

Quantitative structure-property relationship (QSPR) model for predicting acidities of ketones

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ABSTRACT

Ketones are one of the most common functional groups, and ketone-containing compounds are essential in both the nature and the chemical sciences. As such, the acidities (pK_a) of ketones provide valuable information for scientists to screen for biological activities, to determine physical properties or to study reaction mechanisms. Direct measurements of pK_a of ketones are not readily available due to their extremely weak acidity. Hence, a quantitative structure-property relationship (QSPR) model that can predict the acidities of ketones and their acidity order is highly desirable. The establishment of an acidity scale in dimethyl sulfoxide (DMSO) solution by Bordwell *et al.* made such an effort possible. By utilizing the pK_a values of forty-eight ketones determined in DMSO as the training set, a QSPR model for predicting acidities of ketones was built by stepwise multiple linear regression analysis. The established model showed statistical significance and predictive power ($r^2 = 0.91$, $q^2 = 0.86$, $s = 1.42$). Moreover, the QSPR model also gave reasonable acidity predictions for five ketones in an external prediction set that were not included in the model generation phase ($r^2 = 0.92$, $s = 1.618$). Overall, the reported QSPR model for predicting acidities of ketones provides a useful tool for both biologists and chemists in understanding the biophysical properties and reaction rates of different classes of ketones.

Keywords: QSPR; Acidity; Ketones; Linear Regression

1. INTRODUCTION

Ketones play a crucial role in nature. For example, metabolism of carbohydrates, fatty acids and amino acids

in humans and most vertebrates generates acetone, acetoacetate and beta-hydroxybutyrate, which are known as ketone bodies in biochemistry. Acetoacetate and beta-hydroxybutyrate are important fuels for many tissues. For example, it was reported that acetoacetate contributes over 90% to the energy required for respiration in the sheep heart, 85% in the sheep kidney cortex and 74% in the sheep diaphragm [1]. Ketone bodies are also found to have therapeutic values for neurological diseases such as Alzheimer's disease [2,3] and Parkinson's disease [3, 4]. Hasebe and Hauptman *et al.* have discovered their function in reducing epileptic seizures as well [5,6]. Additionally, it was reported that monoacetoacetin (glycerol monoacetoacetate) has the potential to decrease growth of human gastric cancer cells [7].

Being inspired by their imperative role in nature, ketones are also commonly applied scientifically and commercially, especially in the field of chemistry. Not only are they massively produced as solvents in industry, but base-catalyzed condensation reactions with ketones are also employed on a daily base in organic synthesis labs. According to the reaction mechanism, in the presence of a base, the chemoselectivity (relative reaction rates) or regioselectivity (preferred reaction site), is primarily determined by the acidities (pK_a) of different ketones, *i.e.* the acidities of alpha hydrogen atoms at different positions (**Figure 1**). Since ketones are extremely weak acids, direct measurements of their acidities in hydroxylic solvents seem to be impossible. Although measurements of deuterium exchange rates along with some other methods have determined the equilibrium acidities for a number

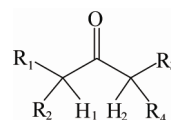


Figure 1. Alpha hydrogens at different positions may have different acidities due to the effect of different R group substituents ($pK_a(H_1) \neq pK_a(H_2)$).

of ketones [8-11], the accuracy and applicability were not very satisfying. The establishment of an acidity scale in dimethyl sulfoxide (DMSO) solution by Bordwell [12] was undoubtedly a milestone in this respect. It provided a large number of pK_a values for a variety of weak acids in DMSO, including ketones. With these pK_a values as well as the oxidation potentials of ketones and their conjugate bases, Bordwell *et al.* were able to predict both the acidities of the radical cations formed from the parent acids [13] and the homolytic bond dissociation energies (BDEs) of their acidic C-H bonds [14]. BDEs are very useful in terms of studying reaction mechanisms and assessing stabilities of radicals [15]. Due to its simplicity and general applicability, this method of calculating BDEs is still being used today since it was first introduced about twenty years ago [16].

Several groups have described QSPR models to predict pK_a values of acids, alcohols, phenols, chlorinated phenols and amines [17-23]. To our knowledge, no such effort has been focused on the acidities of ketones yet. Because ketones are so important both in nature and science and because experimental determination of pK_a values is an exhausting process, the development of a computational model that can accurately predict their acidities is both valuable and timely. More specifically, by utilizing such a QSPR model, biologists can easily explore a variety of ketones that may carry comparable pK_a with the aforementioned ketone bodies to address the same diseases based on the concept of "bioisosterism", whereas chemists can predict a reaction mechanism where ketones are involved in order to design a more reliable synthetic route for their target compounds. In this report, we present a quantitative structure-property relationship (QSPR) model to predict acidities of ketones in DMSO. The effects of different functional groups and substitution patterns on their acidity as represented by five descriptors in the model are discussed.

