

Correlation for the solubilities of pharmaceutical compounds in supercritical carbon dioxide

Chie-Shaan Su, Yan-Ping Chen*

Department of Chemical Engineering, National Taiwan University, Taipei, Taiwan, ROC

Received 23 December 2006; received in revised form 28 February 2007; accepted 1 March 2007

Available online 6 March 2007

Abstract

The solid solubilities of pharmaceutical compounds in supercritical carbon dioxide were correlated using the regular solution model with the Flory–Huggins equation. The pharmaceutical compounds include steroids, antioxidants, antibiotics, analgesics and specific functional drugs. The molar volumes of these solid solutes in supercritical carbon dioxide were taken as the empirical parameters in this study. They were optimally fitted for each pharmaceutical compound using the experimental solid solubility data from literature. The logarithms of the molar volumes of these solutes were then correlated as a linear function of the logarithms of the densities for supercritical carbon dioxide. With one or two parameters in this linear equation, satisfactory solid solubilities were calculated that were comparable to those from the commonly used semi-empirical equations with more adjustable parameters. The parameters of this model were further generalized as a function of the properties of the pharmaceutical compounds. It was observed that the prediction of solubilities of pharmaceutical compounds in supercritical carbon dioxide was within acceptable accuracy for more than 50% of the systems investigated in this study.

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Keywords: Correlation; Solid solubility; Pharmaceutical compounds; Supercritical carbon dioxide

1. Introduction

Many applications of supercritical fluid technology have continuously been developed for the processing of food, pharmaceutical, polymer and specific chemicals [1,2]. Supercritical fluid technology for pharmaceutical compounds includes the modifications of particle size, shape, morphology and surface structure. It maximizes the drug efficiency and leads to a better option than conventional manufacturing processes. Carbon dioxide is the most commonly used supercritical fluid owing to its mild critical properties, nontoxic and inflammatory characteristics. There are several methods of pharmaceutical particle formation using supercritical CO₂, such as RESS, SAS, SEDS and PGSS [3]. The major criterion for choosing different processes depends on the solubility of pharmaceutical compound in supercritical CO₂. Experimental measurements of the solubilities of these substances in supercritical CO₂ provided essential information for engineering processing. Increasing data are

appearing in recent literature and it was the purpose of this study to develop a useful correlation based on these data of pharmaceutical compounds.

Calculations for the solubilities of pharmaceutical solid compounds were presented using equation of state, solution model, and semi-empirical equation methods. The equation of state approach was limited by the uncertain critical properties and sublimation pressures of complex pharmaceutical molecules. Semi-empirical equations were mostly often employed in literature. For example, Chrastil [4] derived an equation that was based on molecular association. Mendez-Santiago and Teja [5] developed a relationship for solid solubility that incorporated the Clausius–Clapeyron equation. Zhong et al. [6] proposed a model that the solute–solvent clusters were in chemical equilibrium with the free solute and solvent molecules. These correlation equations contained three or four constants that were empirically adjusted for each pure compound. No generalization of these parameters with the solute properties was attempted.

An alternative and feasible method for correlating the solubilities of complex pharmaceutical compounds in supercritical CO₂ was the application of solution model. In this approach, the supercritical CO₂ was taken as the liquid solvent, and an

* Corresponding author. Fax: +886 2 2362 3040.

E-mail address: yphen@ntu.edu.tw (Y.-P. Chen).

infinite dilution activity coefficient was employed to account for the nonideal behavior of solid–liquid equilibrium. Iwai et al. [7] correlated the high boiling point components in supercritical CO₂ using the regular solution model coupled with the Flory–Huggins term. Bush and Eckert [8] presented a predictive model based on linear solvation energy relationship (LSER), but relatively large error around 100% existed in their model for polar compounds. In our previous study [9], solid solubility of biological compounds including steroids, antioxidants and xanthenes in supercritical carbon dioxide were correlated as an extension of the model of Iwai et al. [7]. In recent years, more experimental data have been published in literature for the solubilities of pharmaceutical compounds in supercritical CO₂. It was our intention to revise our previous correlation and investigate a better generalization of the model parameters. This correlation involved new experimental data of complex pharmaceutical components, and provided useful information for engineering applications. The feasibility for the prediction of the solubilities of pharmaceutical compounds in supercritical CO₂ was also investigated.

2. Method of calculation

Applying the solution model, supercritical CO₂ was treated as an expanded liquid. The equilibrium solubility of a solid solute (component 2), y_2 , in supercritical CO₂ (component 1) was expressed as:

$$y_2 = \frac{f_2^s}{\gamma_2^\infty f_2^l} \quad (1)$$

where γ_2^∞ was the activity coefficient of the solid solute at infinite dilution, f_2^s and f_2^l were the fugacities of pure solute in solid phase and supercritical phase, respectively. The ratio of these two fugacities was approximated as:

$$\ln \frac{f_2^s}{f_2^l} = \frac{\Delta H_2^f}{R} \left(\frac{1}{T_{2,m}} - \frac{1}{T} \right) \quad (2)$$

where ΔH_2^f was the molar heat of fusion of the pharmaceutical compound, $T_{2,m}$ was its melting temperature, and R was the gas constant. The infinite dilution activity coefficient γ_2^∞ was expressed by the modified regular solution model coupled with the Flory–Huggins term:

$$\ln \gamma_2^\infty = \left(\frac{v_2}{RT} \right) (\delta_1 - \delta_2)^2 + 1 - \left(\frac{v_2}{v_1} \right) + \ln \left(\frac{v_2}{v_1} \right) \quad (3)$$

