relation was improved further by partitioning T_R into branching and size parameters. The best relation, Eq. (37), could explain 99.3% of the variance. However, due to insufficient occurrence of quaternary carbon fragment (N_X) type) in the compounds, LOO could not be applied for this equation. Ethyl cyclopropyl ether was an outlier for this equation. T_R and bulk showed positive contributions, while branching and shape factors showed negative impact. Functionality did not show any contribution.

Specific contributions of tertiary and quaternary type carbons, and molecular bulk are evident from the respective regression coefficients.

Table 5. Relations of hydrophobicity (log *P*) of ethers with various indices

 $+$ LOO could not be applied due to insufficient occurrence of N_X fragment (quaternary carbon)

Table 6. Relations of hydrophobicity (log *P*) of halocarbons with various indices

Eq	Index	Regression coefficient(s) and constant Statistics								
	Type	β_1	β_2	β_3	β_4	α	Q^2	R_a^2	S	AVRES
		se	se	se	se	se	(SDEP)	$(r \text{ or } R)$	(F(df))	(n)
38	MCI	$-0.222*1^{\nu}$				2.605	-1.010	-0.157	0.276	0.193
		0.392				0.932	(0.333)	(0.273)	(0.322(1, 4))	(6)
39	MN(I)	0.038 I				1.778	0.424	0.748	0.129	0.090
		0.009				0.092	(0.178)	(0.894)	(15.843(1, 4))	(6)
40	TAU	0.118 T				2.219	0.640	0.849	0.100	0.070
		0.022				0.048	(0.141)	(0.938)	(29.122(1, 4))	(6)
41	TAU	$0.347 T_R$				1.421	0.564	0.831	0.105	0.078
		0.069				0.137	(0.155)	(0.930)	(25.652(1, 4))	(6)
42	TAU	$-0.208 F$	1.449 B			2.675	$-$ ⁺	0.897	0.083	0.039
		0.032	0.594			0.095	$-$ ⁺	(0.968)	(22.663(2, 3))	(6)
43	TAU	$-0.208 F$	$0.265 N_B$			2.675	$ \!^+$	0.897	0.083	0.039
		0.032	0.109			0.095	$\overline{}^+$	(0.968)	(22.663(2, 3))	(6)

* Insignificant at 90% level; + LOO could not be applied due to insufficient occurrence of branching

3.5 QSPR of Halocarbons $(n = 6)$

In case of halocarbons, inferior relations were obtained. These relations, Eqs. (38)–(43) are listed in Table 6. Molecular connectivity could not give any acceptable relation (explained variance 27.2%, insignificant β –coefficient) while molecular negentropy explained 74.8% variance (predicted variance 42.4%). The composite topochemical index *T* could explain 84.9% of the variance (predicted variance 64.0%). The topochemical skeletal index T_R explained almost similar to *T* (83.1% of the variance), but the predicted variance reduced to 56.4%. On further partitioning of

TR, TAU scheme generated equations 42 and 43, both explaining 89.7% of the variance. However, due to insufficient occurrence of branching, PRESS statistics could not be obtained for these two equations. There was no outlier for the equations. Functionality was found to have negative contribution while branching showed positive impact. However, the number of data points of halocarbon compounds was insufficient for regression with two predictor variables. The results obtained are treated only as preliminary ones.

	rapic <i>i</i> . Kelations of hydrophobieny (log <i>r</i>) or hydrocarbons with various murces										
Eq	Index		Regression coefficient(s) and constant					Statistics			
	Type	β_1	β_2	β_3	β_4	α	Q^2	R_a^2	S	AVERES	
		se	se	se	se	se	(SDEP)	$(r \text{ or } R)$	(F(df))	(n)	
44	MCI	$0.936^{1} \chi^{\nu}$				0.244	0.949	0.953	0.125	0.103	
		0.036				0.099	(0.130)	(0.977)	(673.244(1, 32))	(34)	
45	MN(I)	0.064 I				1.772	0.455	0.507	0.407	0.311	
		0.011				0.181	(0.421)	(0.722)	(34.920(1, 32))	(34)	
46	TAU	0.564 T				1.300	0.511	0.550	0.389	0.314	
		0.088				0.237	(0.399)	(0.751)	(41.312(1, 32))	(34)	
47	TAU	$0.856 T_R$	$-0.245 F$			0.198	0.742	0.768	0.279	0.235	
		0.082	0.085			0.261	(0.290)	(0.884)	(55.615(2, 31))	(34)	
48	TAU	$-0.270 F$	$-0.854 B$	$0.453 N_V$			0.788	0.816	0.248	0.211	
		0.083	0.654	0.011			(0.263)	(0.910)	(1449.764(3, 31))	(34)	
49	TAU	$-0.477 F$	$0.722 N_P$	$0.456 N_I$		-0.254	0.977	0.981	0.080	0.061	
		0.026	0.020	0.012		0.083	(0.086)	(0.991)	(560.968(3, 30))	(34)	