2. MATERIALS AND METHODS

2.1. Data Set

Fifty-eight ketones with experimental pK_a data [16,24, 25] were subjected to initial data screening. Three of the ketones were first discarded to avoid incongruence of data. Among them, two have markedly different structures (one is a quaternary ammonium salt and another is a chromium tricarbonyl complex), while the pK_a of the third ketone was acquired under different condition. The final set of fifty-five ketones (**Figure 2**) can be categorized into three groups based on their structures. Group A is composed of aliphatic noncyclic ketones **1 - 6**. Group B consists of cyclic ketones **7 - 17**. The remaining ketones **18 - 55**, which typically contain at least one phenyl ring in their structures and are exocyclic with respect to

the ketone, form the group C. Group C can be further divided into five subgroups: C1, $\text{CH}_3\text{COCHR}_1\text{R}_2$ (R_1, R_2 can be either same or different), **18 - 20**; C2, CH_3COR (R is substituted or non-substituted aromatic ring), **21 - 33**; C3, PhCOCH_2R (R can be either aliphatic or aromatic), **34 - 51**; C4, $\text{PhCOCHR}_1\text{R}_2$ (R_1, R_2 can either be independent or form a cyclic ring), **52 - 54**; C5, **55**, which falls into none of the above groups.

In order to evaluate how well a model to be built can predict the acidities of ketones, an external prediction set (PSET) that includes one or more members from each group is considered necessary. The criteria for building such a PSET were: 1) ketones which have either highest (**7**) or lowest (**39**) pK_a values are not eligible for the PSET because a model cannot reliably predict properties out of the range it was built, *i.e.* extrapolation; 2) the qualified candidates for the PSET should be able to represent at least several of their group members or their counterparts from other groups that share the same moieties in their structures. For example, the effect on alpha hydrogen acidity by replacing one hydrogen atom with a methyl group can be calculated from comparing ketones **34** and **31**. Similarly ketone **5** is qualified to enter the PSET as long as ketone **1** remains in the training set (TSET). In other words, those that have unique structures were not considered for inclusion in the PSET. Based on these guidelines and the size of each group, ketone **5** from group A, ketone **10** from group B, and ketones **20**, **30**, and **35** from group C were selected for the PSET. The remaining fifty ketones were used as the training set to generate the QSPR model.

2.2. Computational Details

The structures of the selected fifty-five ketones were sketched and energy-minimized by SYBYL 8.1 [26] using the Tripos Force Field and Gasteiger-Hückel charges with a 0.05 kcal/(mol \times Å) energy termination gradient, dielectric constant $\epsilon = 1.0$, and an 8.0 Å nonbonded interaction (NB) cutoff. Molecular descriptors used for describing the acidity and generating the QSPR equation were calculated for each molecule using MDL QSAR version 2.2.0.365. The stepwise multiple linear regression method was used to build the model. The number of descriptors (n) in the equation was limited to no more than the square root of the number of ketones in the TSET minus 2 ($n \leq (\text{TSET})^{0.5} - 2$), which is 5 in this case. The following criteria were considered when selecting the descriptors: 1) higher F -statistic value introduced first; 2) absolute t -statistic value not less than 3.5; 3) descriptors should not be highly correlated with each other (intercorrelation coefficient below 0.7); 4) no descriptors with only a few non-zero (or different) values.