where δ was the solubility parameter, and v was the molar volume:

$$\delta_i = \left(\frac{\Delta U_i^{\text{vap}}}{v_i} \right)^{0.5} \quad (4)$$

where ΔU^{vap} in Eq. (4) was the molar internal energy of vaporization. Incorporating this infinite dilution activity coefficient and the fugacity ratio, the solubility of solid solute in supercrit-

ical phase was:

$$\ln y_2 = \frac{\Delta H_2^f}{R} \left(\frac{1}{T_{2,m}} - \frac{1}{T} \right) - \left(\frac{v_2}{RT} \right) (\delta_1 - \delta_2)^2 - 1 + \left(\frac{v_2}{v_1} \right) - \ln \left(\frac{v_2}{v_1} \right) \quad (5)$$

The heat of fusion, ΔH_2^f , in Eq. (5) was either taken from literature or estimated by the method of Yalkowsky [10]. The value of δ_1 was directly calculated using the Peng–Robinson equation of state [11]. The molar volume of supercritical carbon dioxide, v_1 , was estimated by Jacobsen and Stewart EOS with 32 constants regressed by Ely et al. [12]. δ_2 was determined using v_2 and ΔU_2^{vap} , and the latter was estimated by the group contribution method developed by Fedor [13]. v_2 was taken as the adjustable parameter and was regressed for each solid solute by minimizing the objective function over all data points j :

$$\text{obj.} = \sum \frac{|y_{2,j}^{\text{cal}} - y_{2,j}^{\text{exp}}|}{y_{2,j}^{\text{exp}}} \quad (6)$$

The superscripts cal and exp denoted the calculated and experimental results, respectively.

To compare the calculated solubilities of pharmaceutical compounds from the solution model with those from semi-empirical correlation, the following equations presented in literature were employed. The Chrastil equation [4] was:

$$\ln c = k \ln \rho_1 + \frac{a}{T} + b \quad (7)$$

where c was the concentration of solute in supercritical fluid with the unit of (kg/m³), ρ_1 the density (kg/m³) of supercritical CO₂, k , a , and b were three adjustable parameters. The Mendez-Santiago and Teja equation [5] was:

$$T \ln(y_2 P) = a + b \rho_1 + c T \quad (8)$$

where a , b and c were also adjustable parameters. Finally, a simple two-parameter (a and b) equation cited in previous literature [14–16] was applied:

$$\ln y_2 = a \ln \rho_1 + b \quad (9)$$

The adjustable parameters in these semi-empirical equations were also optimally fitted in this study by minimizing the objective function of Eq. (6):

3. Results and discussion

Solid solubilities in supercritical carbon dioxide for 60 pharmaceutical compounds containing steroids, antioxidants, antibiotics, analgesics and specific functional drugs were correlated in this study. Table 1 lists these pharmaceutical compounds and their thermodynamic properties. The data sources for these pharmaceutical components are shown in Table 2. In this study, the solid solubilities were correlated using Eq. (5) where the molar volume v_2 of the solid solute in supercritical CO₂ was taken as an adjustable parameter. The optimally fitted values of

Table 1
Data references and physical properties for pharmaceutical compounds in this study