Table 7. Relations of hydrophobicity (log *P*) of hydrocarbons with various indices

3.6 QSPR of Hydrocarbons $(n = 34)$

Eqs. (44)–(49) relating log *P* of hydrocarbons with topological indices are shown in Table 7. First order molecular connectivity index could explain 95.3% of the variance (predicted variance 94.9%) while molecular negentropy explained only 50.7% of the variance (predicted variance 45.5%). Though the composite topochemical index *T* did not give satisfactory results (explained variance 55.0%, predicted variance 51.1%), on partitioning into different factors like functionality, branching and size, TAU scheme gave an highly acceptable relation, Eq. (49), with 98.1% explained variance and 97.7% predicted variance. This relation contains N_P as the shape parameter and *NI* as the size parameter. 1,6–Heptadiyne showed outlier behaviour for the equation. Another relation (eq. 48), which is somewhat inferior in quality (explained variance 81.6%, predicted variance 78.8%) than that of Eq. (49), does not have any outlier. From the relations, it is evident that bulk (N_V) of the molecules has positive impact while branching and functionality have negative impact on log *P* values.

3.**7 QSPR of Ketones** (*n* **= 15**)

Table 8 shows the relations of log *P* values of ketones with topological parameters. First order molecular connectivity index and composite topochemical index *T* explained nearly to the same extent (96.7% and 95.8% respectively). However, MCI predicted somewhat better than *T* (94.3% *vs*. 92.9% predicted variance). On the other hand, molecular negentropy gave inferior relation,

explaining 83.0% of the variance (predicted variance 79.0%) and having intercept significant at 90% level. When *T* was factored into *F*, *B* and *NI*, Eq. (53), explained variance rose to 98.2% (predicted variance 96.8%). This equation carries information about specific contributions of the parameters *F* (functionality), *B* (branching) and N_I (size factor). Intercept of this equation is significant at 90% level. This relation shows that log *P* value of ketones increases with increase in size and decreases as branching and functionality value rise. Acetone showed outlier behavior in all the equations.

	rative of intrations of hydrophobicity (log <i>r</i>) or iccones with various murces										
Eq	Index		Regression coefficient(s) and constant					Statistics			
	Type	β_1	β_2	β_3	β_4	α		R_a^2	S	AVERES	
		se	se	se	se	se	(SDEP)	$(r \text{ or } R)$	(F(df))	(n)	
50	MCI	$0.912 \frac{1}{\chi^{\nu}}$				-1.257	0.943	0.967	0.141	0.098	
		0.045				0.131	(0.177)	(0.984)	(405.934(1, 13))	(15)	
51	MN(I)	0.096 I				$-0.458*$	0.790	0.830	0.317	0.229	
		0.012				0.223	(0.341)	(0.918)	(69.277(1, 13))	(15)	
52	TAU	0.896 T				-0.735	0.929	0.958	0.158	0.111	
		0.050				0.119	(0.198)	(0.980)	(319.367(1, 13))	(15)	
53	TAU	$-2.241 F$	5.404 B	$0.476 N_I$		1.508*	0.968	0.982	0.103	0.069	
		0.682	0.500	0.019		0.711	(0.133)	(0.993)	(254.377(3, 11))	(15)	

Table 8. Relations of hydrophobicity (log *P*) of ketones with various indices

* Significant at 90% level

.**8 QSPR of Composite Set** (*n* **= 168**)

Table 9 lists the relations of log *P* values of all compounds (composite set) with topological indices. For the composite set, MCI and molecular negentropy could explain only 63.9% and 39.7% respectively of the variance (predicted variance 63.3% and 38.6% respectively). The composite topochemical index *T* singularly explained 49.7% of the variance (predicted variance 46.4%). When *T* was partitioned into T_R and *F*, then the relation could explain 58.6% of the variance (predicted variance 55.9%). On further partitioning of T_R , different relations, Eqs. (58), (60), (61), (62) showing 63–65% explained variance (60–63% predicted variance) were obtained.