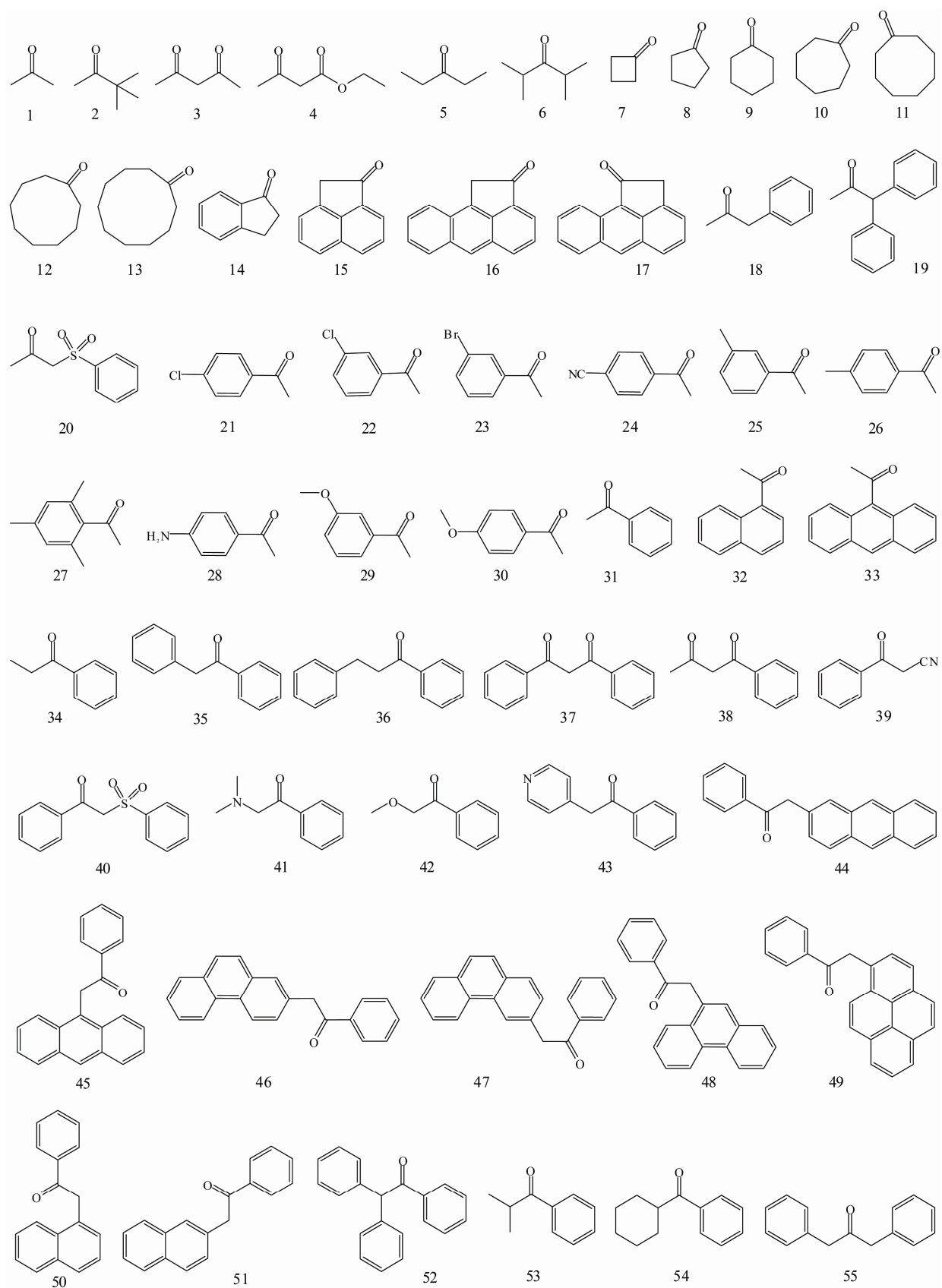


Figure 2. Structures of the ketones employed in this study.

3. RESULTS AND DISCUSSION

3.1. QSPR Model Building

A QSPR model for predicting the pK_a of ketones in DMSO was generated using the method described above by utilizing the fifty ketones in the training set:

$$pK_a = -15.95 \times \text{Hmin} - 6.931 \times \text{SdssC_acnt} + 5.091 \times \text{Qv} - 23.49 \times \text{MaxNeg} - 1.434 \times \text{nelem} + 29.0743 \quad (1)$$

$$(n = 50, r^2 = 0.86, q^2 = 0.80, s = 1.92, F = 54.19, P = 2.3E-5)$$

where Hmin is an atom-type electrotopological state (E-state) descriptor encoding the minimum hydrogen E-state value (HS) in a molecule [27]. The calculation of a HS (HS_i) is given as follows:

$$\text{HS}_i = 0.2 + (\delta_i^v - \delta_i) / N^2 - \sum_j [(I_j - I_i) / r_{ij}^2]$$

where δ^v is all the valence electrons associated with the atom i ; δ is the non-hydrogen bonded sigma electron count; N is the principal quantum number; the intrinsic state value I is defined as:

$$I = [(2/N)^2 \cdot \delta^v + 1] / \delta$$

The HS tends to be the smallest for hydrogen which is bonded to an element of low electronegativity. SdssC_acnt is an atom-type count that represents the number of all non-aromatic sp^2 hybridized carbons ($=C<$) in the molecule (such as $O=C<$, $S=C<$). Qv is a whole-molecule E-state polarity index that decreases as the polarity increases [27]. It encodes the existence of heteroatoms and polar functional groups and is given by:

$$Qv = \sum_i I_i^{\max} \cdot \sum_i I_i^{\text{alkane}} / (\sum_i I_i)^2$$

where I_i^{\max} = the intrinsic state value of the atom where the following replacements have been made: 1) all terminal atoms replaced by -F; 2) all divalent atoms replaced by -O-; 3) all trivalent atoms replaced by >N-; 4) all quaternary atoms replaced by >C<. MaxNeg reflects the largest partial negative charge over the atoms in a molecule. Nelem is the total number of different elements in the molecule.