Compound	Formula	M_w (g/mol)	T range (K)	P range (MPa)	Data points	T_m (K)	ΔH^f (kJ/mol)	$\Delta U^{\text{vap},298.15\text{K}}$ (kJ/mol)	Data references ^a
Amical-48	C ₈ H ₈ O ₂ SI ₂	422.02	318–338	10–30	18	453.15	25.60	97.38	12
9,10-Anthraquinone	C ₁₄ H ₈ O ₂	208.21	308–318	8–31	17	559.15	31.59	99.62	31
Artemisinin	C ₁₅ H ₂₂ O ₅	282.33	310–338	10–27	36	429.65	24.27	88.89	1
Ascorbyl palmitate	C ₂₂ H ₃₈ O ₇	414.53	308–313	13–20	8	389.65	79.09	206.42	25
Aspirin	C ₉ H ₈ O ₄	180.16	308–328	12–25	24	407.36	23.01	82.24	9
Beclomethasone dipropionate	C ₂₈ H ₃₇ ClO ₇	521.04	338–358	21–39	21	391.15	26.19	211.15	21
Benzocaine	C ₉ H ₁₁ NO ₂	165.19	308–348	12–36	40	363.05	20.51	67.80	8
Bisacodyl	C ₂₂ H ₁₉ NO ₄	361.39	308–348	12–36	39	408.27	23.06	152.00	3
(Rac) Boc-piperazine	C ₁₄ H ₂₇ N ₃ O ₃	285.38	308–328	9–20	19	381.05	21.52	106.73	5
(S) Boc-piperazine	C ₁₄ H ₂₇ N ₃ O ₃	285.38	308–328	9–20	21	376.33	21.26	106.73	5
Budesonide	C ₂₅ H ₃₄ O ₆	430.53	338–358	21–39	21	499.65	28.22	197.13	21
Caffeine	C ₈ H ₁₀ N ₄ O ₂	194.19	313–368	8–35	56	510.28	22.52	88.14	16, 34, 35
Carbamazepine	C ₁₅ H ₁₂ N ₂ O	236.27	308–348	12–36	39	464.22	26.22	107.61	20
Chlorothalonil	C ₈ Cl ₄ N ₂	265.91	318–338	10–30	23	523.40	29.56	129.16	12
Cholesterol	C ₂₇ H ₄₆ O	386.65	313–333	10–25	24	421.16	28.19	147.38	23, 26
Cholesteryl acetate	C ₂₉ H ₄₈ O ₂	428.69	308–328	9–24	24	387.94	25.97	140.29	23
Cholesteryl benzoate	C ₃₄ H ₅₀ O ₂	490.76	308–328	12–27	20	421.75	28.23	167.51	23
Cholesteryl butyrate	C ₃₁ H ₅₂ O ₂	456.76	308–328	10–24	20	372.15	24.91	150.16	23
Codeine	C ₁₈ H ₂₁ NO ₃	299.36	308–348	12–36	45	429.40	24.25	124.81	20
<i>p</i> -Coumaric acid	C ₉ H ₈ O ₃	164.16	313–333	9–50	29	486.48	27.48	97.95	19
DDT	C ₁₄ H ₉ Cl ₅	354.48	313–333	10–21	18	381.65	51.56	114.43	27
Diazepam	C ₁₆ H ₁₃ ClN ₂ O	284.74	308–348	12–36	45	401.90	22.70	123.66	20
7,8-Dihydroxy flavone	C ₁₅ H ₁₀ O ₄	254.24	308–318	9–25	9	558.10	34.50	153.80	7
Eflucimibe	C ₂₉ H ₄₃ NO ₂ S	469.73	308–318	8–30	20	403.15	60.72	209.97	14
Erythromycin	C ₃₇ H ₆₇ NO ₁₃	733.93	313–333	10–30	8	464.40	26.23	347.88	16
Ferulic acid	C ₁₀ H ₁₀ O ₄	194.18	313–333	12–28	18	445.15	25.14	106.00	4
Flurbiprofen	C ₁₅ H ₁₃ FO ₂	244.26	303–323	8–25	27	383.90	21.68	105.29	2
Ketoprofen	C ₁₆ H ₁₄ O ₃	254.28	313–332	9–25	25	367.15	20.74	116.96	17, 30
Medroxyprogesterone acetate	C ₂₄ H ₃₄ O ₄	386.54	308–348	11–36	48	480.60	27.15	141.73	13
Methimazole	C ₄ H ₆ N ₂ S	114.17	308–348	12–36	40	418.82	23.66	49.02	15
Methyl gallate	C ₈ H ₈ O ₅	184.15	313–333	10–50	27	475.32	26.85	143.99	18
Methylparaben	C ₈ H ₈ O ₃	152.15	308–348	12–36	40	402.27	24.31	84.41	3
Metronidazole benzoate	C ₁₃ H ₁₃ N ₃ O ₄	275.27	308–348	12–36	40	375.15	21.19	123.66	8
1,4-Naphthoquinone	C ₁₀ H ₆ O ₂	158.15	318–343	10–36	18	399.15	15.41	77.99	28
Naproxen	C ₁₄ H ₁₄ O ₃	230.26	308–348	9–36	58	427.24	31.50	97.36	8, 24
Nifedipine	C ₁₇ H ₁₈ N ₂ O ₆	346.33	333–373	13–30	29	446.00	25.20	215.52	29
Nimesulide	C ₁₃ H ₁₂ N ₂ O ₅ S	308.31	313–332	13–22	8	421.65	23.82	116.46	30
Nimodipine	C ₂₁ H ₂₆ N ₂ O ₇	418.44	313–333	10–25	21	398.15	26.65	153.53	6
Penicillin G	C ₁₆ H ₁₈ N ₂ O ₄ S	334.39	313–333	10–35	18	736.31 ^b	41.59	151.25	22
3,3,4,5,7-Pentahydroxy flavone	C ₁₅ H ₁₀ O ₇	302.24	308–318	10–25	8	587.15	40.90	243.17	7
Phenazopyridine	C ₁₁ H ₁₁ N ₅	213.24	308–348	12–36	45	412.15	23.28	97.19	15
Piroxicam	C ₁₅ H ₁₃ N ₃ O ₄ S	331.35	313–332	10–22	9	469.15	26.51	167.84	30
Progesterone	C ₂₁ H ₃₀ O ₂	314.46	313–333	9–24	11	400.15	26.89	119.47	32
Propranolol	C ₁₆ H ₂₁ NO ₂	259.35	308–348	12–36	45	369.15	28.57	121.21	15
Propyl gallate	C ₁₀ H ₁₂ O ₅	212.20	313–333	15–25	8	423.15	23.91	130.02	25
Protocatechualdehyde	C ₇ H ₆ O ₃	138.12	313–333	10–50	24	426.48	24.09	112.84	18
Protocatechuic acid	C ₇ H ₆ O ₄	154.12	313–333	10–50	24	472.98	26.72	119.12	18
<i>p</i> -Quinone	C ₆ H ₄ O ₂	108.09	308–318	9–29	18	390.15	22.04	58.03	31
Salicylic acid	C ₇ H ₆ O ₃	138.12	313–328	9–25	11	432.24	24.41	89.33	17
Stigmasterol	C ₂₉ H ₄₈ O	412.69	308–333	9–30	19	435.15	29.14	155.94	33
Sulfadimethoxine	C ₁₂ H ₁₄ N ₄ O ₄ S	310.33	313–333	13–49	19	475.90	26.88	139.54	10
Sulfamerazine	C ₁₁ H ₁₂ N ₄ O ₂ S	264.30	313–333	15–47	18	509.75	31.60	128.14	10
Syringic acid	C ₉ H ₁₀ O ₅	198.17	313–333	9–50	27	478.90	27.05	105.44	11
Tebuconazole	C ₁₆ H ₂₂ ClN ₃ O	307.83	323–338	10–30	12	377.85	21.34	142.40	12
Theobromine	C ₇ H ₈ N ₄ O ₂	180.16	313–353	19–35	23	620.00	41.11	90.96	35
Theophylline	C ₇ H ₈ N ₄ O ₂	180.16	313–353	20–35	24	547.50	29.71	90.96	35
Uracil	C ₄ H ₄ N ₂ O ₂	122.09	313–333	10–30	12	609.65	34.44	61.13	16
Vanillic acid	C ₈ H ₈ O ₄	168.15	313–333	9–50	28	482.61	27.26	97.38	11
Vitamin C	C ₆ H ₈ O ₆	176.12	313	13–20	4	465.15	26.28	128.49	25
Zopiclone	C ₁₇ H ₁₇ ClN ₆ O ₃	388.81	313–333	10–25	21	451.15	25.48	175.50	6