NoCompound		log P		radics To. Observed and calculated molecular hydrophobicity (log NoCompound	uuu	log P	
	Exp^a	Calc	Calc ⁱ		$Exp^{\overline{a}}$	Calc	$Calc^i$
1 Methanol	-0.66	-0.948^{b}	-0.049	2 Ethanol	-0.32	-0.223^{b}	$0.55\overline{5}$
3 n -Propanol	0.34	0.265^{b}	0.995	4 n -Butanol	0.88	0.754^{b}	1.434
5 <i>n</i> -Pentanol	1.40	1.242^{b}	1.873	6 <i>n</i> -Hexanol	1.84	1.730^{b}	2.312
7 <i>n</i> -Heptanol	2.34	2.218^{b}	2.752	8 <i>n</i> -Octanol	2.84	2.707^{b}	3.191
9 n -Nonanol	3.15	3.195^{b}	3.630	10 Isopropanol	0.14	0.097^b	0.538
11 Isobutanol	0.61	0.479^{b}	0.904	12 tert-Butanol	0.37	0.518^{b}	0.741
13 Isopentanol	1.14	0.968^{b}	1.344	142-Methyl butanol	1.14	0.968^{b}	1.344
15 1-Methyl butanol	1.14	1.073^{b}	1.417	163-Pentanol	1.14	1.073^{b}	1.417
173-Methyl-2-butanol	0.91	0.799^{b}	0.887	182-Methyl-2-butanol	0.89	1.006^b	1.180
19 2,2-Dimethyl-1-propanol	1.36	0.839^{b}	1.064	202-Hexanol	1.61	1.561^{b}	1.856
213-Hexanol	1.61	1.561^{b}	1.856	22 3-Methyl-3-pentanol	1.39	1.495^{b}	1.620
23 2-Methyl-2-pentanol	1.39	1.494^{b}	1.620	24 2-Methyl-3-pentanol	1.41	1.288^{b}	1.327
253-Methyl-2-pentanol	1.41	1.288^{b}	1.327	264-Methyl-2-pentanol	1.41	1.288^{b}	1.327
27 2,3-Dimethyl-2-butanol	1.17	1.220^{b}	1.090	28 3,3–Dimethyl–1–butanol	1.86	1.327^{b}	1.503
29 3,3-Dimethyl-2-butanol	1.19	1.158^{b}	1.047	302-Methyl-2-hexanol	1.87	1.982^{b}	2.059
31 3-Methyl-3-hexanol	1.87	1.983^{b}	2.059	323-Ethyl-3-pentanol	1.87	1.983^{b}	2.059
33 2,3-Dimethyl-2-pentanol	1.67	1.709^{b}	1.530	34 2,3-Dimethyl-3-pentanol	1.67	1.709^{b}	1.530
35 2,4-Dimethyl-2-pentanol	1.67	1.709^{b}	1.530	36 2,4-Dimethyl-3-pentanol	1.71	1.502^{b}	1.237
37 2,2-Dimethyl-3-pentanol	1.69	1.647^{b}	1.487	38 2,2,3–Trimethyl–3–pentanol	1.99	2.068^{b}	1.689
39 Cyclohexanol	1.23	1.561^{b}	1.856	40 4-Penten-1-ol	1.04	0.914^{b}	1.645
413 -Penten-2-ol	0.81	0.715^{b}	1.168	42 1-Penten-3-ol	0.81	0.763^{b}	1.201
43 1– $Hexen-3-o1$	1.31	1.251^{b}	1.640	44 2-Hexen-4-ol	1.31	1.203^{b}	1.607
45 2-Methyl-4-penten-3-ol	1.11	0.977^b	1.111	46 2,2,2-Trifluoroethanol		$0.41 - 0.334^b$	0.248
47 Ethylene glycol	-1.93	-0.307^{b}	0.597	48 Allyl alcohol		$0.17 - 0.062^b$	0.767
49 sec-Butanol	0.61	0.585^{b}	0.978	502,3-Butanediol	-0.92	0.333^{b}	0.563
51 1-Ethynylcyclohexanol	1.73	2.471^{b}	2.498	52 Methyl amine		-0.57 -0.642 ^c -0.073	
53 Ethyl amine		$-0.13 -0.162^c$	0.538	54 n -Propyl amine	0.48	0.317^{c}	0.978
55 n -Butyl amine	0.75	0.796^{c}	1.417	56 n -Pentyl amine	1.49	1.276^{c}	1.856
57 n –Hexyl amine	1.98	1.755^{c}	2.295	58 n -Heptyl amine	2.57	2.234^{c}	2.735
59 Isobutyl amine	0.73	0.652^{c}	0.887	60 sec-butyl amine	0.74	0.652^{c}	0.964
612-Amino octane	2.82	2.569^{c}	2.721	62 Cyclohexyl amine	1.49	1.610^{c}	1.842
63 Isopropyl amine	0.26	0.173^c	0.524	64 Methyl ethylamine	0.15	0.317^{c}	0.684
65 Di $-n$ -propyl amine	1.67	1.755^{c}	2.123	66 Triethylamine	1.44	1.610^{c}	1.462
67 Di $-n$ -butyl amine	2.68	2.713^{c}	3.002	68 Diethylamine	0.57	0.796^{c}	1.245
69 <i>n</i> -Propyl- <i>n</i> -butyl amine	2.12	2.234^{c}	2.563	70 Methyl- n -butyl amine	1.33	1.276^{c}	1.562
71 Piperidine	0.85	1.276^{c}	1.684	72 Ethyl-isopropyl amine	0.93	1.131^{c}	1.209
73 n-Propyl-sec-butylamine	1.91	2.090^{c}	2.087	74 n -Propyl-isobutylamine	2.07	2.090^{c}	2.033
75 Trimethylamine	0.27		0.173^{c} -0.153	76 Dimethyl- n -butylamine	1.70	1.610^{c}	1.264
77 Acetic acid	-0.17	-0.204^{d}	0.305	78 Propionic acid	0.25	0.316^{d}	0.763
79 Butyric acid	0.79	0.825^{d}	1.202	80 Hexanoic acid	1.88	1.844^{d}	2.081
81 Decanoic acid	4.09	3.881^{d}	3.838	82 Ethyl formate	0.23	0.321^{d}	1.057