The statistical parameters that describe the quality of the regression **Eq.1** such as squared correlation coefficient (r^2), predictive squared correlation coefficient (q^2), standard error of estimation (s), Fisher's F -value using the F statistic (F), and P -value using the F statistic (P) are given below **Eq.1**.

As shown in the plot of the calculated pK_a against experimental pK_a (**Figure 3**), **Eq.1** poorly predicted the pK_a of four ketones, which are **28**, **37**, **39**, and **52**, especially for ketone **39**, with an absolute residual between the prediction and experimental data of nearly 6 log units. There

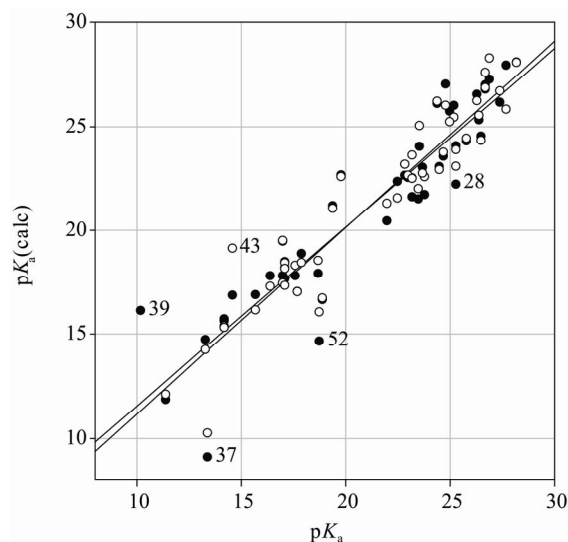


Figure 3. Plot of calculated pK_a vs. experimental pK_a for **Eq.1** (●) and **Eq.2** (○).

are two ketones containing cyano groups, **24** and **39**. The influence of the cyano group on the acidity of **39** is more profound than it is on **24**, since the cyano group is directly attached to the methylene group in **39**. However, among the five descriptors, only Hmin partially reflected this distance difference between the cyano group and the alpha hydrogen atoms. This could be the cause of poor acidity prediction for **39**. Since these descriptors are favorable for most of the members in the TSET, a second model was thus built without ketone **39** to test this hypothesis by using the same method mentioned above to give **Eq.2**:

$$pK_a = -12.46 \times \text{Hmin} - 6.337 \times \text{SdssC_acnt} + 7.187 \times \text{Qv} - 23.43 \times \text{MaxNeg} - 2.634 \times \text{xc3} + 21.2905 \quad (2)$$

$$(n = 49, r^2 = 0.90, q^2 = 0.85, s = 1.58, F = 74.54, P = 1.6E-5)$$

where xc3 is the simple 3rd order chi cluster connectivity index and it is defined for a single branch point ("Y" type) and encodes the number and branching environments of such points [28]. For example, acetone (**1**) has only one such a branching point, whereas 3,3-dimethylbutan-2-one (**2**) has five. A more detailed illustration of the xc3 calculation is found in **Figure 4** for ketones **1** and **2**.

By leaving out ketone **39**, not only the r^2 value is improved, but more importantly, the cross-validation indicated a more robust model. However, the prediction of ketone **43** by **Eq.2** was far from acceptable, with a residual of -4.5 log units (**Figure 3**). The nitrogen atom in the pyridine ring of **43** has the same electron-withdrawing effect as a nitro group, but none of the five descriptors can reveal this feature. Additionally, the absolute t -

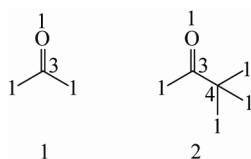


Figure 4. Illustration of the xc3 descriptor calculation. The digit (δ) near each atom indicates the number of non-hydrogen atoms that is attached to it. The xc3 descriptor for each molecule is then calculated by the following function:

$$xc3 = \sum (\delta_i \delta_j \delta_k \delta_l)^{-0.5}$$

For ketone **1**, $xc3 = (1 \times 3 \times 1 \times 1)^{-0.5} = 0.57735$, whereas for ketone **2**, $xc3 = (1 \times 3 \times 1 \times 4)^{-0.5} + (1 \times 1 \times 4 \times 1)^{-0.5} + (1 \times 1 \times 4 \times 3)^{-0.5} + (1 \times 1 \times 4 \times 3)^{-0.5} + (1 \times 1 \times 4 \times 3)^{-0.5} = 1.6547$.