^a Data reference are listed in Table 2.

^b Melting temperature estimate from the group contribution method of Joback et al. [17].

Table 2
Data sources for solubility data

Reference number	Source
1	J. Chem. Eng. Data 48 (2003) 330–332
2	J. Chem. Eng. Data 49 (2004) 449–452
3	J. Chem. Eng. Data 48 (2003) 61–65
4	J. Chem. Eng. Data 46 (2001) 1255–1257
5	J. Chem. Eng. Data 49 (2004) 1560–1564
6	J. Chem. Eng. Data 46 (2001) 1211–1214
7	J. Chem. Eng. Data 48 (2003) 1040–1043
8	J. Chem. Eng. Data 49 (2004) 709–712
9	J. Chem. Eng. Data 49 (2004) 1323–1327
10	J. Chem. Eng. Data 44 (1999) 1222–1225
11	J. Chem. Eng. Data 49 (2004) 779–782
12	J. Chem. Eng. Data 48 (2003) 541–547
13	J. Supercrit. Fluids 30 (2004) 111–117
14	J. Supercrit. Fluids 31 (2004) 133–140
15	J. Pharm. Biomed. Anal. 32 (2003) 181–187
16	Fluid Phase Equilib. 220 (2004) 57–69
17	J. Chem. Eng. Data 45 (2000) 161–165
18	J. Supercrit. Fluids 23 (2002) 113–121
19	J. Supercrit. Fluids 27 (2003) 239–245
20	J. Chem. Eng. Data 46 (2001) 451–455
21	J. Supercrit. Fluids 33 (2005) 21–25
22	J. Supercrit. Fluids 15 (1999) 183–190
23	J. Supercrit. Fluids 30 (2004) 25–39
24	Ind. Eng. Chem. Res. 32 (1993) 1471–1481
25	J. Supercrit. Fluids 14 (1999) 139–144
26	Ind. Eng. Chem. Res. 30 (1991) 2476–2482
27	Ind. Eng. Chem. Res. 33 (1994) 2757–2763
28	J. Chem. Eng. Data 31 (1986) 204–212
29	J. Chem. Eng. Data 40 (1995) 216–220
30	J. Chem. Eng. Data 41 (1996) 1083–1086
31	J. Chem. Eng. Data 42 (1997) 463–466
32	Ind. Eng. Chem. Res. 35 (1996) 4718–4726
33	Biotechnol. Prog. 2 (1986) 29–39
34	Fluid Phase Equilib. 68 (1991) 263–280
35	Fluid Phase Equilib. 95 (1994) 215–226

In v_2 were observed as a linear function of the logarithm of density for supercritical CO₂. The similar trend was also presented by Iwai et al. [7] and our previous study [9]. Fig. 1 shows an example for the plot of the optimal $\ln v_2$ values of Diazepam, a drug for the symptomatic relief of tension and anxiety, against the logarithm of density of supercritical CO₂. A linear relationship was demonstrated and the similar results were obtained for other pharmaceutical compounds in this study. A simple correlation was then proposed:

$$\ln v_2 = \alpha \ln \rho_1 + \beta \quad (10)$$

where α and β were two temperature independent parameters for each pharmaceutical compound.

Table 3 presents the optimally fitted α and β parameters in the two-parameter model shown in Eq. (10) for various pharmaceutical compounds. Applying Eq. (10) into Eq. (5), the calculated absolute relative deviation in solid solubility (ARDY) was 16.5%. Over 70% of the systems investigated in this study had ARDY less than 20% that was within the possible experimental accuracy.