Tables 10. Observed and calculated molecular hydrophobicity (log *P*) data

^{*a*} Taken from refs. [37] and [38] ; ^{*b*} As per Eq. (14); ^{*c*} As per Eq. (19); ^{*d*} As per Eq. (26); ^{*e*} As per Eq. (37); ^{*f*} As per Eq. (42); *^g* As per Eq. (49); *^h* As per Eq. (53); *ⁱ* As per Eq. (64).

The best TAU relation, Eq. (64), could explain 65.9% of the variance (predicted variance 63.1%). Ethylene glycol, 2,3–butanediol, 2,4–dimethylpentane, chloroform, methyl iodide, ethyl iodide and 1–bromopropane showed outlier behaviour for most of the equations. When the four

halogen compounds from this list of outliers were deleted, the statistical quality of the equation, Eq. (65), rose significantly (explained variance 81.2% and predicted variance 80.3%). Positive impact of carbon skeleton (T_R) and bulk (N_V) factors and negative impacts of functionality, branching and shape factors were found with specific quantitative contribution pattern.

4 CONCLUSIONS

This study shows that though composite topochemical index *T* does not always provide better model for log *P* of heterofunctional acyclic compounds in comparison to molecular connectivity and negentropy, TAU scheme can generate statistically superior relations when the composite index is partitioned into different components like skeletal index, size and shape factors, branching and functionality. Moreover, TAU indices can unravel specific contributions of molecular bulk (size), functionality, branching and shape parameters to the lipophilicity of diverse functional compounds. In general, lipophilicity increases with increase in molecular bulk and skeletal index value, and decreases with increase in branching and functionality. However, the halocarbons show some aberrant behavior and behave as outliers in the composite set.

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