statistic values for both MaxNeg and xc3 in **Eq.2** are below 3.5 (data not shown). This is important because the t -statistic indicates the significance of each individual descriptor in the linear regression equation. A third model (**Eq.3**) was thus built after leaving out **43** to improve the t -statistic by following the same procedure stated above as:

$$pK_a = -11.42 \times Hmin - 6.365 \times SdssC_acnt + 7.487 \times Qv - 3.274 \times xc3 - 24.12 \times MaxNeg + 20.5577 \quad (3)$$

($n = 48$, $r^2 = 0.91$, $q^2 = 0.86$, $s = 1.42$, $F = 89.54$, $P = 3.6E-6$)

Although there was not much difference for r^2 and q^2 between **Eq.2** and **Eq.3**, the F statistic is modestly improved along with the t -statistic for each descriptor (**Table 1**). Among the five descriptors, the $|t|$ values for Hmin, SdssC_acnt and Qv are each above 4.0 and all $|t|$ values are ≥ 3.5 , which implies that these descriptors contribute significantly to the model. Furthermore, to check the validity of the selected descriptor set (Hmin, SdssC_acnt, Qv, xc3, and MaxNeg), 100 randomizations of the dependent variable values among the training set were carried out. Values of the multiple r^2 were computed for each of corresponding regressions. The mean of r^2 was 0.11. The mean square deviation of r^2 value was 0.058, indicating that the model was not arrived at merely by chance.

High F , low s , a P value near 0, and r^2 and q^2 values near 1 all indicate a reasonable QSPR model. In general, a QSPR model is considered significant when $P < 0.001$ [29]. The established QSPR model (**Eq.3**) thus shows a significant statistical quality, both in a reliability ($r^2 = 0.91$) and a predictability ($q^2 = 0.86$). The following discussion will therefore focus only on **Eq.3**.

A correlation plot of the calculated pK_a against experimental pK_a for **Eq.3** is shown in **Figure 5**. The calculated pK_a values for each ketone in the TSET and correlation matrix for the five descriptors can be found in **Tables 2** and **3** respectively. The absolute value of the

highest intercorrelation coefficient between any two of the five descriptors in **Table 3** is 0.6345 (Hmin to xc3), which is below 0.7. As shown in **Table 2**, the residuals between calculated pK_a and experimental pK_a for over 70% of the TSET (thirty-four ketones out of forty-eight in total) are smaller than standard error of estimation. In general, the equation gave better prediction for group C2 (CH_3COR), followed by group B (cyclic ketones) and group C3 ($PhCOCH_2R$). This is not surprising since the group sizes of these three groups are much larger than others, which let them take a leading role in selecting descriptors that are more favorable for them. In most of the groups, pK_a values for ketones that show a distinguishable structure than other members are not very well predicted by the model (e.g. ketones **2**, **15**, and **28**). In addition, it seems that the effect of substitutions is not additive: if two identical functional groups are present in a molecule, the pK_a doesn't simply change twice as much compared to a molecule containing only one of such. This is illustrated by the two series of ketones **1**→**18**→**55** and **3**→**38**→**37**.

3.2. Interpretation of Ketone Acidity

As pointed out by Bordwell *et al.* [25], acidity changes observed for ketones by different substituents are mainly

Table 1. Mean, standard deviation (SD) and t -statistic (t) for variables in **Eq.3**.

	pK_a	Qv	MaxNeg	xc3	SdssC_acnt	Hmin
Mean	21.67	1.13	-0.419	0.798	1.064	0.701
SD	4.47	0.19	0.0370	0.339	0.247	0.227
t	NA ^a	4.62	-3.472	-3.557	-7.467	-7.474

^aNA = Not applicable.

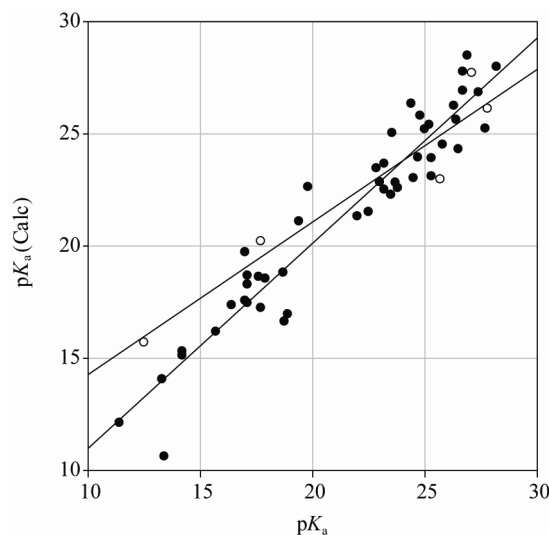


Figure 5. Plot of calculated pK_a (**Eq.3**) vs. experimental pK_a for training set (\bullet) and predicting set (\circ).