A further simplification for the model parameters was attempted by setting either α or β as a constant. An average

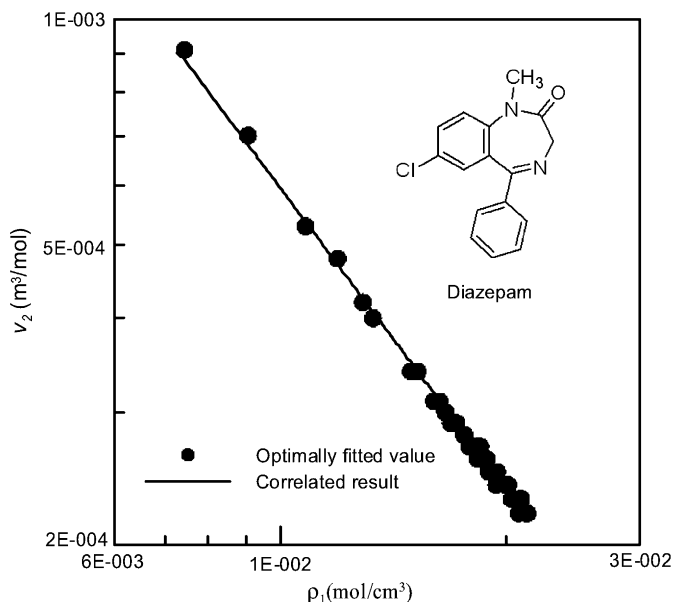


Fig. 1. Plot of the optimally fitted solid molar volume of diazepam (2) in supercritical CO₂ (1) against density of pure CO₂.

value of $\beta = -12.89$ for all compounds was applied in this study, and α was left as a single adjustable parameter. The results are shown in Table 3 as the one-parameter model with an ARDY of 23.7%. With only one parameter, over 60% of the compounds had an ARDY less than 25% that was acceptable for the complex pharmaceutical molecules.

Fig. 2 shows the calculated solubility of Vanillic acid (an antioxidant) using either the two- or one-parameter solution model in this study. The ARDY were 12% and 20% from each model, respectively. The results in Fig. 2 showed that the solution model was feasible in correlating the experimental solubility data at various temperatures. Fig. 3 presents another example for a more complex molecule of nimodipine. The ARDY from the

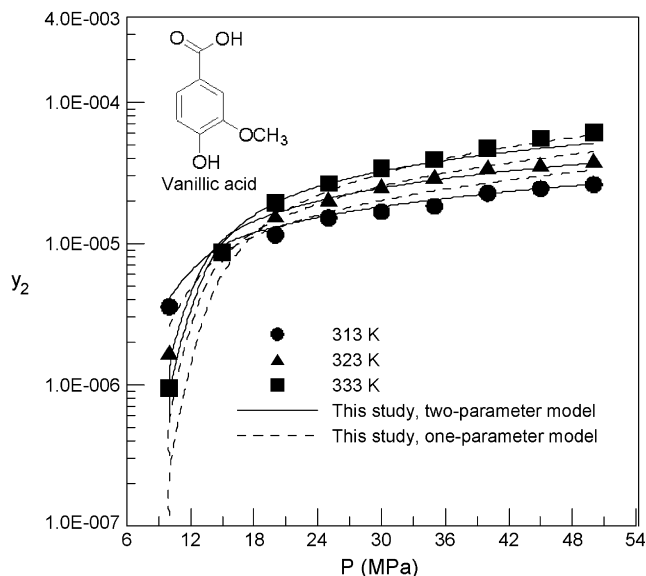


Fig. 2. Comparison of the experimental and calculated solid solubilities of vanillic acid (2) in supercritical CO₂ (1) at various temperatures.

Table 3

Calculation results for the solid solubility of pharmaceutical compounds in supercritical carbon dioxide using various models