Table 2. Calculated descriptor and pK_a values (Eq.3) for ketones employed in this study.

Compd.	Qv	MaxNeg	xc3	SdssC_ aent	Hmin	pK_a (calc)	pK_a (exp)	Residual	Set Type	
A	1	1.18837	-0.363773	0.57735	1	0.495833	24.3131	26.5	2.187	TSET
	2	1.71171	-0.362911	1.6547	1	0.447569	25.2345	27.7	2.466	TSET
	3	0.921785	-0.362257	0.816497	2	0.589958	14.0578	13.3	-0.7578	TSET
	4	0.910359	-0.362154	0.696923	2	0.523632	15.1185	14.2	-0.9185	TSET
	5	1.38932	-0.363201	0.288675	1	0.411125	27.7148	27.1	-0.6148	PSET
	6	1.72295	-0.362625	0.859117	1	0.441347	27.9874	28.2	0.2126	TSET
B	7	0.926293	-0.363181	0.288675	1	0.440625	23.912	25.3	1.388	TSET
	8	1.04132	-0.363182	0.288675	1	0.462847	24.5195	25.8	1.28	TSET
	9	1.13987	-0.363182	0.288675	1	0.430569	25.6259	26.4	0.7741	TSET
	10	1.22499	-0.363182	0.288675	1	0.443069	26.1192	27.8	1.6808	PSET
	11	1.29913	-0.363182	0.288675	1	0.427722	26.8508	27.4	0.5492	TSET
	12	1.42173	-0.363182	0.288675	1	0.427536	27.7709	26.7	-1.071	TSET
	13	1.51875	-0.363182	0.288675	1	0.428395	28.4874	26.9	-1.587	TSET
	14	0.992438	-0.444381	0.538452	1	0.677375	22.845	23	0.155	TSET
	15	0.979818	-0.449867	0.816229	1	0.874486	19.723	17	-2.723	TSET
	16	1.01571	-0.452228	1.08839	1	0.951111	18.2827	17.1	-1.183	TSET
	17	1.01571	-0.452207	1.09401	1	0.92667	18.5429	17.9	-0.6429	TSET
C1	18	1.13081	-0.363127	0.612372	1	0.594142	22.6294	19.8	-2.829	TSET
	19	1.14996	-0.362471	0.777778	1	0.692451	21.093	19.4	-1.693	TSET
	20	0.817223	-0.360885	1.60306	1	0.706621	15.6978	12.5	-3.1978	PSET
C2	21	1.03258	-0.440311	0.788675	1	0.646047	22.5859	23.8	1.214	TSET
	22	1.03258	-0.440311	0.788675	1	0.652433	22.513	23.2	0.687	TSET
	23	1.14927	-0.448196	0.788675	1	0.644333	23.6693	23.2	-0.4693	TSET
	24	1.04142	-0.457447	0.788675	1	0.642647	21.3198	22	0.6802	TSET
	25	1.22222	-0.444696	0.788675	1	0.550583	25.2015	25	-0.2015	TSET
	26	1.22222	-0.444696	0.788675	1	0.532923	25.4031	25.2	-0.2031	TSET
	27	1.50044	-0.439044	1.20452	1	0.548923	25.8057	24.8	-1.006	TSET
	28	1.04142	-0.457447	0.788675	1	0.642647	23.1043	25.3	2.196	TSET
	29	1.09917	-0.435697	0.704124	1	0.665776	23.0247	24.5	1.475	TSET
	30	1.09917	-0.428801	0.704124	1	0.655976	22.9683	25.7	2.7317	PSET
	31	1.0651	-0.439062	0.5	1	0.628361	23.9463	24.7	0.7537	TSET
	32	1.09295	-0.445689	0.772166	1	0.680545	22.8278	23.7	0.8722	TSET
	33	1.10829	-0.448904	1.04994	1	0.732728	21.515	22.5	0.985	TSET
C3	34	1.13081	-0.43878	0.402369	1	0.488934	26.343	24.4	-1.943	TSET
	35	1.05257	-0.438425	0.606493	1	0.915111	20.2119	17.7	-2.5119	PSET
	36	1.09714	-0.444395	0.606493	1	0.775684	22.2841	23.5	1.216	TSET
	37	0.908075	-0.443194	0.804738	2	1.05672	10.6166	13.4	2.783	TSET
	38	0.917396	-0.443207	0.810617	2	0.650357	15.3071	14.2	-1.107	TSET
	40	0.830623	-0.44186	1.59718	1	1.20175	12.1195	11.4	-0.7195	TSET
	41	1.25082	-0.443047	0.810617	1	0.573968	25.037	23.55	-1.487	TSET
	42	1.02249	-0.441973	0.402369	1	0.676746	23.4647	22.85	-0.6147	TSET
	44	1.08644	-0.438425	1.27316	1	0.996096	17.3608	16.4	-0.9608	TSET
	45	1.08644	-0.438425	1.14395	1	1.06836	16.9588	18.9	1.941	TSET
	46	1.08644	-0.438425	1.21199	1	0.996096	17.5611	17	-0.5611	TSET
	47	1.08644	-0.438425	1.21199	1	1.00519	17.4573	17.1	-0.3573	TSET
C3	48	1.08644	-0.438425	1.14395	1	1.04392	17.2378	17.7	0.4622	TSET
	49	1.08031	-0.438425	1.42734	1	1.05132	16.1796	15.7	-0.4796	TSET
	50	1.0727	-0.438425	0.871785	1	0.991736	18.6218	17.6	-1.022	TSET
	51	1.0727	-0.438425	0.939826	1	0.967295	18.6781	17.1	-1.578	TSET
C4	52	1.08506	-0.43778	0.803561	1	1.19256	16.6292	18.75	2.121	TSET
	53	1.26181	-0.444132	0.69245	1	0.511156	26.2495	26.3	0.05051	TSET
	54	1.2096	-0.444113	0.525783	1	0.465708	26.9227	26.7	-0.2227	TSET
C5	55	1.09714	-0.36248	0.696923	1	0.880892	18.8109	18.7	-0.1109	TSET

Table 3. Correlation matrix (r values) for descriptors in Eq.3.

	pK_a	Qv	MaxNeg	xc3	SdssC_acnt	Hmin
pK_a	1					
Qv	0.628	1				
MaxNeg	0.2247	0.3353	1			
xc3	-0.5299	0.0181	-0.3723	1		
SdssC_acnt	-0.4585	-0.3003	0.21	-0.01812	1	
Hmin	-0.7548	-0.4545	-0.5028	0.6345	-0.1316	1

a balance among three effects: 1) steric effect on resonance and solvation of the anion; 2) stabilizing effect on the enolate ion either through delocalization or induction; and 3) lone-pair-lone-pair electron repulsions.

Among the five descriptors in Eq.3, Qv is positively correlated with pK_a , and the other four descriptors contribute negatively to the pK_a value, especially Hmin. Being developed to encode both the electronic and steric attributes of atoms in a molecule, two indices might be expected to successfully capture the features influencing pK_a as noted in the previous paragraph. Indeed, the E-state index Hmin was selected as one of the most significant descriptors in the model. As shown in Table 2, except for **44** and **46**, the ketones have unique Hmin values. Furthermore, Hmin is significantly inversely correlated with pK_a (Table 3) and ketones that have Hmin values larger than 0.8 (e.g. **15** - **17**, **37**, **40**, etc.) tend to be more acidic (observed pK_a values are among 11.4 to 18.9). Therefore these compounds are well predicted. A more specific example could be illustrated by comparing **21** to **22**. Having identical values for the other four descriptors, the differences in their Hmin properties decided the variations in their pK_a values. The *meta*-chloro group in **22** generates a stronger induced electron withdrawing effect on the enolate ion than the *para*-chloro group in **21** does, and hence **22** is more acidic than **21**. The steric effect reflected by Hmin is exemplified by comparing **16** to **17** (although **16** and **17** don't have exactly matching MaxNeg and xc3 values, the role of both descriptors is quite insignificant comparing to Hmin, in this case). As suggested by their 3D structures, atoms C2, C2a and C3 are not in the same plane as the C4-C10 atoms, and this generates a more hindered environment for the methylene group in **17** than the one in **16**. Since steric effect contributes negatively to the acidity, **16** is more acidic than **17**. On the other hand, Hmin seemed not sufficient to evaluate the acidity of polycyclic aromatic ketones. For example, although **48** is more acidic than **45**, **45** shows a higher Hmin value than **48** in spite of the fairly strong inverse relationship between pK_a and Hmin (see Table 2).

The impact of SdssC_acnt on ketones acidities can be easily observed for **3** - **4** and **37** - **38** compared to the rest of the ketones. The SdssC_acnt values for these four ke-

tones are 2, two times of those for other ketones (Table 2), which makes them quite acidic as demonstrated by the lower pK_a of **3** and **38** than **1** and **31** respectively. This was due to the additional electron withdrawing effect contributed by the second carbonyl group.