Component	This study					Semi-empirical correlation equations		
	Two-parameter model			One-parameter model, $\beta = -12.89$		I	II	III
	α	β	ARDY (%)	α	ARDY (%)	ARDY (%)	ARDY (%)	ARDY (%)
Amical-48	-1.31	-13.72	8.1	-1.10	28.0	8.8	10.0	32.0
9,10-Anthraquinone	-0.96	-12.16	16.9	-1.14	39.2	12.8	12.3	34.1
Artemisinin	-1.20	-13.16	8.1	-1.13	9.2	6.2	6.3	23.1
Ascorbyl palmitate	-1.08	-11.83	7.2	-1.35	29.0	3.6	3.9	13.6
Aspirin	-1.24	-13.71	11.5	-1.04	19.7	5.2	4.7	30.3
Beclomethasone dipropionate	-1.25	-12.61	11.0	-1.32	13.6	10.8	10.7	24.8
Benzocaine	-1.21	-13.61	18.9	-1.03	32.3	10.7	11.7	39.1
Bisacodyl	-1.21	-12.77	16.9	-1.25	20.2	12.0	9.5	40.4
(Rac) Boc-piperazine	-1.30	-13.51	17.5	-1.14	30.2	11.6	10.5	30.6
(S) Boc-piperazine	-1.21	-13.01	24.7	-1.18	26.0	20.0	20.0	33.2
Budesonide	-1.22	-12.50	11.5	-1.31	17.3	11.5	11.2	26.7
Caffeine	-1.03	-12.57	20.4	-1.10	31.6	21.3	28.1	48.1
Carbamazepine	-1.15	-13.04	18.4	-1.11	18.8	13.4	13.8	43.6
Chlorothalonil	-1.20	-12.78	22.6	-1.23	23.7	19.8	19.6	28.1
Cholesterol	-1.21	-12.76	7.0	-1.24	8.6	6.0	6.2	24.6
Cholesteryl acetate	-1.18	-12.67	15.4	-1.24	24.9	10.1	9.3	30.8
Cholesteryl benzoate	-1.20	-12.63	14.7	-1.27	18.7	6.8	6.9	38.9
Cholesteryl butyrate	-1.16	-12.48	7.3	-1.25	28.5	6.4	7.3	29.4
Codeine	-1.23	-13.01	14.2	-1.20	16.6	12.6	11.3	42.3
<i>p</i> -Coumaric acid	-1.07	-13.17	28.5	-1.00	30.2	20.4	19.0	49.1
DDT	-1.12	-12.57	13.3	-1.20	26.5	2.8	7.6	25.4
Diazepam	-1.26	-13.14	11.3	-1.19	15.8	11.6	12.3	32.0
7,8-Dihydroxy flavone	-1.41	-13.42	13.5	-1.28	26.8	4.5	8.5	29.1
Eflucimibe	-1.24	-12.54	16.8	-1.32	33.3	14.0	13.1	43.6
Erythromycin	-1.41	-12.51	12.8	-1.50	51.2	14.4	17.8	39.6
Ferulic acid	-1.22	-13.43	6.1	-1.08	15.2	5.4	6.6	41.9
Flurbiprofen	-1.17	-13.14	21.0	-1.11	23.8	8.4	9.8	38.6
Ketoprofen	-1.16	-12.98	19.9	-1.14	20.2	11.1	11.6	40.2
Medroxyprogesterone acetate	-1.16	-12.57	17.2	-1.24	22.7	17.5	16.8	37.5
Methimazole	-0.33	-11.34	12.0	-0.72	21.1	12.7	10.7	36.6
Methyl gallate	-1.25	-13.14	13.3	-1.18	17.2	10.8	8.7	45.1
Methylparaben	-1.27	-13.73	11.5	-1.06	26.2	9.5	9.3	39.1
Metronidazole benzoate	-1.21	-12.92	16.7	-1.20	16.9	17.2	14.6	37.2
1,4-Naphthoquinone	-1.16	-13.19	12.7	-1.09	21.7	7.4	11.3	27.0
Naproxen	-1.17	-13.22	14.5	-1.09	20.2	14.6	14.9	45.5
Nifedipine	-1.22	-13.00	15.7	-1.19	18.5	14.2	16.8	35.9
Nimesulide	-1.12	-12.75	20.0	-1.15	22.4	8.0	8.3	38.4
Nimodipine	-1.31	-13.21	11.5	-1.23	25.7	7.4	9.2	31.9
Penicillin G	-1.20	-12.45	31.2	-1.31	36.7	24.9	25.0	48.7
3,3,4,5,7-Pentahydroxy flavone	-1.37	-12.71	8.0	-1.41	23.5	4.4	5.1	25.7
Phenazopyridine	-1.12	-13.09	15.8	-1.08	17.5	10.7	8.9	43.2
Piroxicam	-1.20	-12.58	8.1	-1.28	24.9	8.3	8.1	33.4
Progesterone	-1.07	-12.37	7.9	-1.19	38.1	3.2	7.9	24.6
Propranolol	-1.28	-13.32	25.5	-1.17	35.0	17.9	16.9	52.3
Propyl gallate	-1.32	-13.53	6.6	-1.16	21.6	3.4	4.8	36.9
Protocatechualdehyde	-1.22	-13.38	25.4	-1.10	29.2	18.3	15.3	49.2
Protocatechuic acid	-1.16	-13.19	27.0	-1.09	30.7	15.6	12.8	46.6
<i>p</i> -Quinone	-1.12	-13.15	21.6	-1.05	23.0	15.3	17.3	31.7
Salicylic acid	-1.24	-13.42	8.0	-1.11	24.4	4.7	4.7	26.0
Stigmasterol	-1.24	-12.87	20.7	-1.24	20.9	13.2	12.3	46.7
Sulfadimethoxine	-0.92	-11.92	39.0	-1.17	46.6	28.5	27.8	29.5
Sulfamerazine	-0.89	-11.82	29.8	-1.16	50.3	28.6	29.5	41.9
Syringic acid	-1.03	-12.71	14.8	-1.08	17.3	8.9	19.4	39.6
Tebuconazole	-1.19	-12.74	32.7	-1.23	33.9	26.9	13.7	47.5
Theobromine	-1.03	-12.65	15.9	-1.09	16.4	10.3	11.7	29.6
Theophylline	-1.11	-13.04	12.6	-1.08	13.9	5.2	5.6	20.7
Uracil	-0.89	-12.56	40.7	-0.97	43.9	29.8	28.0	54.3
Vanillic acid	-1.21	-13.40	12.0	-1.09	20.0	10.2	11.4	33.3
Vitamin C	-1.22	-12.99	2.0	-1.20	3.7	2.0	1.4	2.0
Zopiclone	-1.37	-13.24	5.4	-1.28	22.1	6.6	7.6	30.1
Grand			16.5		23.7	12.7	13.0	37.3

I: $\ln c = k \ln \rho_1 + a/T + b$ (Chrastil equation [4]); II: $T \ln (y_2 P) = a + b \rho_1 + cT$ (Mendez-Santiago and Teja equation [5]); III: $\ln y_2 = a \ln \rho_1 + b$ (model employed in Refs. [14–16]).

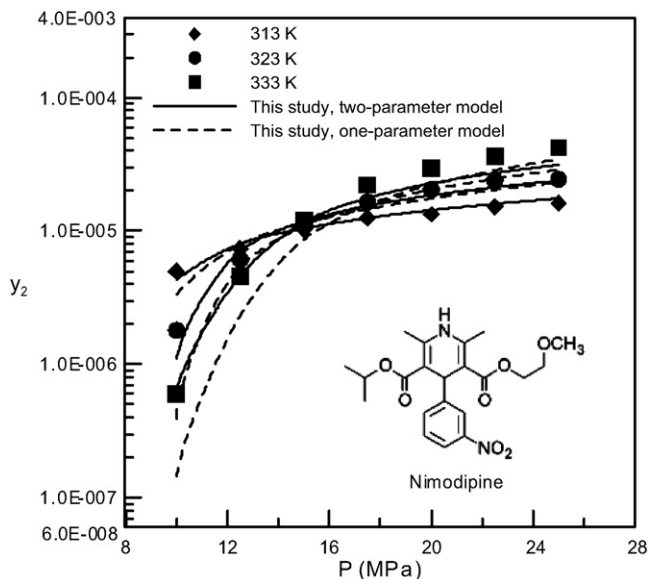


Fig. 3. Comparison of the experimental and calculated solid solubilities of nimodipine (2) in supercritical CO₂ (1) at various temperatures.

two- or one-parameter model were 11.5% and 25.7%, respectively. It is depicted that the single parameter model yielded relatively larger deviation than that from the dual parameter approach, but still gave the acceptable saturated solid solubility at higher-pressure range.

The calculation results from this study were compared with those from the semi-empirical equations, as shown in Table 3. Empirical model I was the Chrastil equation [4] with three adjustable parameters. The ARDY from the Chrastil equation was 12.7% that was comparable to the solution model with two parameters in this study. Empirical model II was the Mendez-Santiago and Teja equation [5], also with three adjustable parameters. It resulted in the similar accuracy (an ARDY of 13%) as that from the Chrastil equation. Empirical model III [14–16] was a two-parameter equation that gave an ARDY of 37.3%. It is demonstrated that with the same number of parameters, the solution model of this study yielded better calculation accuracy. The solution model of this study could further be simplified to a single parameter equation without significant increase in calculation error.

One advantage of the solution model approach is that even the single parameter α can be generalized and the solid solubility can be predicted. This study extended our previous work [9] and correlated the single parameter α as a function of the physical properties of the solid pharmaceutical compounds. Fig. 4 shows a plot of the α values from the one-parameter model in Table 3 against the logarithm of ΔU_2 from Table 1 for each pure pharmaceutical component. A linear relationship was observed:

$$\alpha = 0.3579 - 0.3185 \ln \Delta U_2 \quad (11)$$

This correlation included many new solubility data of pharmaceutical compounds that were not considered in our previous work [9]. For the pharmaceutical systems in this study, the coefficient of determination (the r^2 -value) for the generalization of α from Eq. (11) was closer to unity ($r^2 = 0.88$) than that from

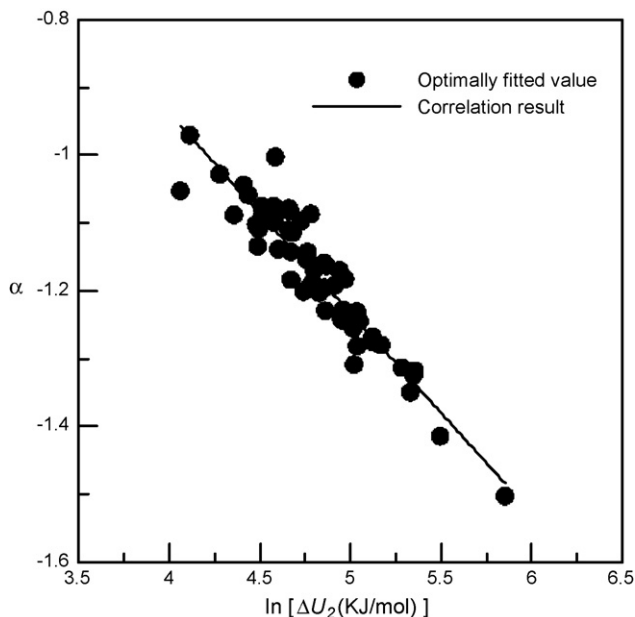


Fig. 4. Plot of the optimally fitted α values against the logarithm of internal energy change of vaporization for various compounds.

the previous generalization ($r^2 = 0.77$) [9]. Fig. 5 compared the calculated solubility of Amical-48 (an organic biocide) in supercritical CO₂ using Eq. (11) and the generalized equation in our previous work [9]. Since the model parameters were generalized, both calculations were predictive in nature. It was demonstrated that Eq. (11) was feasible in predicting the solid solubility, and yielded improved results to those from our previous study [9].

The prediction and comparison for solid solubility of bisacodyl (a laxative drug) in supercritical CO₂ is shown in Fig. 6. The ARDY for this prediction was 25.6% and was comparable to that from the single-parameter result shown in Table 3. The generalized correlation shown in Eq. (11) has further been

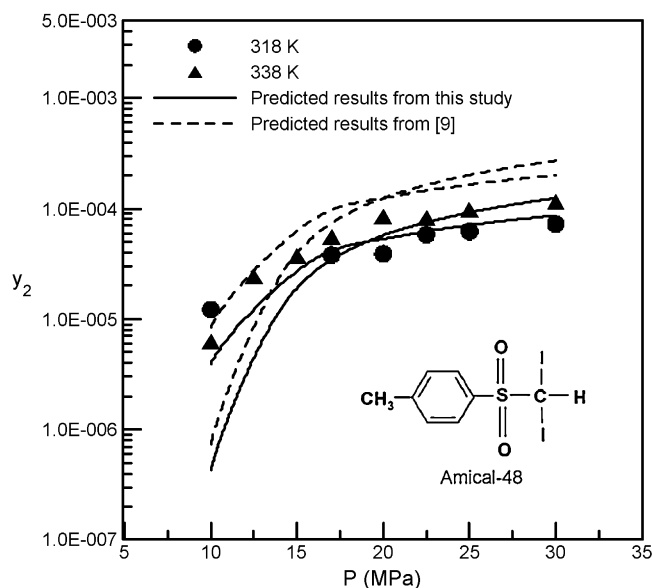


Fig. 5. Comparison for the predicted solid solubility of Amical-48 (2) in supercritical CO₂ (1) using two solution models.