Not surprisingly, geometric and positional isomers have the same Qv values (for example, **44** - **48**). Among the forty-eight ketones, only **3** - **4**, **7**, **37** - **38**, and **40** have Qv values less than 0.95. It is not difficult to understand that **3** - **4**, and **37** - **38** are more polar due to the presence of second carbonyl groups. Similarly, the sulfonyl group in **40** makes the molecule more polar. These moieties are electron-withdrawing groups, which have a stabilizing effect on the enolate ion through their inductive stabilizing effect, and therefore the ketones containing these moieties are more acidic. Cyclobutanone **7** is the most polar compound in the aliphatic cycloketones category, and has the lowest pK_a amongst them. On the other hand this descriptor as well as others would not be able to distinguish among different conformation of ketones (such as *cis* vs. *trans*, *chair* vs. *boat*) and its influence to the acidity of the ketones.

Descriptor xc3 is an indicator of the degree of third order branching, and thus implicates the effect of substitution in a molecule. A molecule that is relatively compact at some point(s) will have a higher xc3 value. There are eleven ketones of which xc3 values are larger than 1 in the TSET. A critical aspect will have to be considered when xc3 is involved to explain the acidities of ketones in addition to the hindrance effect it causes, that is, whether the branching at certain position(s) can stabilize the enolate ion. This factor is perfectly demonstrated by ketones **16** - **17**, **44** - **49** and **40**. The enolate ion for ketone **40** is stabilized through inducing effect by sulfonyl group, whereas the delocalization of the anion (the negative charge is distributed to the phenyl rings through resonance) is achieved for ketones **16** - **17** and **44** - **49**. In contrast, the increased branching in ketone **2** can't attain either of the above effects, and this counts for its decreased acidity, compared to the less branching counterpart **1**. For ketone **2**, the steric hindrance for the solvation of its anion is the determining factor for pK_a .

MaxNeg is a charge index. Most of the ketones carry similar MaxNeg values. Interestingly, no matter what the size of the cycloketones is, they share the same MaxNeg value. More importantly, the MaxNeg values for ketones in which the carbonyl groups are directly attached to a phenyl ring are around -0.44. The MaxNeg values for the remaining ketones are approximately -0.36. The repulsion between the negatively charged carbonyl oxygen and the aromatic *pi*-bonds, which is unfavorable for the stability of the enolate ion, might be the reason that MaxNeg contributes negatively to the pK_a of the ketones, a similar effect that lone-pair-lone-pair electron repul-

sions have.

3.3. Ketone Acidity Prediction

To further validate the built QSPR model, the generated regression **Eq.3** was used to predict the pK_a of the five ketones in the external prediction set (**Table 2**). A correlation plot of the calculated pK_a against experimental pK_a for the PSET is shown in **Figure 5**. A linear regression was performed for the calculated pK_a and the experimental pK_a . The statistics r^2 and s are 0.92 and 1.618 respectively, which was considered to be satisfactory. **Table 2** showed that the QSPR model estimated the acidity for those ketones with acceptable values while the best prediction was obtained for compound **5**. Compound **20** is one of the only two ketones that carry a sulfonyl group, providing an explanation for the relatively poor prediction.

The pK_a shows a parabolic relationship with the ring size of cycloketones 4 - 9, with the pK_a of cycloheptanone **10** being the highest. However, cycloheptanone **10** wasn't in the TSET when the model was built to reveal this characteristic and hence none of the five descriptors in the QSPR model **Eq.3** can actually reflect this particular feature of cycloketones. Having a small number of members among the whole training set also likely confounded the prediction of the relative acidities of ketones in this series, although the residuals for most cycloketones are acceptable.

4. CONCLUSION

Ketones are important in both biochemistry and organic chemistry, and information about their pK_a properties will be beneficial for both biologists and chemists. The direct measurements of pK_a of ketones are not available due to their extremely weak acidity. Hence, a QSPR model which can be used to predict the acidities of ketones is highly desirable. Fifty-five ketones of which the pK_a in DMSO were determined using the method developed by Bordwell were used to build such a QSPR model. By leaving out two ketones (**39** and **43**) that show unique structures from others, the training set of forty-eight ketones in three main classes covering most functional groups with an overall pK_a in DMSO ranging from 11.4 to 28.2 is very well described by the statistically significant regression **Eq.3** ($r^2 = 0.91$, $q^2 = 0.86$, $s = 1.42$). Steps have been taken to ensure the quality of the generated QSPR model in this paper. Importantly, the five descriptors used to build the model are largely chemically intuitive and in agreement with the proposed theory that describes the acidity of ketones, which further strengthened the significance of the model. Moreover, the QSPR model can reasonably predict the acidity of the five ketones in the external prediction set ($r^2 = 0.92$, $s =$

1.618). We anticipate that the model obtained will be useful for prediction of ketone acidity that may be related to their reactivity, reaction mechanism, and possibly some biophysical properties in biological systems.

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