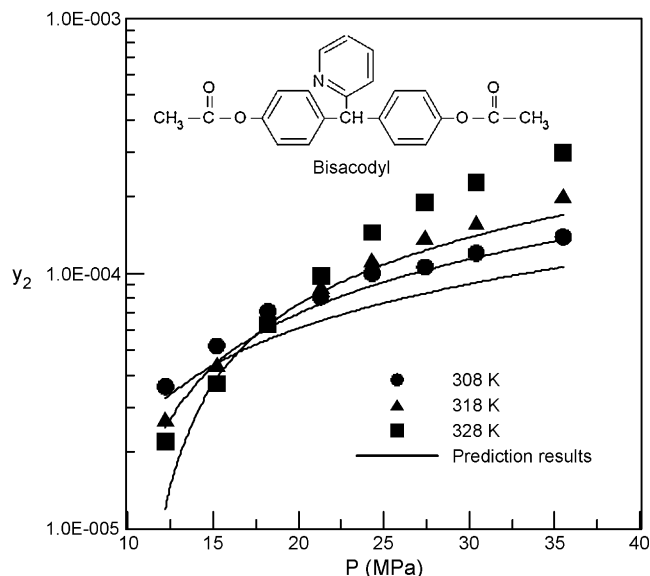


Fig. 6. Predicted solid solubility of bisacodyl (2) in supercritical CO₂ (1) using the generalized correlation in this study.

employed to predict the solid solubility of pharmaceutical compounds that were not originally included in Table 1. Fig. 7 shows the prediction results of flavone at different temperatures. The ARDY for this prediction was 49.2% where only limited estimated values of pure solid properties were used in this calculation. Due to the diversification and complexity of pharmaceutical compounds, the simple generalized correlation shown in Eq. (11) yielded acceptable accuracy (ARDY = 58%) for 30 systems listed in Table 1. The generalized equation from our previous study [9] could also predict the solid solubilities for half of the systems listed in Table 1 with an ARDY of 64%. Eq. (11) of this study, however, gave smaller peak deviations for pharmaceutical compounds. Further classification of pharmaceutical compounds according to specific functional groups, and

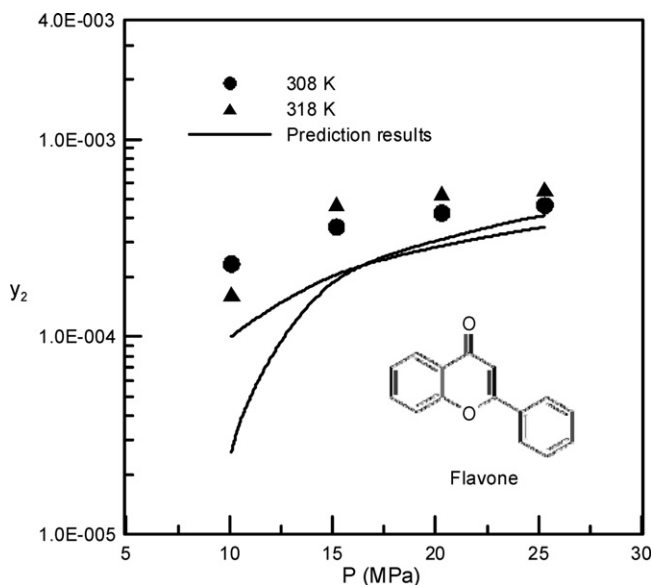


Fig. 7. Predicted solid solubility of flavone (2) in supercritical CO₂ (1) using the generalized correlation in this study.

generalized correlation equations for each group are required when more experimental data are reported. Applying Eq. (11), it also showed that the solid solubilities in supercritical CO₂ could be predicted within the correct order of magnitude for 70% of the systems listed in Table 1. These results demonstrated the superior prediction capability of the solution model where the semi-empirical methods shown in Table 3 did not provide.

Based on our calculation results, correlation for the solubilities of pharmaceutical compounds in supercritical CO₂ with only one adjustable parameter α is the best choice. If we have one or two experimental solid solubility data, the α value can be determined and further used in estimating the solubilities at other temperature or pressure conditions.

4. Conclusion

The solid solubilities of pharmaceutical compounds in supercritical CO₂ were correlated using the solution model. The correlation results were also compared with those from three commonly used semi-empirical equations. The solution model with dual or single adjustable parameter yielded absolute relative deviation (ARDY) in solid solubility of 16.5% and 23.7%, respectively. These results were better than those from the semi-empirical equations with the same number of adjustable parameters. The optimally fitted parameters of this study were further generalized and used for predicting the solubilities of pharmaceutical compounds in supercritical CO₂. The prediction results showed an ARDY of 58% for half of the systems in this study. Further improvement for the generalized correlation based on specific functional groups is required when more experimental data are reported.

Acknowledgement

The authors are grateful to the support of this study from the National Science Council, Republic of China.

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